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Efficiency of spanish mutual fund companies

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EFFICIENCY OF SPANISH MUTUAL FUND COMPANIES

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Efficiency of Spanish Mutual Fund Companies

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PhD Dissertation

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Motivation

The mutual fund industry has witnessed remarkable growth worldwide during the last decade. A significant effect on the savings patterns of the countries has been promoted by investment alternatives offered by mutual funds. Mutual fund management companies have suffered both great expansion and contraction periods in the same manner as other financial institutions. The need to analyse the efficiency of financial institutions has instigated an important line of research in the financial literature using alternative approaches. However, the emphasis of these studies has been typically on the banking and insurance industry rather than other specific companies related to the investment business. In addition, research on the mutual fund industry has been primarily focused on mutual funds rather than on their management companies. For that reason, this thesis aims to fill this gap in the empirical knowledge of the efficiency obtained by mutual fund management companies. In this sense, the thesis will contrast the efficiency level achieved by companies in the management and distribution of mutual funds to establish their core competence. The thesis analyses the potential interaction between this core competency and the overall efficiency results obtained by the fund company. This research will analyse the Spanish case, which is one of the most relevant fund industries in the Euro market. Therefore, to better understand the rationale behind the mutual fund industry, this study offers new insights into issues that have not been studied before in the European market.

This thesis develops an innovative model that considers different management stages of mutual fund companies, thereby overcoming the traditional dispute between the different approaches used in banking and insurance research. However, the variables considered in this model are unique and specific to mutual fund companies rather than a mere replication of the variables used in the banking and insurance literature.

Data envelopment analysis (DEA) is the methodology used to evaluate this model. This frontier method is one of the most widely used methods in studies of technical efficiency, but the specific technique used in this study is one of the most recent innovations in this

field and overcomes the problem of the high market concentration in the Spanish mutual fund industry. Specifically, the thesis is one of the first empirical applications of slack-based measure (SBM) variations, which represents an important and recent contribution to DEA methodology. Additionally, the persistence analysis of the efficiency achieved by the different management stages considered in the fund company model is revealed to be an original contribution. Moreover, non-parametric statistics based on contingency tables have been developed based on an innovative cluster analysis of efficiency scores.

This dissertation consists of two parts. The first part initially includes the basic concepts, a brief explanation of the basic DEA models, and a review of the most important applications to financial institutions. We then formulate the multi-management stage model and the unique set of fund industry-specific variables used in this research. Subsequently, the efficiency of Spanish mutual fund companies is evaluated using the SBM approach. Finally, this part concludes and summarises the major findings of the efficiency analysis.

The second part of the thesis reviews the major concepts of the variations in the original SBM approach and the more accurate empirical results of these SBM variants in the Spanish mutual fund industry. Subsequently, this part illustrates the persistence phenomenon to further determine whether any relevant factors may drive the efficiency results previously obtained. Finally, the second part presents conclusions and a summary of the major findings.

Finally, the thesis presents an overview of the main results obtained by the research and proposes several ideas for further research.

**PART I: THE EFFICIENCY OF
SPANISH MUTUAL FUND
COMPANIES: A SLACKS-BASED
MEASURE APPROACH**

1 Introduction

From the early 1990s through the middle of 2007, the mutual fund industry saw remarkable growth worldwide, reaching a maximum of 26.13 trillion U.S. dollars in December 2007 (source: European Fund and Asset Management Association, EFAMA). This successful expansion involved both the proliferation of Mutual Fund Management Companies and the inception of a large number of mutual funds, both of which placed greater demands on the business structure of the industry. Then, when the markets experienced the subprime mortgage financial crisis, the mutual fund industry suffered a significant decline in the total assets under management, especially in 2008, primarily driven by a fall in financial investments by households due to both increased risk aversion sparked by the high market volatility and a drop in confidence in financial instruments as a result of the crisis.¹ This market picture stopped the creation of new management companies and, in some cases, led to the merger or closure of other firms, thereby rearranging the competitive map of the industry. This drop in assets under management and the number of investors is explained not only by the poor image of mutual funds but also by a lack of confidence in the global markets where these funds allocate their primary investment positions. Unfortunately, despite all financial advice about diversification, equity investors had no place to hide in this crisis (Bartram and Bodnar, 2009; Ait-Sahalia et al., 2012). In many cases, this situation forced financial institutions to rethink their production structures and to create new products not only on their own initiative but also driven by new legislative and regulatory requirements imposed by supervisory authorities who were trying to restore confidence in financial markets (Bernanke, 2009; McCauley et al., 2012; Kirkpatrick, 2009).

More recently, mutual fund assets worldwide have been slowly returning to the successful figures achieved before the crisis, i.e., 28.4 trillion U.S. dollars managed in September 2012, representing over one third of world GDP. This industry employs skilled

¹ Mutual fund net assets worldwide decreased by approximately 27% in 2008. Europe and America experienced quite similar trends, while Asia and the Pacific showed even more dramatic decreasing rates. Source: EFAMA, International Statistical Releases.

labour, has spillover effects on other sectors and tax returns, and provides important liquidity to the financial system and wealth for retail and institutional investors. The analysis of the efficiency of mutual fund industry is quite similar to the extensive literature addressed in other financial sectors such as banking and insurance. If banks and insurance companies are more productive, then these financial sectors should obtain better performance, thereby offering new and safer products to their customers at lower prices. According to this argument, Mutual Fund Management Companies should work to offer a wide range of top-performing funds with diverse investment characteristics for different types of investors while keeping fees and expenses as low as possible.² The expansion of these appropriate standards of management should result in higher levels of overall efficiency in the mutual fund industry. It is worth noting that mutual fund activities reduce the exposure of banks to financial-services industry risk and increase scale economies and bank profitability, thereby improving the operating performance of banks (Gallo et al., 1996; Asaftei, 2008). In addition, the analysis of productivity differences across these companies in recent years could make it possible to identify the success or failure of management initiatives and might also highlight the different strategies undertaken by the companies during the financial crisis.

While extensive research has been devoted to productivity in financial institutions, as far as we know, only Zhao and Yue (2010) and Medeiros (2010) have studied the efficiency of mutual fund companies and pension fund companies, respectively. On one hand, Zhao and Yue (2010) examine the core efficiency of Chinese fund firms by analysing both the investment/research and the marketing/service subsystems. On the other hand, Medeiros (2010) analyses the changes in total productivity of a sample of Portuguese pension fund companies from 1994 to 2007 by means of a DEA-Malmquist index. A potential explanation for this scarce literature may be the difficulty of identifying specific variables for the appropriate evaluation of these companies without merely replicating the previous studies focused on financial institutions, such as banks and insurance companies. To

² Referring to this last subject, there has been a gradual reduction in the annual management fees of funds since the mid-90s, leading to a major impact on fund-manager income during the last financial crisis due to the drastic fall in assets managed. For instance, in the Spanish market, the management fees dropped from 1.65% in 1994 to 0.83% in 2009. Source: INVERCO, Impact of Subprime crisis in the IIC, June 2010.

develop appropriate evaluation models for mutual fund companies, it should be desirable to have a range of possibilities for specific industry variables that would complement those models analysed in other financial sectors. Therefore, these fund industry-specific proposals should contain relevant management inputs/outputs for mutual fund companies instead of merely using the general approach previously considered in banking and insurance.

This study fills this gap in the financial literature and aims to shed additional light by analysing the efficiency of mutual fund companies in Spain, which is one of the most relevant fund industries in the Euro market. The important market concentration of Spanish mutual fund companies is the most challenging feature when obtaining an appropriate evaluation for this industry. That is, the coexistence of a few, very large and well-diversified Mutual Fund Management Companies together with a huge number of small managers specialised in fund strategy per sector and/or geographical area makes it difficult to obtain appropriate evaluations of the industry as a whole because of the striking differences between competitors. Therefore, a question arises as to the selection of an accurate methodology and management variables to appropriately analyse so heterogeneous a set of Spanish mutual fund companies.

To conduct this analysis, we apply Data Envelopment Analysis (DEA), which has been one of the most popular methods over the last decades for evaluating efficiency in the financial industry (e.g., Berg et al., 1991; Berg et al., 1993; Schaffnit et al., 1997; Mlima and Hjalmarsson, 2002; Cummins et al., 2004; Casu et al., 2004; Cummins and Xie, 2008; Cummins et al., 2010; Holod and Lewis, 2011)³, and we examine the performance of institutional portfolios as an alternative approach to the traditional performance measures; portfolio performance works with the functional relationships between return and risk associated with behavioural assumptions (e.g., Murthi et al., 1997; Basso and Funari, 2001; Gregoriou et al., 2005; Eling, 2006; Lozano and Gutiérrez, 2008a, 2008b).

³ Some other studies are Drake and Howcroft (1994); Yeh (1996); Thompson et al. (1997); Athanassopoulos (1997); Sherman and Rupert (2006); Tortosa-Ausina et al. (2008); Chen et al. (2009); Eling and Luhn (2010).

This first part of study employs an original model and a unique set of fund industry-specific variables that complement the traditional models in banking and insurance, thereby allowing for an accurate and comprehensive evaluation of the overall efficiency of Mutual Fund Management Companies.

We used this innovative approach to address a number of questions regarding the efficiency of Mutual Fund Management Companies in the Spanish market and to further discuss the implications of the results obtained in this analysis. What are the key management stages driving the efficiency of a Mutual Fund company? Is efficiency robust across the different management stages within a mutual fund company? How does scale affect the efficiency results?

This first part is organised as follows: Section 2 provides a review of the early DEA literature, a brief explanation of the basic models, a short review of major contributions to efficiency in financial institutions, and a discussion of the two most popular approaches. Section 3 describes the proposed theoretical model and the variables used in our analysis. Section 4 illustrates the data, the empirical analysis and results, the influence of the variable-returns-to-scale, and robustness analyses. Finally, Section 5 concludes and summarises the primary findings.

2 Background

2.1 Basic Concepts

In this section, we provide a brief overview of the conceptual framework that underpins efficiency and productivity measurement. Essentially, the productivity of an organisation is defined as the ratio of the outputs that it produces to the inputs that it uses.

$$productivity = output/input \quad (I.1)$$

If the production unit uses a single input to produce a single output, the efficiency is easy to calculate, but it is more likely that the unit uses multiple inputs to produce multiple outputs. Both inputs and outputs must be combined in an economically reasonable way to maintain the productivity ratio of two scalars. Differences in production technology, differences in the efficiency of the production process, and differences in the settings in which production occurs are all potential issues in an industry analysis.

The comparison between the observed values of outputs and inputs (the production unit efficiency) can take the form of the ratio of the maximum potential output obtainable from the inputs or the ratio of the minimum potential input required to produce the outputs. The optimal result is defined in terms of production possibilities, and the efficiency is technical. Economic efficiency is obtained through the comparison of the optimal revenue, cost, profit or any other data that consider an appropriate quantity and price, and it is defined as the behavioural goal of the production unit.

Tangen (2005) has provided a useful description of the terms “productivity”, “efficiency” and “effectiveness”, which are often interchangeable but are quite distinct from each other. According to this study, productivity is closely related to the use of resources, meaning that a company's productivity is reduced if its resources are not

properly used. Secondly, productivity is also strongly connected to the creation of value. Thus, high productivity is accomplished when activities and resources in the manufacturing transformation process add value to the produced outputs. Because productivity is the productive capability of the resources consumed in organisations, it can be measured for each production resource separately, i.e., single factor productivity, as well as for all resources jointly.⁴

With respect to the concepts of efficiency and effectiveness, Tangen (2005) states that both concepts may be confused with productivity.⁵ Efficiency, in an organisational context, is related to the utilisation of inputs during the transformation process; effectiveness is concerned with the correctness and enhancement of the output, i.e., higher quantity and/or quality of output. It is important to note that productivity, efficiency, and effectiveness are relative concepts because it cannot be said to be increased or decreased unless a comparison is made with either variations from a "standard" at a certain point in time (which can be based on, for instance, a competitor or another production unit) or from changes over time.

The productivity ratio is an indicator of the efficiency with which a firm converts its resources (inputs) into finished goods or services (outputs). An increase in productivity can firstly be driven by producing more output with the same level of input (maximum potential output). Productivity can also be increased by producing the same output with fewer inputs (minimum potential input), or it can be a combination between both approaches. One of the primary problems when measuring productivity is the identification of the most appropriate inputs and outputs and determining how they will be measured.⁶ In Figure I-1 shows a scheme for the relationship between these terms.

⁴ This measure is known as the total productivity factor (TPF).

⁵ Saari (2006) states that efficiency is a general concept related to economic activity, and that productivity and profitability are typically specific concepts of efficiency.

⁶ Any of the traditional factors of production — land, labour, or capital — can be used as the denominator for the ratio, although productivity calculations are rarely made for land or capital in practice because their capacity is difficult to measure. Labour is in most cases easily quantified — for instance in R&D, where workers are engaged in developing a particular product.

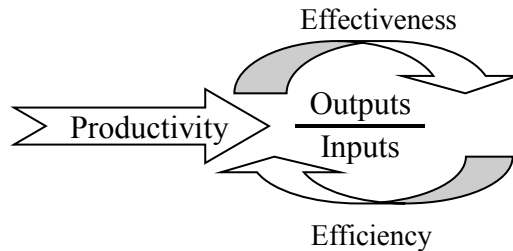


Figure I-1 Relationships between productivity, efficiency and effectiveness

Certainly, productivity is based on both efficiency and effectiveness; therefore, it is possible to talk in terms of productive efficiency when the orientation model is based on possible changes in the inputs and in terms of productive effectiveness when the orientation model is based on the possible changes in the outputs. Basically, one should select the orientation according to the managers' ability to control the inputs or/and outputs. As discussed in the next subsection, there are models that do not consider any orientation, which may be a useful approach when it is not clear what the managers can better control and influence. In any case, most of the studies over the last 30 years have referred to efficiency rather than productivity or performance (see Emrouznejad et al., 2008).⁷

A simple commonly accepted definition is that efficiency refers to how firms are performing relative to the existing technology in an industry, whereas productivity refers to the evolution of technology over time (Cummins and Weiss, 2001). For this reason, in this first part we refer to efficiency, but considering the issues clarified above

Moreover, it is essential to be familiar with the concept and utility of the production function (production frontier). The production function is basically the total product P , related to the labour L , capital C , and terrain T , along with the other inputs that are combined to produce it.

$$P = f(L, C, T \dots) \quad (\text{I.2})$$

⁷ Performance is another term extensively used in the literature. Performance is the umbrella term for excellence in financial studies and includes profitability and productivity as well as other non-cost factors such as quality, speed, delivery and flexibility.

The technological relationship is expressed in the function. The function expounds the maximum output obtainable at the existing state of technological knowledge from given quantities of factor inputs. In other words, a production function is a set of recipes or techniques for combining inputs to produce outputs, and the goal must be to maximise outputs or minimise inputs. Production functions are related to the level of the individual firms and the macro economy at large. At the micro level, economists use production functions to generate cost functions and input demand schedules for the firm. The renowned profit-maximising conditions of optimal factors derive from these micro-economic functions. At the level of the macro economy, analysts use aggregate production functions to explain the determination of factor income shares and to specify the relative contributions of technological progress to the expansion of factor supplies to economic development.

Figure I-2 shows a simple production process in which a single input is used to produce a single output. The line $0F'$ denotes a production frontier that defines the relationship between the input and the output. Firms in this industry that operate on the frontier are technically efficient, and those that are below the frontier are inefficient. Point B represents an inefficient point, while points A and C represent efficient points. In other words, the distance from one firm to the frontier identifies the degree of inefficiency, and, therefore, any improvement in the efficient use of inputs will result in a movement toward the frontier (e.g., the optimisation of capital and technological resources or improvements in organisational practices). A firm operating at point B could technically increase the output quantity to the level of point C without more input.

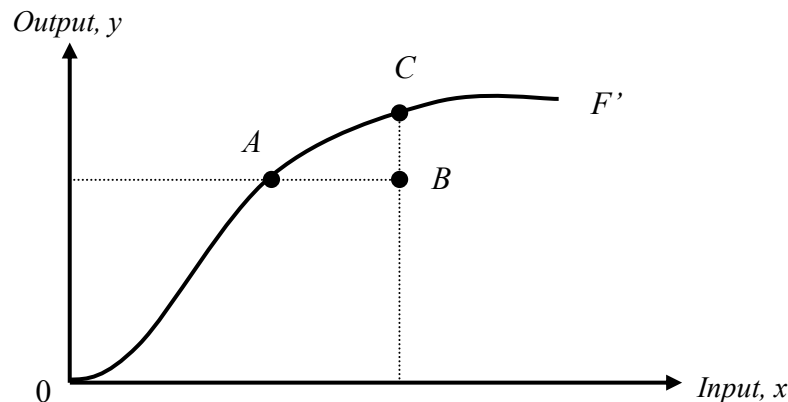


Figure I-2 Production frontier and technical efficiency

2.2 Basic DEA Models

Farrell (1957) made some important contributions to the issue of efficiency and productivity, specifically speaking of “productive efficiency”.⁸ This author states that there has been a total neglect of the theoretical side of efficiency measurement because for a long time, it was considered adequate to measure the average productivity specifically of labour and to use it as a measure of efficiency. This measure is patently unsatisfactory for Farrell because it ignores all other inputs aside from labour. Farrell (1957) proposes a satisfactory measure of productive efficiency for U.S. agricultural production. This measure takes into account all inputs and yet avoids index number problems. One of the most important features of this method is the distinction between technical efficiency and allocative efficiency. The former measures the firm’s success in producing maximum output from a given set of inputs, while the latter, called price efficiency by Farrell (1957), measures the success in choosing an optimal set of inputs. These two measures are then combined to afford a measure of total economic efficiency, called overall efficiency by Farrell (1957).

Measures in decision making efficiency were initially applied to evaluate public programs, where each program had a collection of decision making units (DMUs) with common inputs and outputs and focused on decision making rather than profits.⁹ For instance, in an educational program, the efficiency of various schools (DMUs) was measured by reference to outputs involving standard education categories such as cognitive or affective skills. The inputs could similarly range between fairly easy-to-measure quantities such as the number of teachers or the time spent in program activities by community leaders or parents. The measure of efficiency was obtained as the maximum of the ratio of weighted outputs to weighted inputs, subject to the condition

⁸ There were other interesting statistical contributions before Farrell (1957). We could mention Cobb and Douglas (1928), who offered what is commonly known as the Cobb-Douglas production function. These authors describe the relationship between the time series for manufacturing output, labour input, and capital input. Their equation displayed constant returns to scale, assumed unchanged technology, and omitted land and material inputs. Malmquist (1953) proposed total factor productivity (TFP), which is a measure of the changes in total output relative to inputs as well as the distance function approach. TFP is also a productivity measure that combines labour and capital as inputs.

⁹ Fried et al. (2008) presents a summary of the primary empirical applications in public areas such as health care, transportation, among others.

that similar ratios for each DMU should be less than or equal to unity (Charnes et al., 1978).

Based on Farrell (1957), Charnes et al. (1978) developed the well-known “Data Development Analysis” (DEA). This method identifies the production frontier as mentioned in the paragraph above; that is, it is necessary to find the maximum set of outputs that can be obtained from a given set of inputs. This frontier is a function that envelops or is restricted to the data sample. Essentially, DEA aimed at assessing the relative performance of a number of comparable DMUs, which may be within a company or across different companies within an industry.

There are many areas of application for DEA, from public programs to private financial studies. There are also many different techniques for implementing DEA and a wide range of different factors (inputs and outputs) depending on what is being evaluated or measured. In financial institutions, these efficiency methods are useful in a variety of contexts, from agency problems to control methodologies for regulators and policy makers. The application of DEA to mutual funds has usually been focused on the excess mean returns versus standard deviation and other risk measures. In most DEA models, it is common to find that the scalar efficiency measure is between zero (poor performance) and one (good performance) after applying a linear programming model, i.e., the efficiency score is the distance between the assessed DMU and a composite benchmark or target DMU that dominates it and lies on the efficient frontier (Cooper et al., 2000).

Schmidt (1985) classified the DEA as a non-parametric approach. Efficiency studies can be divided into those that measure technical efficiency using non-parametric techniques such as DEA and FDH (free disposal hull) and those that measure economic efficiency using parametric approaches such as SFA (stochastic frontier analysis), TFA (thick frontier approach) and DFA (distribution free approach).¹⁰ The parametric

¹⁰ The Free Disposal Hull (FDH) approach is a more flexible model that is considered to be a special case of DEA, where the points on lines connecting the DEA vertices are not included in the frontier. The econometric stochastic frontier analysis (SFA) is a method that assumes a two-part or composed error term. Efficiency is

methods adopt a specific functional form of the cost, profit or production functions. In contrast, the non-parametric methods do not place any restrictions on the functional form of the relationship between inputs and outputs. In the case of financial institutions, it is not a simple task to specify this functional relationship in the production process, which explains why non-parametric methods are predominantly applied for calculating the efficiency of financial institutions.¹¹

There are several types of DEA, but the two models most extensively applied are the CCR model proposed by Charnes et al. (1978), which has an input orientation and assumes constant returns to scale (CRS), and the BCC model proposed by Banker et al. (1984), which considers variable returns to scale (VRS). Since these well-known studies, other papers have considered alternative sets of assumptions in recent years: the Pareto-Koopmans model by Charnes et al. (1985), the Russell measure by Färe and Lovell (1978), the slacks-based measure (SBM) introduced by Tone (2001), and the recent variants of the SBM approach (Tone, 2010). These two latter papers are the methodological basis of this thesis.

Let us briefly review the basic ideas behind the most popular DEA models. We begin with the input-oriented CCR model because this approach was the first to be widely applied in the DEA literature.

Based on the explanations of Cook and Seiford (2009), we consider a set of n DMUs, with each DMU j , ($j = 1, 2, 3, \dots, n$) using the same m inputs x_{ij} ($i = 1, 2, 3, \dots, m$), possibly in different amounts, and producing the same s outputs y_{rj} ($r = 1, 2, 3, \dots, s$), also possibly in different amounts. If the prices or multipliers \bar{u}_r and \bar{v}_i related to outputs r and inputs i , correspondingly, are known, then using the conventional

assumed to follow an asymmetric distribution, usually half normal, while the random error is assumed to follow a standard symmetric distribution. The thick frontier approach (TFA) compares the average efficiencies of groups of firms instead of estimating the frontier edge. The distribution free approach (DFA) employs the average residuals of the cost function estimated using panel data to construct a measure of frontier-efficiency cost. DFA does not impose a specific form on the distribution of efficiency but assumes that there is a core efficiency or average efficiency for each firm that is constant over time. Lovell (1993) provides an excellent introduction to the different methods used over the last 40 years.

¹¹ These issues will be further discussed in the next sections.

benefit/cost concept, the efficiency of DMU_j could be expressed as the ratio of weighted outputs to weighted inputs, i.e.,

$$\sum_r \bar{u}_r y_{rj} / \sum_i \bar{v}_i x_{ij} \quad (\text{I.3})$$

If the multipliers are unknown, Charnes et al. (1978) proposed deriving appropriate multipliers for a given DMU by solving a particular non-linear programming problem. Thus, the fractional programming problem, with an input-oriented model would be

$$\begin{aligned} \max \theta &= \sum_r u_r y_{ro} / \sum_i v_i x_{io} \\ \text{s.t.} \\ \sum_r u_r y_{rj} - \sum_i v_i x_{ij} &\leq 0, \quad \text{for all } j \\ u_r, v_i &\geq \varepsilon, \quad \text{all } r, i. \end{aligned} \quad (\text{I.4})$$

where y_{ro} is the output vector, x_{io} is the input vector of the DMU evaluated, and ε is a non-archimedian value designed to enforce strict positivity on the output and input multipliers (u_r and v_i).¹² The objective of this problem is to minimise inputs while producing at least the given output levels, i.e., to obtain values (weights) for u_r and v_i that maximise the ratio.¹³ The constraints mean that this ratio should not exceed 1 for every DMU. By virtue of these constraints, the optimal objective value θ^* is at most 1.

Problem (I.4) can be converted to an equivalent linear programming model applying the theory of fractional programming (See Charnes and Cooper, 1962; Charnes et al.,

¹² In the original paper of Charnes et al. (1978), the authors simply restricted the variables to be non-negative ($\varepsilon=0$); the imposition of a strictly positive lower limit ($\varepsilon > 0$) was added by Charnes et al. (1981).

¹³ We could invert this ratio and solve the corresponding output-oriented model, which attempts to maximise outputs while using no more than the observed amount of any input.

1978), making the following changes in the variables: $u_r = tu_r$ and $u_i = tu_i$, where $t = (\sum_i v_i x_{i0})^{-1}$. Thus, we have the *multiplier* model:

$$\begin{aligned}
 & \max \theta = \sum_r u_r y_{r0} \\
 & \text{s.t.} \\
 & \sum_r v_r x_{i0} = 1 \\
 & \sum_r u_r y_{rj} - \sum_i v_i x_{ij} \leq 0, \quad \text{all } j \\
 & u_r, v_i \geq \varepsilon, \quad \text{all } r, i.
 \end{aligned} \tag{I.5}$$

and finally, this multiplier model can be expressed with a real variable θ and a non-negative set of variables λ ($\lambda_1, \dots, \lambda_n$), making it equivalent to the linear programming problem; we now have the *envelopment* or primal problem:

$$\begin{aligned}
 & \min \theta_o - \varepsilon \left(\sum_r s_r^+ + \sum_i s_i^- \right) \\
 & \text{s.t.} \\
 & \theta_o x_{i0} = \sum_j \lambda_j x_{ij} + s_i^-, \quad i = 1, \dots, m \\
 & y_{r0} = \sum_j \lambda_j y_{rj} - s_r^+, \quad r = 1, \dots, s \\
 & \lambda_j, s_i^-, s_r^+ \geq 0, \quad \text{all } i, j, r \\
 & \theta_o \text{ unconstrained}
 \end{aligned} \tag{I.6}$$

Note that the optimal solutions show the occurrence of both excess inputs and shortfalls in outputs, which are called *slacks* (s_i^-, s_r^+). If $\theta < 1$, the objective DMU is inefficient. If an optimal solution satisfies $\theta = 1$ and $s^+ = 0$ and $s^- = 0$, it is called CCR-efficient. These conditions taken together are also described as Pareto-Koopmans efficient and denote that a DMU is efficient if and only if it is not possible to improve any input or output without worsening some other input or output.

Then, the constraint space of equation (I.6) defines the following *production possibility set* P :

$$P = \left\{ (x, y) \mid x \geq \sum_j X_j \lambda_j, \quad y \leq \sum_j Y_j \lambda_j, \quad \lambda_j \geq 0 \right\} \quad (\text{I.7})$$

Any semi positive linear combination of activities in P belongs to P , and it assumes constant returns-to-scale. The CCR model appears to be appropriate when the entire set of firms is operating at an optimal scale. However, imperfect competition, government regulations, constraints on finance, etc., may cause a DMU to not operate at optimal scale. The BCC DEA model by Banker et al. (1984) suggests adjusting the CCR DEA model to account for variable returns to scale (VRS). According to Coelli et al. (2005), the use of constant returns to scale (CRS, named the CCR model here) when not all DMUs are operating at the optimal scale results in measures of *technical efficiency* that are mistaken for *scale efficiencies*. Thus, the use of the VRS specification allows for appropriate calculations of *technical efficiency*.

Banker et al. (1984) simply added a convexity constraint on λ_j . That is, the BCC model has its production frontiers P now spanned by the convex hull of all existing DMUs, namely $\sum_j \lambda_j = 1$ for the CRS linear programming problem (equation I.6), to provide the following model:

$$\min \theta_o - \varepsilon \left(\sum_i s_i^- + \sum_r s_r^+ \right)$$

s.t.

$$\theta_o x_{io} = \sum_j \lambda_j x_{ij} + s_i^-, \quad i = 1, \dots, m$$

$$y_{ro} = \sum_j \lambda_j y_{rj} - s_r^+, \quad r = 1, \dots, s$$

$$\sum_j \lambda_j = 1$$

$$\lambda_j, s_i^-, s_r^+ \geq 0, \quad \text{all } i, j, r$$

$$\theta_o \text{ unconstrained} \tag{I.8}$$

This model essentially ensures that the projected point of the inefficient DMU on the BCC DEA frontier is a convex combination of the observed DMUs. In this situation, the λ -weights do not sum to a value less than or greater than one.¹⁴

In Figure I-3, the difference between both production sets can be observed. Efficiency estimates are generally higher with BCC, and rankings can differ in the two specifications. Note that equation (I.8) differs from equation (I.6) in that it has the additional convexity constraint on λ_j , $\sum_j \lambda_j = 1$. The frontier BCC determines the different returns to scale: the portion of the frontier from point *A* up to (but not including) *B* establishes increasing returns to scale; point *B* has a constant returns to scale¹⁵; and all the points on the frontier BCC to the right of *B*, in this case only point *D*,

¹⁴ The BCC model aims to somehow determine a benchmark between the points from an inefficient DMU alongside DMUs of a similar size.

¹⁵ Point *B* is the point of maximum possible productivity, the point representing the technically optimal scale. Operation at any other point on the production frontier results in lower productivity. The DMUs *A* and *D* are

represent decreasing returns to scale. As with the CCR model, a DMU is BCC-efficient in the VRS model if the solution of equation (I.8) obtains a value of $\theta_o = 1$ and all the slacks s_i^- and s_i^+ have a zero value. Hence, any CCR-efficient DMU can also be BCC-efficient. However, it is possible to project the inefficient point C to the frontiers (points C' and C''), and we can see that it is a horizontal projection; an output-oriented model would involve a vertical projection from DMU C up to the frontier.

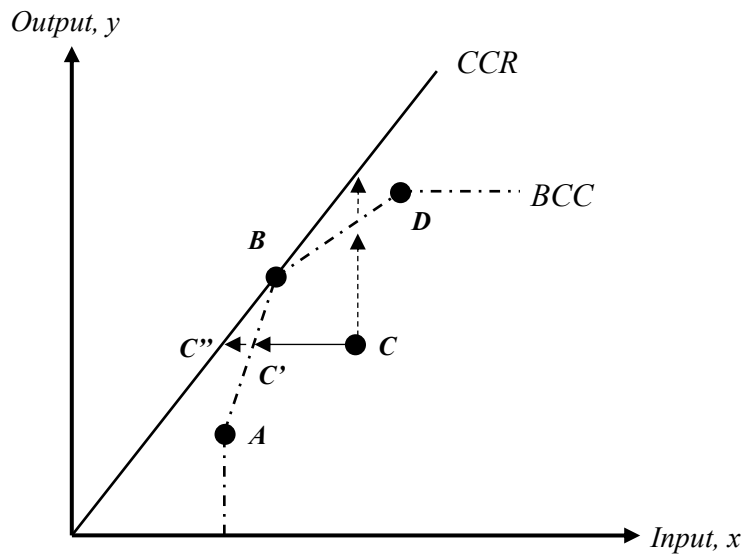


Figure I-3 Differences of returns to scale

Charnes et al. (1985) proposed combining both models and formulated the Pareto-Koopmans (PK) model. In this approach, the inputs are proportionally reduced while the outputs remain fixed, and the outputs are proportionally increased while the inputs remain unchanged. This additive model has the same convex production as the BCC model, and a DMU is PK efficient if and only if it is VRS-efficient and all of its slacks are equal to zero. Specifically, the model is

efficient but could have higher or lower productivity, i.e., at these points, the firm may still be able to improve its productivity by exploiting scale economies.

$$P_o = \max \sum_r s_r^+ + \sum_i s_i^-$$

s.t.

$$x_{io} = \sum_j \lambda_j x_{ij} + s_i^-, \quad i = 1, \dots, m$$

$$y_{ro} = \sum_j \lambda_j y_{rj} - s_r^+, \quad r = 1, \dots, s$$

$$\sum_j \lambda_j = 1$$

$$\lambda_j, s_i^-, s_r^+ \geq 0, \quad \text{all } i, j, r \quad (\text{I.9})$$

This model has no scalar measure or efficiency ratio per se, unlike the CCR and BCC models. The additive PK model can discriminate between efficient and inefficient DMUs by the existence of slacks, but it does not measure the depth of inefficiency, which is similar to θ_o in the CCR and BCC models.

To address the above shortcoming, Tone (2001) defines the slacks-based measure (SBM), which is invariant to the units of measurement and is monotonic increasing in each input and output slack.¹⁶ This scalar measure directly addresses the input excesses and the output shortfalls of the decision making unit (DMU) evaluated.

¹⁶ The Russell measure proposed by Färe and Lovell (1978) and later revised by Pastor et al. (1999) is considered to be equivalent to SBM according to Cook and Seiford (2009).

Tone (2001) states that the following properties should be considered when designing the efficiency measures:

- (P1) *Invariant units*: The measure should be invariant with respect to the units of data.
- (P2) *Monotone*: The measure should be monotone decreasing in each slack in input and output.
- (P3) *Translation invariant*: The measure should be invariant under a parallel translation of the coordinate system applied.
- (P4) *Reference set dependent*: The measure should be determined only by consulting the reference set of the DMU concerned.¹⁷

Tone (2001) maintains that the SBM model satisfies the properties (P1), (P2) and (P3). He also states that because the beginning DEA employs piece-wise linear efficient frontiers whose area is spanned by efficient DMUs and because an inefficient unit is “inefficient” with respect to the DMUs in its reference set, the measure of efficiency should be determined by the reference set dependent values as in the CCR and BCC models. In addition, the measure should not be influenced by extreme values, e.g., the minimum and the maximum of the data set.

Based on the CCR and BCC models, which consider that n DMUs have inputs $X = (x_{ij}) \in R^{m \times n}$ and outputs $Y = (y_{rj}) \in R^{s \times n}$ ¹⁸, the SBM model initially assumes that the data set is positive, i.e., $X > 0$ and $Y > 0$, although this assumption can be relaxed. Thus, production possibility P is defined as in the former expression (I.7):

$$P = \left\{ (x, y) \mid x \geq \sum_j X_j \lambda_j, \quad y \leq \sum_j Y_j \lambda_j, \quad \lambda_j \geq 0 \right\} \quad (\text{I.7})$$

being λ_j a non-negative vector in R^n , constrains λ , such as $\sum_j \lambda_j = 1$.

¹⁷ The reference set R_o for an inefficient DMU_o is defined as $R_o = \{j \mid \lambda_j^* > 0\}$; ($j = 1, 2, 3, \dots, n$)

¹⁸ X is a matrix with n columns (DMUs being analysed) and m rows (inputs). Y is a matrix with n columns (DMUs being analysed) and s rows (outputs).

A certain DMU (x_o, y_o) is described as

$$x_o = X\lambda + s^-, \quad (\text{I.10})$$

$$y_o = Y\lambda - s^+, \quad (\text{I.11})$$

with $\lambda \geq 0, s^- \geq 0$ and $s^+ \geq 0$. The vectors $s^- \in R^m$ and $s^+ \in R^s$ indicate the *input excesses* and *output shortfalls* of expressions (I.10) and (I.11), respectively,¹⁹ as we have seen the so-called *slacks*. From the conditions $X > 0$ and $\lambda \geq 0$, it holds that

$$x_o \geq s^-. \quad (\text{I.12})$$

Using s^- and s^+ , Tone (2001) defines an index p as follows:

$$p = \frac{1 - (1/m) \sum_{i=1}^m s_i^- / x_{io}}{1 + (1/s) \sum_{r=1}^s s_r^+ / y_{ro}} \quad (\text{I.13})$$

According to Cooper et al. (2000), expression (I.13) could be interpreted as the ratio of mean input and output mix inefficiencies, that is, the relative proportional reduction rate of inputs and the relative proportional expansion of outputs. Here, it can be verified that p satisfies the properties (P1) = units invariant because the numerator and denominator are measured in the same units for every item and (P2) = monotone because an increase in either s^- or s^+ (the rest remains constant) will decrease p in a strictly monotone manner. Additionally, from expression (I.12), it holds that

$$0 < p \leq 1 \quad (\text{I.14})$$

¹⁹ s^- and s^+ are points in the n dimensional vector space R^n .

Therefore, the final fractional program in λ , s^- , s^+ of Tone (2001) to estimate the efficiency under constant returns to scale (CRS) of (x_o, y_o) is as follows:

$$\begin{aligned}
 & \text{[SBM]} \\
 & \text{minimise} \quad p = \frac{1 - (1/m) \sum_{i=1}^m s_i^- / x_{io}}{1 + (1/s) \sum_{r=1}^s s_r^+ / y_{ro}} \\
 & \text{s. t.} \\
 & \quad x_o = X\lambda + s^-, \\
 & \quad y_o = Y\lambda - s^+, \\
 & \quad \lambda, s^-, s^+ \geq 0
 \end{aligned} \tag{I.15}$$

Then, Tone (2001) multiplies a scalar variable $t (> 0)$ by both the denominator and the numerator of expression (I.15). This transformation is performed because the goal is to minimise the numerator, and it does not cause changes in p .

$$\begin{aligned}
 & \text{[SBMt]} \\
 & \text{minimise} \quad \tau = t - \frac{1}{m} \sum_{i=1}^m t s_i^- / x_{io} \\
 & \text{s. t.} \\
 & \quad 1 = t + \frac{1}{s} \sum_{r=1}^s t s_r^+ / y_{ro}, \\
 & \quad x_o = X\lambda + s^-, \\
 & \quad y_o = Y\lambda - s^+, \\
 & \quad \lambda \geq 0, \quad s^- \geq 0, \quad s^+ \geq 0, \quad t > 0
 \end{aligned} \tag{I.16}$$

The model given above is a nonlinear programming model because it contains the nonlinear term $t s_r^+ (r = 1, \dots, s)$. However, Tone (2001) transforms it into a linear program using the Charnes-Cooper transformation, as follows. The transformation is defined as $S^- = t s^-$, $S^+ = t s^+$ and $\Lambda = t \lambda$.

Thus, $[SBMt]$ becomes the following linear program in t , S^- , S^+ and Λ :

[LP]

$$\text{minimise} \quad \tau = t - \frac{1}{m} \sum_{i=1}^m S_i^- / x_{i0}$$

s. t.

$$1 = t + \frac{1}{s} \sum_{r=1}^s S_r^+ / y_{r0},$$

$$tx_o = X\Lambda + S^-,$$

$$ty_o = Y\Lambda - S^+,$$

$$\Lambda \geq 0, S^- \geq 0, S^+ \geq 0, t > 0 \quad (\text{I.17})$$

Allow the optimal solution of [LP] to be

$$(\tau^*, t^*, \Lambda^*, S^{-*}, S^{+*}).$$

This optimal solution is defined by

$$p^* = \tau^*, \lambda^* = \Lambda^* / t^*, s^{-*} = S^{-*} / t^*, s^{+*} = S^{+*} / t^* \quad (\text{I.18})$$

Based on this optimal solution, Tone (2001) determines that a DMU (x_o, y_o) is *SBM-efficient* if $p^* = 1$. This condition is equivalent to $s^{-*} = 0$ (no input excesses) and $s^{+*} = 0$ (no output shortfalls) for any optimal solution.

For an SBM inefficient DMU (x_o, y_o) , it holds that $x_o = X\lambda^* + s^{-*}$, and $y_o = Y\lambda^* - s^{+*}$

This inefficient DMU (x_o, y_o) can be improved and become efficient by deleting the input excess and augmenting the output shortfall as follows:

$$x_o \leftarrow x_o - s^{-*}, \quad (\text{I.19})$$

$$y_o \leftarrow y_o + s^{+*}. \quad (\text{I.20})$$

Based on λ^* , Tone (2001) defines the reference set to (x_o, y_o) as the set of indices corresponding to positive λ_j^* s.²⁰ The *reference set* R_o is therefore

$$R_o = \{j | \lambda_j^* > 0, j = 1, \dots, n\} \quad (\text{I.21})$$

Now, (x_{io}, y_{ro}) can be expressed in terms of the reference set R_o as follows:

$$x_o = \sum_{j \in R_o} x_j \lambda_j^* + s^{-*} \quad (\text{I.22})$$

$$y_o = \sum_{j \in R_o} y_j \lambda_j^* - s^{+*} \quad (\text{I.23})$$

Tone (2001) adds some bounds on the slacks.²¹ This operation is called the *SBM-projection*. The projection of the DMU is obtained as follows:

$$\bar{x}_o = x_o - s^{-*} = \sum_{j \in R_o} x_j \lambda_j^* \quad (\text{I.24})$$

$$\bar{y}_o = y_o + s^{+*} = \sum_{j \in R_o} y_j \lambda_j^* \quad (\text{I.25})$$

SBM p^* depends only on s^{-*} and s^{+*} , i.e., the reference set dependent values. That is, p^* is not affected by values attributed to other DMUs that are not included in the reference set R_o .

In summary, this section has presented the basic ideas behind the most extensively used DEA models. Every DEA model allows us to analyse multiple outputs and multiple inputs without assuming functional forms between these factors, which is an important difference with respect to conventional regression-based methods. This section has also

²⁰ In the case of multiple optimal solutions, there may be different reference sets.

²¹ These bounds are subject to $s^- \leq s_0^{-B}$ and $s^+ \leq s_0^{+B}$, where the vector $s_0^{-B} (s_0^{+B})$ is the upper bound of input reduction (output enlargement) of the DMU and should be specified for each DMU.

highlighted that there are three basic orientation types: input-oriented, which focuses on reducing the input amounts while keeping the current output levels; output-oriented, which maximises output levels under the current inputs; and a third, based on both the additive and the SBM models and addressing excess inputs and output shortfalls simultaneously, i.e., the unoriented model. In any case, the efficiency determination for a set of DMUs will have the same goal as these methods. Table I-1 shows a recap of some of the important topics behind the basic DEA models treated briefly in this section.

It is important to note that further DEA models were developed in the subsequent years through multilevel DEA models, multiplier restrictions, and considerations regarding the status of variables and data variation (see Cook and Seiford, 2009).²² The refinement of these methodologies during the search for more appropriate evaluations is a challenging issue in the literature.

Table I-1 Comparison between basic DEA models (Cooper et al., 2000)

Model	CCR	BCC	ADD	SBM
Data X	Semi-p	Semi-p	Free	Semi-p
Data Y	Free	Free	Free	Free
Trans invariance X	No	No	Yes	No
Trans invariance Y	No	Yes	Yes	No
Units invariance	Yes	Yes	No	Yes
θ^*	[0,1]	[0,1]	No	[0,1]
Tech. or Mix	Tech.	Tech.	Mix	Mix
Returns to scale	CRS	VRS	C(V)RS	C(V)RS

Semi-p = semi positive, means nonnegative with at least one positive element in the data for each DMU. Free permits negative, zero or positive data. Tech. or Mix indicates whether the model measures use technical efficiency or mix efficiency.²³ CRS and VRS represent constant and variable returns to scale, respectively. The variable returns to scale of ADD and SBM depend on the added constraint $\lambda = 1$.

²² With respect to data variation, the subject of sensitivity analysis is of great interest as it can influence the efficiency status of DMUs. Among these issues, both ranking problems and the size problems of DMUs are important. One approach to the ranking problem is provided by the super efficiency model of Andersen and Petersen (1993). This model involves executing the CRS or VRS models, but under the assumption that the DMU being evaluated is excluded from the reference set. It can also be thought of as a measure of stability, evaluating changes over time for specific input data.

²³ According to Tone (2001), the SBM model (I.15) can be transformed into $p = \left(\frac{1}{m} \sum_{i=1}^m \frac{x_{io} - s_i^-}{x_{io}} \right) \left(\frac{1}{s} \sum_{r=1}^s \frac{y_{ro} + s_r^+}{y_{ro}} \right)^{-1}$, where the ratio $\frac{x_{io} - s_i^-}{x_{io}}$ evaluates the relative reduction rate of input i and, therefore, the first term corresponds to the mean proportional reduction rate of inputs or input mix inefficiencies. This interpretation is similar for the second term with regard to the outputs of the model. Therefore, SBM p can be interpreted as the ratio of mean input and output mix inefficiencies.

2.3 Applications to Financial Institutions

2.3.1 Context of Study

According to Cummins and Weiss (2001), frontier efficiency methods are useful in a variety of contexts for financial institutions. One primary use is for testing economic assumptions. These methods may explore the success of firms using different scenarios such as cost minimisation or profit maximisation under different economic assumptions. Other issues related to organisational form, distribution systems, corporate governance, and vertical integration have also been analysed by these frontier-methodologies.

A second significant application of frontier methodologies is to provide regulators and policymakers with answers to problems and advances in different industries or for the overall economy. For instance, frontier methodologies might be used to determine whether mergers and acquisitions in both banking and insurance industries have positive or negative consequences on the overall efficiency of the sector (in terms of price and quality of services).

A third application of frontier efficiency would be the comparison of economic efficiency between different countries. Several studies have addressed macroeconomic and social differences, which should allow multilateral agencies and governments to make better decisions.

A fourth application, and not the least important, is to warn management units about the strategies tracked by competitors. That is, frontier analysis can be used not only to understand the evolution of a company's productivity over time but also to compare the performance of a set of companies and to analyse the performance of different management units within a company. Frontier analysis can contribute by providing more meaningful evidence than conventional ratios and survey analyses.

2.3.2 Brief Literature Review

Over the last decades, the landscape of the financial sector has seen massive structural changes. The internationalisation and deregulation of markets (Edison et al., 2002; Fernández de Guevara et al., 2007; Salas and Saurina, 2003; Chen, 2007; King, 2012), technological advances, and European integration have changed the financial industry.²⁴ Financial institutions have been forced to diversify their business to remain in the market. For instance, banks have diversified into non-interest earning activities such as insurance and mutual fund sales, private banking and asset management (Goddard et al., 2007). This diversification could potentially entail changes in the efficiency and productivity of banks and in the diverse business units. That is why in recent decades, there have been many studies of efficiency in financial institutions that basically focus on banks, branches, savings and loan companies, and the insurance industry, leaving to assessing specific business units such as investment companies in themselves. This section provides a brief review of the seminal and most current frontier studies in financial institutions, particularly in banking and insurance firms.²⁵

In the banking area, Benston (1965) and Greenbaum (1967) are the first studies that address both cost analyses and banking structure to evaluate their returns to scale. However, not until the decades of the 1980s and 1990s were a great number of empirical studies on cost structure and efficiency in banking conducted. Initially, the most common inputs used in the DEA and FDH methodology were labour, machines, materials, buildings and borrowings, and on the side of the outputs we find incomes, short and long-term loans, number of branches, the opening of new accounts, and other banking services (Aly et al., 1990; Elyasiani and Mehdián, 1990, 1992; Berg et

²⁴ In the case of banks, the total sector assets in the five largest European economies (France, Germany, Italy, Spain and the UK) experienced an increase rate of 340% in nominal terms between 1985 and 2004 (Goddard et al., 2007).

²⁵ For our purpose, we only consider those studies that use non-parametric methodologies, such as DEA and FDH, and parametric models such as SFA, DFA and TFA.

al., 1991; Berg et al. 1993).²⁶ At the same time, we find several studies that apply parametric methodologies (Berger and Humphrey, 1991; Berger et al., 1993).²⁷

The treatment of deposits was one of the major discrepancies between both methodologies. The non-parametric approaches are usually based on the production or service provision approach and take deposits to be an output based on positive consumption of labour and materials. The parametric approaches are based on the asset approach and regard deposits as an input for the production of loans and other financial assets. This issue was further discussed by Berger et al. (1993) and then by Pastor et al. (1997). In this latter study, the efficiency levels of different European and US banking systems were compared and the deposits were taken as an output, i.e., this paper adopted a production approach.

Berger and Humphrey (1997) provide a review of early works on some banks and other financial institutions. This paper finds that efficiency rankings differ depending on which frontier approach is implemented and by how financial institution outputs are measured – “as a transaction-based flow, a stock of numbers of accounts, or a stock of value in these accounts”.²⁸

In the last decade we do not find major changes in the inputs and outputs used in banking studies. In essence, the changes have been linked to the use of other balance sheet values but within the same model approach and with the variables mentioned above. For instance, inputs such as risk measures, equity capital, reserves, cost of

²⁶ Some other studies are Fukuyama (1993), Grabowski et al. (1993), and Tulkens (1993).

²⁷ Some other studies are Bauer et al. (1993) and Pi and Timme (1993).

²⁸ Berger and Humphrey (1997) find that more than half of the studies use the non-parametrical approach, especially DEA, although they also conclude that “The efficiency estimates from non-parametric (DEA and FDH) studies are similar to those from parametric frontier models (SFA, DFA, and TFA), but the non-parametric methods generally yield slightly lower mean efficiency estimates and seem to have greater dispersion than the results of the parametric models”. This debate is still open because other authors consider that there are not significant differences in the results obtained by both approaches. Berger and Humphrey (1997) also identify the potential cost savings from maximising productive efficiency through the comparison of the results of the non-parametric and parametric approach in their large sample of studies. As is known, there are two aspects to productive (operational) efficiency: technical (x-efficiency) involves avoiding input waste by achieving the maximum possible output from a given set of inputs, and allocative efficiency involves selecting the most cost-effective mixture of inputs given an actual set of input prices.

capital, credit losses, and other price ratios, and outputs such as cash, fees and commissions, securities, contingent liabilities, net provisions, purchased funds, and others price ratios can be used (Lang and Welzel, 1996; Chaffai, 1997; Kumbhakar et al., 2002; Mlima and Hjalmarsson, 2002; Casu et al. 2004; Boning et al., 2005; Sherman and Rupert, 2006; Tortosa et al., 2008; Berger et al., 2009; among other studies).

Several studies were conducted collaterally in the insurance industry. The evidence is more recent but there have been few variations in the nature of inputs and outputs in relation to banking. Weiss (1991) uses worldwide surveys to compare productivity in the property-liability insurance industries in the U.S. market and four European countries, applying index numbers for total factor productivity. Later, Delhausse et al. (1995) and Diacon (2001), among others, analysed cross-country efficiency in the European Union using DEA methodology. Studies have also been conducted on Asian insurance companies applying DEA models since Boonyasai et al. (1999). The inputs and outputs included in these studies are summarised in a general overview²⁹ by Eling and Luhn (2010), who develop an additional proposal of study.³⁰

Moreover, in the last decade, many studies in the insurance industry have focused on analysing scope economies from the hypotheses derived by Berger et al. (2000). These authors take labour, materials, business services, and financial equity capital as inputs and use different output proxies related to risk-pooling and risk-bearing. They find that the conglomeration hypothesis dominates for some types of financial service providers and that the strategic focus hypothesis dominates for other types.³¹ Later, we find Cummins et al.'s (2004) study applying DEA and Greene and Segal's (2004)

²⁹ It is notable that more than 2/3 of the studies reviewed by Eling and Luhn (2010) applied DEA models.

³⁰ Eling and Luhn (2010) apply DEA and SFA by taking proxy variables from inputs such as labour and business services, debt capital, and equity capital and outputs such as non-life claims + additions to reserves, life benefits + additions to reserves, and investments.

³¹ The conglomeration hypothesis argues that owning and operating a broad range of services can add value from exploiting cost scope economies by sharing inputs in joint production. In contrast, proponents of the strategic focus hypothesis argue that firms can maximise value by focusing on core services and core competencies.

study using SFA.³² The inputs are similar in both papers: labour, business services, financial debt capital, equity capital, and other balance sheet variables. However, in Cummins et al. (2004), the outputs are the risk-pooling, risk-bearing, “real” financial services relating to insured losses and intermediations while in Greene and Segal (2004), the outputs are life policies, annuities, accident and health (A&H) and investment income. Later, Fenn et al. (2008) apply the SFA method to the European insurance industry, considering total capital and reserves, technical provisions, debt capital, and labour as inputs. On the outputs side, this research considers the product itself, thereby using proxy variables such as the insurer’s estimation of incurred claims (claims paid together with the reserves set against future payments).

The aforementioned literature included in this section indicates that labour in its several forms of calculation³³ is present in most research, and other inputs such as materials and physical capital, funding and equity capital are also common. However, different approaches are taken toward the outputs. In general, we find fee-based income, investment income, consumer and commercial loans, and branches as outputs. As mentioned in some paragraphs above, one of the major differences is the consideration of deposits because they can be found as an input or an output, depending on the efficiency approach considered (asset approach or production approach).

In the most innovative applications of frontier methods to the financial industry, it is worth noting that DEA models have been applied to the performance analysis of mutual funds and pension funds since the DPEI index originally developed by Murthi et al. (1997).³⁴ Interest in DEA methodology in this area has increased because DEA models do not require any functional form between inputs and outputs, while this is

³² Cummins et al. (2004) analyse the effect of organisational structure on efficiency in the Spanish insurance industry. In a later study, Cummins et al. (2010) use similar variables, but they are applied to the US insurance industry. Greene and Segal (2004) also analyse efficiency in the US life insurance industry.

³³ Some of the measures of labour found in the literature are administrative or agent labour, personnel expenses, the number of employees, man-hour expenses, and ratios such as personnel expenses divided by total assets, among other measures of labour.

³⁴ The DPEI index considers mutual fund returns as the only output and the standard deviation and transactional costs as inputs.

required by well-known classical performance indexes such as the Sharpe and Jensen ratios (See Basso and Funari, 2001). This type of application is different from that in the extensive existing literature of DEA empirical studies because the DMUs evaluated are financial portfolios rather than companies or business units. Choi and Murthi (2001) indicate that the Sharpe index is similar to a CRS convex frontier model applied to a single input (risk) and a single output (return). For this reason, many authors have used CRS models to evaluate portfolio management, but there are also many others that have implemented VRS models. This controversy remains in new studies (see Glawischnig and Sommersguter-Reichmann, 2010), although Kerstens et al. (2011) recently justify the use of CRS frontiers in the analysis of these collective portfolios.

Lozano and Gutiérrez (2008b) provide a complete review of the empirical studies that use DEA frontiers to assess the performance of mutual funds and hedge funds. The most common inputs in these studies are the standard deviation of fund returns and the expense ratio, while the fund return is the most common output. The latest studies fine-tune these basic variables using subscription and redemption costs and fund size as new inputs (Basso and Funari, 2001; Daraio and Simar, 2006) and an ethical score as an output (Basso and Funari, 2003). Recently, Andreu et al. (2013) include some of these variables in a SBM model to evaluate the efficiency of the strategic asset allocation of a sample of pension funds.

Despite the increasing interest in DEA methodology in mutual fund studies, there is very scarce evidence of evaluations of efficiency in Mutual Fund Management Companies. As far as we know, only Zhao and Yue (2010) explore the efficiency of mutual fund companies in the Chinese market by proposing a multi-subsystem fuzzy DEA. These authors divide the core competence³⁵ of a mutual fund company into an investment subsystem and a marketing and service subsystem. That is, the mutual fund company manages financial assets to obtain returns derived from assuming

³⁵ According to Prahalad and Hamel (1990), the core competence of a company represents the knowledge and experience accumulated in the firm.

certain levels of risk, but the mutual fund company also constantly pursues an increased company market size by obtaining larger money inflows into the mutual funds managed by the firm. This paper provides a concept model with variables that aim to appropriately represent both subsystems.

This study fills this gap in the studies in the literature by implementing a two-stage concept model based on Holod and Lewis (2011) for the first time to a major European mutual fund industry. Our model includes a set of more detailed variables that better represent the two subsystems proposed originally by Zhao and Yue (2010) to properly evaluate the efficiency interactions between the different management subsystems present in the core competence of a Mutual Fund Management Company.

2.3.3 Brief Discussion on the Production and Asset Approaches

The literature review addressed in the previous section highlights the apparent controversy between the two major conceptual approaches to measuring the efficiency of financial institutions, especially in banking: the production approach and the asset approach. It is necessary to know the main characteristics of these different conceptual models to better understand the conceptual system used in our analysis. In this section, we briefly discuss the main points of these different approaches to the production process in financial institutions. Let us briefly review these two major theoretical models found in the literature.

The production approach or service provision approach: a business valuation method based on the production value of a going concern. Here, banks are treated as companies that use capital and labour to produce different categories of deposit and loan accounts. In other words, banks provide services to customers by administering the customers' financial transactions, keeping customer deposits, issuing loans, cashing cheques and managing other financial assets (Berg et al. 1991; Berg et al. 1993; Parson et al. 1993; and Schaffnit et al. 1997). Efficiency and productivity can be analysed by comparing the quantity of services offered with the quantity of

resources used by the bank. It is widely known that Berg et al. (1991) identified five activities performed by a bank: (1) supplying demand, facilitating deposit services, (2) short and long-term loan services, (3) brokerage and other services, (4) property management and (5) the provision of safe deposit boxes. These authors also noted that a bank incurs positive operating costs in terms of (1) labour, (2) machines, (3) materials and (4) buildings.

The asset approach or intermediation approach: a business valuation method based on the net asset value of a going concern. Here, the banks are viewed as intermediaries of financial services rather than as producers of loan and deposit account services. In this approach, the bank accepts deposits from customers and transforms them into loans to clients. The inputs are labour, materials and deposits and the outputs are loans and other income generating activities, namely banking services (Mester, 1997). The banks perform the two major roles of mobilising and distributing resources efficiently to ease investment activities in the economy. Colwell and Davis (1992) noted that the major disadvantage of this approach is the absence of trust operations, which causes increases in the unit costs of large banks.

The asset approach has two major sub-groups: (1) the profit approach and (2) the risk management approach. In the profit approach, the bank manager's purpose is to maximise the bank's profit function. The bank manager must evaluate all types of costs and the income generated in the production process. This approach simultaneously measures inefficiency on the input and output side, which reduces problems associated with mis-specification and mis-measurement (Berger et al. 1993; Thompson et al. 1997). In the second sub-group based on risk-management, it is necessary to evaluate the risks linked to various forms of assets in a bank. In risk-management, banks take some risks to produce acceptable returns. A bank's performance will affect its valuation in the market, its ability to acquire other banks or to be acquired at a good price, and therefore its ability to be funded in deposits and financial markets (Mester 1996; Battese et al. 2000). Therefore, the risk-management approach considers both management's decision-making process and the

implementation of these decisions as inputs and shareholders' value and bank profit as outputs (the well-known agency problem).

As noted in the above subsection, non-parametric models are usually based on the production or service provision approach and take deposits to be the output based on the positive consumption of labour and materials. Parametric models are based on the asset approach and observe deposits as an input for the production of loans and other financial assets.

Mlima and Hjalmarsson (2002) find that efficiency scores are very sensitive to the choice of input and output variables, similar to Berg et al.'s (1991) findings from their original study. Berger and Humphrey (1997, p. 179) alleged that "it is not possible to determine which of the two major approaches dominates the other since the true level of efficiency is unknown. The solution, in our opinion, lies in adding more flexibility to the parametric approaches and introducing a degree of random error into the non-parametric approaches". Bauer et al. (1998) note that it is not necessary to reach consensus, instead finding consistency to be the most useful approach for regulators and other decision makers. In a study of productivity changes³⁶ in European banking, Casu et al. (2004) state that both approaches sometimes identify conflicting findings for the sources of productivity for individual years. However, the two different approaches generally do not yield markedly different results in terms of identifying the components of productivity growth.

Holod and Lewis (2011) argue that the primary confusion in the literature has been the disagreement among researchers about the appropriate input and output selection and therefore the conceptual approach used. According to these authors, banks use their employees and fixed assets to obtain deposits and make investments and loans with the purpose of generating profitability. In the first stage, the deposits obtained

³⁶ Productivity change is decomposed into technological change, which reflects improvement or deterioration in the performance of best-practice decision making units (DMUs), and technical efficiency change, which reflects the convergence towards or divergence from best practices on the part of the remaining DMUs. This decomposition provides useful information on the sources of the overall productivity change.

serve as the principal resource of funding for a bank's lending activity in the second stage. So, the result of the deposits on bank efficiency depends on the efficiency at both stages. That is, in contrast to the extensive literature that treats deposits as a pure input or a pure output, the primary contribution of the Holod and Lewis' (2011) model is the double role of deposits in the bank production process, where deposits are an intermediate product, an output from the first stage and an input to the second stage. As a result, this hypothesis does not impose judgment on whether either larger or smaller values of deposits are more desirable. Furthermore, another essential aspect in this paper is that the proposed DEA methodology evaluates banking efficiency using an unoriented model, thereby identifying a mix of input reductions and output increases that lead to efficiency.

The next section will further discuss our conceptual model and the specific variables used to analyse the efficiency of Spanish mutual fund companies. This model will be based on the interactions between the inputs and outputs of the double-stage production approach of Holod and Lewis (2011).

3 Model and Variables

3.1 Multi-Management Stages Model in a Mutual Fund Management Company (MFMC)

To establish an appropriate evaluation of the efficiency of an MFMC, it is necessary to have adequate indicators that describe the primary management activities of these firms. However, defining the set of management inputs/outputs is not an easy issue, as we have found from the previous literature on banking and insurance companies. The well-known production and asset approaches considered different models to explore financial institutions, which could report misleading evidence due to the different nature of the core competence of the institutions considered by each of these conceptual models.

As noted in the above section, our proposal for explaining the activity of a MFMC is based on the recent approach of Holod and Lewis (2011), which aims to resolve the discussion of the production and asset models in the banking literature. These authors make a more suitable proposal in which bank deposits are considered to be an intermediate product in the management process of a bank.³⁷ The model is solved using an unoriented DEA methodology to emphasise the importance of simultaneously decreasing inputs and increasing outputs through a two-stage approach, called a network DEA model.³⁸

³⁷ Holod and Lewis (2011) state that deposits have a dual role in the bank production process and propose the novel idea of analysing a banking organisation through different management sub-stages, i.e., deposits are seen as an intermediate product, that is, an output obtained by the first stage of the bank production process that is then considered to be an input to the second stage. Therefore, the effect of the amount of deposits on bank efficiency depends on the efficiency at both stages of the bank production process. A bank aims to reduce its inputs (employee and fixed assets) and to increase its outputs (loans and other earning assets), given a certain amount of deposits. Holod and Lewis (2011) emphasise that their model has no need to assume whether having a higher or lower value of deposits is better for bank efficiency. Instead, the effect of deposits on overall bank efficiency is determined by the relative efficiency at each stage of production.

³⁸ The network DEA model allows the analyst to look inside the DMU, thereby offering the best perception as to the sources of organisational efficiency. Each DMU in this model can be comprised of two or more sub-stages. The network DEA model can be input-oriented, output-oriented, or unoriented, and additionally, it may consider different returns-to-scale (CRS or VRS).

According to Zhao and Yue (2010), the core competence of a MFMC is the application of expert knowledge and experience to managing the money collected from the market to pursue both the benefit of fund unit-holders and increasing assets under management. This overview means that the efficiency of these companies could be analysed using a production approach or an asset approach depending on the consideration of the money managed by the company, which might have a very similar interpretation to bank deposits in the earlier literature on the efficiency of banking institutions. This issue could affect the selection of different inputs/outputs to run the DEA model, thereby affecting the interpretation of the efficiency scores obtained by the model. To overcome this problem, we apply the called multi-management stages approach of Holod and Lewis (2011) by considering the interaction between the different management stages within a MFMC.

Based on a stakeholder approach (see Figure I-4), the model of Berkowitz and Qiu (2003) implies three relevant management interactions related to the primary stakeholders in a MFMC, the mutual fund unitholders and the company shareholders. The first management interaction is similar to the first management stage considered by Zhao and Yue (2010), and it refers to the portfolio management competence of a MFMC. This interaction is the most intuitive because it is described as the central activity of the MFMC by the Official Business Registries in Europe. Fund managers should take investment decisions based on professional and technical criteria to obtain good return records with the lowest levels of risk as possible. These returns will be provided to mutual fund unitholders after subtracting the corresponding management fees and other expenses. Secondly, closely linked to this first stage, the fund companies should have an appropriate distribution system for their mutual funds because distribution is a key factor in explaining the increase of assets managed by the MFMC, i.e., it is necessary to sell these mutual funds to gain both money and investor inflows into the mutual funds. As a consequence of the management fees charged by the company, the income derived from these fees will be higher when larger amounts of assets are under management.³⁹ Finally,

³⁹ Díaz-Mendoza et al. (2012) find a residual presence of performance-based fees in the Spanish fund industry, where most of the management fees charged by mutual funds are calculated as a percentage of the assets of the fund.

the MFMC shareholders aim to obtain higher profits through the final results of the entire activity of the company, i.e., after taking into account the income and cost structure of the MFMC. This overall stage is included in the management picture of Berkowitz and Qiu (2003) as a key interaction with the owners of the company because it responds to shareholder demands for higher levels of overall efficiency. This latter management interaction between the competence of the company and the shareholders was omitted by Zhao and Yue (2010), but in our opinion, it is also important when evaluating the efficiency of the MFMC as a DMU.

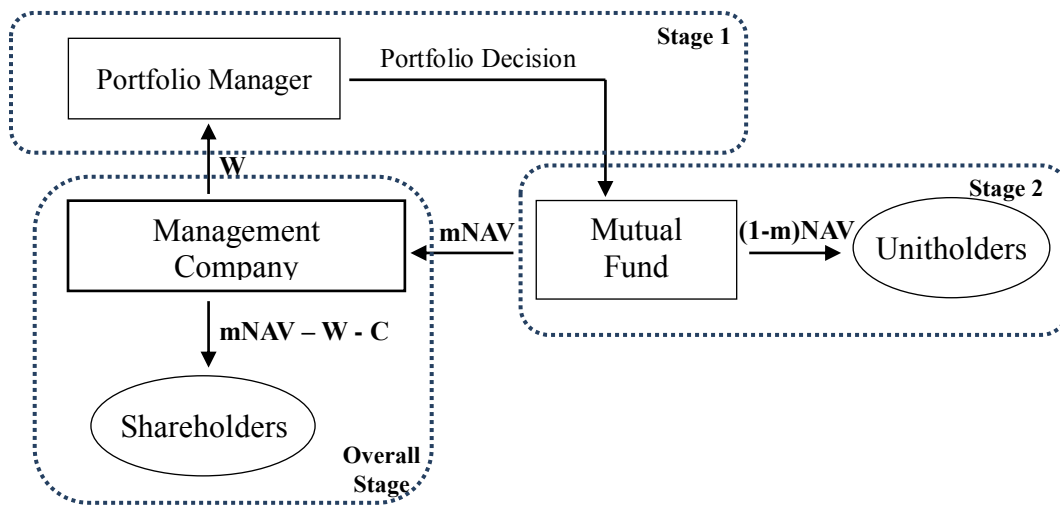


Figure I-4 MFMC interactions based on Berkowitz and Qiu (2003).

where m is the fees ratio of the company, NAV is the net asset value of the mutual fund, W is company wages, and C is total operating expenses. The dotted lines represent Stage 1 (*Portfolio Management stage*), Stage 2 (*Marketing and Service stage*), and the Overall Stage (*Overall Efficiency*).

Based on the unoriented model⁴⁰, which was proposed by Holod and Lewis (2011) to end the debate between the production and asset approach, as applied to the banking industry and according to the different management stages within a MFMC proposed by Berkowitz and Qiu (2003), we propose a specific model to capture the interaction between the different demands from the stakeholders of a MFMC and the different sub-

⁴⁰ As stated above, Holod and Lewis (2011) claim that the primary advantage of their model is that it does not require a choice between a production and an intermediation approach, therefore leading to less controversial efficiency estimates.

DMUs within the company to appropriately evaluate the efficiency of these financial institutions. Figure I-5 illustrates our conceptual model.

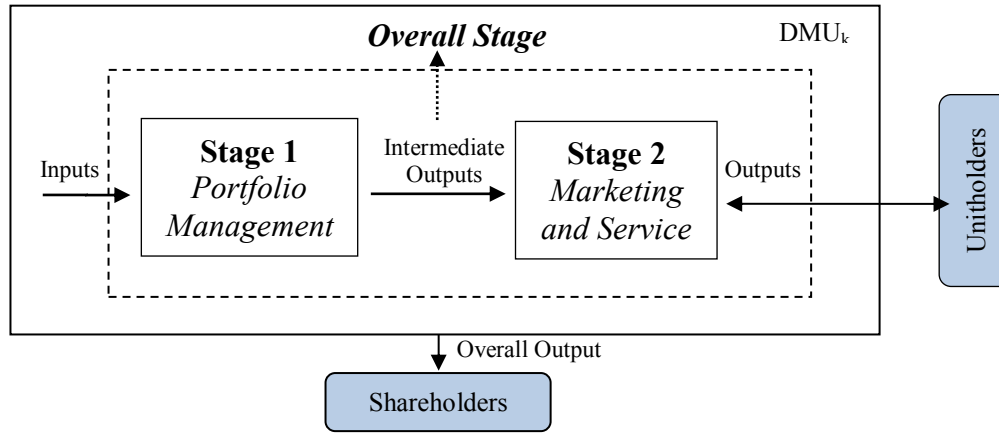


Figure I-5 Multi-management stages model for a MFMC

Our unoriented model considers two primary steps in the management process of each MFMC analysed (Decision-Making-Unit, DMU_k). The first stage in this in-company management process would correspond to the *Portfolio Management stage*. In this stage, the fund managers of the company assume a specific level of risk to obtain higher gross returns in as many mutual funds and types as possible. The rationale behind this first stage is that a company with efficient portfolio management skills is able to obtain better returns before fees and other expenses with controlled levels of risk for a large and well-diversified selection of mutual funds without assuming extra personnel expenses and financial resources.

According to the sub-stages framework of Holod and Lewis (2011), the outputs of the first stage in-company may be considered to be intermediate outputs of the MFMC, thereby being inputs into the second stage, *Marketing and Service*. In this second step of the management process, efficient distribution attracts both unitholders and money net inflows into every fund managed by the company, thereby generating higher new incomes because of the asset-based fees charged by the company. The resources to obtain the goal of this distribution stage are represented by the company's offer to the

market, that is, the intermediate outputs generated by the *Portfolio Management stage* to appropriately sell the mutual funds.

Both stages of our unoriented model are similar in nature to the proposal of Zhue and Yue (2010), but our original contribution is to define a clear interaction between these two subsystems within the core competence of a MFMC, thereby overcoming the aforementioned problem of the conceptual orientation model used in most banking studies.

Finally, the *Overall Efficiency stage* evaluates the final result reported to the MFMC shareholders as a result of the entire activity of each MFMC or DMU_k . This overall stage includes both the *Portfolio Management stage* and the *Marketing stage* as a whole, thereby considering the final profits to be a consequence of the core competence of the MFMC (Stage 1 and Stage 2). These profits will be a more useful indicator for company shareholders and will include the income and cost structure of all resources necessary to manage the capital collected from the market through different mutual funds and investment categories.

3.2 Variables of the Model

Once we have described the conceptual framework to evaluate the efficiency of the MFMC, it is necessary to define how we measure the variables, which appropriately represent the inputs and outputs and capture the ideas displayed by Figure I-5.

Figure I-6 illustrates the variables selected to run the model, and Table I-2 lists the inputs and outputs included in this multi-management stage approach to run the different models from 2005 to 2009. All of the data necessary to establish these variables have been obtained from the Iberian Balance Sheets Analysis System (SABI), the Spanish Official Business Registry (Registro Mercantil), and the Spanish Securities Exchange Commission (CNMV).

Table I-2 Set of Inputs and Outputs

Portfolio Management	Labour: L_k is the number of employees of company k at 31 st December.	Assets Managed: AM_k is the total assets managed by company k at 31 st December.
	Shareholders' Equity: SE_k is the equity capital including reserves and profits of company k at 1 st January. ⁴¹	Number of Funds: NF_k is the number of funds managed by company k at 31 st December.
	Portfolio Risk: PR_k is the fund size-weighted average of the normalised value ⁴² of the standard deviation of the daily gross returns of all funds managed by company k at 31 st December.	Fund Types: FT_k is the number of fund categories according to the official classification of investment objectives ⁴³ covered by company k at 31 st December. Gross Returns: GR_k is computed by the fund size-weighted average of the normalised value ⁴⁴ of the daily average gross returns of all funds managed by company k at 31 st December.

⁴¹ This variable considers the equity capital at the end of the previous year, i.e., this input is not influenced by the results of the company during the analysed year.

⁴² We agree with Zhue and Yue (2010) that returns weighted by size do to some extent represent the Mutual Fund Management Companies' investment skills. However, the different size and return patterns between the different fund types could bias the weighted returns and the levels of risk associated with the mutual fund companies due to the assorted fund types managed by these companies. For instance, a company with more assets in equity funds than in bond funds would obtain upwards biased size-weighted returns in years with bullish stock markets compared with a company much more focused on bond funds, and the opposite could be found in bearish stock markets. Zhue and Yue (2010) solve this potential problem by using a membership function to characterise fund types. In our case, we compute the normalised standard deviation of the daily gross returns for each mutual fund existing at 31st December with respect to all funds in the market included in the same category and during the same time period. This normalisation provides a value between 0 and 1 that reports more insightful information for the risk skills of the fund with respect to fund competitors with the same investment objective.

⁴³ These official classifications are reported by the Spanish Securities Exchange Commission (CNMV)

⁴⁴ The reason for rejecting a fund size-weighted average of the returns obtained by the different funds offered by a company is similar to that addressed in the measure of risk. We obtain the daily average gross return for each mutual fund from 1st January to 31st December with respect to all funds in the market included in the same official fund category and during the same time period. We then compute the normalised value between 0 and 1 of these average gross returns to obtain the size-weighted value for every mutual fund company.

Table I-2 Set of Inputs and Outputs (Continued)

Marketing and Service	<p>Assets Managed: AM_k</p> <p>Number of Funds: NF_k</p> <p>Fund Types: FT_k</p> <p>Net Returns: NR_k is the fund size-weighted average of the normalised value⁴⁵ of the daily average net returns of all funds managed by company k at 31st December.</p>	<p>Unitholders Net Flows: UNF_k represents the normalised value of unitholder inflows minus unitholder outflows for company k from 1st January to 31st December.</p> <p>Money Net Flows: MNF_k represents the normalised value of the implied net money flows⁴⁶ for company k from 1st January to 31st December.</p> <p>New Incomes: NI_k computes the new management fees received by the MFMC as a consequence of the net money flows into company k from 1st January to 31st December.⁴⁷</p>
Overall Efficiency	<p>Assets Managed: AM_k</p> <p>Number of Funds: NF_k</p> <p>Fund Types: FT_k</p> <p>Shareholders' Equity: SE_k</p>	<p>Profits: P_k is the normalised value of the profits obtained by the MFMCk from 1st January to 31st December.</p>

⁴⁵ Daily management and custodial fees charged by the company to the fund unitholders have been subtracted from the daily gross returns obtained by each fund.

⁴⁶ Implied net flows have been defined as monthly changes in the total assets of each fund net of fund returns. As we do not know the exact moment of flows, we follow the usual approach of the literature and assume that these flows occur at the end of the month for which we are computing this measure. Zheng (1999) determined that this standard approach is robust with other assumptions about the timing of these implied flows.

⁴⁷ This variable is proxied by the product of the asset-based management fees of each fund and the implied money flows obtained by MNF_k

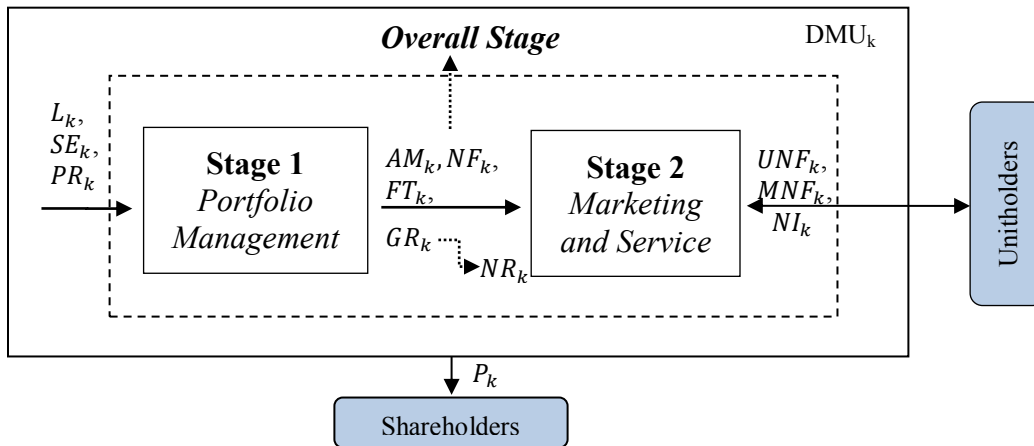


Figure I-6 Multi-management stages model for a MFMC (including variables)

The rationale behind the selection of these variables is based on the previous DEA literature. According to our conceptual approach to the *Portfolio Management stage* in a fund management company, we should define inputs, which capture the labour, capital and risk assumed in managing the mutual fund portfolios. Labour should be a proxy of the human resources required to appropriately develop the major activity of the MFMC, i.e., the management of financial portfolios. In that sense, the number of employees has been extensively used in the DEA literature to address efficiency (Lang and Welzel, 1996; Hussels and Ward, 2007; Chen et al, 2009). We also find this variable measured as the total expenditure of wages at the end of each fiscal period (Berg et al, 1991; Berg et al, 1993; among other initial studies); as the value of wages as a percentage of total expenses (Cummins et al., 2010); as the weighted average of annual wages per employee (Berger et al., 2009)⁴⁸; and finally as a normalised variable dividing wages by assets under management for the mutual funds (Casu et al., 2004). The availability of the number of employees in our dataset provides more detailed information about the quality of the human resources for a company than other measures related to personnel

⁴⁸ Berger et al. (2009) do not consider the DEA methodology. They estimate efficiency levels by specifying the translog functional form for the cost and profit functions based on the SFA method.

expenses, which could be distorted by the mispricing of these resources, i.e., significantly high or low wages could affect these measures.⁴⁹

Since Murthi et al. (1997), other studies have applied variables related to expense ratio as inputs (e.g., Daraio and Simar, 2006). The expense ratio determines the effort associated with portfolio management. However, these studies are related to portfolio performance, and they are not associated with efficiency in management companies as in our case. We also consider the use of the capital variable in our model. Capital has been measured in diverse forms in the DEA efficiency literature linked to the banking and insurance industry (e.g., Berg et al., 1993; Cummins et al., 2004; Hussels and Ward, 2007). It is important to note that in these studies, the amount of capital is defined as the long-term investment or as the book value of equity plus reserves at the beginning of each period. Alternatively, Berg et al. (1993) and Brockett et al. (1997) also used furniture and equipment variables as a proxy for capital to evaluate bank efficiency using DEA methodology.⁵⁰

Finally, and according to the growing body of literature using the DEA methodology to evaluate portfolio performance, it is necessary to compute the risk assumed by the portfolios as a representative input into portfolio management. Firstly, we find studies that consider the standard deviation of mutual fund returns as an appropriate risk measure (Murthi et al., 1997; Choi and Murthi, 2001; Basso and Funari, 2001, 2003; Chang, 2004; Daraio and Simar, 2006; Lozano and Gutiérrez, 2008a) and secondly, we find much more sophisticated and asymmetric risk measures applied to hedge funds (Gregoriou, 2003; Gregoriou et al., 2005; Nguyen-Thi-Thanh, 2008)⁵¹.

⁴⁹ The consideration of personnel expenses as an alternative input for labour is not a relevant issue for the efficiency DEA rankings in the *Portfolio Management stage*. Detailed results will be provided in the next empirical section when robustness tests are performed.

⁵⁰ Some robustness tests are developed in the next empirical section, in which we analyse the efficiency scores obtained by both equity capital and equipment measures as variables representing the capital effort of the mutual fund management company.

⁵¹ Measures such as MVaR (Modified Value at Risk) and CVaR (Conditional Value at Risk) are examples of the measures applied to hedge funds.

The outputs obtained by this *Portfolio Management stage* should represent the success of the MFMC in managing the greatest amount of assets (total assets managed by the company) with the best (gross) return record possible (Murthi et al., 1997; Choi and Murthi, 2001), enabling potential diversification through the entire set of mutual funds (number of mutual funds) and investment categories (fund types) offered to the market. Then, an efficient MFMC should assume controlled levels of labour, capital and risk to obtain winning return records for a wide-range of mutual funds covering as many fund types as possible to meet the demands of different sets of unitholders with assorted return-risk preferences.

The results of the *Portfolio Management stage* should be an appropriate indicator of the quality of the mutual funds sold during the *Marketing and Service stage*. Unitholders' interest in recent return records as a significant factor to explain money and investor flows into mutual funds has been a well-known hypothesis in the literature since the evidence of Sirri and Tufano (1998), followed by the comparative study of Del Guercio and Tkac (2002). In addition to these return records, the existence of a large and well-diversified set of mutual funds should help marketing management to better collect money and investors from the market. The only relevant difference between the intermediate outputs obtained from the *Portfolio Management stage* and the intermediate inputs of the *Marketing and Service stage* is the nature of the returns because the return records offered to the market to gain both unitholders and money inflows into the company are net of management fees and other expenses. Both unitholders and money net flows into the company should represent the final success or failure of the marketing and distribution of the mutual fund records obtained by the *Portfolio Management stage*. In addition, the new management fees received by the company as a consequence of new money flows into the mutual funds is a step further than the mere consideration of money flows because it also captures the so-called extended asset-based management fee structure in Europe, and especially in the Spanish fund industry. The most accurate measure to proxy for these new fee incomes is the product of the asset-based management fees of each fund and the new implied money

flows obtained by the company, which is a unique and useful measure for evaluating the efficiency of the marketing and distribution stage of the MFMC.

Finally, *Overall Efficiency* should consider the final results provided to the shareholders after all of the production processes of the MFMC. Similarly to other banking studies, the most robust variable for measuring the output of this overall stage should include the entire income and cost structure of the company, which is mostly represented by the profits obtained by the shareholders (Hussels and Ward, 2007; Berger et al., 2009; Chen et al., 2009).

Furthermore, the inputs of the *Overall Efficiency* should include a set of indicators that represent the main characteristics of the mutual funds managed by the company. However, we must define these variables with caution to avoid including inputs that could already be included in the final profits, i.e., expenses or other money variables. This *Overall Efficiency* model might be considered to be an extension of the return-on-equity ratio for the MFMC after including these relevant measures for the mutual funds managed by the company as additional inputs to the capital effort of the stockholders (shareholders' equity). We use total assets, the number of funds and the different fund types managed by the company as relevant variables representing the major characteristics of the mutual fund offering to the industry.

Finally, and according to Coelli et al. (2005), it is important to note that our set of variables meets the DEA convention that the minimum number of DMUs in our study (93 in 2008) must be greater than three times the sum of the number of inputs and outputs considered in the different management stages included in our models.

4 Data and Empirical Analysis

4.1 Data

We work with all Spanish MFMCs registered in the Spanish Securities Exchange Commission (CNMV) on 31st December of each year included in our time horizon 2005-2009. The number of MFMCs ranges from a minimum of 93 in 2008 to a maximum of 102 in 2005. Those companies with more than 15% of assets in hedge funds were excluded to limit the sample to companies primarily focused on the management and distribution of mutual funds. We also exclude those companies with recent inception dates to avoid a possible inception bias in the variables used in our model. In any case, the bias of both exclusions is quite residual in terms of the economic relevance of the sample.

Table I-3 shows the main descriptive statistics for our data set. In general, these statistics show a large dispersion of the data, thereby indicating the assorted characteristics of the companies competing in the Spanish fund industry. Figure I-7 shows in aggregate terms the yearly evolution of some of the major magnitudes for the companies included in our sample.

Table I-3 Descriptive statistics of the data (31st December)

Year	Num. of MFMCs	Measure	Num. of Employees (L _k)	Shareholders' Equity * (SE _k)	Assets Managed** (AM _k)	Num. of Funds (NF _k)	New Incomes** (NI _k)	Profits** (P _k)	Num. of Unitholders
2000	102	Mean	20	9,855,448	2,568,985	26	1,117	3,842	82,809
00		Std. Dev.	27	18,504,585	7,965,647	43	4,062	10,450	255,312
00		Minimum	2	333,081	4,217	1	-24,529	-1,431	102
05		Maximun	200	130,379,000	61,873,451	293	14,478	68,546	1,747,614
2000	102	Mean	21	10,748,921	2,651,042	28	-488	4,549	84,684
00		Std. Dev.	27	20,316,255	7,963,606	48	6,286	12,050	260,916
00		Minimum	2	322,079	4,424	1	-51,255	-1,069	85
06		Maximun	193	146,469,000	61,663,399	308	11,487	81,358	1,858,235
2000	95	Mean	23	12,800,419	2,682,037	31	-3,558	5,146	84,749
00		Std. Dev.	29	22,973,569	7,415,818	53	14,847	13,390	241,650
00		Minimum	2	598,000	4,869	1	-127,020	-1,196	100
07		Maximun	180	167,646,000	53,063,849	324	9,770	85,724	1,690,245
2000	93	Mean	23	14,829,848	1,887,768	31	-7,661	3,422	63,659
00		Std. Dev.	29	26,704,624	5,286,616	54	24,669	9,642	177,150
00		Minimum	2	483,941	3,734	1	-212,890	-1,749	100
08		Maximun	161	186,439,000	34,237,954	298	1,563	65,054	1,284,565
2000	95	Mean	23	15,570,010	1,794,888	27	105	1,806	57,631
00		Std. Dev.	27	28,842,879	4,848,550	42	3,742	5,204	154,350
00		Minimum	2	208,337	4,457	1	-15,094	-3,163	95
09		Maximun	145	191,655,197	32,580,875	206	24,911	36,126	1,107,698

(*) At beginning of the year; (**) thousand Euros

From a sample of 102 MFMCs in 2005, in economic terms, we found initially that these MFMCs represented an approximate average of 2,569 million Euros of assets managed by approximately 2,700 mutual funds with 8,446 thousand unitholders (see Figure I-7). However, 57.84% of total assets were concentrated in the five largest companies, while 73.88% were managed by the top ten. The distribution of the number of unitholders is similar, with some minor changes in the ranking of the largest companies. In 2005, the 5 largest MFMCs took 52.26% of the sector profits, while the 10 largest took 69.68%. This figure highlights the great concentration in the sector, which may affect the efficiency of the mutual fund industry.

In 2006, with the same number of MFMCs, the mean size increased to 2,651 million Euros, adding 191,220 unitholders since 2005 and with 2,825 mutual funds managed (see Figure I-7). However, the major market share is still held by the largest companies with some minor changes. The top ten largest MFMCs manage approximately 73% of total assets and take approximately 70% of the sector profits. In addition, the average shareholders' equity is approximately 91 million Euros higher than in the previous year.

During 2007, the analysis of 95 MFMC illustrates a slowdown in the mean assets under management (2,682 million Euros) and in global terms, a loss of 15,612 million Euros (see Figure I-7), primarily impacting the largest companies (from a maximum value delivered of 61,663 million Euros to 53,063 in 2007). There was also an important reduction of unitholders in global terms of approximately 586,639 (see also Figure I-7). Moreover, there is a loss in the market share of the top five and top ten largest companies, dropping to 54.78% and 71.54%, respectively. The profits remain at a similar level, but with better participation for the top five and top ten companies, at 56.85% and 73.94%, respectively. However, the MFMCs increased the number of funds to 2,908 compared to 2,825 in 2006 (see also Figure I-7). The positive trend in shareholders' equity remains, reaching a mean of 12,800 thousand Euros since 10,749 in 2006 (in global terms, 119,650 thousand Euros).

For the starting period of the financial crisis (2008), 93 MFMC were analysed and exhibited an important decrease in the asset mean to approximately 1,887 million Euros (in global terms, the companies saw a drop from 254,793 million Euros under management to 175,562 million Euros), primarily due to an outflow of 2.1 million unitholders (see Figure I-7). The market share improved for the top five companies, with 56.69% of assets, and for the top ten, with 73.71%, thereby increasing the concentration of the industry. Sector profits fell by approximately 170 million Euros (see also Figure I-7), but interestingly, the percent of participation for the top five and top ten improved to reach approximately 60% and 79%, respectively. Obviously, one of the most important findings from 2008 are both the dramatic fall in new incomes with a mean value of -7,661 thousand Euros and the outflow of 2,130 million unitholders from 2007 to the end

of 2008. The mean number of employees remains quite similar to previous years, although this variable slightly increases as a whole but decreases in the largest companies.

Finally, we analyse 95 MFMC in 2009, managing a mean of approximately 1,794 million Euros (in global terms, 170 MM Euros) with 2,530 mutual funds, in contrast to 2,893 funds in 2008, and with 445,421 fewer unitholders than in 2008 (see Figure I-7). The assets managed by the top five and top ten companies underwent a slight decrease, to 54.82% and 71.38%, respectively. However, the five and ten largest companies took approximately 60% and 83% of the profits, which were low at only 171 million Euros for the entire industry. When reviewing the number of employees, the downward mean trend is even more important in the largest employers (changing from 161 employees in 2008 to 145 in 2009).

Summarising the descriptive analysis above, we find a relevant concentration in the sample and predominantly represented by the banking industry. We also find a significant effect from the financial crisis starting in 2008 on the number of unitholders, and therefore on the profits, assets and funds managed by the MFMCs. Further research could focus on how the crisis may have strengthened the concentration of the Spanish fund industry.

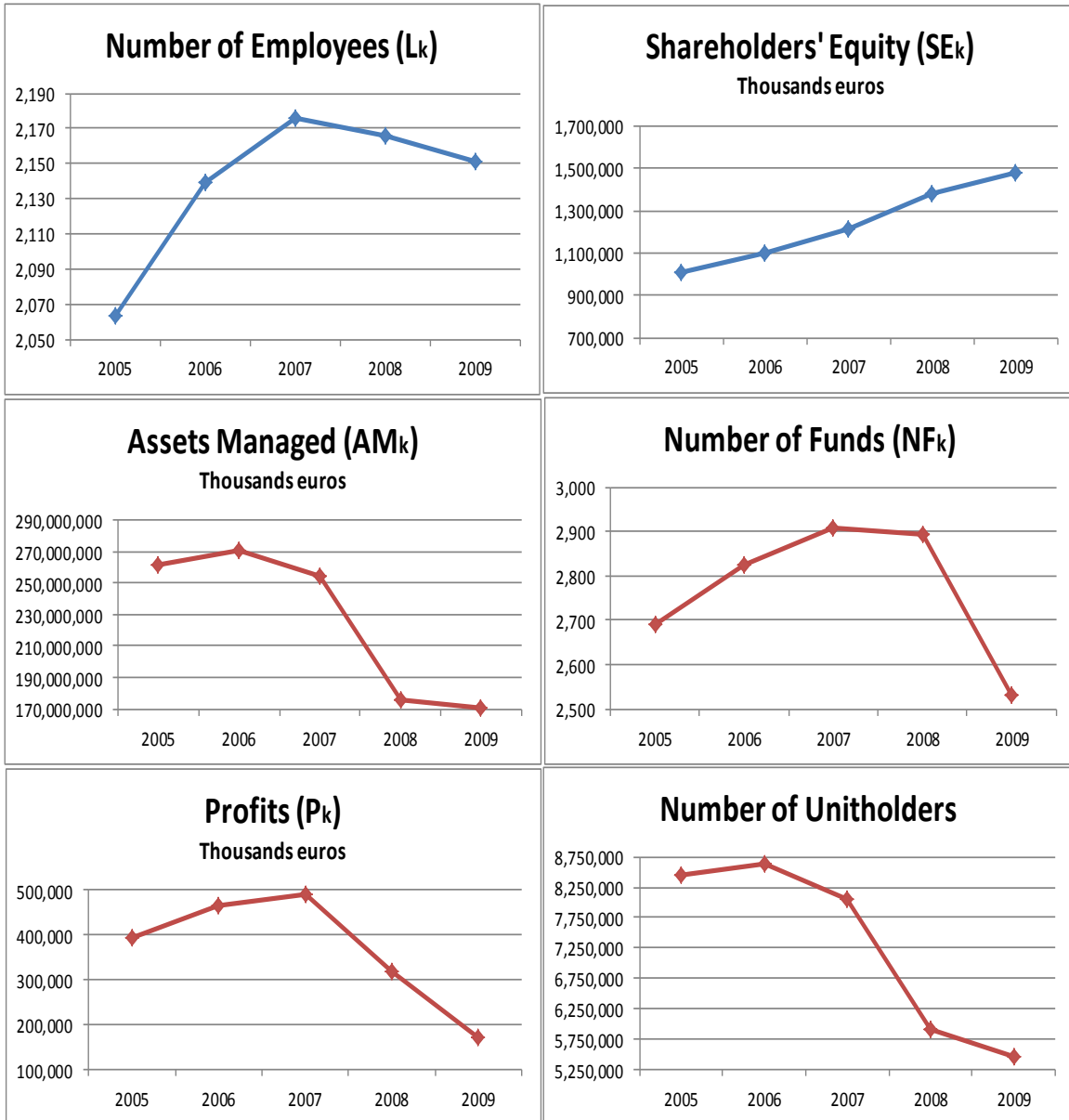


Figure I-7 Evolution of the mutual fund industry (2005-2009)

4.2 Empirical Analysis

We formulate three major assumptions to evaluate the efficiency of the MFMCs based on our unoriented sub-stages model (Figure I-5). First, we consider a portfolio management system to be technically and operationally efficient when it works with fewer employees, lower capital resources and more controlled levels of risk than the competition and at the same time is able to manage more funds with different investment vocations to obtain better return records than the competition. We call this first subsystem *Portfolio management*. Second, we consider a good marketing system to be one that is able to capture as many new unitholders and money flows into the mutual funds managed by the company as possible to generate higher new fee-based incomes. Furthermore, the outputs obtained from the portfolio management subsystem will subsequently determine the quality of the mutual funds sold by the company. We call this second or intermediate subsystem *Marketing and Service stage*. Finally, we consider a company to be globally (overall) efficient for its shareholders if it obtains higher profits than the competition. These profits should be obtained with a precise offer of mutual funds and assets under management, diverse investment categories and suitable capital equity. We call this subsystem *Overall Efficiency*.

First, we run the original SBM-efficiency model under constant returns to scale (equation I.15) for each stage considered in our conceptual model developed in Section 3. The primary results obtained by our multi-management approach for Stage 1: *Portfolio Management*, Stage 2: *Marketing and Service*, and for the *Overall Efficiency* of the company are shown in Table I-4 for each year analysed from 2005 to 2009 (Appendix A includes complete results in Table I-A-1, Table I-A-2, Table I-A-3, Table I-A-4, and Table I-A-5).⁵²

⁵² The entire empirical analysis has been run using R project software and the specific package "nonparaeff". This package illustrates Nonparametric Methods for Measuring Efficiency and Productivity, version 0.5-8. This package contains functions for measuring the efficiency and productivity of decision making units (DMUs) under the framework of Data Envelopment Analysis (DEA) and its variations, thereby including the Slacks Based Measure (SBM) proposed by Tone (2001). Details for this package are available at <http://cran.r-project.org/web/packages/nonparaeff/index.html>.

Table I-4 Efficient companies by stage and variable (%)

Year	MMS	Eff.	Direct	Assets	Funds	Unitholders
		MFMCs	Jobs			
2005	Stage 1	28	31.90	56.48	44.56	58.63
	Stage 2	7	2.42	0.09	0.41	0.08
	Overall Eff.	10	17.45	41.87	17.91	41.69
2006	Stage 1	22	28.28	53.27	41.95	56.13
	Stage 2	8	3.55	0.17	0.78	0.10
	Overall Eff.	8	7.48	18.07	7.01	22.01
2007	Stage 1	21	27.34	51.23	42.68	53.38
	Stage 2	6	1.79	0.17	0.58	0.46
	Overall Eff.	6	3.72	1.53	0.52	0.57
2008	Stage 1	13	28.81	58.22	46.18	56.27
	Stage 2	2	0.32	0.01	0.07	0.00
	Overall Eff.	5	2.95	0.95	0.45	0.55
2009	Stage 1	20	18.08	34.12	30.20	37.13
	Stage 2	7	8.46	10.22	8.34	8.39
	Overall Eff.	5	2.93	1.54	0.75	0.63

MMS = Multi-Management Stage

Let us first analyse some aggregate figures shown in Table I-4 before further commenting on some detailed efficiency patterns displayed by the extended tables included in Appendix A. Table I-4 includes the relative percentages of the efficient MFMCs in terms of assets managed, unitholders' accounts, direct jobs, and number of mutual funds as the major variables to describe the characteristics of these efficient MFMCs initially obtained by the CRS-SBM technique.

Table I-4 shows 28 efficient companies at the *Portfolio Management stage* in 2005; these companies managed 56.48% of assets and 58.63% of the unitholders' accounts, represent 44.56% of the number of funds, and maintained approximately 32% of the direct jobs in the Spanish fund industry. These high efficiency percentages were sustained until 2009, when these percentages decreased significantly as a potential consequence of the financial crisis in 2008.

There were significantly fewer efficient companies at the *Marketing and Service stage* than in the previous Stage 1. The relative efficiency percentages in terms of assets, unitholders, and funds remained lower than 1% from 2005 to 2008, but these figures increased to 10.22%, 8.39%, and 8.34%, respectively, in 2009. The explanation of this situation is that some of the largest companies ceased to be inefficient this year. That is, the largest companies appeared to adapt much better to the crisis of confidence in financial markets, at least in terms of selling their mutual funds. However, this conclusion should be considered with caution because further analysis beyond 2009 would be required to confirm this hypothesis.

Table I-4 shows that the overall efficient companies represented an important percentage in 2005 in terms of assets managed and unitholders' accounts, although the number of these overall efficient MFMCs is much lower than in Stage 1. This result provides evidence that large companies in 2005 were efficient overall. However, these large companies ceased to be efficient in the subsequent years, thereby contributing to the dramatic fall of the *Overall Efficiency* percentages. Quite similar trends can be observed in terms of direct jobs and the number of funds managed.

Furthermore, if we extend this analysis to those companies that remain efficient at the different management stages, Table I-A-1 (Appendix A) shows that in 2005, only 3 out of the 28 efficient units at the first stage remain efficient at the second stage. Therefore, only these companies have marketing departments capable of appropriately selling their efficiently managed mutual funds, but these companies managed only a residual percentage of the assets in the Spanish fund industry in December 2005. Table I-A-1 also shows 7 efficient companies at the *Portfolio Management stage* that remain efficient overall; the interesting evidence about these units is that they managed more than 42% of assets and unitholders' accounts and approximately 16% of the direct jobs in the sample. In addition, 4 out of the 7 efficient companies in the marketing stage remain efficient overall, and they only manage 0.06% of the assets, 0.07% of unitholders, and 1.26% of the direct jobs. Finally, we find that there are only three

companies that are efficient through the entire multi-stage model previously proposed, but their economic relevance in the market figures is very low.

Again, if we extend this analysis to 2006, Table I-A-2 (Appendix A) shows that only 1 out of the 22 efficient units at the first stage remains efficient at the second stage. Therefore, this is the only company with a marketing department capable of appropriately selling their efficiently managed mutual funds. Similar to 2005, this company managed a residual percentage of the assets in the Spanish fund industry as of December 2006. Table I-A-2 also shows that only 3 companies efficient at the *Portfolio Management stage* remain efficient overall, managing more than 16% of assets and unitholders' accounts and approximately 4% of the direct jobs of the sample. In addition, only 2 out of the 8 efficient companies in the marketing stage remain efficient overall. These overall efficient companies only manage 0.004% of the assets, 0.003% of unitholders, and 0.61% of the direct jobs in the sample. Finally, we find that there is only one company that is efficient throughout the entire multi-stage model, and its participation in the industry is also residual.

Table I-A-3 (Appendix A) illustrates that only 2 out of the 21 efficient units from the first stage in 2007 remain efficient at the second stage. Similar to previous findings, these companies managed a residual percentage of the assets in the Spanish fund industry as of December 2007. Table I-A-3 also shows that only 2 efficient companies at the *Portfolio Management stage* remain efficient overall, managing a residual percentage of the assets, unitholders' accounts, and direct jobs. Finally, 3 out of the 6 efficient companies in the marketing stage remain efficient overall, with a little relevance in economic terms. Finally, we find that there is only one company, the same unit as in 2006 that is efficient for the entire multi-stage model.

Repeating the analysis for 2008, we find evidence that the effect of the financial crisis on the mutual fund industry is important: Table I-A-4 shows that no company that is efficient at the first stage remains efficient at the second stage. Therefore, the crisis meant that marketing departments were not able to sell their efficiently managed mutual

funds. Table I-A-4 also shows that only 1 company that is efficient at the *Portfolio Management stage* remains efficient overall, with a residual percentage of the 2008 figures. In addition, 2 out of the 2 efficient companies in the marketing stage remain efficient overall and also shared a residual percentage of the 2008 figures. Obviously, no company was efficient for the entire multi-stage model for this financial crisis year.

Lastly, Table I-A-5 shows that only 2 out of the 20 companies efficient at the first stage remain efficient at the marketing stage, although managing a residual percentage of the assets in the Spanish fund industry as of December 2009. If we compare the 20 efficient companies at the *Portfolio Management stage* with the 5 overall efficient companies, we find 3 units that are efficient at both stages, with a residual role in the industry in terms of the variables described. In addition, only 1 out of the 7 companies efficient in the marketing stage remains efficient overall. This company is the only efficient unit for our entire multi-stage model and represents a residual percentage of the fund industry figures.

The previous findings show that companies are better portfolio managers than sellers, which appears to affect the *Overall Efficiency* results obtained by our sample. In addition, the detailed explanation in Appendix A demonstrates the difficulty of achieving efficiency at every management stage proposed by our model.

Despite the extended information provided by Appendix A for the efficiency scores, it is also necessary to further discuss the interaction between the different management stages included in our conceptual model to better understand these subsystems as a whole for the company. The Spearman rank correlation is used as a non-parametric tool⁵³ to verify the robustness of these efficiency rankings across the different

⁵³ Spearman's rank correlation measures the relationship or dependence between two rankings. If the Spearman correlation obtains values of +1 or -1, each ranking is a perfect monotone function of the other. The measure is obtained as $r_s = 1 - \frac{6 \sum d^2}{n(n^2-1)}$, where d^2 is the difference in paired ranks squared and n is the number of cases.

management stages of the MFMCs. Table I-5 shows this correlation for the different rankings obtained by the CRS-Slacks Based Measures of efficiency.⁵⁴

Table I-5 Average CRS-SBM scores and Spearman's rank correlation

Year	MMS	Average CRS-SBM Scores	Eff.	Spearman Rank Corr. (Scores)		
				Stage 1	Stage 2	Overall Eff.
2005	Stage 1	0.53359	28	1	-0.405**	-0.194
05	Stage 2	0.26553	7		1	0.560**
5	Overall Eff.	0.25697	10			1
2006	Stage 1	0.48751	22	1	-0.498**	-0.293**
06	Stage 2	0.32233	8		1	0.427**
6	Overall Eff.	0.26854	8			1
2007	Stage 1	0.51105	21	1	-0.269**	-0.204*
07	Stage 2	0.34870	6		1	0.370**
7	Overall Eff.	0.19564	6			1
2008	Stage 1	0.37089	13	1	-0.640**	-0.214*
08	Stage 2	0.13860	2		1	0.525**
8	Overall Eff.	0.19781	5			1
2009	Stage 1	0.45114	20	1	-0.348**	-0.107
09	Stage 2	0.26990	7		1	0.708**
9	Overall Eff.	0.19333	5			1

MMS = Multi-management stages
Eff. = Number of efficient companies

* Correlation is significant at the 0.05 level (2-tailed)
** Correlation is significant at the 0.01 level (2-tailed)

Table I-5 Table I-5 reports some major findings from our analysis⁵⁵. The efficiency scores in the *Portfolio Management stage* are the highest for any year analysed. This finding proves that companies are much better portfolio managers than sellers. Moreover, the rank correlation coefficient demonstrates a negative relationship between portfolio management and the marketing stage of the companies from 2005 to 2009, which significantly suggests that the commercial skills of the companies are inversely related to the portfolio management abilities of the mutual fund managers. Furthermore, we also find similar evidence, but less significant, after comparing the efficiency ranks

⁵⁴ The major interest of this first part of the study is in analysing the efficiency ranks across the different management stages of the company, while the next second part focuses attention on the ranks of the SBM scores along our time horizon from 2005 to 2009 to test the persistence phenomenon for the efficiency of the different stages in our multi-stage management model.

⁵⁵ The conclusions are also quite similar after applying the Kendall rank correlation coefficient. Results are available upon request.

between portfolio management and the overall efficiency of the companies from 2005 to 2009. Both previous results provide evidence that pure portfolio management skills are not directly related to either the marketing distribution system or to the overall efficiency of a fund company.

However, we find a significant and positive rank relationship between the commercial and the overall efficiency for the entire time horizon. That is, the ability of a company to sell mutual funds proves to be a much more relevant factor for explaining the profitability of a fund company instead of pure portfolio management skills.

These results show a very interesting efficiency pattern in the Spanish fund industry. In general terms, those companies that best manage mutual funds are not the most efficient distributors of their mutual funds, thereby considerably reducing the company's overall profits. These findings should question the importance attributed to the major activity of the company (portfolio management) in creating the profits generated for the shareholders, i.e., it appears to be much more important to sell than to manage mutual funds. This evidence could severely erode the relevance of the portfolio managers' role in a fund company.

Moreover, during the starting impact of the financial crisis in 2008, marketing efficiency was the lowest for the whole time horizon, thereby highlighting the important negative effects of the global financial crisis on the confidence of mutual fund investors.

4.3 Influence of the Variable>Returns-to-Scale on the Efficiency Rankings

As previously discussed in subsection 2.2, the BCC DEA model (equation I.8) proposed by Banker et al. (1984) was suggested for adjusting the CCR DEA model (equation I.6) into variable returns to scale (VRS). The rationale of this VRS approach is that the use of the CCR DEA model results in mistaken measures of scale efficiencies if not all DMUs are operating at the optimal scale. In this section, we extend the analysis of the scale effect in the efficiency rankings obtained by the original SBM approach

(equation I.15) applied to our multi-management stages approach (Figure I-5), i.e., we check for the effect of companies with not optimal scale on the efficiency scores and rankings previously obtained in subsection 4.2. Table I-B-1, Table I-B-2, Table I-B-3, Table I-B-4, and Table I-B-5 included in Appendix B show the results.

A preliminary analysis of these tables shows the existence of some changes in the efficiency rankings using SBM under variable returns to scale (VRS) from those previously obtained under the constant returns to scale assumption. The Spearman rank correlation will be applied to better understand the significance of these changes between the different management stages; see Table I-6.

Table I-6 Average VRS-SBM scores and Spearman rank correlation

Year	MMS	Average VRS-SBM Scores	Eff.	Spearman Rank Corr. (Scores)		
				Stage 1	Stage 2	Overall Eff.
2005	Stage 1	0.69497	41	1	0.211*	0.056
	Stage 2	0.52572	25		1	0.451**
	Overall Eff.	0.35588	17			1
2006	Stage 1	0.62197	33	1	0.081	-0.008
	Stage 2	0.55330	26		1	0.319**
	Overall Eff.	0.37623	18			1
2007	Stage 1	0.61534	29	1	0.131	0.005
	Stage 2	0.52311	16		1	0.418**
	Overall Eff.	0.28568	11			1
2008	Stage 1	0.54304	27	1	0.269**	0.094
	Stage 2	0.41883	14		1	0.246*
	Overall Eff.	0.33042	11			1
2009	Stage 1	0.58171	26	1	0.168	0.111
	Stage 2	0.50108	20		1	0.410**
	Overall Eff.	0.31910	15			1

MMS = Multi-management stages * Correlation is significant at the 0.05 level (2-tailed)
Eff. = Number of efficient companies ** Correlation is significant at the 0.01 level (2-tailed)

Firstly, Table I-6 shows the average SBM scores under the VRS hypothesis and the Spearman rank correlation between the multi-management stages of our model.⁵⁶ This table also shows similar trends to those found in Table I-5 for the CRS-SBM scores, i.e., the highest scores are obtained by the portfolio managers of the company while the

⁵⁶ The conclusions are also quite similar after applying the Kendall rank correlation coefficient. Results are available upon request.

overall efficiency provides the lowest scores for any year analysed. Moreover, we also find the expected result of higher VRS-SBM average scores than under the CRS assumption as a consequence of the VRS model better capturing the scale. A comparison of Table I-5 and Table I-6 shows that the most important increments in the average scores are obtained for the marketing stage, which may prove the relevance of the fund company's scale to its ability to successfully sell funds. This finding is especially relevant in a concentrated market with heavy bank participation such as the Spanish fund industry.

A more detailed look at Table I-B-1, Table I-B-2, Table I-B-3, Table I-B-4, and Table I-B-5 (Appendix B) shows the increase in the number of efficient companies for every multi-management stage of our model due to better capturing of the scale effect. However, the Spearman rank coefficients prove that the VRS hypothesis does not significantly affect the major findings displayed by Table I-5, i.e., commercial skills appear to be a much more relevant factor for explaining the overall profitability of a MFMC instead of pure portfolio management abilities. Table I-7 provides additional evidence of the similarity in the efficiency rankings obtained by CRS-SBM and VRS-SBM methods for the *Portfolio Management stage* and for the overall results of the company. The lowest correlations detected in the marketing stage provide robustness to the conclusions obtained from the comparison between the average CRS-SBM (Table I-5) and VRS-SBM (Table I-6) scores, that is, size appears to be an important determinant in successfully commercialising funds. This result is even more evident during a period with more difficulties selling funds, such as 2008.⁵⁷

Table I-7 Spearman rank correlation between CRS-SBM and VRS-SBM eff. scores

MMS	2005	2006	2007	2008	2009
Stage 1	0.619**	0.685**	0.756**	0.615**	0.713**
Stage 2	0.373**	0.379**	0.591**	-0.055	0.319**
Overall Eff.	0.857**	0.858**	0.862**	0.779**	0.685**

** Correlation is significant at the 0.01 level (2-tailed)

⁵⁷ The conclusions are also quite similar after applying the Kendall rank correlation coefficient. Results are available upon request.

The previous findings could provide evidence that not all companies are operating at the optimal scale at the commercial stage, which could result in biased efficiency estimates under constant returns to scale. The importance of scale to selling funds justifies the use of more appropriate benchmarks in the SBM approach to find more accurate results. This issue motivates the application, in the next part of the thesis, of those variants to the original SBM recently proposed by Tone (2010).

4.4 Robustness of the More Controversial Variables of the Model

It is necessary to evaluate the impact on the efficiency evaluation of new measures for those variables included in our unoriented multi-management stages model. We must control for the changes in the new SBM efficiency rankings to verify the robustness of the variables initially included in our analysis.

According to the review of variables previously included in the financial literature (see section 3 of this first part), the most controversial variables that have been measured in a number of different ways in the efficiency studies are labour, capital and portfolio risk. Regarding this latter variable, portfolio risk was included as an input in the *Portfolio Management* stage, as is done by nearly all DEA studies on mutual funds. This variable has been extensively measured using the standard deviation of fund returns; while other asymmetric measures have been applied to hedge fund studies (see the detailed literature review of Lozano and Gutiérrez, 2008b). According to the construction of our sample previously described in subsection 4.1, this study focuses on those management companies mostly working with mutual funds, which supports the use of the standard deviation of mutual fund returns to measure risk instead of running additional models using asymmetric risk measures that are more suitable to evaluating the efficiency of hedge funds.

Regarding labour, which is used as an input in our *Portfolio Management* stage, we believe that this variable should be a proxy of the human resources necessary to appropriately develop the major activity of the MFMC, i.e., portfolio management. As

was discussed in subsection 2.3.2 of this first part, the number of employees has been commonly applied in the DEA efficiency studies (Lang and Welzel, 1996; Hussels and Ward, 2007; Chen et al, 2009). This variable has also been measured as the total expenditure on wages at the end of each fiscal period (Berg et al, 1991; Berg et al, 1993; among other initial studies); as the percentage of wages from total expenses (Cummins et al., 2010); as the weighted average of annual wages per worker (Berger et al., 2009); and finally as a normalised variable of wages considering the assets under management (Casu et al., 2004). According to this literature, we use personnel expenses instead of number of employees as a new measure for the input *labour* to evaluate the efficiency of the *Portfolio Management stage* in our sample of fund companies.

Table I-8 shows the comparison between the new efficiency scores and those initially obtained by our original model in subsection 4.2.⁵⁸

Table I-8 Stage 1-Robustness analysis based on labour data

Year	Data	Average CRS-SBM Scores	Eff.	Spearman Rank Corr.
2005	Number of Employees	0.53359	28	0.964**
	Personnel Expenses	0.54475	26	
2006	Number of employees	0.48751	22	0.970**
	Personnel Expenses	0.49152	24	
2007	Number of Employees	0.51105	21	0.975**
	Personnel Expenses	0.50506	21	
2008	Number of Employees	0.37089	13	0.900**
	Personnel Expenses	0.34447	13	
2009	Number of Employees	0.45114	20	0.955**
	Personnel Expenses	0.44970	20	

Eff. = Number of efficient companies in each model

** Correlation is significant at the 0.01 level (2-tailed)

⁵⁸ The comparative scores are shown in Appendix C (Table I-C-1, Table I-C-2, Table I-C-3, Table I-C-4, and Table I-C-5). These scores are obtained assuming the Constant>Returns-to-Scale hypothesis.

Table I-8 shows that both the number of efficient MFMCs and the average SBM scores are very similar, thereby providing evidence for the robustness of our variable *Labour*. However, the lower correlations found in 2008 and 2009 appear to be explained by the lower value of the commissions paid to personnel caused by the financial crisis, which should affect the robustness of both measures of labour. In any case, the correlation is higher than 0.9 and still quite significant.

With respect to *Capital*, we use shareholders' equity value as a measure to represent the capital effort of the company to appropriately manage portfolios. To verify the robustness of this input in our *Portfolio Management* stage, we use equipment as a new proxy for this capital effort. This measure has been extensively applied in the DEA banking literature (Berg et al., 1993; Brockett et al., 1997).

Table I-9 shows a comparison between the new efficiency scores for the *Portfolio Management* stage using furniture and equipment and those initially obtained using shareholders' equity.⁵⁹

Table I-9 Stage 1- Robustness analysis based on the capital effort data

Year	Data	Average CRS-SBM Scores	Eff.	Spearman Rank Corr.
2005	Shareholders' equity	0.53359	28	0.748**
	Equipment	0.38171	21	
2006	Shareholders' equity	0.48751	22	0.668**
	Equipment	0.43914	26	
2007	Shareholders' equity	0.51105	21	0.637**
	Equipment	0.42300	21	
2008	Shareholders' equity	0.37089	13	0.718**
	Equipment.	0.30537	13	
2009	Shareholders' equity	0.45114	20	0.742**
	Equipment	0.37362	20	

Eff. = Number of efficient companies in each model

** Correlation is significant at the 0.01 level (2-tailed)

⁵⁹ The comparative scores are shown in Appendix D (Table I-D-1, Table I-D-2, Table I-D-3, Table I-D-4, and Table I-D-5). These scores are obtained assuming the Constant>Returns-to-Scale hypothesis.

We find positive and significant rank correlations of approximately 0.7, which rejects any significant changes in the SBM rankings obtained from using equipment instead of shareholders' equity value. However, these changes are more important than those detected in Table I-8, which may be because several companies do not reflect a high value of furniture and equipment in their balance sheet account considering the amount of assets under management. These surprising results could show that some companies use leasing and other financial instruments to invest in the equipment necessary to manage their portfolios. In our view, this potential bias is overcome by a proxy for the financial effort of the company's shareholders, which is traditionally measured in the literature using the shareholders' equity value.⁶⁰

Finally, *Capital* was also included as an input to the *Overall Efficiency* stage. We then run the SBM model using equipment instead of shareholders' equity value to control for the robustness of this variable in the overall stage. The high and significant rank correlations displayed by Table I-10 provide evidence of the robustness of this variable in the *Overall Efficiency* stage, i.e., measuring *Capital* using shareholder's equity value or furniture and equipment is not a very important choice when evaluating the overall efficiency of the fund company.⁶¹

⁶⁰ In accounting terms, all leasing operations should have been recognised as an intangible asset before December 2008. Currently, if a tangible asset is financed by leasing, this asset must be accounted for as a tangible asset.

⁶¹ The comparative scores are shown in Appendix E (Table I-E-1, Table I-E-2, Table I-E-3, Table I-E-4, and Table I-E-5). These scores are obtained assuming the Constant>Returns-to-Scale hypothesis.

Table I-10 Overall efficiency-robustness analysis based on the capital effort data

Year	Capital as:	Average CRS-SBM Scores	Eff.	Spearman Rank Corr.
2005	Shareholders' equity	0.25697	10	0.952**
	Equipment	0.12418	4	
2006	Shareholders' equity	0.26854	8	0.953**
	Equipment	0.13192	5	
2007	Shareholders' equity	0.19564	6	0.916**
	Equipment	0.10073	5	
2008	Shareholders' equity	0.19781	5	0.915**
	Equipment	0.08937	3	
2009	Shareholders' equity	0.19333	5	0.894**
	Equipment	0.11569	2	

Eff. = Number of efficient companies in each model

** Correlation is significant at the 0.01 level (2-tailed)

The above robustness analyses provide reliable measurement for the variables that have traditionally been subject to greater variability in measurement in the DEA efficiency literature, such as *Labour* and *Capital*. That is, the use of the most common measures for these controversial variables does not drive changes to the efficiency evaluations that would significantly affect the findings previously obtained through the original SBM approach in subsection 4.2.

5 Conclusion and Summary

To our knowledge, this study is the first analysis of the efficiency of Mutual Fund Management Companies in a relevant Euro fund industry. Based on the sub-DMUs approach of Holod and Lewis (2011), the thesis develops a model that includes three interacting management subsystems within a mutual fund company originally proposed by Berkowitz and Qiu (2003): *Portfolio Management*, *Marketing and Service*, and *Overall Efficiency*. The interaction between the inputs and outputs of these different units attempts to overcome the traditional debate between the production versus the intermediation approach in the financial industry, especially in banking and insurance companies.

We discuss a specific set of variables to be included in this multi-subsystem model. The measures of these variables are not a mere replication of those inputs/outputs traditionally considered in the banking and insurance studies because it is necessary to take into account the particular characteristics of the fund companies to develop an appropriate model. The consistency of the efficiency results obtained using different measures for the more controversial variables in the literature proves the robustness of our findings.

The application of a non-oriented frontier approach using slacks as proposed by Tone (2001) has been discussed in detail to justify using this technique as an appropriate tool to obtain efficiency scores for Mutual Fund Management Companies.

The efficiency rankings provide evidence for the low impact of portfolio management abilities on the efficiency of the marketing and sales process of the mutual funds managed by the company. Furthermore, the lower efficiency scores found in this commercial stage appear to affect to the overall profits reported to the company shareholders. That is, the results support the evidence that the best-managed funds are not the most efficiently sold in the commercial stage of the company, thereby

considerably reducing the overall profits for shareholders. This major finding from this first part is quite robust to the scale effects in the companies and along all of the years considered in our time horizon. However, the influence of returns-to-scale suggests that size appears to be an important variable for successfully selling funds. This issue is especially relevant in a concentrated market with heavy bank participation such as the Spanish fund industry.

Further research to complement the major findings of this first part of the thesis is included in the next second part of the thesis: 1) the application of recent variants to the non-oriented slacks-based measure of efficiency (Tone, 2010) to overcome some of the limitations potentially present in this frontier methodology as a consequence of inappropriate benchmarking of the companies analysed; 2) the persistence of the efficiency results across the time horizon of the study; and 3) a further discussion of the major factors that could potentially drive the efficiency results obtained by the companies.

Appendix A

Table I-A-1 SBM-efficiency under CRS for 2005

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency		
	SBM	Ref.	Rank	SBM	Ref.	Rank	SBM	Ref.	Rank
1	0.33795	12 161	66	0.08966	59	91	0.04431	196	99
2	0.40030	12 177	59	0.11600	59	69	0.04597	100	97
4	0.42223	14 161	55	0.11224	59	74	0.14556	125	55
6	1.00000	6	1	0.12635	59	61	0.09447	196	71
7	1.00000	7	1	0.11275	59	72	0.07625	100	82
9	0.24942	12 177	78	0.18803	59	47	0.27668	100	25
12	1.00000	12	1	0.01733	59	102	1.00000	12	1
14	1.00000	14	1	0.12632	59	62	1.00000	14	1
15	0.54453	12 161	44	0.08926	59	92	0.13830	200	57
20	0.49983	12 161	48	0.08596	59	95	0.26639	196	28
21	0.21632	161	80	0.12651	59	60	0.20771	125	36
24	0.41055	12 177	56	0.10505	59	78	0.06403	100	89
29	0.45102	12 43 177	53	0.10109	59	83	0.08061	100	79
31	0.64661	12 177	36	0.14563	59	54	0.07928	100	80
34	0.20845	12 177	81	0.15740	59	49	0.10354	196	68
35	0.81735	12 40 160 176 177	30	0.14150	59	55	0.08897	100	75
36	0.17644	12 174	85	0.48623	46	14	0.34542	100	19
37	0.32688	12 40 174 177	68	0.26371	46	30	0.17881	100	43
38	0.12944	12 160 176	89	0.11694	59	67	0.19097	100	40
40	1.00000	40	1	1.00000	40	1	1.00000	40	1
42	0.11366	12 55 174	93	0.48345	46	15	0.33274	100	20
43	1.00000	43	1	0.08761	59	93	0.05890	100	90
45	0.55011	12 174 177	43	0.15939	59	48	0.21755	100	35
46	0.06900	12 152	96	1.00000	46	1	0.58714	100	14
47	0.48304	12 176	49	0.12083	59	64	0.07059	100	84
49	0.38867	12 40 174 177	60	0.26299	46	31	0.19546	100	39
50	0.14370	12 177	88	0.83008	46 59 205	8	0.22464	100	33
51	0.58078	12 40 160 161	39	0.21977	59	39	0.13051	100	59
53	1.00000	53	1	0.59927	46 196	11	0.26866	100	27
55	1.00000	55	1	0.04260	59	101	0.30758	196	21
57	0.28221	12 174	74	0.20929	46	42	0.15238	100	51
58	0.70651	12 161	35	0.11267	59	73	0.18955	196	41
59	0.02740	12 161	99	1.00000	59	1	0.14996	125	53
60	0.00646	12 40	102	0.58960	46	12	0.35570	196	18
61	0.31073	55 161	70	0.09937	59	84	0.03392	196	100
62	0.64389	12 160 161 176	37	0.15399	59	51	0.15798	100	49
63	0.76928	12 43 154 160	31	0.10850	59	75	0.04542	100	98
69	0.11333	12 174 177	94	0.41402	46 196	18	0.12470	100	61
71	1.00000	71	1	0.10392	59	79	0.10305	100	70
73	0.70830	12 174 177	33	0.20019	59	44	0.17174	100	45
75	0.47698	12 161	50	0.07912	59	97	0.05309	100	94
76	1.00000	76	1	0.09364	59	86	0.26174	100	29
78	0.35690	12 174 177	63	0.23072	59	37	0.12188	100	62
83	0.61178	12 55 161	38	0.09589	59	85	0.07843	100	81
84	0.45669	14 161	52	0.09259	59	87	0.49052	125 196	15
85	0.46772	12 161	51	0.06204	59	100	0.15520	196	50
86	0.50307	161	46	0.09105	59	88	0.23166	100	32
93	0.19591	12 55 161	83	0.26485	59	29	0.13984	196	56
95	0.73830	55 161	32	0.11970	59	66	0.28886	100	24
98	0.38578	12 43 160	62	0.08995	59	90	0.18462	100	42
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
102	0.03736	12 152	98	0.76026	40 196	9	0.65068	100 196	13
103	1.00000	103	1	0.15074	59	52	1.00000	103	1
105	0.41006	12 163 177	57	0.08666	59	94	0.01924	100	101
110	0.56571	12 161	40	0.10239	59	81	0.08687	100	77
113	0.17183	12 14 161	86	0.10707	59	76	0.01282	196	102
115	0.15611	12 100 174	87	0.32630	46	24	0.17119	100	46
121	0.27680	161	75	0.09003	59	89	0.08832	196	76
123	0.12551	12 160 177	91	0.19068	59	46	0.11286	100	63
125	0.01309	14 161	101	0.44341	46	16	1.00000	125	1
126	0.35117	12 177	65	0.19487	59	45	0.09324	100	72

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127	1.00000	127	1	0.07950	59	96	0.10429	100	67
128	1.00000	128	1	0.10130	59	82	0.10884	196	65
130	1.00000	130	1	0.10245	59	80	0.06501	100	86
131	0.12109	12 152	92	0.32630	46	25	0.29562	100	22
132	0.26751	12 161 177	76	0.14965	59	53	0.08999	100	74
133	0.28577	12 40 174	73	0.38862	46	22	0.24787	100	31
137	0.30621	12 174	72	0.21974	59	40	0.20631	100	37
138	0.18856	12 152	84	1.00000	138	1	0.73150	100 152	11
139	0.20468	12 177	82	0.28047	59	28	0.41251	196	16
140	1.00000	140	1	0.12058	59	65	0.05620	100	92
142	0.35601	161	64	0.07777	59	98	0.08564	196	78
152	1.00000	152	1	0.41096	196	20	1.00000	152	1
154	1.00000	154	1	0.13565	59	58	0.04904	100	96
156	0.31727	12 55 174	69	0.24016	59	35	0.11235	100	64
159	0.32720	12 55 174 177	67	0.25823	59	33	0.10316	100	69
160	1.00000	160	1	0.25883	46	32	0.15886	100	48
161	1.00000	161	1	0.07408	59	99	0.07363	100	83
162	0.70769	12 55 154 176 177	34	0.13594	59	57	0.12509	100	60
163	1.00000	163	1	0.13043	59	59	0.05362	100	93
164	0.06441	12 174	97	0.72657	40 196	10	0.69829	125 196	12
168	0.23644	12 55 174	79	0.23932	59	36	0.15203	100	52
173	0.30974	12 177	71	0.11520	59	71	0.05016	100	95
174	1.00000	174	1	0.35947	46	23	0.37122	100 205	17
176	1.00000	176	1	0.10591	59	77	0.06601	100	85
177	1.00000	177	1	0.15581	59	50	0.09050	100	73
182	0.55858	12 176 177	41	0.11549	59	70	0.06486	100	87
185	0.38616	12 43 154	61	0.20259	59	43	0.27353	100	26
189	0.12673	12 152	90	0.49939	46 196	13	0.29159	100	23
190	0.55505	12 176 177	42	0.12094	59	63	0.06482	100	88
191	0.40856	12 174 177	58	0.28520	59	27	0.16560	100	47
192	1.00000	192	1	0.14081	59	56	0.10601	100	66
193	1.00000	193	1	0.11656	59	68	0.05747	100	91
194	0.82258	12 43 174 198	29	0.21982	59	38	0.17578	100	44
195	0.53370	12 174 198	45	0.42547	46 59	17	0.24808	100	30
196	0.01433	12 40 152	100	1.00000	196	1	1.00000	196	1
197	0.50042	12 174 177	47	0.25072	46	34	0.13386	100	58
198	1.00000	198	1	0.30823	59	26	0.14766	100	54
200	0.42806	55 174 177	54	0.21853	59	41	1.00000	200	1
202	0.25857	12 40 161 174 177	77	0.41174	46	19	0.22371	100	34
205	1.00000	205	1	1.00000	205	1	1.00000	205	1
206	0.10614	12 152 174	95	0.39816	196	21	0.20326	100	38

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-A-2 SBM-efficiency under CRS for 2006

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency		
	SBM	Ref.	Rank	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.44895	12 177 205	47	0.16883	196	80	0.05102	100	101
4	0.36609	14 51	57	0.20362	196	63	0.20237	103	41
6	0.24471	51 55 161	72	0.19315	196	67	0.13330	100	63
7	0.80473	12 51 53 154 163	26	0.18790	196	68	0.08920	100	82
9	0.21437	140 174	75	0.28269	196	35	0.33738	100	22
12	1.00000	12	1	0.00577	202	102	0.48454	14 103	14
14	1.00000	14	1	0.17761	202	74	1.00000	14	1
15	0.56122	12 51 127	34	0.00899	202	101	0.12844	103	66
20	0.54491	12 51 127	36	0.14657	196	93	0.32461	103	25
21	0.24047	127	73	0.17893	196	72	0.28615	103	31
24	0.42342	12 177 205	51	0.15599	196	88	0.07796	100	88
29	0.41718	12 128 205	52	0.16855	196	81	0.12851	100	65
31	0.84051	12 51 53 154 163	25	0.28779	196	34	0.12455	100	69
34	0.16519	12 177	84	0.47668	196 202	18	0.07962	100	87
35	0.51953	12 174	39	0.19937	196	64	0.10131	100	78
36	0.16393	12 14 174	85	0.45964	203	20	0.36283	100	19
37	0.37920	12 174 205	56	0.25101	196	45	0.19572	100	43
38	0.10118	12 51 163	95	0.20477	196	60	0.30177	100	28
40	0.10521	174	94	1.00000	40	1	0.98297	100 103	9
42	0.14496	14 174	88	0.41011	203	22	0.39169	100	17
43	0.69014	128 154 163 198 205	28	0.15153	196	91	0.06698	100	93
45	0.50592	12 174	41	0.21997	196	54	0.32857	100	23
46	0.12119	12 205	91	0.74057	196 203	10	0.61736	100	11
47	0.41186	12 51	53	0.19877	196	66	0.11071	100	72
49	0.32253	12 174	63	0.26393	196	42	0.25735	100	36
50	0.12725	12 174	90	0.59623	196 203	13	0.22569	100	38
51	1.00000	51	1	0.16411	196	83	0.11605	100	71
53	1.00000	53	1	0.48690	196 203	17	0.29695	100	30
55	1.00000	55	1	0.10510	196	99	0.38660	103	18
57	0.34304	174	60	0.22792	196	50	0.17904	100	47
58	0.53423	12 51 128	38	0.17201	202	78	0.27552	103	34
59	1.00000	59	1	0.71181	196 203	12	0.02049	125	102
60	0.00147	12 205	102	1.00000	60	1	0.56232	100 125	13
61	0.20216	14 161	78	0.16816	196	82	0.09254	100	80
62	1.00000	62	1	0.22136	196	53	0.17945	100	46
63	0.62818	12 53 128 163 205	32	0.14954	202	92	0.05501	100	97
69	0.16152	12 174 205	86	0.29226	203	33	0.15998	100	51
71	1.00000	71	1	0.16032	196	86	0.07392	100	90
75	0.34772	12 51 128	59	0.15477	196	90	0.05768	100	96
76	0.64041	55 174	30	0.16160	196	85	0.32722	100	24
78	0.44933	12 174	46	0.23471	196	47	0.13623	100	60
83	0.73548	12 51 55 127 140	27	0.22539	196	51	0.12697	100	68
84	0.44491	14 51 127	49	0.18098	196	71	0.58563	103	12
85	0.45866	12 51 128	43	0.08541	202	100	0.14745	103	55
86	0.50058	12 51	42	0.12824	196	97	0.30006	100	29
93	0.20408	14 51 55	77	0.27776	196	36	0.15319	100	54
95	0.45104	12 14 51	45	0.19907	196	65	0.43432	100 103	16
98	0.35902	12 51	58	0.15588	196	89	0.27062	100	35
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
102	0.02073	12 174 205	99	0.53768	203	14	0.33771	100	21
103	0.17378	55 127	83	0.22883	196	49	1.00000	103	1
105	0.11507	12 174	92	0.21509	196	55	0.15480	100	53
110	0.51738	12 51	40	0.16956	196	79	0.13768	100	59
113	0.19399	12 14 51	81	0.20468	196	61	0.05241	100	99
115	0.23929	14 174	74	0.26007	203	43	0.17871	100	48
121	0.19729	14 51	79	0.20370	196	62	0.10587	103	75
123	0.13253	12 177 205	89	0.21301	196	56	0.10904	100	74
125	0.01207	14 51	100	0.46693	196	19	1.00000	125	1
126	0.28710	12 174	66	0.22891	196	48	0.10139	100	77
127	1.00000	127	1	0.13583	196	95	0.08865	100	84
128	1.00000	128	1	0.13471	202	96	0.13552	103	61
130	1.00000	130	1	0.17482	196	75	0.07611	100	89
131	0.11444	12 174 205	93	0.32372	196	31	0.27757	100	32
132	0.45200	12 51 177 205	44	0.26697	196	39	0.13370	100	62

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133	0.26298	12 174	70	0.39194	196	24	0.25268	100	37
137	0.28825	12 174	65	0.27432	196	38	0.30851	100	27
139	0.19623	12 51 154	80	0.36027	196	27	0.48390	100	15
140	1.00000	140	1	0.18767	196	69	0.08654	100	85
142	0.26240	12 51	71	0.18135	196	70	0.08910	100	83
152	0.32485	12 174	62	1.00000	152	1	0.34260	100	20
154	1.00000	154	1	0.13960	196	94	0.05299	100	98
156	0.27883	12 14 174	68	0.27650	196	37	0.10983	100	73
159	0.26417	12 174	69	0.33236	196	30	0.09063	100	81
160	1.00000	160	1	0.26484	196	40	0.18193	100	45
161	1.00000	161	1	0.11123	196	98	0.09544	100	79
162	0.54073	14 51 55 140 177	37	0.21295	196	57	0.16337	100	50
163	1.00000	163	1	0.16022	196	87	0.05147	100	100
164	0.08552	12 174 205	97	0.71796	196 203	11	1.00000	164	1
168	0.17788	12 14 174	82	0.49530	196 202	16	0.16436	100	49
173	0.43200	12 177	50	0.16336	196	84	0.07258	100	92
174	1.00000	174	1	0.38481	196	25	0.31727	100	26
176	0.85503	12 140 161 177	24	0.17342	196	76	0.07287	100	91
177	1.00000	177	1	0.17241	196	77	0.12188	100	70
182	0.63326	12 51 53 163 177	31	0.20785	196	59	0.10220	100	76
185	0.28203	51 128	67	0.33509	196 202	29	0.13982	103	57
190	0.44540	12 51 205	48	0.22225	196	52	0.08337	100	86
191	0.39387	12 174	54	0.25457	196	44	0.15846	100	52
192	0.59081	161 174	33	0.17847	196	73	0.12837	100	67
193	0.54975	12 174 177	35	0.24157	196	46	0.06506	100	95
194	1.00000	194	1	0.31866	196	32	0.18850	100	44
195	0.21295	12 174 198	76	0.80293	152 196 202	9	0.21477	100	40
196	0.04463	12 205	98	1.00000	196	1	0.85674	100 125 164	10
197	0.38743	12 174	55	0.26409	196	41	0.13269	100	64
198	1.00000	198	1	0.52403	196 202	15	0.13823	100	58
200	0.33249	55 161	61	0.34650	196	28	1.00000	200	1
202	0.30145	12 177 205	64	1.00000	202	1	0.22387	100	39
203	0.00248	205	101	1.00000	203	1	1.00000	203	1
204	0.86527	12 14 174 177 205	23	0.21243	196	58	0.14578	100	56
205	1.00000	205	1	0.42132	203	21	1.00000	205	1
206	0.15011	12 174	87	0.39985	196 203	23	0.19629	100	42
207	0.09895	12 205	96	1.00000	207	1	0.06510	100	94
210	0.68387	12 174 205	29	0.38090	203	26	0.27657	100	33

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-A-3 SBM-efficiency under CRS for 2007

Stage 1 Portfolio Management				Stage 2 Marketing and Service			Overall Efficiency		
MFMC	SBM	Ref.	Rank	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.51806	7 12 50 198	39	0.27159	50	57	0.04937	100	88
4	0.43624	12 14 51 127	46	0.23952	50	74	0.10488	103	54
6	0.26835	14 174 177	69	0.26111	46	61	0.12678	100	42
7	1.00000	7	1	0.30253	50	37	0.06810	100	75
9	0.25915	12 177	72	0.28129	46	45	0.25633	103	17
12	1.00000	12	1	0.02373	50	95	0.24853	103	19
14	1.00000	14	1	0.09046	50	94	0.29689	103	12
15	0.62492	12 51 127	34	0.18102	50	92	0.05225	103	86
20	0.17673	12 51	82	0.19869	50	90	0.21514	103	23
21	0.29413	12 51 127	66	0.27917	46	50	0.11364	103	49
24	0.39268	12 152	52	0.27726	46	53	0.05693	100	84
29	0.39638	12 50 198	51	0.26095	50	62	0.09364	100	58
31	0.38277	12 152 177 198	56	0.29044	46	42	0.08811	100	59
34	0.16140	12 50	84	0.42903	46 50	17	0.07355	100	70
35	0.74560	12 174 177 198	26	0.25775	46	66	0.06677	100	76
36	0.17889	12 174	80	0.46345	46	13	0.34604	100	11
37	0.35377	12 152 198	60	0.27020	46	58	0.11603	100	48
38	0.09342	12 51	90	0.27542	46	56	0.21607	100	22
40	0.13492	12 174	89	0.50290	46 50	10	0.24673	100	20
43	0.69769	12 50 128 198	28	0.25369	50	67	0.08387	100	63
45	0.65904	12 174 177 198	30	0.22965	46	79	0.26707	100	16
46	0.13909	12 174 210	87	1.00000	46	1	0.54995	100	7
47	0.44036	12 51	45	0.26043	50	64	0.10175	103	56
49	0.29208	12 198	67	0.42126	46 50	18	0.18861	100	29
50	1.00000	50	1	1.00000	50	1	0.19133	100	27
51	1.00000	51	1	0.22873	46	80	0.10175	100	55
53	1.00000	53	1	0.62385	46 100	8	0.22934	100	21
55	1.00000	55	1	0.18614	50	91	0.17281	103	32
57	0.29513	12 198	65	0.28414	46	44	0.16067	100	34
58	0.40114	12 51 127	50	0.21110	50	88	0.20325	103	25
59	0.03667	12 14 51	92	0.74972	46 100	7	0.02239	125	95
61	0.23730	14 51	73	0.26365	50	59	0.06150	100	81
62	0.62576	12 50 127 177	33	0.38723	46 50	25	0.13891	100	40
63	0.35861	12 51	59	0.22016	50	85	0.02930	103	94
69	0.47606	198	42	0.44864	46 50	16	0.12486	100	43
71	0.68783	12 14 177	29	0.27921	50	49	0.14463	100	37
75	0.39004	12 51 128	54	0.22553	50	83	0.05125	103	87
76	0.64337	14 174 177	32	0.26010	46	65	0.21237	100	24
78	1.00000	78	1	0.32589	46 50	33	0.08658	100	61
83	1.00000	83	1	0.23689	50	76	0.08579	103	62
84	0.43146	12 14 51 127	48	0.21743	50	86	0.40844	103	10
85	0.34176	12 51 128	61	0.17989	50	93	0.07036	103	73
86	0.65432	12 50 51 127	31	0.24629	50	70	0.13901	100	39
93	0.19031	12 14 51	78	0.45340	46 50	14	0.15701	100	35
95	0.58126	12 51	36	0.20359	50	89	0.27348	103	15
98	0.23619	12 50	74	0.27866	46	51	0.27933	100	14
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
102	1.00000	102	1	0.49690	46 100	11	1.00000	102	1
103	0.17974	12 51 127 128	79	0.33876	46	30	1.00000	103	1
105	0.01915	12 14 51	93	0.38316	46	26	0.28315	100	13
110	0.45598	12 51	43	0.27668	50	55	0.12210	100	45
113	0.19658	12 14 51	77	0.22404	50	84	0.04300	103	90
115	0.17387	12 174	83	0.40079	46	21	0.14418	100	38
121	0.23576	12 51	75	0.24177	50	72	0.08012	103	64
125	0.00157	14 51	95	0.52661	100	9	1.00000	125	1
126	0.30830	12 152 177 198	63	0.23754	46	75	0.07084	100	72
127	1.00000	127	1	0.24432	50	71	0.07826	100	67
128	1.00000	128	1	0.22772	50	81	0.05711	103	83
130	0.55948	12 50 51 128	38	0.23219	50	77	0.06568	100	79
131	0.14736	12 198	86	0.30184	46	38	0.18744	100	30
132	0.41934	12 51 177	49	0.21567	46	87	0.10706	100	53
133	0.36801	12 174 198	57	0.41140	46	20	0.20135	100	26
137	0.39150	12 174 198	53	0.27974	46	47	0.25393	100	18
139	0.28044	12 51 128 198	68	0.38813	46	24	0.42997	100	9

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140	1.00000	140	1	0.23008	50	78	0.06866	100	74
142	0.29894	12 51	64	0.24925	50	69	0.05337	103	85
152	1.00000	152	1	0.44911	46 100	15	0.18678	100	31
154	0.90413	12 51 128 140 198	22	0.22616	50	82	0.04190	100	91
156	0.34074	12 14 174 177	62	0.37985	46 50	28	0.09471	100	57
159	0.17862	12 152 198	81	0.37814	46 50	29	0.08677	100	60
160	1.00000	160	1	0.39852	46 50	23	0.10742	100	52
161	0.75521	14 51 55 127 174 177	24	0.32144	50	34	0.07592	100	68
162	0.47837	14 51 127	41	0.29792	50	39	0.11256	100	50
163	1.00000	163	1	0.24070	50	73	0.03624	100	93
168	0.22990	12 14 174 177	76	0.32660	46	32	0.13868	100	41
173	0.35901	12 50	58	0.27699	50	54	0.06195	100	80
174	1.00000	174	1	0.40039	46 50	22	0.16349	100	33
176	0.75386	12 51 149 177 198	25	0.30405	50	35	0.06578	100	78
177	1.00000	177	1	0.27741	46	52	0.07277	100	71
182	0.44175	12 51	44	0.28875	50	43	0.07833	100	65
185	0.26586	51 128	70	0.29669	46	40	0.04871	103	89
190	0.56018	12 50 177 198	37	0.27991	50	46	0.05825	100	82
191	0.60545	12 174 198	35	0.25283	46	68	0.12380	100	44
192	0.38297	14 174 177	55	1.00000	192	1	0.12072	100	46
193	0.71789	12 152 177 198	27	0.26127	50	60	0.03958	100	92
194	0.48534	12 198	40	0.27968	46	48	0.11228	100	51
196	0.04396	12 198	91	1.00000	196	1	1.00000	196	1
197	0.43505	12 198	47	0.29052	46	41	0.07830	100	66
198	1.00000	198	1	0.26090	46	63	0.06653	100	77
200	0.15380	51 127	85	0.30270	46	36	0.47176	103 125	8
203	0.00418	174	94	1.00000	203	1	1.00000	203	1
204	0.78538	12 50 177 198	23	0.41182	46 50	19	0.11611	100	47
206	0.26231	12 152 174	71	0.33323	46	31	0.14614	100	36
207	0.13670	12 174	88	0.38133	46	27	0.07377	100	69
210	1.00000	210	1	0.47154	46 50	12	0.18937	100	28

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-A-4 SBM-efficiency under CRS for 2008

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency		
	SBM	Ref.	Rank	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.33993	14 174	40	0.07661	100	70	0.06936	100	71
4	0.49801	14 127 174	22	0.05026	100	88	0.12356	103	45
6	0.15269	14 174	73	0.09354	100	50	0.09404	46	59
7	0.59259	14 15 154 174	16	0.08171	100	64	0.08422	100	65
9	0.16354	14 174	68	0.17328	100	17	0.28667	46	13
12	1.00000	12	1	0.00580	100	93	0.36512	103	10
14	1.00000	14	1	0.02121	100	92	0.53273	46 103	7
15	1.00000	15	1	0.04369	100	91	0.06101	46	82
20	0.27041	14 127 174	48	0.07434	100	73	0.24916	46	23
21	0.29940	14 127	43	0.08726	100	56	0.10910	103	52
24	0.18848	174	63	0.09683	100	47	0.06890	100	72
29	0.16027	174	69	0.08021	100	67	0.05422	100	86
31	0.23317	174	55	0.10037	100	46	0.05397	100	87
34	0.15519	14 174	71	0.11094	100	39	0.05629	46	84
35	0.46813	14 174	24	0.10075	100	43	0.10524	100	56
36	0.08975	14 174	84	0.29986	100	5	0.37877	46	9
37	0.21857	174	58	0.10292	100	42	0.11621	100	48
38	0.06335	14 174	87	0.10059	100	45	0.09978	46	58
40	0.04120	14 174	88	0.22784	100	8	0.07030	46	70
43	0.22706	14 174	57	0.08061	100	66	0.06407	100	79
45	0.28680	14 174	44	0.11144	100	37	0.26292	100	19
46	0.02133	15 210	90	1.00000	46	1	1.00000	46	1
47	0.32840	14 154 174	41	0.08297	100	62	0.08838	46	62
49	0.16823	15 174	67	0.13015	100	32	0.16023	100	35
50	0.15546	14 174	70	0.21096	100	14	0.28164	100	17
51	0.41660	14 127 154 174	32	0.10495	100	40	0.13157	100	43
53	0.21290	210 217	59	0.22162	100	10	0.40558	100	8
55	1.00000	55	1	0.05008	100	89	0.22664	46	26
57	0.24432	174	52	0.13549	100	29	0.11501	100	49
58	0.43570	12 127 160	28	0.04599	100	90	0.28277	103	16
61	0.18555	14 174	65	0.08615	100	59	0.06024	46	83
62	0.36581	14 174	38	0.11135	100	38	0.14484	100	38
63	0.51151	14 154 174	20	0.06001	100	82	0.02524	46	91
69	0.15155	15 174	74	0.21138	100	12	0.23610	100	25
71	0.40797	14 174	33	0.07507	100	71	0.13102	46	44
76	0.64471	14 55 127 174	14	0.07948	100	69	0.13805	100	41
78	0.34477	14 174	39	0.09341	100	51	0.09133	100	61
83	1.00000	83	1	0.05749	100	84	0.06880	46	73
84	0.39343	14 127 174	35	0.05633	100	85	0.26264	46	20
85	0.53103	12 14 127 160	18	0.05201	100	86	0.08376	46	66
86	0.42106	14 154 174	30	0.06724	100	77	0.06671	46	77
93	0.14790	14 174	75	0.18624	100	15	0.18065	46	31
95	0.51312	14 154 174	19	0.06332	100	80	0.22117	46	27
98	0.10369	174	79	0.08690	100	58	0.20722	46	29
100	0.01315	210	92	1.00000	100	1	1.00000	100	1
103	0.08069	14 174	85	0.14642	100	24	1.00000	103	1
105	0.02505	14 174	89	0.15219	100	22	0.00449	46	93
110	0.43663	14 154 174	27	0.07954	100	68	0.11121	46	51
113	0.18537	14 174	66	0.07449	100	72	0.07830	46	67
115	0.10093	14 174	80	0.16853	100	19	0.19021	100	30
121	0.13894	14 174	77	0.08205	100	63	0.07646	46	68
125	0.00112	14 174	93	0.32678	100	4	1.00000	125	1
126	0.25734	14 174	50	0.09261	100	52	0.09270	100	60
127	1.00000	127	1	0.05888	100	83	0.06516	100	78
128	1.00000	128	1	0.05170	100	87	0.06187	46	81
130	0.56401	14 154 174	17	0.06347	100	79	0.05462	100	85
131	0.09758	15 174	82	0.17022	100	18	0.21404	100	28
132	0.15380	174	72	0.10070	100	44	0.10632	46	55
133	0.23740	15 174	54	0.22745	100	9	0.28295	100	15
137	0.23982	174	53	0.16323	100	20	0.28322	100	14
139	0.19390	154 174	62	0.21125	100	13	0.35653	46	11
140	0.64084	14 154 174	15	0.06216	100	81	0.06238	100	80
142	0.37973	14 154 174	37	0.07049	100	75	0.14721	46	37
152	0.20657	174 210 217	60	0.21544	100	11	0.24542	100	24

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154	1.00000	154	1	0.07043	100	76	0.05124	100	88
156	0.26916	14 174	49	0.13949	100	27	0.13365	100	42
159	0.14658	14 15 174	76	0.12133	100	33	0.10815	100	53
160	1.00000	160	1	0.14901	100	23	0.16151	100	34
161	0.40556	14 55 174	34	0.08691	100	57	0.07374	46	69
162	0.28408	14 174	46	0.08763	100	55	0.10303	46	57
163	0.45947	15 154 174	26	0.06445	100	78	0.05120	100	89
168	0.18805	14 174	64	0.13937	100	28	0.16156	100	33
173	0.10768	174	78	0.08971	100	54	0.04383	46	90
174	1.00000	174	1	0.14526	100	25	0.25869	100	22
176	0.47612	14 154 174	23	0.08379	100	61	0.08507	100	64
177	0.46210	174	25	0.09069	100	53	0.10718	100	54
182	0.24487	14 174	51	0.08063	100	65	0.06770	46	75
185	0.43031	12 127 160	29	0.09539	100	49	0.01699	46	92
190	0.49808	154 174	21	0.07405	100	74	0.06689	100	76
191	0.39324	174	36	0.13382	100	30	0.17292	100	32
192	0.28529	14 174	45	0.09595	100	48	0.12119	46	47
193	0.28315	174	47	0.08454	100	60	0.06802	100	74
194	0.23248	174	56	0.11563	100	35	0.08797	100	63
196	0.01937	15 174	91	0.56400	100	3	0.56694	46	6
197	0.32582	15 174	42	0.11451	100	36	0.13962	100	39
198	1.00000	198	1	0.10335	100	41	0.13824	100	40
200	0.10051	14 174	81	0.18564	100	16	0.34346	103	12
203	0.06892	15 174	86	0.22949	100	7	0.26175	100	21
204	0.41712	14 174	31	0.12076	100	34	0.11325	100	50
206	0.19500	15 174	61	0.13368	100	31	0.15593	100	36
207	0.09353	14 174	83	0.16322	100	21	0.12223	100	46
210	1.00000	210	1	0.14436	100	26	0.26583	100	18
217	1.00000	217	1	0.27604	100	6	1.00000	217	1

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-A-5 SBM-efficiency under CRS for 2009

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency		
	SBM	Ref.	Rank	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.40521	7 14 76	42	0.14137	196	76	0.06684	100	74
4	1.00000	4	1	0.09835	196	90	0.07747	100	64
6	0.13156	14 50 76	82	0.21464	196	34	0.09868	100	55
7	1.00000	7	1	0.15481	196	66	0.08650	100	59
9	0.23876	76	65	0.24172	196	23	0.23953	100	19
12	0.63083	14 46	25	0.00009	196	94	0.00002	100	95
14	1.00000	14	1	0.00003	196	95	0.24104	100	18
15	0.24576	14 46	61	1.00000	15	1	0.05933	100	83
20	0.21855	14 46 51	69	0.18649	196	49	0.23018	100	20
21	0.23710	14 46	66	0.19100	196	45	0.11611	100	49
24	0.25751	14 46 76	57	0.14132	196	77	0.06891	100	72
29	0.38650	45 76 131	45	0.14545	196	71	0.07943	100	62
31	0.51517	7 14 50 51	32	0.29884	196	18	0.07714	100	65
34	0.14753	14 76	79	0.15801	196	64	0.08981	100	57
35	0.56053	46 76 210	28	0.15662	196	65	0.10771	100	52
36	0.06559	46 76	89	0.44315	196	11	0.31329	100	14
37	0.24667	14 46 76	60	0.17962	196	51	0.11910	100	48
38	0.04226	14 50 76	90	0.14600	196	70	0.06778	100	73
40	0.12595	45 46 76 100	84	0.34411	196	15	0.29908	100	15
43	0.64348	7 14 76 128	24	0.18516	196	50	0.07326	100	68
45	1.00000	45	1	0.19426	196	43	1.00000	45	1
46	1.00000	46	1	1.00000	46	1	0.60339	100	7
47	0.38456	14 46 50 76	46	0.17026	196	53	0.08880	100	58
49	0.18292	46 76	75	0.21968	196	30	0.14272	100	37
50	1.00000	50	1	0.40832	196	13	0.26275	100	16
51	1.00000	51	1	0.23329	196	26	0.13888	100	38
53	0.03428	45 76	92	1.00000	53	1	0.35490	100	11
55	1.00000	55	1	0.11516	196	89	0.12805	100	42
57	0.36334	46 76 210	48	0.21488	196	33	0.18196	100	27
58	0.39576	14 46 51	43	0.13294	196	84	0.16127	100	33
61	0.19978	14 50 76	73	0.13415	196	82	0.00117	100	94
62	1.00000	62	1	0.21576	196	32	0.17493	100	28
63	0.45249	14 46 51	38	0.09070	196	91	0.01846	100	93
69	0.03591	45 46 76	91	0.78075	46 53 196	10	0.32852	100	13
71	0.70489	7 14 50 76	22	0.15199	196	68	0.12169	100	46
76	1.00000	76	1	0.16795	196	54	0.12508	100	44
78	0.30618	46 76	53	0.19213	196	44	0.14369	100	36
83	0.37017	14 46 50 76	47	0.12291	196	87	0.04583	100	90
84	0.46416	14 46 51	36	0.19749	196	40	0.17140	100	30
85	0.52118	14 46 51	31	0.08841	196	92	0.07218	100	69
86	0.48328	14 46 51	34	0.17899	196	52	0.06392	100	77
93	0.14737	14 50 76	80	0.29418	196	20	0.19697	100	26
95	0.60825	7 14 46 50 51 140	26	0.14857	196	69	0.06981	100	70
98	0.15652	7 14 76	78	0.22425	196	29	0.09003	100	56
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
103	0.18388	76	74	0.15821	196	63	1.00000	103	1
105	0.01502	14 46	94	0.21197	196	35	0.04605	100	89
110	0.34448	7 14 50 76	49	0.14326	196	73	0.07889	100	63
113	0.17287	14 51	76	0.13521	196	81	0.04774	100	87
115	0.08306	46 50 76 100	86	0.32520	196	16	0.17193	100	29
121	0.24066	14 46 50	64	0.14466	196	72	0.06331	100	79
125	0.00176	14 46	95	0.92911	46 53 196	8	1.00000	125	1
126	0.24345	46 76	63	0.16061	196	61	0.12112	100	47
127	1.00000	127	1	0.14041	196	78	0.06583	100	75
128	1.00000	128	1	0.02592	196	93	0.02722	100	92
130	0.85387	7 14 51 128 140	21	0.15991	196	62	0.05414	100	84
131	1.00000	131	1	0.18828	196	48	0.83185	45 100 217	6
132	0.25683	76	58	0.16460	196	59	0.06399	100	76
133	0.20976	45 46 76	71	0.30729	196	17	0.25524	100	17
137	0.25320	46 76	59	0.20871	196	37	0.21064	100	23

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139	0.45351	46 76 210	37	0.24574	196	22	0.34998	100	12
140	1.00000	140	1	0.13995	196	79	0.04534	100	91
142	0.28848	14 46 51	56	0.12292	196	86	0.06106	100	82
152	1.00000	152	1	0.43377	196	12	0.37313	100	10
154	0.65242	14 46 76	23	0.15444	196	67	0.06351	100	78
156	0.29329	50 76	55	0.18926	196	46	0.12481	100	45
159	0.21585	14 46 76	70	0.18860	196	47	0.11241	100	50
160	0.31310	14 46 76	51	0.21135	196	36	0.13288	100	41
161	0.54754	14 50 76	30	0.14168	196	75	0.06268	100	80
162	0.47966	14 46 76	35	0.12123	196	88	0.07697	100	66
163	0.41834	7 14 50 76	40	0.13196	196	85	0.05326	100	85
168	0.20792	46 76	72	0.20162	196	38	0.16093	100	34
173	0.40896	14 50 51	41	0.16598	196	57	0.04681	100	88
174	1.00000	174	1	0.28618	196	21	0.22081	100	22
176	0.58569	14 50 51 76	27	0.16460	196	58	0.07694	100	67
177	0.12601	14 50 76	83	0.23625	196	25	0.12704	100	43
182	0.38700	14 50 76	44	0.14183	196	74	0.06908	100	71
185	0.32710	14 51	50	0.13941	196	80	0.04822	100	86
190	0.49875	7 14 50 76	33	0.13383	196	83	0.06186	100	81
191	0.55086	46 76 174	29	0.21636	196	31	0.22495	100	21
192	0.15836	14 50 76	77	0.19708	196	41	0.10456	100	54
193	1.00000	193	1	0.16678	196	56	0.08055	100	61
194	0.42958	46 76 210	39	0.16364	196	60	0.10909	100	51
195	0.30678	46 76 210	52	0.16712	196	55	0.08512	100	60
196	0.02243	14 46 76	93	1.00000	196	1	0.53158	100	8
197	0.30023	45 76 174	54	0.24011	196	24	0.15332	100	35
198	0.22702	14 46 50 76	68	0.20008	196	39	0.10700	100	53
200	0.11143	14 46	85	0.29435	196	19	0.16554	100	32
203	0.07050	45 46 76	88	0.37516	196	14	0.20074	100	25
204	0.24552	14 46 76	62	0.22747	196	28	0.13595	100	39
206	0.14670	45 46 76	81	0.23200	196	27	0.16866	100	31
207	0.08097	14 46 76	87	1.00000	207	1	0.13347	100	40
210	1.00000	210	1	0.19581	196	42	0.20425	100	24
217	1.00000	217	1	0.90675	46 53 196	9	1.00000	217	1
221	0.23553	50 76 174	67	1.00000	221	1	0.45852	100 103	9

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Appendix B

Table I-B-1 SBM-efficiency under VRS for 2005

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency		
	SBM	Ref.	Rank	SBM	Ref.	Rank	SBM	Ref.	Rank
1	0.39734	50 161	87	0.11066	59 185	102	0.05293	196 200	100
2	0.43260	12 53 177	81	0.37702	59 142	61	0.04981	100 205	101
4	0.45433	14 161	76	0.76897	58 59 185	27	0.25024	84 103	44
6	1.00000	6	1	0.18046	59 103	94	0.11683	125 196	75
7	1.00000	7	1	0.15148	59 103	100	0.11135	100 200	78
9	0.34252	12 50 174 177	92	0.22094	59 103	84	0.34215	196 200	35
12	1.00000	12	1	1.00000	12	1	1.00000	12	1
14	1.00000	14	1	1.00000	14	1	1.00000	14	1
15	0.63988	12 161	53	0.32240	58 59	71	0.16568	103 200	60
20	0.51692	12 161	65	0.39990	58 59	53	0.50334	84 103	26
21	0.28622	14 50 161	94	0.19362	59 103	92	0.21484	125 196	53
24	0.43524	12 177	80	1.00000	24	1	0.07816	196 200	94
29	0.46306	12 43 177	74	0.20280	59 103	89	0.17983	196 200	58
31	0.68268	12 53 177	50	0.38048	59 103	60	0.08929	100 205	90
34	0.25772	174 177	97	0.54161	59 142	37	0.10711	196 200	82
35	0.85435	12 53 160 176 177	42	0.20474	59 103	87	0.10534	100 196	84
36	1.00000	36	1	0.52594	40 46	40	0.37625	100 196	31
37	0.47087	12 53 174 177 205	73	0.66741	40 103 205	31	0.29904	196 200	38
38	0.27465	12 50 51 154 161	96	0.38737	59 142	59	0.52222	125 196 200	25
40	1.00000	40	1	1.00000	40	1	1.00000	40	1
42	0.54122	50 100 174	61	0.53447	46 60	39	0.34557	100 196	34
43	1.00000	43	1	0.16138	59 103	98	0.08579	100 200	91
45	0.59579	12 174 177	57	0.21028	59 103	86	0.54787	196 200	23
46	1.00000	46	1	1.00000	46	1	1.00000	46	1
47	0.48506	161 176	71	0.50658	59 185	41	0.09238	196 200	88
49	0.50229	12 40 53 174	69	0.29527	40 46	75	0.23770	100 196	48
50	1.00000	50	1	1.00000	50	1	0.23416	100 189	49
51	1.00000	51	1	0.35387	59 103	64	0.14536	100 196	69
53	1.00000	53	1	0.66795	46 60 196	30	0.31657	100 189	36
55	1.00000	55	1	1.00000	55	1	0.62299	14 84 103	21
57	0.33242	12 174	93	0.45973	40 46	47	0.19057	100 196	57
58	0.73505	12 161	46	1.00000	58	1	0.27174	84 103	40
59	1.00000	59	1	1.00000	59	1	1.00000	59	1
60	0.15875	46 50	101	1.00000	60	1	0.43364	125 196	29
61	0.39978	50 161	86	0.71508	59 60 142	29	0.03559	100 196	102
62	1.00000	62	1	0.20441	59 103	88	0.24854	196 200	45
63	0.81975	12 43 161	44	0.55047	58 59	36	0.05347	100 200	99
69	1.00000	69	1	0.45020	46 60 196	48	0.25188	100 189	42
71	1.00000	71	1	0.19410	59 185	91	0.15232	100 200	67
73	0.71778	12 43 174 177	47	0.39312	59 142	57	0.19414	100 205	55
75	0.50137	12 161	70	0.16392	58 59	97	0.06363	196 200	96
76	1.00000	76	1	0.17941	59 103	95	0.47644	196 200	28
78	0.44371	12 174 177	78	0.33894	40 46	68	0.16206	100 196	61
83	0.64544	12 55 128 161	52	0.56196	58 59 185	35	0.10991	196 200	79
84	0.50361	14 161	67	0.43501	58 185	49	1.00000	84	1
85	0.52525	12 161	64	0.39907	59 142	55	0.19952	103 200	54
86	0.50323	161	68	0.23866	59 185	81	0.47850	103 200	27
93	0.51361	50 51 53 161	66	0.46400	59 142	45	0.16004	46 100	64
95	0.77266	50 55 161	45	0.64414	58 59 185	32	0.52372	103 200	24
98	0.42659	12 50 154 160	83	0.18227	59 103	93	0.36779	196 200	33
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
102	1.00000	102	1	1.00000	102	1	0.65068	100 196	19
103	1.00000	103	1	1.00000	103	1	1.00000	103	1
105	0.43223	12 50 154 177	82	1.00000	105	1	1.00000	105	1
110	0.63046	50 161	55	0.28034	59 103	76	0.12674	196 200	72
113	0.25329	50 161	98	0.53917	59 103 185	38	1.00000	113	1
115	0.43550	174	79	0.36456	46 60	62	0.19154	100 196	56
121	0.27722	161 176	95	0.19692	59 185	90	0.10675	196 200	83
123	0.23727	12 50 53	99	0.35161	40 46	65	0.11453	100 196	76
125	0.12736	14 50	102	0.47350	46 59	44	1.00000	125	1
126	0.45257	12 174 177	77	0.24914	40 46	80	0.12967	100 196	71

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127	1.00000	127	1	0.21689	58 59	85	0.16059	196 200	63
128	1.00000	128	1	0.47383	58 59	43	0.13142	103 200	70
130	1.00000	130	1	0.31322	59 185	73	0.10521	100 200	85
131	0.22256	12 205	100	0.33505	40 46	70	0.63202	125 196 200	20
132	0.39645	12 50 176	88	0.22471	59 103	83	0.15230	196 200	68
133	0.60323	53 174 205	56	0.39218	40 46	58	0.26029	100 196	41
137	0.38743	12 174	89	0.27511	40 46	77	0.39163	196 200	30
138	1.00000	138	1	1.00000	138	1	0.74146	100 152 205	18
139	0.39993	12 50 53 177	85	0.31554	40 59	72	0.61212	196 200	22
140	1.00000	140	1	1.00000	140	1	0.09337	100 200	87
142	0.36315	50 161	91	1.00000	142	1	0.11308	196 200	77
152	1.00000	152	1	0.42990	60 196	50	1.00000	152	1
154	1.00000	154	1	0.39942	58 59	54	0.06079	100 200	97
156	0.46183	12 174	75	0.25291	40 46	79	0.11686	100 196	74
159	1.00000	159	1	0.46162	40 46 59	46	0.10955	100 189	80
160	1.00000	160	1	0.34062	40 46	67	0.16094	100 196	62
161	1.00000	161	1	0.22820	59 142	82	0.10748	196 200	81
162	1.00000	162	1	0.73793	58 59 103 163	28	0.22061	100 200	51
163	1.00000	163	1	1.00000	163	1	0.07851	100 200	93
164	0.64881	40 46 53 100 174	51	1.00000	164	1	1.00000	164	1
168	0.41372	12 174	84	0.35529	59 60	63	0.24481	125 196	46
173	0.38499	12 50 53 177	90	0.12893	59 142	101	0.05623	100 196	98
174	1.00000	174	1	0.50328	40 46	42	0.37185	100 205	32
176	1.00000	176	1	0.17768	59 103	96	0.09046	100 200	89
177	1.00000	177	1	0.34543	59 103	66	0.11894	100 200	73
182	0.58070	12 53 176 177	58	0.15336	59 103	99	0.10488	196 200	86
185	0.54027	12 50 154 160	62	1.00000	185	1	0.31444	196 200	37
189	0.47989	100 205	72	1.00000	189	1	1.00000	189	1
190	0.63170	12 50 53 176 177	54	0.42191	59 103	51	0.08577	196 200	92
191	0.53021	12 174	63	0.31025	40 46	74	0.24122	100 196	47
192	1.00000	192	1	1.00000	192	1	0.17446	196 200	59
193	1.00000	193	1	0.27291	59 103	78	0.06393	100 196	95
194	0.82670	12 43 152 174 198	43	0.39790	59 103	56	0.22137	100 200	50
195	1.00000	195	1	1.00000	195	1	0.28930	100 196	39
196	1.00000	196	1	1.00000	196	1	1.00000	196	1
197	0.69078	53 174 177	49	0.33874	40 46	69	0.15604	100 196	66
198	1.00000	198	1	0.40775	192 205	52	0.15979	100 196	65
200	0.57467	14 50 53 161 174	60	0.81630	40 59 103	26	1.00000	200	1
202	0.69576	50 53	48	0.59748	40 46 59	34	0.25188	100 189	43
205	1.00000	205	1	1.00000	205	1	1.00000	205	1
206	0.57683	53 152 174 205	59	0.60397	40 46 196	33	0.22016	100 196	52

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under VRS (Equation I.15 with the restriction $\sum_j \lambda_j = 1$ proposed by Banker et al. (1984) in the called BCC model). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-B-2 SBM-efficiency under VRS for 2006

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency		
	SBM	Ref.	Rank	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.45883	12 174 177	61	0.31221	105 196	79	0.06105	100 205	98
4	0.43956	14 51 127	64	1.00000	4	1	0.20302	40 103	60
6	0.34865	14 174	84	0.40960	105 196	58	0.17303	100 125	62
7	0.81226	51 53 154 163	37	0.19971	83 196	98	0.12452	100 103	79
9	0.21509	12 140 174	96	0.38067	105 203	63	0.37212	100 103	35
12	1.00000	12	1	1.00000	12	1	0.73274	14 103	20
14	1.00000	14	1	1.00000	14	1	1.00000	14	1
15	0.67038	12 51 127 128 161	41	1.00000	15	1	0.13084	40 103	77
20	0.55084	12 51 127	50	0.24677	83 202	93	0.32489	100 103	39
21	0.24959	14 51 53	91	1.00000	21	1	0.29431	40 103	42
24	0.43343	12 177 205	66	0.54860	105 196	41	0.10677	100 200	88
29	1.00000	29	1	0.28144	83 196	86	0.28382	100 200	43
31	0.84179	12 53 154 163 205	35	1.00000	31	1	0.16161	100 103	69
34	0.18767	12 205	99	1.00000	34	1	0.09654	46 100	92
35	0.53036	12 174 177	52	0.30998	105 196	80	0.11282	100 200	84
36	0.61302	46 100 174 205	44	0.46132	196 203	48	0.48417	100 200	27
37	0.40213	12 174 205	72	0.56139	196 204	40	0.35981	100 200	36
38	0.21763	12 53 128	95	0.32626	105 196	74	0.54696	100 125 200	24
40	0.48331	14 46 100 205	56	1.00000	40	1	1.00000	40	1
42	0.46173	46 100 174	59	0.42326	105 203	55	0.57538	100 125 200	22
43	1.00000	43	1	0.15924	83 196	101	0.08950	100 200	94
45	0.52066	12 174 177	54	0.22665	105 196	96	1.00000	45	1
46	1.00000	46	1	1.00000	46	1	1.00000	46	1
47	0.42104	12 51	68	0.26011	105 196	91	0.13144	100 103	76
49	0.33546	12 174	85	0.26657	196 203	89	0.39234	100 200	32
50	1.00000	50	1	0.78554	60 105 196	27	0.31128	100 205	40
51	1.00000	51	1	0.28786	105 203	85	0.17301	100 200	63
53	1.00000	53	1	0.51272	60 196 203	44	0.38294	100 205	34
55	1.00000	55	1	0.43451	15 196	52	0.38677	40 103	33
57	0.34962	12 174 205	83	0.62217	196 204	34	0.24408	100 200	49
58	0.74253	12 51 128 161	38	0.67805	14 83 202	31	0.27602	40 103	45
59	1.00000	59	1	1.00000	59	1	1.00000	59	1
60	0.19902	203 205	98	1.00000	60	1	1.00000	60	1
61	0.25872	14 51 53	90	0.52730	105 196	42	0.12012	100 125	81
62	1.00000	62	1	0.45747	105 196	50	0.27787	100 200	44
63	0.73606	12 51 154 163 205	40	1.00000	63	1	0.05821	100 103	101
69	0.41408	174 205	70	0.31333	196 203	78	0.33400	100 205	37
71	1.00000	71	1	0.49325	105 196	47	0.08687	100 103	95
75	0.41580	12 51	69	0.40463	105 196	59	0.05873	100 103	100
76	0.65384	55 174 177	42	0.31517	105 196	77	0.48498	100 103	26
78	0.47813	12 174 198	58	0.34898	196 204	67	0.23014	100 200	53
83	0.82218	12 51 55 127	36	1.00000	83	1	0.14160	100 103	72
84	0.52870	14 51 127	53	0.42498	83 202	53	1.00000	84	1
85	0.57413	12 14 51 127	48	1.00000	85	1	0.14950	40 103	71
86	0.59759	12 51 154 163	46	0.59068	15 105 196	38	0.46582	100 103	29
93	0.35722	14 53 174	80	0.59413	105 203	36	0.16837	46 100	66
95	0.48326	12 14 51	57	0.32196	105 196	75	0.45918	100 103	30
98	0.35958	12 51 53	79	0.15592	196	102	0.43898	100 200	31
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
102	1.00000	102	1	0.59282	60 203	37	1.00000	102	1
103	0.23983	51 53 55	92	1.00000	103	1	1.00000	103	1
105	0.15206	12 53 205	101	1.00000	105	1	1.00000	105	1
110	0.51753	12 51	55	0.21797	83 196	97	0.18258	100 103	61
113	0.23335	12 14 51	94	0.24035	105 196	94	0.05560	125 164	102
115	0.35240	174	82	0.27513	196 203	88	0.22152	100 200	55
121	0.23817	12 51	93	0.26370	83 202	90	0.10752	100 103	87
123	0.21151	12 53	97	0.34524	105 203	68	0.12162	100 200	80
125	0.15186	53 59	102	0.49352	105 203	46	1.00000	125	1
126	0.35556	12 53 174 205	81	0.38279	105 196	62	0.17156	100 200	65
127	1.00000	127	1	0.15938	105 196	100	0.11483	100 103	83
128	1.00000	128	1	0.25199	83 202	92	0.13745	40 103	74
130	1.00000	130	1	0.18229	196 202	99	0.10772	100 103	86
131	0.17911	12 205	100	0.32898	196 203	72	1.00000	131	1
132	0.45214	12 51 177 205	63	0.44244	83 196	51	0.23492	100 200	52
133	0.42338	46 174 205	67	0.40250	196 203	60	0.25349	100	48
137	0.29418	12 174	87	0.27940	196 203	87	0.53930	100 200	25

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139	0.39906	14 53 59	73	0.36266	196 203	64	0.59504	100 200	21
140	1.00000	140	1	0.51538	83 202	43	0.13933	100 103	73
142	0.27431	12 51	88	0.71522	15 34 105 196	29	0.10126	100 200	90
152	1.00000	152	1	1.00000	152	1	0.47039	100 200	28
154	1.00000	154	1	0.33533	83 196	71	0.06088	100 103	99
156	0.36219	12 174	78	0.34349	105 196	70	0.11174	100 200	85
159	0.31987	12 174 205	86	0.35959	196 204	66	0.09153	100 200	93
160	1.00000	160	1	0.30180	196 204	81	0.21533	100 200	57
161	1.00000	161	1	0.68683	15 105 196	30	0.13068	100 103	78
162	0.54232	12 14 51 140 177	51	0.50372	105 196	45	0.25963	100 103	46
163	1.00000	163	1	0.41717	83 196	57	0.07382	100 103	97
164	0.57498	46 205	47	1.00000	164	1	1.00000	164	1
168	0.27328	50 174	89	0.78285	34 105 152	28	0.30055	100 200	41
173	0.43382	12 177 205	65	0.30042	83 196	82	0.10138	100 200	89
174	1.00000	174	1	0.38546	196 203	61	0.32698	100 205	38
176	0.86840	51 55 140 174 177	34	0.34456	105 196	69	0.09672	100 200	91
177	1.00000	177	1	0.23325	105 196	95	0.20325	100 200	59
182	0.64430	12 51 154 163 177	43	0.29979	83 196	83	0.17212	100 103	64
185	0.39705	12 53 59	74	1.00000	185	1	0.16245	40 103	68
190	0.46094	12 51 177 205	60	0.36108	83 196	65	0.11544	100 200	82
191	0.45374	12 174	62	0.32728	105 203	73	0.20589	100 200	58
192	0.60508	14 55 161 174	45	0.45913	196 204	49	0.21577	100 200	56
193	1.00000	193	1	0.29092	105 196	84	0.07620	100 200	96
194	1.00000	194	1	0.65016	83 196 204	32	0.25779	100 103	47
195	0.39469	46 174 205	75	1.00000	195	1	0.24303	100 200	50
196	1.00000	196	1	1.00000	196	1	0.91253	100 164	19
197	0.41165	12 174	71	0.31801	196 204	76	0.13588	100 205	75
198	1.00000	198	1	0.62416	83 196 202	33	0.15961	100 205	70
200	0.57188	14 50 53 174	49	0.41843	103 196	56	1.00000	200	1
202	1.00000	202	1	1.00000	202	1	0.23816	100 205	51
203	1.00000	203	1	1.00000	203	1	1.00000	203	1
204	1.00000	204	1	1.00000	204	1	0.16524	100 103	67
205	1.00000	205	1	0.60200	60 203	35	1.00000	205	1
206	0.37016	174 205	77	0.42407	196 203	54	0.22677	100 200	54
207	0.38073	46 202 205	76	1.00000	207	1	1.00000	207	1
210	0.73788	174 205	39	0.57338	196 204	39	0.55536	100 205	23

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under VRS (Equation I.15 with the restriction $\sum_j \lambda_j = 1$ proposed by Banker et al. (1984) in the called BCC model). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-B-3 SBM-efficiency under VRS for 2007

Stage 1 Portfolio Management				Stage 2 Marketing and Service			Overall Efficiency		
MFMC	SBM	Ref.	Rank	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.53472	7 12 50 198	47	0.31012	40 50	74	0.07813	100 103	80
4	0.50662	12 51 55 127	51	0.24517	12 50	92	0.11067	100 103	65
6	0.35797	14 50 177	73	0.35801	40 46	62	0.17805	100 125	45
7	1.00000	7	1	0.31046	40 50	73	0.08777	100 103	74
9	0.29805	12 50 174 177	80	0.28527	46 103	79	0.28458	100 103	28
12	1.00000	12	1	1.00000	12	1	0.66053	14 103	12
14	1.00000	14	1	1.00000	14	1	1.00000	14	1
15	0.64789	12 51 127	42	0.42336	12 50	56	0.06074	100 103	88
20	0.26622	12 51	84	0.24698	12 50	91	0.21900	100 103	38
21	0.32822	12 50 51	76	0.50820	12 40	39	0.13702	100 103	59
24	0.39535	12 152	68	0.44813	31 40	52	0.06249	100 103	87
29	0.39905	12 50 198	67	0.29356	40 50	78	0.18259	100 103	44
31	0.40268	12 50 152	65	1.00000	31	1	0.09867	100 103	67
34	0.20994	12 50	91	0.62742	40 50 161	24	0.07564	100 103	82
35	0.74567	12 152 174 177 198	33	0.27663	40 46	84	0.08945	100 103	71
36	1.00000	36	1	0.55598	57 100	30	0.50495	100 103	15
37	0.41139	12 50 152	64	0.63188	46 57	22	0.23170	100 103	34
38	0.15447	12 50	94	0.27942	46 103	81	0.32278	100 103	26
40	0.27170	12 50	83	1.00000	40	1	0.41863	100 103	19
43	0.71184	7 12 50 128 198	37	0.26926	50 103	87	0.14735	100 103	54
45	0.68430	12 152 174 177	38	0.26316	46 103	88	0.52468	100 103	14
46	1.00000	46	1	1.00000	46	1	1.00000	46	1
47	0.46196	12 51	56	0.45831	12 40	48	0.10355	100 103	66
49	0.39464	12 50 198	69	0.55462	46 50 103	31	0.32807	100 103	25
50	1.00000	50	1	1.00000	50	1	0.20350	100	41
51	1.00000	51	1	0.24173	46 57	93	0.16409	100 103	48
53	1.00000	53	1	0.63850	46 100 105	21	0.44545	100 102	18
55	1.00000	55	1	0.52758	12 50	35	0.17717	100 103	46
57	0.33581	12 152 198	74	1.00000	57	1	0.22552	100 103	36
58	0.44104	12 51 127	62	0.45585	12 50	49	0.20454	100 103	40
59	1.00000	59	1	1.00000	59	1	1.00000	59	1
61	0.25363	14 50 51	86	0.27729	40 50	83	0.06815	100 103	85
62	0.62613	12 50 127 177	43	0.45100	40 46 50	51	0.18927	100 103	43
63	0.47019	12 51	55	0.61199	12 50	26	0.03937	100 103	95
69	0.72944	53 152 210	34	0.70115	46 50 57 102	19	0.60801	102 207	13
71	1.00000	71	1	0.53182	12 40 50	33	0.16319	100 103	49
75	0.44829	12 51 127	59	0.28476	12 50	80	0.05955	100 103	90
76	0.68177	14 174 177	39	0.27753	46 103	82	0.22398	100 103	37
78	1.00000	78	1	0.56097	46 50 103	29	0.16483	100 103	47
83	1.00000	83	1	0.34974	50 103	65	0.09166	100 103	68
84	0.52118	14 51 55 127	48	0.34101	12 50	67	1.00000	84	1
85	0.59249	12 51 127	45	0.41610	12 50	57	0.07662	100 103	81
86	0.67705	7 12 51 127 163	40	0.44507	12 40	54	0.14951	100 103	53
93	0.30902	12 50	78	0.52315	40 46 50	38	0.16014	100 103	50
95	0.72377	12 51 127 140	35	0.52639	12 40	36	0.27605	100 103	29
98	0.23747	12 50	88	0.34048	40 46	68	0.33234	100 103	24
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
102	1.00000	102	1	1.00000	102	1	1.00000	102	1
103	0.21627	12 50 127	89	1.00000	103	1	1.00000	103	1
105	0.11383	12 50	95	1.00000	105	1	0.37980	100 125	22
110	0.45724	12 14 51	57	0.38024	31 40	60	0.13187	100 103	63
113	0.21013	12 51	90	0.27075	40 50	86	0.04445	100 103	94
115	0.45071	152 174	58	0.40322	46 105	58	0.34616	100 103	23
121	0.27893	12 51	82	0.33012	12 40	71	0.08962	100 103	70
125	0.16271	46 50	93	0.87468	100 105	17	1.00000	125	1
126	0.38053	12 50 152 198	70	0.23877	46 105	94	0.13530	100 103	61
127	1.00000	127	1	0.30885	40 50	75	0.08517	100 103	75
128	1.00000	128	1	0.30422	12 50	76	0.06399	100 103	86
130	0.72165	12 51 140 163	36	0.33259	12 40	70	0.07227	100 103	83
131	0.23936	12 152	87	0.34452	46 57	66	0.38450	100 103	21
132	0.42802	12 50 51 177	63	0.21899	46 103	95	0.14085	100 103	58
133	0.59230	46 50 152	46	0.50292	46 57	40	0.24999	100 103	33
137	0.44129	12 152 198	61	0.32158	46 103	72	0.40933	100 103	20
139	0.33125	50 128	75	0.43419	46 103	55	0.45230	100 103	17
140	1.00000	140	1	0.44676	50 103	53	0.08431	100 103	76
142	0.30717	12 51	79	0.50033	12 40	41	0.05969	100 103	89

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152	1.00000	152	1	0.52855	46 100 102	34	0.30614	100 103	27
154	0.94650	12 51 140 163 198	30	0.25960	40 50	90	0.04640	100 103	93
156	0.47794	12 50 174	54	0.48083	46 50 57	45	0.11274	100 103	64
159	0.31312	12 50 152	77	0.49184	40 46 50	42	0.14318	100 103	56
160	1.00000	160	1	0.57922	46 50 57	27	0.13508	100 103	62
161	1.00000	161	1	1.00000	161	1	0.08893	100 103	73
162	0.48062	14 51 127	53	0.56542	40 50 161	28	0.13616	100 103	60
163	1.00000	163	1	0.35558	50 103	63	0.05035	100 103	91
168	0.37464	12 50 174	71	0.33824	46 105	69	0.26687	100 103	30
173	0.36004	12 50 51	72	0.63161	50 57 103	23	0.07129	100 103	84
174	1.00000	174	1	0.46193	46 50 57	47	0.22709	100 103	35
176	0.75805	12 50 51 127 140 177	32	0.54508	50 103	32	0.07997	100 103	78
177	1.00000	177	1	0.30167	40 46	77	0.14722	100 103	55
182	0.44513	12 50 51	60	0.48956	12 31 40	43	0.09090	100 103	69
185	0.27914	50 51 128	81	0.35504	46 103	64	0.08942	100 103	72
190	0.60713	12 51 177 198	44	0.46622	50 103	46	0.07861	100 103	79
191	0.66772	12 152 174	41	0.26225	46 57	89	0.20997	100 103	39
192	0.40238	14 50 174 177	66	1.00000	192	1	0.20013	100 103	42
193	1.00000	193	1	0.27329	40 50	85	0.04831	100 103	92
194	0.50928	12 50 198	50	0.36047	46 103	61	0.15354	100 103	51
196	1.00000	196	1	1.00000	196	1	1.00000	196	1
197	0.51857	12 152 198	49	0.39410	46 57	59	0.08052	100 102	77
198	1.00000	198	1	0.52337	46 57	37	0.15239	100 102	52
200	0.18653	50 127	92	0.69303	12 40 46	20	0.50396	100 103	16
203	1.00000	203	1	1.00000	203	1	1.00000	203	1
204	0.80813	12 50 177 198	31	0.45206	40 46 50	50	0.14163	100 103	57
206	0.49552	152	52	0.61382	46 57 100	25	0.25225	100 103	32
207	0.26542	12 152	85	0.48471	46 57	44	1.00000	207	1
210	1.00000	210	1	0.72708	46 50 57	18	0.26662	100 102	31

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under VRS (Equation I.15 with the restriction $\sum_j \lambda_j = 1$ proposed by Banker et al. (1984) in the called BCC model). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-B-4 SBM-efficiency under VRS for 2008

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency		
	SBM	Ref.	Rank	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.35450	15 174 198	53	0.15839	83 100	81	0.15121	45 46	63
4	0.50020	14 127	42	0.63029	12 100	18	0.12664	46 103	71
6	0.22888	14 174	75	0.30001	83 100	51	0.15328	46 125	62
7	0.69705	154 174 198	32	0.20038	100 103	73	0.17458	46 103	57
9	0.19491	14 174 217	82	0.21948	37 100	70	0.38559	46 103	24
12	1.00000	12	1	1.00000	12	1	0.62204	14 103	13
14	1.00000	14	1	1.00000	14	1	1.00000	14	1
15	1.00000	15	1	0.43931	12 100	29	0.09377	46 103	86
20	0.29217	127 174	62	0.40205	12 100	35	0.25296	46 103	43
21	0.33957	14 127 174	56	0.23364	100 103	66	0.11110	46 103	78
24	0.24908	14 174	70	0.41024	83 100	33	0.09542	46 103	85
29	0.22837	14 15 174	76	0.21373	83 100	72	0.10188	46 103	83
31	0.29791	14 174	61	0.15646	83 100	83	0.14451	40 100	64
34	0.16639	14 174	87	0.19510	83 100	74	0.18575	40 100	52
35	0.77408	154 174 177 193	29	0.18147	83 100	77	0.17492	45 46	56
36	1.00000	36	1	0.55918	83 100	20	0.61028	46 103	14
37	0.25407	14 174	68	1.00000	37	1	0.23079	46 103	45
38	0.14387	14 174	90	0.24811	83 100	62	0.15543	46 125	61
40	0.28663	174 217	63	1.00000	40	1	1.00000	40	1
43	0.27245	14 15 174	65	0.25778	83 100	58	0.10466	46 103	81
45	0.32624	14 174	58	0.14237	83 100	88	1.00000	45	1
46	1.00000	46	1	1.00000	46	1	1.00000	46	1
47	0.39462	14 174	51	0.46808	12 100 103	26	0.14006	46 103	69
49	0.17073	15 174	86	0.15618	83 100	84	0.32039	46 103	29
50	1.00000	50	1	0.24121	40 100	64	0.48814	46 103	21
51	1.00000	51	1	0.33802	83 100	46	0.30597	46 103	32
53	1.00000	53	1	0.24435	40 100	63	1.00000	53	1
55	1.00000	55	1	0.54017	12 83 100	22	0.28299	46 103	37
57	0.24737	15 174	73	0.65513	83 100 103 192	17	0.11766	40 100	77
58	0.46243	12 127	44	1.00000	58	1	0.28280	46 103	38
61	0.20956	14 174	79	0.18033	83 100	78	0.07130	46 103	91
62	0.40313	14 174	48	0.54068	83 100	21	0.29017	46 103	36
63	0.58162	14 127 154	37	0.43692	12 100	31	0.02567	46 100	93
69	0.81256	53 174 196 217	28	0.28416	40 100	53	0.26802	40 100	40
71	0.71378	14 76 177	31	0.29769	100 103	52	0.26404	46 103	41
76	1.00000	76	1	0.35235	100 103	44	0.31451	46 103	30
78	0.34679	14 15 174	54	0.23910	83 100	65	0.16522	46 103	59
83	1.00000	83	1	1.00000	83	1	0.13315	46 103	70
84	0.45443	14 127 174	45	0.45349	12 100	28	0.27677	46 103	39
85	0.62510	14 127	35	0.36975	12 100	40	0.10777	46 103	79
86	0.45412	14 154 174	46	0.39068	83 100	36	0.12626	46 103	72
93	0.25066	174 217	69	0.25479	100 103	61	0.19930	46 103	50
95	0.64120	14 154	34	0.35896	12 100	43	1.00000	95	1
98	0.11395	15 174	92	0.30296	83 100	48	0.55087	45 46 103	16
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
103	0.08225	14 127 174	93	1.00000	103	1	1.00000	103	1
105	0.16609	174 217	88	0.18955	83 100	76	1.00000	105	1
110	0.44384	14 154 174	47	0.36392	100 103	42	0.22731	46 103	46
113	0.20802	14 174	80	0.11591	83 100	91	0.10143	46 103	84
115	1.00000	115	1	0.22695	100 192	67	0.49892	46 103	20
121	0.22471	14 174	77	0.37307	12 100	39	0.10356	46 103	82
125	0.19855	196 217	81	0.34792	40 100	45	1.00000	125	1
126	0.26668	14 174	67	0.09743	83 100	93	0.18466	46 103	53
127	1.00000	127	1	0.36909	83 100	41	0.14315	46 103	65
128	1.00000	128	1	0.49501	12 100	25	0.08206	46 103	89
130	0.65405	14 15 154	33	0.27748	12 100	54	0.10546	46 103	80
131	0.17218	174 217	85	0.27090	83 100	55	0.52334	46 103	19
132	0.18146	14 174	84	0.22574	83 100	68	0.19756	46 103	51
133	0.53391	174 217	39	0.26130	83 100	57	0.44793	46 103	23
137	0.24868	174 217	71	0.30217	83 100	50	0.80058	45 46 103	12
139	0.49912	14 174 196	43	0.26666	100 103	56	0.52367	46 103	18
140	0.75233	14 15 127 128 154	30	0.43718	83 100	30	0.14061	46 103	68
142	0.40239	14 154 174	49	0.22025	12 100	69	0.22440	46 103	47
152	1.00000	152	1	0.25736	40 100	59	0.26375	46 100	42
154	1.00000	154	1	0.11273	12 100	92	0.08012	46 103	90
156	0.27772	14 174	64	0.17114	83 100	80	0.20856	46 103	49

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159	0.18165	14 174	83	0.15800	83 100	82	0.15863	46 103	60
160	1.00000	160	1	0.15392	83 100	86	0.18106	46 103	55
161	1.00000	161	1	0.43251	83 100	32	0.14086	46 103	67
162	0.37056	14 154 174	52	0.30249	83 100	49	0.21679	46 103	48
163	1.00000	163	1	0.38345	83 100	38	0.12269	45 46	75
168	0.21910	174	78	0.15578	83 100	85	0.33406	46 103	26
173	0.16180	14 174	89	0.31568	83 100	47	0.05454	40 100	92
174	1.00000	174	1	0.60208	37 100	19	0.52867	45 46 53	17
176	0.59752	14 154 174	36	0.14808	83 100	87	0.16796	46 103	58
177	1.00000	177	1	0.38525	83 100	37	0.32291	46 103	28
182	0.34271	14 154 174	55	1.00000	182	1	0.09287	46 103	87
185	0.50956	14 154 160	40	0.49563	100 103	24	0.30836	40 105	31
190	0.56458	154 174 198	38	0.21676	83 100	71	0.14313	46 103	66
191	0.40090	14 174	50	0.51205	37 100	23	0.32910	45 46	27
192	0.31048	14 174	60	1.00000	192	1	0.29656	46 103	33
193	1.00000	193	1	0.25719	83 100	60	0.12401	45 46	73
194	0.27151	15 174	66	0.12192	83 100	90	0.08920	46 100	88
196	1.00000	196	1	1.00000	196	1	0.57496	46 100	15
197	0.32877	15 174	57	0.18999	83 100	75	0.23400	45 46	44
198	1.00000	198	1	0.12534	83 100	89	0.18140	46 100	54
200	0.13390	14 174 196	91	0.40705	100 103	34	0.35104	46 103	25
203	0.31545	174 217	59	0.70986	37 100 206	16	0.29407	46 100	35
204	0.50569	14 174	41	0.45865	83 100	27	0.11891	46 103	76
206	0.24767	174 217	72	1.00000	206	1	0.29464	46 103	34
207	0.24003	174 217	74	0.17539	83 100	79	0.12393	46 100	74
210	1.00000	210	1	0.80966	37 100 206	15	0.48170	45 46 53	22
217	1.00000	217	1	1.00000	217	1	1.00000	217	1

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under VRS (Equation I.15 with the restriction $\sum_j \lambda_j = 1$ proposed by Banker et al. (1984) in the called BCC model). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-B-5 SBM-efficiency under VRS for 2009

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency		
	SBM	Ref.	Rank	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.49366	7 50 76	50	0.28606	163 196	69	0.07278	100 103	88
4	1.00000	4	1	0.77482	14 63 128 196	22	0.09058	103 125	76
6	0.38080	14 50 76	69	0.21638	163 196	78	0.11434	100 125	65
7	1.00000	7	1	0.17241	31 196	89	0.11250	100 103	66
9	0.36710	14 50 76 174 217	74	0.50058	63 196	35	0.35390	100 103	25
12	0.65098	14 46	30	1.00000	12	1	1.00000	12	1
14	1.00000	14	1	1.00000	14	1	1.00000	14	1
15	0.38669	14 46	66	1.00000	15	1	0.06580	100 103	92
20	0.22348	14 46 51	89	0.38735	15 196	50	0.35570	100 103	24
21	0.31174	14 51	83	1.00000	21	1	0.14973	100 103	57
24	0.31166	50 76	84	0.16051	163 196	94	0.07853	100 103	85
29	0.38248	45 76 131 217	68	0.15359	163 196	95	0.08399	45 100	81
31	0.63533	14 50 51 196	31	1.00000	31	1	1.00000	31	1
34	0.24760	50 76	86	0.16733	31 196	90	0.10130	46 100	72
35	0.62663	50 76 174	32	0.18766	31 196	87	0.10941	46 100	70
36	0.56101	100 196 217 221	35	0.55882	163 196	31	0.34123	100 125	26
37	0.38608	14 50 76	67	0.41286	163 196	45	0.17374	100 103	50
38	0.19126	14 51	93	0.35894	163 196	54	0.10037	100 125	74
40	0.31824	50 174 217	81	0.44726	31 196	40	0.40951	100 125	21
43	1.00000	43	1	0.63853	15 31	27	0.12531	100 103	61
45	1.00000	45	1	0.20398	163 196	82	1.00000	45	1
46	1.00000	46	1	1.00000	46	1	1.00000	46	1
47	0.38674	14 46 50 76	65	0.20994	163 196	80	0.12354	100 103	62
49	0.32537	50 76	80	0.43134	31 196	43	0.14991	100 103	56
50	1.00000	50	1	0.57667	46 163 196	30	0.28761	46 100	29
51	1.00000	51	1	0.29490	163 196	68	0.16103	100 103	54
53	1.00000	53	1	1.00000	53	1	0.53502	46 100 217	19
55	1.00000	55	1	0.30076	128 196	65	0.21466	100 103	40
57	0.50571	76 217	47	0.28363	31 196	70	0.24496	100 103	34
58	0.40513	14 46 51	60	0.16446	15 196	93	0.21319	100 103	41
61	0.23876	14 50	88	0.20141	163 196	83	1.00000	61	1
62	1.00000	62	1	0.54634	163 196	32	0.21274	100 103	42
63	0.48459	14 46 51 127	54	1.00000	63	1	0.33965	46 61	27
69	0.52300	46 196 217	42	1.00000	69	1	0.72186	46 105 217	16
71	0.74767	14 50 76 140	27	0.30902	31 196	62	0.21485	100 103	39
76	1.00000	76	1	0.63984	15 31	26	0.21934	100 103	38
78	0.46832	50 76	56	0.33212	31 196	57	0.15635	100 103	55
83	0.39232	14 50 51 76	64	0.38804	128 196	49	0.06712	100 103	91
84	0.50705	14 46 51 127	46	0.71348	15 31 196	25	0.26746	100 103	31
85	0.54235	14 46 51	39	0.58485	128 196	28	0.10122	100 103	73
86	0.53105	7 14 51 127	40	0.49635	15 31	37	0.10450	100 103	71
93	0.34772	14 50 217	78	0.30179	163 196	64	0.22078	46 100	36
95	0.72751	7 14 46 76 140	28	0.19845	15 196	84	0.11090	100 103	68
98	0.22035	14 46 51	90	0.34044	163 196	55	0.19212	100 125	47
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
103	0.21496	14 50 76	91	1.00000	103	1	1.00000	103	1
105	0.15130	14 50	95	0.44205	31 196	41	1.00000	105	1
110	0.40325	14 50 51	61	0.75853	163 185 196	24	0.14598	100 103	58
113	0.20854	14 46 50 51	92	0.23831	163 196	74	0.06517	100 125	93
115	0.42630	50 174 217	58	0.45508	163 196	38	0.19983	100 125	45
121	0.24182	14 46 50	87	0.33228	15 196	56	0.08170	100 103	84
125	0.15351	46 50 196	94	0.95921	46 53 69 196	21	1.00000	125	1
126	0.37473	76 174	72	0.36162	31 196	53	0.16464	100 103	53
127	1.00000	127	1	0.40841	163 196	46	0.08722	100 103	79
128	1.00000	128	1	1.00000	128	1	0.02859	100 103	95
130	1.00000	130	1	0.16453	31 196	92	0.07067	100 103	89
131	1.00000	131	1	0.30196	31 196	63	1.00000	131	1
132	0.27014	50 76	85	0.30981	31 196	61	0.07679	100 103	86
133	0.49312	76 217	52	0.38304	163 196	52	0.27377	46 100	30
137	0.37597	76 174	71	0.23812	163 196	75	0.38803	100 103	22
139	0.54358	14 46 76 217	38	0.29894	163 196	66	0.55187	100 103	18
140	1.00000	140	1	0.19659	31 196	85	0.06871	100 103	90
142	0.31514	14 46 51	82	0.18666	31 196	88	0.08555	100 103	80
152	1.00000	152	1	0.51258	69 196	34	0.47555	100 131	20
154	0.65955	14 46 76 128 140	29	0.22411	31 196	76	0.06359	100	94
156	0.49467	76 174	49	0.20629	31 196	81	0.13465	100 103	59

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159	0.51473	50 76 128 210	44	0.45279	31 196	39	0.12255	46 100	63
160	0.52963	50 76 174	41	0.22078	163 196	77	0.17972	46 100	48
161	0.54829	14 50 76	37	0.31272	163 196	59	0.08365	100 103	82
162	0.49604	14 50 76	48	0.18895	163 196	86	0.12971	100 103	60
163	0.48417	7 14 46 51 128	55	1.00000	163	1	0.08298	100 103	83
168	0.38066	76 174	70	0.25571	163 196	73	0.25961	100 103	32
173	0.40898	7 14 50 51	59	1.00000	173	1	0.09736	46 61	75
174	1.00000	174	1	0.38985	163 196	48	0.22299	100 217	35
176	0.58991	14 50 51 76	34	0.49979	31 196	36	0.11224	100 103	67
177	0.35088	76 174	76	0.76668	53 163 196	23	0.22030	100 125	37
182	0.39349	14 50 76	63	0.16491	31 196	91	0.08757	100 103	78
185	0.36622	14 46 51	75	1.00000	185	1	0.36445	46 61	23
190	0.51067	7 14 50 76	45	0.31498	31 196	58	0.07434	100 103	87
191	0.55785	76 174 217	36	1.00000	191	1	0.32272	100 103	28
192	0.32577	14 174	79	0.53253	163 196	33	0.16683	100 125	52
193	1.00000	193	1	0.39059	31 196	47	0.08861	45 100	77
194	0.51672	50 76 210	43	0.27329	31 196	71	0.11031	46 100	69
195	0.40166	46 76 210	62	0.43569	31 196	42	1.00000	195	1
196	1.00000	196	1	1.00000	196	1	0.60130	46 100	17
197	0.43436	76 174 217	57	0.38592	31 196	51	0.20472	100 217	44
198	0.59498	50 51 128 210	33	0.27306	163 196	72	0.11634	46 100	64
200	0.49351	14 51 196	51	0.58243	163 196	29	0.17872	100 125	49
203	0.37215	76 217	73	0.41956	163 196	44	0.25638	46 100	33
204	0.48731	50 76	53	0.31205	163 196	60	0.16710	46 100	51
206	0.34988	76 217	77	0.29535	163 196	67	0.19768	100 125	46
207	1.00000	207	1	1.00000	207	1	1.00000	207	1
210	1.00000	210	1	0.21562	163 196	79	0.20633	100 217	43
217	1.00000	217	1	1.00000	217	1	1.00000	217	1
221	1.00000	221	1	1.00000	221	1	1.00000	221	1

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under VRS (Equation I.15 with the restriction $\sum_j \lambda_j = 1$ proposed by Banker et al. (1984) in the called BCC model). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Appendix C

Table I-C-1 Comparative CRS-SBM scores for 2005 with different measures of labour

Portfolio Management stage

MFMC	Using number of employees			Using personnel expenses		
	SBM	Ref.	Rank	SBM	Ref.	Rank
1	0.33795	12 161	66	0.32145	12 161	71
2	0.40030	12 177	59	0.39290	12 163 177	63
4	0.42223	14 161	55	0.47356	14 161	56
6	1.00000	6	1	1.00000	6	1
7	1.00000	7	1	1.00000	7	1
9	0.24942	12 177	78	0.22913	177	81
12	1.00000	12	1	1.00000	12	1
14	1.00000	14	1	1.00000	14	1
15	0.54453	12 161	44	0.54301	12 161	43
20	0.49983	12 161	48	0.53238	12 161	45
21	0.21632	161	80	0.18903	161	87
24	0.41055	12 177	56	0.41069	12 177	62
29	0.45102	12 43 177	53	0.48605	12 163 177	50
31	0.64661	12 177	36	0.63676	12 177	37
34	0.20845	12 177	81	0.14575	12 163	90
35	0.81735	12 40 160 176 177	30	0.87243	12 40 160 161 177	29
36	0.17644	12 174	85	0.22517	12 55 174	82
37	0.32688	12 40 174 177	68	0.33562	12 40 174 177	70
38	0.12944	12 160 176	89	0.12828	7 12 160	92
40	1.00000	40	1	1.00000	40	1
42	0.11366	12 55 174	93	0.21497	55 95 100 174	83
43	1.00000	43	1	1.00000	43	1
45	0.55011	12 174 177	43	0.64783	12 55 174 177	36
46	0.06900	12 152	96	0.07305	12 174	97
47	0.48304	12 176	49	0.51298	161 177	47
49	0.38867	12 40 174 177	60	0.35029	112 40 74 177	67
50	0.14370	12 177	88	0.19264	12 71 163 177	85
51	0.58078	12 40 160 161	39	0.57287	12 40 160 161	41
53	1.00000	53	1	1.00000	53	1
55	1.00000	55	1	1.00000	55	1
57	0.28221	12 174	74	0.26984	12 174	76
58	0.70651	12 161	35	0.84183	12 127 128 161	31
59	0.02740	12 161	99	0.02453	12	99
60	0.00646	12 40	102	0.00630	12 40	102
61	0.31073	55 161	70	0.30275	161	73
62	0.64389	12 160 161 176	37	0.67030	7 160 161	35
63	0.76928	12 43 154 160	31	0.73092	12 43 154 160	32
69	0.11333	12 174 177	94	0.09783	12 163 174 177	95
71	1.00000	71	1	1.00000	71	1
73	0.70830	12 174 177	33	0.59738	12 163 174 177	39
75	0.47698	12 161	50	0.45885	12 161	58
76	1.00000	76	1	0.87324	95 161 174	28
78	0.35690	12 174 177	63	0.38484	12 163 174	65
83	0.61178	12 55 161	38	0.71653	12 127 161	33
84	0.45669	14 161	52	0.44921	14 161	59
85	0.46772	12 161	51	0.48488	12 161	51
86	0.50307	161	46	0.52822	161	46
93	0.19591	12 55 161	83	0.18416	12 161	88
95	0.73830	55 161	32	1.00000	95	1
98	0.38578	12 43 160	62	0.36293	12 43 160	66
100	1.00000	100	1	1.00000	100	1
102	0.03736	12 152	98	0.07409	12 100 174	96
103	1.00000	103	1	1.00000	103	1
105	0.41006	12 163 177	57	0.41198	12 163 177	61
110	0.56571	12 161	40	0.55575	161	42
113	0.17183	12 14 161	86	0.16095	12 14 161	89
115	0.15611	12 100 174	87	0.18958	12 14 100 174	86
121	0.27680	161	75	0.24320	161	80
123	0.12551	12 160 177	91	0.12921	12 160 177	91

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125	0.01309	14 161	101	0.01252	14 161	101
126	0.35117	12 177	65	0.39237	12 174 177	64
127	1.00000	127	1	1.00000	127	1
128	1.00000	128	1	1.00000	128	1
130	1.00000	130	1	1.00000	130	1
131	0.12109	12 152	92	0.12573	12 174	93
132	0.26751	12 161 177	76	0.27706	7 12 161 163	75
133	0.28577	12 40 174	73	0.26080	12 40 174	78
137	0.30621	12 174	72	0.42526	12 174	60
138	0.18856	12 152	84	0.46963	12 100 152 174	57
139	0.20468	12 177	82	0.21329	154 163 177	84
140	1.00000	140	1	1.00000	140	1
142	0.35601	161	64	0.31699	161	72
152	1.00000	152	1	1.00000	152	1
154	1.00000	154	1	1.00000	154	1
156	0.31727	12 55 174	69	0.48112	95 174	52
159	0.32720	12 55 174 177	67	0.34084	12 55 162 174	68
160	1.00000	160	1	1.00000	160	1
161	1.00000	161	1	1.00000	161	1
162	0.70769	12 55 154 176 177	34	1.00000	162	1
163	1.00000	163	1	1.00000	163	1
164	0.06441	12 174	97	0.05693	12 174	98
168	0.23644	12 55 174	79	0.26409	12 55 174	77
173	0.30974	12 177	71	0.28514	12 163 177	74
174	1.00000	174	1	1.00000	174	1
176	1.00000	176	1	0.85200	7 12 154 160 161	30
177	1.00000	177	1	1.00000	177	1
182	0.55858	12 176 177	41	0.58053	7 12 163 177	40
185	0.38616	12 43 154	61	0.34024	12 43 154	69
189	0.12673	12 152	90	0.11684	12 152	94
190	0.55505	12 176 177	42	0.53645	12 160 163 177	44
191	0.40856	12 174 177	58	0.50925	12 174	48
192	1.00000	192	1	0.62273	53 161 177	38
193	1.00000	193	1	0.70928	12 71 174 177	34
194	0.82258	12 43 174 198	29	0.93880	12 163 174 198	27
195	0.53370	12 174 198	45	0.47889	12 174 198	54
196	0.01433	12 40 152	100	0.01446	12 40 152	100
197	0.50042	12 174 177	47	0.47555	12 163 174 177	55
198	1.00000	198	1	1.00000	198	1
200	0.42806	55 174 177	54	0.50677	95 161 174	49
202	0.25857	12 40 161 174 177	77	0.24407	12 40 174 177	79
205	1.00000	205	1	1.00000	205	1
206	0.10614	12 152 174	95	0.48094	95 100 174	53

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-C-2 Comparative CRS-SBM scores for 2006 with different measures of labour

Portfolio Management stage

MFMC	Using number of employees			Using personnel expenses		
	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.44895	12 177 205	47	0.46489	12 53 163 174	44
4	0.36609	14 51	57	0.46362	14 127	45
6	0.24471	51 55 161	72	0.30032	53 161 174	65
7	0.80473	12 51 53 154 163	26	1.00000	7	1
9	0.21437	140 174	75	0.16838	140 174	86
12	1.00000	12	1	1.00000	12	1
14	1.00000	14	1	1.00000	14	1
15	0.56122	12 51 127	34	0.58281	12 51 127	33
20	0.54491	12 51 127	36	0.56044	12 51 127	36
21	0.24047	127	73	0.18393	127	83
24	0.42342	12 177 205	51	0.34341	12 53	59
29	0.41718	12 128 205	52	0.51418	163 174 198 205	40
31	0.84051	12 51 53 154 163	25	0.76332	12 51 53 154 163	26
34	0.16519	12 177	84	0.14001	12 53	89
35	0.51953	12 174	39	0.48647	12 174	42
36	0.16393	12 14 174	85	0.19107	12 174	82
37	0.37920	12 174 205	56	0.38004	12 53 174	56
38	0.10118	12 51 163	95	0.09838	12 51 163	95
40	0.10521	174	94	0.09864	140 174	94
43	0.14496	14 174	88	0.38078	55 100 174	55
45	0.69014	128 154 163 198 205	28	0.74099	163 198 205	27
46	0.50592	12 174	41	0.51060	14 174	41
47	0.12119	12 205	91	0.15967	12 174 205	87
49	0.41186	12 51	53	0.41732	14 53 161	50
50	0.32253	12 174	63	0.29871	12 174	66
51	0.12725	12 174	90	0.17204	12 14 71 174	85
53	1.00000	51	1	1.00000	51	1
55	1.00000	53	1	1.00000	53	1
57	1.00000	55	1	1.00000	55	1
58	0.34304	174	60	0.29760	174	67
59	0.53423	12 51 128	38	0.55523	12 51 128	37
61	1.00000	59	1	1.00000	59	1
62	0.00147	12 205	102	0.00142	12 53	102
63	0.20216	14 161	78	0.17845	12 14 127	84
69	1.00000	62	1	1.00000	62	1
71	0.62818	12 53 128 163 205	32	0.58582	12 53 128 163 205	32
75	0.16152	12 174 205	86	0.19461	12 174	81
76	1.00000	71	1	1.00000	71	1
78	0.34772	12 51 128	59	0.33210	12 51 127	61
83	0.64041	55 174	30	0.61961	161 174	31
84	0.44933	12 174	46	0.46257	12 174	47
85	0.73548	12 51 55 127 140	27	0.65946	12 51 127	30
86	0.44491	14 51 127	49	0.46346	13 51 127	46
93	0.45866	12 51 128	43	0.47604	12 51 128	43
95	0.50058	12 51	42	0.56373	12 51	35
98	0.20408	14 51 55	77	0.20056	140 161 53	78
100	0.45104	12 14 51	45	1.00000	95	1
102	0.35902	12 51	58	0.32851	12 53	62
103	1.00000	100	1	1.00000	100	1
105	0.02073	12 174 205	99	0.03257	174	99
110	0.17378	55 127	83	0.07304	51 128	97
113	0.11507	12 174	92	0.11794	12 174	92
115	0.51738	12 51	40	0.51552	12 53 127	39
121	0.19399	12 14 51	81	0.19870	12 51	79
125	0.23929	14 174	74	0.28386	14 174	70
126	0.19729	14 51	79	0.15939	12 51	88
127	0.13253	12 177 205	89	0.12591	12 53	90
128	0.01207	14 51	100	0.01253	14 51	100
130	0.28710	12 174	66	0.30671	12 174	63
131	1.00000	127	1	1.00000	127	1
132	1.00000	128	1	1.00000	128	1
133	1.00000	130	1	1.00000	130	1

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137	0.11444	12 174 205	93	0.11996	12 174	91
139	0.45200	12 51 177 205	44	0.37526	12 53 154	57
140	0.26298	12 174	70	0.22118	12 174	75
142	0.28825	12 174	65	0.30638	174	64
152	0.19623	12 51 154	80	0.21397	12 140 154	77
154	1.00000	140	1	1.00000	140	1
156	0.26240	12 51	71	0.23388	12 51	74
159	0.32485	12 174	62	0.27817	12 174	71
160	1.00000	154	1	1.00000	154	1
161	0.27883	12 14 174	68	0.42966	14 55 174	49
162	0.26417	12 174	69	0.26863	12 174	72
163	1.00000	160	1	1.00000	160	1
168	1.00000	161	1	1.00000	161	1
173	0.54073	14 51 55 140 177	37	0.41261	12 51 53 154	51
174	1.00000	163	1	1.00000	163	1
176	0.08552	12 174 205	97	0.07647	12 174	96
177	0.17788	12 14 174	82	0.19703	12 174	80
182	0.43200	12 177	50	0.37391	12 53 154	58
185	1.00000	174	1	1.00000	174	1
190	0.85503	12 140 161 177	24	0.67773	12 53 140 154	29
191	1.00000	177	1	1.00000	177	1
192	0.63326	12 51 53 163 177	31	0.57446	12 51 53 163	34
193	0.28203	51 128	67	0.25644	51 128	73
194	0.44540	12 51 205	48	0.39979	12 53 154	53
196	0.39387	12 174	54	0.38420	12 174	54
197	0.59081	161 174	33	0.52594	53 140 161 174	38
198	0.54975	12 174 177	35	0.44945	12 53 154 174	48
200	1.00000	194	1	1.00000	194	1
203	0.21295	12 174 198	76	0.21985	12 174	76
204	0.04463	12 205	98	0.04226	12 205	98
206	0.38743	12 174	55	0.40529	12 174	52
207	1.00000	198	1	1.00000	198	1
210	0.33249	55 161	61	0.34201	53 127 161	60

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-C-3 Comparative CRS-SBM scores for 2007 with different measures of labour

Portfolio Management stage

Using number of employees				Using personnel expenses		
MFMC	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.51806	7 12 50 198	39	0.53588	7 78	39
4	0.43624	12 14 51 127	46	1.00000	4	1
6	0.26835	14 174 177	69	0.28480	83 127 174	67
7	1.00000	7	1	1.00000	7	1
9	0.25915	12 177	72	0.21653	7 174	76
12	1.00000	12	1	1.00000	12	1
14	1.00000	14	1	1.00000	14	1
15	0.62492	12 51 127	34	0.64370	12 51 127 128	32
20	0.17673	12 51	82	0.18961	12 14 51	81
21	0.29413	12 51 127	66	0.24835	51 127	71
24	0.39268	12 152	52	0.33647	7 53	58
29	0.39638	12 50 198	51	0.40170	7 198	52
31	0.38277	12 152 177 198	56	0.32762	7 53	62
34	0.16140	12 50	84	0.16847	7 12 50	84
35	0.74560	12 174 177 198	26	0.71677	7 174	25
36	0.17889	12 174	80	0.22020	12 174	75
37	0.35377	12 152 198	60	0.38018	12 53 78	53
38	0.09342	12 51	90	0.09902	12 51	89
40	0.13492	12 174	89	0.18550	7 12 174	82
43	0.69769	12 50 128 198	28	0.65573	7 128 198	30
45	0.65904	12 174 177 198	30	0.69743	7 174 177	26
46	0.13909	12 174 210	87	0.14997	12 102 210	85
47	0.44036	12 51	45	0.43671	7 50	48
49	0.29208	12 198	67	0.29921	12 198	65
50	1.00000	50	1	1.00000	50	1
51	1.00000	51	1	1.00000	51	1
53	1.00000	53	1	1.00000	53	1
55	1.00000	55	1	0.64596	12 14 51 83	31
57	0.29513	12 198	65	0.36073	12 198	54
58	0.40114	12 51 127	50	0.42236	12 51 127	50
59	0.03667	12 14 51	92	0.02283	12 51	92
61	0.23730	14 51	73	0.17934	12 51 127	83
62	0.62576	12 50 127 177	33	0.50301	7 12 51	42
63	0.35861	12 51	59	0.33326	12 51	60
69	0.47606	198	42	0.57991	12 174 198	35
71	0.68783	12 14 177	29	0.60046	7 83 174	34
75	0.39004	12 51 128	54	0.35884	12 51 128	55
76	0.64337	14 174 177	32	0.53637	7 83 174	38
78	1.00000	78	1	1.00000	78	1
83	1.00000	83	1	1.00000	83	1
84	0.43146	12 14 51 127	48	0.48103	12 14 51 83	43
85	0.34176	12 51 128	61	0.33368	12 51 128	59
86	0.65432	12 50 51 127	31	0.66097	12 50 51 127	29
93	0.19031	12 14 51	78	0.20008	7 50 127	79
95	0.58126	12 51	36	0.71697	7 83 127	24
98	0.23619	12 50	74	0.22109	7 12 50	74
100	1.00000	100	1	1.00000	100	1
102	1.00000	102	1	1.00000	102	1
103	0.17974	12 51 127 128	79	0.08077	51 128	90
105	0.01915	12 14 51	93	0.02273	12 14 53	93
110	0.45598	12 51	43	0.44396	7	46
113	0.19658	12 14 51	77	0.19945	12 51	80
115	0.17387	12 174	83	0.25761	14 174	70
121	0.23576	12 51	75	0.20063	12 51	78
125	0.00157	14 51	95	0.00145	14 51	95
126	0.30830	12 152 177 198	63	0.33316	7 12 78 174	61
127	1.00000	127	1	1.00000	127	1
128	1.00000	128	1	1.00000	128	1
130	0.55948	12 50 51 128	38	0.54276	7 12 50 51	37
131	0.14736	12 198	86	0.14810	12 53	86
132	0.41934	12 51 177	49	0.35142	7 12	56
133	0.36801	12 174 198	57	0.31475	12 174 198	64

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137	0.39150	12 174 198	53	0.44342	12 174	47
139	0.28044	12 51 128 198	68	0.27718	7 53	68
140	1.00000	140	1	1.00000	140	1
142	0.29894	12 51	64	0.27161	12 51	69
152	1.00000	152	1	1.00000	152	1
154	0.90413	12 51 128 140 198	22	0.72932	7 50 51	23
156	0.34074	12 14 174 177	62	0.43167	7 14 174	49
159	0.17862	12 152 198	81	0.20578	12 174 198	77
160	1.00000	160	1	1.00000	160	1
161	0.75521	14 51 55 127 174 177	24	0.66743	7 50 83 127 174	28
162	0.47837	14 51 127	41	0.54722	7 50 127	36
163	1.00000	163	1	1.00000	163	1
168	0.22990	12 14 174 177	76	0.24531	7 12 14 174	72
173	0.35901	12 50	58	0.32617	12 50	63
174	1.00000	174	1	1.00000	174	1
176	0.75386	12 51 149 177 198	25	0.61769	7 50	33
177	1.00000	177	1	1.00000	177	1
182	0.44175	12 51	44	0.42101	12 50	51
185	0.26586	51 128	70	0.22346	51 128	73
190	0.56018	12 50 177 198	37	0.44652	7 12	45
191	0.60545	12 174 198	35	0.51237	12 78 174	41
192	0.38297	14 174 177	55	0.34289	7 174	57
193	0.71789	12 152 177 198	27	0.67350	7 53 78 174 198	27
194	0.48534	12 198	40	0.47500	12 198	44
196	0.04396	12 198	91	0.03811	12 198	91
197	0.43505	12 198	47	0.51600	12 78 174 198	40
198	1.00000	198	1	1.00000	198	1
200	0.15380	51 127	85	0.12812	50 127	88
203	0.00418	174	94	0.00343	102	94
204	0.78538	12 50 177 198	23	0.76028	7 53	22
206	0.26231	12 152 174	71	0.29213	12 53 78 174	66
207	0.13670	12 174	88	0.13746	12 174	87
210	1.00000	210	1	1.00000	210	1

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-C-4 Comparative CRS-SBM scores for 2008 with different measures of labour

Portfolio Management stage

MFMC	Using number of employees			Using personnel expenses		
	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.33993	14 174	40	0.28302	14 174	40
4	0.49801	14 127 174	22	0.51778	14 127	19
6	0.15269	14 174	73	0.14590	14 174	66
7	0.59259	14 15 154 174	16	0.67021	14 127 174	14
9	0.16354	14 174	68	0.08658	14 174	79
12	1.00000	12	1	1.00000	12	1
14	1.00000	14	1	1.00000	14	1
15	1.00000	15	1	1.00000	15	1
20	0.27041	14 127 174	48	0.25956	127 174	43
21	0.29940	14 127	43	0.20925	127	50
24	0.18848	174	63	0.12290	174	71
29	0.16027	174	69	0.11933	174	74
31	0.23317	174	55	0.14881	174	65
34	0.15519	14 174	71	0.12114	174	72
35	0.46813	14 174	24	0.30147	174	37
36	0.08975	14 174	84	0.06627	174	85
37	0.21857	174	58	0.15638	174	63
38	0.06335	14 174	87	0.06834	14 174	84
40	0.04120	14 174	88	0.04951	174	88
43	0.22706	14 174	57	0.18636	14 174	54
45	0.28680	14 174	44	0.15522	174	64
46	0.02133	15 210	90	0.01975	15 174	91
47	0.32840	14 154 174	41	0.28935	14 127 174	38
49	0.16823	15 174	67	0.13104	174	70
50	0.15546	14 174	70	0.13282	14 174	69
51	0.41660	14 127 154 174	32	0.40076	14 127 154 174	29
53	0.21290	210 217	59	1.00000	53	1
55	1.00000	55	1	0.46004	14 127 174	23
57	0.24432	174	52	0.19104	174	52
58	0.43570	12 127 160	28	0.44668	14 127 160	24
61	0.18555	14 174	65	0.17441	14 127 174	57
62	0.36581	14 174	38	0.35091	14 174	36
63	0.51151	14 154 174	20	0.47376	14 127 160	22
69	0.15155	15 174	74	0.13528	15 174	67
71	0.40797	14 174	33	0.35457	14 174	35
76	0.64471	14 55 127 174	14	0.56062	14 127 174	16
78	0.34477	14 174	39	0.26316	174	42
83	1.00000	83	1	0.40667	127 174	28
84	0.39343	14 127 174	35	0.38909	14 127 174	30
85	0.53103	12 14 127 160	18	0.49329	14 127 160	20
86	0.42106	14 154 174	30	0.41893	14 127 174	26
93	0.14790	14 174	75	0.10649	174	77
95	0.51312	14 154 174	19	0.53242	14 127 174	18
98	0.10369	174	79	0.08455	174	81
100	0.01315	210	92	1.00000	100	1
103	0.08069	14 174	85	0.03762	127	89
105	0.02505	14 174	89	0.02742	14 174	90
110	0.43663	14 154 174	27	0.38586	14 127 174	33
113	0.18537	14 174	66	0.21438	14 127 174	49
115	0.10093	14 174	80	0.11028	174	75
121	0.13894	14 174	77	0.10823	14 174	76
125	0.00112	14 174	93	0.00113	14 174	93
126	0.25734	14 174	50	0.20483	174	51
127	1.00000	127	1	1.00000	127	1
128	1.00000	128	1	1.00000	128	1
130	0.56401	14 154 174	17	0.55809	14 154 174	17
131	0.09758	15 174	82	0.09485	15 174	78
132	0.15380	174	72	0.08394	174	82
133	0.23740	15 174	54	0.16016	174	60
137	0.23982	174	53	0.17655	174	56
139	0.19390	154 174	62	0.15675	14 174	62
140	0.64084	14 154 174	15	0.64631	127 154 174	15
142	0.37973	14 154 174	37	0.36662	127	34

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152	0.20657	174 210 217	60	0.15870	174 217	61
154	1.00000	154	1	1.00000	154	1
156	0.26916	14 174	49	0.16486	174	59
159	0.14658	14 15 174	76	0.13415	174	68
160	1.00000	160	1	1.00000	160	1
161	0.40556	14 55 174	34	0.22776	14 174	46
162	0.28408	14 174	46	0.22662	14 174	47
163	0.45947	15 154 174	26	0.47424	154 174	21
168	0.18805	14 174	64	0.11942	174	73
173	0.10768	174	78	0.06551	174	86
174	1.00000	174	1	1.00000	174	1
176	0.47612	14 154 174	23	0.38750	14 154 174	31
177	0.46210	174	25	0.28557	174	39
182	0.24487	14 174	51	0.19078	14 174	53
185	0.43031	12 127 160	29	0.38642	14 127 160	32
190	0.49808	154 174	21	0.43746	154 174	25
191	0.39324	174	36	0.23167	174	45
192	0.28529	14 174	45	0.23386	14 174	44
193	0.28315	174	47	0.21984	174	48
194	0.23248	174	56	0.18601	174	55
196	0.01937	15 174	91	0.01738	15 174	92
197	0.32582	15 174	42	0.26318	174	41
198	1.00000	198	1	1.00000	198	1
200	0.10051	14 174	81	0.07604	127 174	83
203	0.06892	15 174	86	0.06311	15 174	87
204	0.41712	14 174	31	0.41643	14 174	27
206	0.19500	15 174	61	0.16735	174	58
207	0.09353	14 174	83	0.08551	174	80
210	1.00000	210	1	1.00000	210	1
217	1.00000	217	1	1.00000	217	1

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-C-5 Comparative CRS-SBM scores for 2009 with different measures of labour

Portfolio Management stage

Using number of employees				Using personnel expenses		
MFMC	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.40521	7 14 76	42	0.38819	7 14 76	44
4	1.00000	4	1	1.00000	4	1
6	0.13156	14 50 76	82	0.16045	14 50 76	78
7	1.00000	7	1	1.00000	7	1
9	0.23876	76	65	0.22978	76	67
12	0.63083	14 46	25	0.69551	14 46	23
14	1.00000	14	1	1.00000	14	1
15	0.24576	14 46	61	0.26034	14 46	59
20	0.21855	14 46 51	69	0.23557	14 46 51	64
21	0.23710	14 46	66	0.24590	14 46	62
24	0.25751	14 46 76	57	0.27218	76	57
29	0.38650	45 76 131	45	0.40643	45 76 131	41
31	0.51517	7 14 50 51	32	0.44427	7 14 46	35
34	0.14753	14 76	79	0.15063	76	81
35	0.56053	46 76 210	28	0.54203	50 76 174	28
36	0.06559	46 76	89	0.07549	76 174	88
37	0.24667	14 46 76	60	0.27455	14 46 76	56
38	0.04226	14 50 76	90	0.05607	7 14 50	91
40	0.12595	45 46 76 100	84	0.15476	45 76 174	80
43	0.64348	7 14 76 128	24	0.59058	7 14 46 128	26
45	1.00000	45	1	1.00000	45	1
46	1.00000	46	1	1.00000	46	1
47	0.38456	14 46 50 76	46	0.39978	7 14 50	42
49	0.18292	46 76	75	0.18258	46 76	73
50	1.00000	50	1	1.00000	50	1
51	1.00000	51	1	1.00000	51	1
53	0.03428	45 76	92	0.04477	45 76 100 174	92
55	1.00000	55	1	1.00000	55	1
57	0.36334	46 76 210	48	0.34954	46 76 210	49
58	0.39576	14 46 51	43	0.41342	14 46 51	39
61	0.19978	14 50 76	73	0.17760	7 14	75
62	1.00000	62	1	1.00000	62	1
63	0.45249	14 46 51	38	0.43186	14 46 51	37
69	0.03591	45 46 76	91	0.05638	45 50 76 174	90
71	0.70489	7 14 50 76	22	0.76189	7 14 50 76	22
76	1.00000	76	1	1.00000	76	1
78	0.30618	46 76	53	0.31866	46 76	52
83	0.37017	14 46 50 76	47	0.49336	7 127 140	32
84	0.46416	14 46 51	36	1.00000	84	1
85	0.52118	14 46 51	31	0.53493	14 46 51	29
86	0.48328	14 46 51	34	0.56848	14 46 51	27
93	0.14737	14 50 76	80	0.14435	50 76	84
95	0.60825	7 14 46 50 51 140	26	0.62757	7 14 46 127 140	24
98	0.15652	7 14 76	78	0.14605	7 14	82
100	1.00000	100	1	1.00000	100	1
103	0.18388	76	74	0.13271	46 128	85
105	0.01502	14 46	94	0.02238	14 46 50	94
110	0.34448	7 14 50 76	49	0.37690	7 14 50 76	46
113	0.17287	14 51	76	0.20992	14 50	69
115	0.08306	46 50 76 100	86	0.11679	76 100 174	86
121	0.24066	14 46 50	64	0.23113	14 46 127	65
125	0.00176	14 46	95	0.00175	14 46	95
126	0.24345	46 76	63	0.24475	50 76	63
127	1.00000	127	1	1.00000	127	1
128	1.00000	128	1	1.00000	128	1
130	0.85387	7 14 51 128 140	21	0.81420	7 14 51 128 140	21
131	1.00000	131	1	1.00000	131	1
132	0.25683	76	58	0.23040	50 76	66
133	0.20976	45 46 76	71	0.17856	45 46 76	74
137	0.25320	46 76	59	0.27671	76 174	55
139	0.45351	46 76 210	37	0.44078	45 46 76	36
140	1.00000	140	1	1.00000	140	1
142	0.28848	14 46 51	56	0.28020	14 46 51	54

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152	1.00000	152	1	0.16647	45 76 100 174	77
154	0.65242	14 46 76	23	0.61763	7 14 46 76	25
156	0.29329	50 76	55	0.35208	76 174	48
159	0.21585	14 46 76	70	0.22536	14 46 76	68
160	0.31310	14 46 76	51	0.32353	46 76	51
161	0.54754	14 50 76	30	0.53026	50 76	30
162	0.47966	14 46 76	35	0.48333	50 76	33
163	0.41834	7 14 50 76	40	0.41213	7 14 46 76	40
168	0.20792	46 76	72	0.20245	46 76	70
173	0.40896	14 50 51	41	0.36590	7 14 46	47
174	1.00000	174	1	1.00000	174	1
176	0.58569	14 50 51 76	27	0.52523	7 14 46 50	31
177	0.12601	14 50 76	83	0.15941	76 174	79
182	0.38700	14 50 76	44	0.38750	14 46 50	45
185	0.32710	14 51	50	0.29825	7 14 46	53
190	0.49875	7 14 50 76	33	0.46920	7 14 46 76	34
191	0.55086	46 76 174	29	0.39083	45 76 174	43
192	0.15836	14 50 76	77	0.17071	76 174	76
193	1.00000	193	1	1.00000	193	1
194	0.42958	46 76 210	39	0.43137	46 76 210	38
195	0.30678	46 76 210	52	0.34296	46 76 174	50
196	0.02243	14 46 76	93	0.02336	14 46 76	93
197	0.30023	45 76 174	54	0.26811	45 76	58
198	0.22702	14 46 50 76	68	0.25106	14 46 50 76	61
200	0.11143	14 46	85	0.14519	14 46 127	83
203	0.07050	45 46 76	88	0.06514	45 46 76	89
204	0.24552	14 46 76	62	0.25271	14 50 76	60
206	0.14670	45 46 76	81	0.19499	76 174	71
207	0.08097	14 46 76	87	0.08387	14 46 76	87
210	1.00000	210	1	1.00000	210	1
217	1.00000	217	1	1.00000	217	1
221	0.23553	50 76 174	67	0.19116	50 76 174	72

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Appendix D

Table I-D-1 Comparative CRS-SBM scores for 2005 with different measures of capital

Portfolio Management stage

MFMC	Using shareholders' equity			Using furniture and equipment		
	SBM	Ref.	Rank	SBM	Ref.	Rank
1	0.33795	12 161	66	0.53279	58 76 161	26
2	0.40030	12 177	59	0.12208	14 161	73
4	0.42223	14 161	55	0.39042	14 161	36
6	1.00000	6	1	1.00000	6	1
7	1.00000	7	1	0.46946	58 161	27
9	0.24942	12 177	78	0.11632	55 161	75
12	1.00000	12	1	1.00000	12	1
14	1.00000	14	1	1.00000	14	1
15	0.54453	12 161	44	0.42234	14 161	31
20	0.49983	12 161	48	0.36019	14 161	40
21	0.21632	161	80	0.19796	161	60
24	0.41055	12 177	56	0.20284	161	59
29	0.45102	12 43 177	53	0.17641	58 76	64
31	0.64661	12 177	36	0.20623	14 161	58
34	0.20845	12 177	81	0.16511	58 76	68
35	0.81735	12 40 160 176 177	30	0.41603	76 161	33
36	0.17644	12 174	85	0.11162	58 76	76
37	0.32688	12 40 174 177	68	0.05238	14 161	88
38	0.12944	12 160 176	89	0.08748	14 58 161	80
40	1.00000	40	1	1.00000	40	1
42	0.11366	12 55 174	93	0.08123	58 76	81
43	1.00000	43	1	1.00000	43	1
45	0.55011	12 174 177	43	0.32369	58 76	43
46	0.06900	12 152	96	0.01116	14 161	98
47	0.48304	12 176	49	0.26208	161	50
49	0.38867	12 40 174 177	60	1.00000	49	1
50	0.14370	12 177	88	0.10637	58 76 161	77
51	0.58078	12 40 160 161	39	0.26209	40 161	49
53	1.00000	53	1	0.03129	76 161 40	93
55	1.00000	55	1	1.00000	55	1
57	0.28221	12 174	74	0.07292	76 161	84
58	0.70651	12 161	35	1.00000	58	1
59	0.02740	12 161	99	0.02628	14	95
60	0.00646	12 40	102	0.00536	58 161	101
61	0.31073	55 161	70	0.32169	161	44
62	0.64389	12 160 161 176	37	0.29555	58 161	47
63	0.76928	12 43 154 160	31	0.33739	14 161	41
69	0.11333	12 174 177	94	0.03724	14 76 161	90
71	1.00000	71	1	0.75293	76 158 161	22
73	0.70830	12 174 177	33	0.25351	76 161	51
75	0.47698	12 161	50	0.38013	58 161	38
76	1.00000	76	1	1.00000	76	1
78	0.35690	12 174 177	63	0.11863	58 76	74
83	0.61178	12 55 161	38	0.37250	14 55 161	39
84	0.45669	14 161	52	0.45139	14 161	30
85	0.46772	12 161	51	0.39786	14 161	35
86	0.50307	161	46	0.38555	58 161	37
93	0.19591	12 55 161	83	0.17589	58 161	65
95	0.73830	55 161	32	0.60206	14 161	24
98	0.38578	12 43 160	62	0.07966	14 161	82
100	1.00000	100	1	1.00000	100	1
102	0.03736	12 152	98	0.01015	76	99
103	1.00000	103	1	1.00000	103	1
105	0.41006	12 163 177	57	0.10630	40 58	78
110	0.56571	12 161	40	0.56575	58 161	25
113	0.17183	12 14 161	86	0.14956	14 161	72
115	0.15611	12 100 174	87	0.06802	76 100	86
121	0.27680	161	75	0.31817	58 76	45
123	0.12551	12 160 177	91	0.02579	14 161	96

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125	0.01309	14 161	101	0.02950	40 58	94
126	0.35117	12 177	65	0.15364	58 76	70
127	1.00000	127	1	1.00000	127	1
128	1.00000	128	1	1.00000	128	1
130	1.00000	130	1	0.46603	58 161	28
131	0.12109	12 152	92	0.03915	58 76	89
132	0.26751	12 161 177	76	0.23848	58 161	54
133	0.28577	12 40 174	73	0.07821	76 161	83
137	0.30621	12 174	72	0.15272	58 76	71
138	0.18856	12 152	84	0.01647	58 76	97
139	0.20468	12 177	82	0.18147	40 58 76	61
140	1.00000	140	1	0.41887	161	32
142	0.35601	161	64	0.27316	161	48
152	1.00000	152	1	0.06910	76 100	85
154	1.00000	154	1	0.41481	14 161	34
156	0.31727	12 55 174	69	0.21961	58 76	56
159	0.32720	12 55 174 177	67	1.00000	159	1
160	1.00000	160	1	1.00000	160	1
161	1.00000	161	1	1.00000	161	1
162	0.70769	12 55 154 176 177	34	0.46184	161	29
163	1.00000	163	1	1.00000	163	1
164	0.06441	12 174	97	0.03274	58 76	91
168	0.23644	12 55 174	79	0.17642	58 76	63
173	0.30974	12 177	71	0.21370	58 76 161	57
174	1.00000	174	1	1.00000	174	1
176	1.00000	176	1	0.65023	76 161	23
177	1.00000	177	1	1.00000	177	1
182	0.55858	12 176 177	41	0.31532	58 161	46
185	0.38616	12 43 154	61	0.22769	58 161	55
189	0.12673	12 152	90	0.00719	58 76	100
190	0.55505	12 176 177	42	0.24303	14 161	52
191	0.40856	12 174 177	58	0.17693	14 76 161	62
192	1.00000	192	1	1.00000	192	1
193	1.00000	193	1	1.00000	193	1
194	0.82258	12 43 174 198	29	0.16771	58 76	67
195	0.53370	12 174 198	45	0.17460	58 76	66
196	0.01433	12 40 152	100	0.00165	14 161	102
197	0.50042	12 174 177	47	0.15706	58 76 161	69
198	1.00000	198	1	0.10180	58 161	79
200	0.42806	55 174 177	54	0.32734	161 174	42
202	0.25857	12 40 161 174 177	77	0.24248	40 58 76	53
205	1.00000	205	1	0.03209	58 76	92
206	0.10614	12 152 174	95	0.05562	58 76	87

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-D-2 Comparative CRS-SBM scores for 2006 with different measures of capital

Portfolio Management stage						
MFMC	Using shareholders' equity			Using furniture and equipment		
	SBM	Ref.	Rank	SBM	Ref.	Rank
2						
4	0.44895	12 177 205	47	0.10996	14 161	82
6	0.36609	14 51	57	0.34527	14 51	50
7	0.24471	51 55 161	72	0.28396	14 161	56
9	0.80473	12 51 53 154 163	26	0.46688	58 161	38
12	0.21437	140 174	75	0.11370	14 161	81
14	1.00000	12	1	1.00000	12	1
15	1.00000	14	1	1.00000	14	1
20	0.56122	12 51 127	34	0.45221	12 51 59 127	40
21	0.54491	12 51 127	36	0.38326	14 127	46
24	0.24047	127	73	0.25921	14 127	59
29	0.42342	12 177 205	51	0.21261	14 161	62
31	0.41718	12 128 205	52	0.20492	58 192 205	63
34	0.84051	12 51 53 154 163	25	0.33377	14 51 161	51
35	0.16519	12 177	84	0.12525	58 161	80
36	0.51953	12 174	39	0.37968	58 76 161	47
37	0.16393	12 14 174	85	1.00000	36	1
38	0.37920	12 174 205	56	0.06105	14 51 161	90
40	0.10118	12 51 163	95	0.05737	14 205	91
42	0.10521	174	94	1.00000	40	1
43	0.14496	14 174	88	0.18994	76 100 36	67
45	0.69014	128 154 163 198 205	28	0.43122	58 161 192	44
46	0.50592	12 174	41	0.43185	76 174 192	43
47	0.12119	12 205	91	0.02474	14 205	97
49	0.41186	12 51	53	0.29090	161	53
50	0.32253	12 174	63	1.00000	49	1
51	0.12725	12 174	90	0.13339	58 76 161 205	79
53	1.00000	51	1	1.00000	51	1
55	1.00000	53	1	0.03456	14 51	95
57	1.00000	55	1	1.00000	55	1
58	0.34304	174	60	0.08702	14 161	85
59	0.53423	12 51 128	38	1.00000	58	1
60	1.00000	59	1	1.00000	59	1
61	0.00147	12 205	102	0.00154	58 205	101
62	0.20216	14 161	78	0.22262	14 161	60
63	1.00000	62	1	1.00000	62	1
69	0.62818	12 53 128 163 205	32	0.28924	14 51 127	54
71	0.16152	12 174 205	86	0.04714	14 174	93
75	1.00000	71	1	0.72412	58 161	27
76	0.34772	12 51 128	59	0.36397	14 51	49
78	0.64041	55 174	30	1.00000	76	1
83	0.44933	12 174	46	0.16402	14 58 161	73
84	0.73548	12 51 55 127 140	27	0.47999	14 51 127	37
85	0.44491	14 51 127	49	0.55870	14 51 127	29
86	0.45866	12 51 128	43	0.44679	14 51	41
93	0.50058	12 51	42	0.54407	58 161	31
95	0.20408	14 51 55	77	0.18631	161 205	69
98	0.45104	12 14 51	45	0.48983	14 161	34
100	0.35902	12 51	58	0.07141	14 51	89
102	1.00000	100	1	1.00000	100	1
103	0.02073	12 174 205	99	0.00771	76	100
105	0.17378	55 127	83	0.19168	161	66
110	0.11507	12 174	92	1.00000	105	1
113	0.51738	12 51	40	0.48021	58 161	36
115	0.19399	12 14 51	81	0.14310	14 127	78
121	0.23929	14 174	74	1.00000	115	1
123	0.19729	14 51	79	0.22036	161	61
125	0.13253	12 177 205	89	0.02708	14 51	96
126	0.01207	14 51	100	0.04535	40 58 139	94
127	0.28710	12 174	66	0.08598	14 161	87
128	1.00000	127	1	1.00000	127	1
130	1.00000	128	1	1.00000	128	1
131	1.00000	130	1	0.55430	14 62 127	30

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132	0.11444	12 174 205	93	0.05136	58 76	92
133	0.45200	12 51 177 205	44	0.36619	58 161	48
137	0.26298	12 174	70	0.08046	161 174	88
139	0.28825	12 174	65	0.15316	14 58 161	76
140	0.19623	12 51 154	80	1.00000	139	1
142	1.00000	140	1	1.00000	140	1
152	0.26240	12 51	71	0.27109	58 161	58
154	0.32485	12 174	62	1.00000	152	1
156	1.00000	154	1	0.46301	14 51 161	39
159	0.27883	12 14 174	68	0.15089	14 161 174	77
160	0.26417	12 174	69	0.18317	58 76	70
161	1.00000	160	1	1.00000	160	1
162	1.00000	161	1	1.00000	161	1
163	0.54073	14 51 55 140 177	37	0.39332	161	45
164	1.00000	163	1	0.51069	58 127 161	32
168	0.08552	12 174 205	97	0.08689	76 58 161 205	86
173	0.17788	12 14 174	82	0.17425	58 76	71
174	0.43200	12 177	50	0.31460	58 161	52
176	1.00000	174	1	1.00000	174	1
177	0.85503	12 140 161 177	24	0.57588	14 161	28
182	1.00000	177	1	1.00000	177	1
185	0.63326	12 51 53 163 177	31	0.48200	58 62 161	35
190	0.28203	51 128	67	0.20477	14 51 127	64
191	0.44540	12 51 205	48	0.18754	14 161	68
192	0.39387	12 174	54	0.20023	14 174	65
193	0.59081	161 174	33	1.00000	192	1
194	0.54975	12 174 177	35	0.43266	161 174	42
195	1.00000	194	1	1.00000	194	1
196	0.21295	12 174 198	76	0.09168	58 76	83
197	0.04463	12 205	98	0.00936	14 205	99
198	0.38743	12 174	55	0.16609	58 76 161	72
200	1.00000	198	1	0.15397	14 58 161	75
202	0.33249	55 161	61	0.27155	161	57
203	0.30145	12 177 205	64	0.15527	58 161 205	74
204	0.00248	205	101	0.00036	205	102
205	0.86527	12 14 174 177 205	23	0.50161	51 161 174	33
206	1.00000	205	1	1.00000	205	1
207	0.15011	12 174	87	0.08915	40 58 76	84
210	0.09895	12 205	96	0.02457	14 205	98

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-D-3 Comparative CRS-SBM scores for 2007 with different measures of capital

Portfolio Management stage

MFMC	Using shareholders' equity			Using furniture and equipment		
	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.51806	7 12 50 198	39	0.12766	14 50 127	74
4	0.43624	12 14 51 127	46	0.50552	12 14 51 127	31
6	0.26835	14 174 177	69	0.37524	14 152 174 177	42
7	1.00000	7	1	0.53578	14 50 127	27
9	0.25915	12 177	72	0.14459	51 55	70
12	1.00000	12	1	1.00000	12	1
14	1.00000	14	1	1.00000	14	1
15	0.62492	12 51 127	34	0.53020	12 51 127	29
20	0.17673	12 51	82	0.14966	14 51	68
21	0.29413	12 51 127	66	0.35075	14 51 127	44
24	0.39268	12 152	52	0.16639	14 50 177	65
29	0.39638	12 50 198	51	0.19723	95	60
31	0.38277	12 152 177 198	56	0.20128	14 51 177	58
34	0.16140	12 50	84	0.13508	95	73
35	0.74560	12 174 177 198	26	0.60963	95 152 161	26
36	0.17889	12 174	80	0.11796	14 174	76
37	0.35377	12 152 198	60	0.08321	14 51 177	82
38	0.09342	12 51	90	0.08357	14 50	81
40	0.13492	12 174	89	0.04975	14 174	85
43	0.69769	12 50 128 198	28	0.32216	95	46
45	0.65904	12 174 177 198	30	0.83944	152 161	22
46	0.13909	12 174 210	87	0.01078	14 174	92
47	0.44036	12 51	45	0.31184	14 51	47
49	0.29208	12 198	67	0.14381	95 152	72
50	1.00000	50	1	1.00000	50	1
51	1.00000	51	1	1.00000	51	1
53	1.00000	53	1	0.01003	14 51	93
55	1.00000	55	1	1.00000	55	1
57	0.29513	12 198	65	0.11133	14 50 95	77
58	0.40114	12 51 127	50	1.00000	58	1
59	0.03667	12 14 51	92	0.02486	14 51	89
61	0.23730	14 51	73	0.26096	14 50	54
62	0.62576	12 50 127 177	33	1.00000	62	1
63	0.35861	12 51	59	0.28227	12 14 51	51
69	0.47606	198	42	0.04440	14 174 177	86
71	0.68783	12 14 177	29	1.00000	71	1
75	0.39004	12 51 128	54	0.45675	14 51 127	34
76	0.64337	14 174 177	32	1.00000	76	1
78	1.00000	78	1	0.33240	14 95 161 177	45
83	1.00000	83	1	0.68825	12 55 127 128	24
84	0.43146	12 14 51 127	48	0.53546	14 51 127	28
85	0.34176	12 51 128	61	0.43828	12 14 151 127	36
86	0.65432	12 50 51 127	31	1.00000	86	1
93	0.19031	12 14 51	78	0.23442	14 50 95 152	56
95	0.58126	12 51	36	1.00000	95	1
98	0.23619	12 50	74	0.06738	12 14 51	83
100	1.00000	100	1	1.00000	100	1
102	1.00000	102	1	0.01355	14 174	90
103	0.17974	12 51 127 128	79	0.19291	14 95 127	61
105	0.01915	12 14 51	93	0.03419	152 95	88
110	0.45598	12 51	43	0.46672	50 95	33
113	0.19658	12 14 51	77	0.15834	14 50 51	67
115	0.17387	12 174	83	0.21073	152 174 95	57
121	0.23576	12 51	75	0.27867	14 50	52
125	0.00157	14 51	95	0.00596	95	94
126	0.30830	12 152 177 198	63	0.12524	14 174 177	75
127	1.00000	127	1	1.00000	127	1
128	1.00000	128	1	1.00000	128	1
130	0.55948	12 50 51 128	38	0.44063	14 50	35
131	0.14736	12 198	86	1.00000	131	1
132	0.41934	12 51 177	49	0.41650	152 95	38
133	0.36801	12 174 198	57	0.09574	14 174	79
137	0.39150	12 174 198	53	0.17360	14 174	63

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139	0.28044	12 51 128 198	68	0.25274	152 95	55
140	1.00000	140	1	0.64220	14 50 127	25
142	0.29894	12 51	64	0.41284	50 58	39
152	1.00000	152	1	1.00000	152	1
154	0.90413	12 51 128 140 198	22	0.48836	14 50 127	32
156	0.34074	12 14 174 177	62	0.19130	14 174 177	62
159	0.17862	12 152 198	81	0.09006	95 152	80
160	1.00000	160	1	1.00000	160	1
161	0.75521	14 51 55 127 174 177	24	1.00000	161	1
162	0.47837	14 51 127	41	0.52254	14 50 177	30
163	1.00000	163	1	0.41073	50 95	40
168	0.22990	12 14 174 177	76	0.36918	95 152	43
173	0.35901	12 50	58	0.31172	95	48
174	1.00000	174	1	1.00000	174	1
176	0.75386	12 51 149 177 198	25	0.70812	14 50 127 177	23
177	1.00000	177	1	1.00000	177	1
182	0.44175	12 51	44	0.38040	95	41
185	0.26586	51 128	70	0.20105	127	59
190	0.56018	12 50 177 198	37	0.28572	14 127 177	50
191	0.60545	12 174 198	35	0.26171	14 174 177	53
192	0.38297	14 174 177	55	1.00000	192	1
193	0.71789	12 152 177 198	27	0.41897	14 152 161 177	37
194	0.48534	12 198	40	0.10803	14 51	78
196	0.04396	12 198	91	0.01123	14 152	91
197	0.43505	12 198	47	0.16896	95 152 177	64
198	1.00000	198	1	0.14829	14 51 177	69
200	0.15380	51 127	85	0.16180	51 127	66
203	0.00418	174	94	0.00045	174	95
204	0.78538	12 50 177 198	23	0.30341	14 51 177	49
206	0.26231	12 152 174	71	0.06269	14 152 174	84
207	0.13670	12 174	88	0.03710	14 174	87
210	1.00000	210	1	0.14448	14 95 152 177	71

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-D-4 Comparative CRS-SBM scores for 2008 with different measures of capital

Portfolio Management stage

MFMC	Using shareholders' equity			Using furniture and equipment		
	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.33993	14 174	40	0.11392	14 174	64
4	0.49801	14 127 174	22	0.49613	14 127	16
6	0.15269	14 174	73	0.18451	14 76 174	48
7	0.59259	14 15 154 174	16	0.32926	127 174	30
9	0.16354	14 174	68	0.12050	14 174	61
12	1.00000	12	1	1.00000	12	1
14	1.00000	14	1	1.00000	14	1
15	1.00000	15	1	0.37024	12 127	25
20	0.27041	14 127 174	48	0.22664	127	42
21	0.29940	14 127	43	0.24527	127	39
24	0.18848	174	63	0.11657	14 174	63
29	0.16027	174	69	0.10700	76	69
31	0.23317	174	55	0.13988	14 174	56
34	0.15519	14 174	71	0.10824	76 127	67
35	0.46813	14 174	24	0.31579	76 174	31
36	0.08975	14 174	84	0.05197	14 174	80
37	0.21857	174	58	0.06875	14 174	75
38	0.06335	14 174	87	0.04548	14 174	82
40	0.04120	14 174	88	0.01670	14 174	86
43	0.22706	14 174	57	0.15729	127	53
45	0.28680	14 174	44	1.00000	45	1
46	0.02133	15 210	90	0.00248	76	92
47	0.32840	14 154 174	41	0.24072	14 174	40
49	0.16823	15 174	67	0.11990	76 127	62
50	0.15546	14 174	70	1.00000	50	1
51	0.41660	14 127 154 174	32	0.28915	14 127 174	33
53	0.21290	210 217	59	0.00573	14 174	90
55	1.00000	55	1	1.00000	55	1
57	0.24432	174	52	0.08907	14 127 174	72
58	0.43570	12 127 160	28	0.57284	14 127	15
61	0.18555	14 174	65	0.16743	14 174	50
62	0.36581	14 174	38	0.19300	14 174	47
63	0.51151	14 154 174	20	0.35701	14 127	26
69	0.15155	15 174	74	0.01815	14 174	85
71	0.40797	14 174	33	0.48912	76 174	17
76	0.64471	14 55 127 174	14	1.00000	76	1
78	0.34477	14 174	39	0.17621	14 76 174	49
83	1.00000	83	1	0.33031	14 55 174	29
84	0.39343	14 127 174	35	0.37249	14 127 174	24
85	0.53103	12 14 127 160	18	0.45923	14 127	18
86	0.42106	14 154 174	30	0.29132	127	32
93	0.14790	14 174	75	0.13290	76 174	57
95	0.51312	14 154 174	19	0.39267	14 127 174	22
98	0.10369	174	79	0.02021	14 174	84
100	0.01315	210	92	0.00206	76	93
103	0.08069	14 174	85	0.07972	127	74
105	0.02505	14 174	89	0.05409	127	78
110	0.43663	14 154 174	27	0.33701	127 174	28
113	0.18537	14 174	66	0.14933	14 127 174	54
115	0.10093	14 174	80	1.00000	115	1
121	0.13894	14 174	77	0.12597	14 174	59
125	0.00112	14 174	93	0.00710	127	89
126	0.25734	14 174	50	0.11030	14 76 174	65
127	1.00000	127	1	1.00000	127	1
128	1.00000	128	1	1.00000	128	1
130	0.56401	14 154 174	17	0.35436	14 127 174	27
131	0.09758	15 174	82	0.05992	127	77
132	0.15380	174	72	0.14856	76	55
133	0.23740	15 174	54	0.05331	14 174	79
137	0.23982	174	53	0.10837	14 76	66
139	0.19390	154 174	62	0.16001	76 127	52
140	0.64084	14 154 174	15	0.37605	14 127 174	23
142	0.37973	14 154 174	37	0.28503	127	35

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152	0.20657	174 210 217	60	1.00000	152	1
154	1.00000	154	1	0.64112	14 127 174	14
156	0.26916	14 174	49	0.27429	76	37
159	0.14658	14 15 174	76	0.06350	76	76
160	1.00000	160	1	1.00000	160	1
161	0.40556	14 55 174	34	1.00000	161	1
162	0.28408	14 174	46	0.27717	127 174	36
163	0.45947	15 154 174	26	0.26390	127	38
168	0.18805	14 174	64	0.19756	76 127	46
173	0.10768	174	78	0.10007	76	70
174	1.00000	174	1	1.00000	174	1
176	0.47612	14 154 174	23	0.40663	127 174	21
177	0.46210	174	25	0.45249	174	19
182	0.24487	14 174	51	0.20698	127	45
185	0.43031	12 127 160	29	0.28862	12 127 160	34
190	0.49808	154 174	21	0.24055	14 127 174	41
191	0.39324	174	36	0.16052	14 174	51
192	0.28529	14 174	45	0.41315	76 174	20
193	0.28315	174	47	0.12604	14 76	58
194	0.23248	174	56	0.12345	127	60
196	0.01937	15 174	91	0.00485	14 174	91
197	0.32582	15 174	42	0.08853	76	73
198	1.00000	198	1	0.10709	14 174	68
200	0.10051	14 174	81	0.09009	14 127	71
203	0.06892	15 174	86	0.01158	14 76 174	87
204	0.41712	14 174	31	0.20905	14 174	44
206	0.19500	15 174	61	0.04959	14 76 174	81
207	0.09353	14 174	83	0.02378	14 174	83
210	1.00000	210	1	0.22426	14 76 174	43
217	1.00000	217	1	0.00929	14 174	88

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-D-5 Comparative CRS-SBM scores for 2009 with different measures of capital

Portfolio Management stage

MFMC	Using shareholders' equity			Using furniture and equipment		
	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.40521	7 14 76	42	0.14820	14 76 127	64
4	1.00000	4	1	1.00000	4	1
6	0.13156	14 50 76	82	0.14564	14 50 76	66
7	1.00000	7	1	0.60022	12 14 50 76 127	23
9	0.23876	76	65	0.13238	14 50 76	70
12	0.63083	14 46	25	1.00000	12	1
14	1.00000	14	1	1.00000	14	1
15	0.24576	14 46	61	0.19277	14 46	55
20	0.21855	14 46 51	69	0.18583	14 46	57
21	0.23710	14 46	66	0.34658	12 14 46 127	33
24	0.25751	14 46 76	57	0.15320	14 50 76	63
29	0.38650	45 76 131	45	0.09405	76	76
31	0.51517	7 14 50 51	32	0.26786	14 51 76	40
34	0.14753	14 76	79	0.10147	76	74
35	0.56053	46 76 210	28	0.28415	50 76	38
36	0.06559	46 76	89	0.03314	50 76	87
37	0.24667	14 46 76	60	0.06299	14 50 76	83
38	0.04226	14 50 76	90	0.04588	12 14 50 76	85
40	0.12595	45 46 76 100	84	0.04379	50 76	86
43	0.64348	7 14 76 128	24	1.00000	43	1
45	1.00000	45	1	0.29914	76 100	36
46	1.00000	46	1	1.00000	46	1
47	0.38456	14 46 50 76	46	0.20653	14 50	51
49	0.18292	46 76	75	0.22863	76 139	45
50	1.00000	50	1	1.00000	50	1
51	1.00000	51	1	1.00000	51	1
53	0.03428	45 76	92	0.00379	14 50 76	95
55	1.00000	55	1	1.00000	55	1
57	0.36334	46 76 210	48	0.09897	14 50 76	75
58	0.39576	14 46 51	43	0.50144	12 46 127	25
61	0.19978	14 50 76	73	0.18283	14 50 76	58
62	1.00000	62	1	0.20695	14 50 51 76	50
63	0.45249	14 46 51	38	0.30849	14 46 51	35
69	0.03591	45 46 76	91	0.00735	14 50	94
71	0.70489	7 14 50 76	22	0.67773	12 43 50 76	21
76	1.00000	76	1	1.00000	76	1
78	0.30618	46 76	53	0.16103	14 50 76	60
83	0.37017	14 46 50 76	47	0.22730	14 50 76	46
84	0.46416	14 46 51	36	0.47878	14 46 51	26
85	0.52118	14 46 51	31	1.00000	85	1
86	0.48328	14 46 51	34	1.00000	86	1
93	0.14737	14 50 76	80	0.11659	14 50 76	72
95	0.60825	7 14 46 50 51 140	26	0.45715	14 50 127	29
98	0.15652	7 14 76	78	0.02392	14	89
100	1.00000	100	1	1.00000	100	1
103	0.18388	76	74	0.13552	76	69
105	0.01502	14 46	94	0.46509	76 131 139	28
110	0.34448	7 14 50 76	49	0.25858	14 50 76 127	41
113	0.17287	14 51	76	0.15994	14 50	61
115	0.08306	46 50 76 100	86	1.00000	115	1
121	0.24066	14 46 50	64	0.24069	14 50	42
125	0.00176	14 46	95	0.00892	76	92
126	0.24345	46 76	63	0.14008	50 76	67
127	1.00000	127	1	1.00000	127	1
128	1.00000	128	1	0.19289	14 46	54
130	0.85387	7 14 51 128 140	21	0.32377	14 50 127	34
131	1.00000	131	1	1.00000	131	1
132	0.25683	76	58	0.37387	76	31
133	0.20976	45 46 76	71	0.07055	14 50 76	82
137	0.25320	46 76	59	0.08798	50 76	79
139	0.45351	46 76 210	37	1.00000	139	1
140	1.00000	140	1	1.00000	140	1
142	0.28848	14 46 51	56	0.55662	12 43 50 76	24

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152	1.00000	152	1	1.00000	152	1
154	0.65242	14 46 76	23	0.42598	14 50 76	30
156	0.29329	50 76	55	0.21282	50 76	49
159	0.21585	14 46 76	70	0.14749	50 76	65
160	0.31310	14 46 76	51	0.24032	50 76	43
161	0.54754	14 50 76	30	0.46896	76	27
162	0.47966	14 46 76	35	0.35205	14 50 76	32
163	0.41834	7 14 50 76	40	0.20179	12 14 76 127	53
168	0.20792	46 76	72	0.11122	50 76	73
173	0.40896	14 50 51	41	0.16215	14 51	59
174	1.00000	174	1	1.00000	174	1
176	0.58569	14 50 51 76	27	0.61178	12 50 76 127	22
177	0.12601	14 50 76	83	0.08964	50 76	78
182	0.38700	14 50 76	44	0.23744	14 50 76	44
185	0.32710	14 51	50	0.20472	14 50 127	52
190	0.49875	7 14 50 76	33	0.22185	14 50 76	47
191	0.55086	46 76 174	29	0.28638	50 76	37
192	0.15836	14 50 76	77	0.15624	50 76	62
193	1.00000	193	1	1.00000	193	1
194	0.42958	46 76 210	39	0.18687	12 50 76	56
195	0.30678	46 76 210	52	0.28313	76 139	39
196	0.02243	14 46 76	93	0.00822	14 46	93
197	0.30023	45 76 174	54	0.08343	76	81
198	0.22702	14 46 50 76	68	0.09032	14 50 76	77
200	0.11143	14 46	85	0.08670	14 46	80
203	0.07050	45 46 76	88	0.01689	76	90
204	0.24552	14 46 76	62	0.13568	14 50 76	68
206	0.14670	45 46 76	81	0.04676	14 76	84
207	0.08097	14 46 76	87	0.02679	14 50 76	88
210	1.00000	210	1	0.22037	14 50 76	48
217	1.00000	217	1	0.01562	14 50 76	91
221	0.23553	50 76 174	67	0.12272	50 76	71

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Appendix E

Table I-E-1 Comparative CRS-SBM scores for 2005 with different measures of capital

Overall Efficiency

MFMC	Using shareholders' equity			Using furniture and equipment		
	SBM	Ref.	Rank	SBM	Ref.	Rank
1	0.04431	196	99	0.01636	125	91
2	0.04597	100	97	0.01384	125	96
4	0.14556	125	55	0.06605	125	42
6	0.09447	196	71	0.03841	125	64
7	0.07625	100	82	0.02039	125	82
9	0.27668	100	25	0.08061	125	38
12	1.00000	12	1	0.20643	125	17
14	1.00000	14	1	0.22107	125	15
15	0.13830	200	57	0.04611	125	57
20	0.26639	196	28	0.07839	125	39
21	0.20771	125	36	0.09288	125	32
24	0.06403	100	89	0.01842	125	86
29	0.08061	100	79	0.02533	125	77
31	0.07928	100	80	0.01951	125	83
34	0.10354	196	68	0.03391	125	67
35	0.08897	100	75	0.02425	125	78
36	0.34542	100	19	0.17869	125	21
37	0.17881	100	43	0.06041	125	46
38	0.19097	100	40	0.07535	125	41
40	1.00000	40	1	0.26513	125	10
42	0.33274	100	20	0.19157	125	19
43	0.05890	100	90	0.01534	125	94
45	0.21755	100	35	0.06559	125	45
46	0.58714	100	14	0.32322	125	7
47	0.07059	100	84	0.01946	125	84
49	0.19546	100	39	0.10952	125	28
50	0.22464	100	33	0.13513	125	25
51	0.13051	100	59	0.04299	125	58
53	0.26866	100	27	0.17546	100	22
55	0.30758	196	21	0.08741	125	35
57	0.15238	100	51	0.05290	125	51
58	0.18955	196	41	1.00000	58	1
59	0.14996	125	53	0.09780	125	30
60	0.35570	196	18	0.33940	125	6
61	0.03392	196	100	0.01289	125	98
62	0.15798	100	49	0.04718	125	56
63	0.04542	100	98	0.01184	125	100
69	0.12470	100	61	0.08549	100	36
71	0.10305	100	70	0.02828	125	74
73	0.17174	100	45	0.03774	125	65
75	0.05309	100	94	0.01535	125	93
76	0.26174	100	29	0.25594	125	11
78	0.12188	100	62	0.04133	125	60
83	0.07843	100	81	0.02277	125	79
84	0.49052	125 196	15	0.15525	125	23
85	0.15520	196	50	0.04957	125	55
86	0.23166	100	32	0.06599	125	43
93	0.13984	196	56	0.05232	125	52
95	0.28886	100	24	0.08063	125	37
98	0.18462	100	42	0.05357	125	50
100	1.00000	100	1	1.00000	100	1
102	0.65068	100 196	13	0.71453	100 125	5
103	1.00000	103	1	0.24808	125	12
105	0.01924	100	101	0.00713	125	101
110	0.08687	100	77	0.02626	125	75
113	0.01282	196	102	0.00487	125	102
115	0.17119	100	46	0.11405	125	27
121	0.08832	196	76	0.02928	125	73
123	0.11286	100	63	0.04100	125	61

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125	1.00000	125	1	1.00000	125	1
126	0.09324	100	72	0.03274	125	68
127	0.10429	100	67	0.03041	125	72
128	0.10884	196	65	0.03117	125	71
130	0.06501	100	86	0.01642	125	90
131	0.29562	100	22	0.18149	125	20
132	0.08999	100	74	0.03233	125	69
133	0.24787	100	31	0.08768	125	34
137	0.20631	100	37	0.06579	125	44
138	0.73150	100 152	11	0.27757	100	9
139	0.41251	196	16	0.21442	125	16
140	0.05620	100	92	0.01492	125	95
142	0.08564	196	78	0.02607	125	76
152	1.00000	152	1	0.24033	125	13
154	0.04904	100	96	0.01222	125	99
156	0.11235	100	64	0.05047	125	54
159	0.10316	100	69	0.04195	125	59
160	0.15886	100	48	0.05954	125	48
161	0.07363	100	83	0.02086	125	81
162	0.12509	100	60	0.03177	125	70
163	0.05362	100	93	0.01367	125	97
164	0.69829	125 196	12	0.29307	125	8
168	0.15203	100	52	0.06025	125	47
173	0.05016	100	95	0.01666	125	89
174	0.37122	100 205	17	0.08986	125	33
176	0.06601	100	85	0.01775	125	88
177	0.09050	100	73	0.02274	125	80
182	0.06486	100	87	0.01828	125	87
185	0.27353	100	26	0.07611	125	40
189	0.29159	100	23	0.09519	100	31
190	0.06482	100	88	0.01863	125	85
191	0.16560	100	47	0.05370	125	49
192	0.10601	100	66	0.03647	125	66
193	0.05747	100	91	0.01609	125	92
194	0.17578	100	44	0.04062	125	62
195	0.24808	100	30	0.10941	125	29
196	1.00000	196	1	1.00000	196	1
197	0.13386	100	58	0.05123	125	53
198	0.14766	100	54	0.04012	125	63
200	1.00000	200	1	0.22458	125	14
202	0.22371	100	34	0.11967	125	26
205	1.00000	205	1	0.14654	125	24
206	0.20326	100	38	0.19859	125	18

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-E-2 Comparative CRS-SBM scores for 2006 with different measures of capital

Overall Efficiency						
MFMC	Using shareholders' equity			Using furniture and equipment		
	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.05102	100	101	0.01256	125	98
4	0.20237	103	41	0.09519	125	33
6	0.13330	100	63	0.04547	125	54
7	0.08920	100	82	0.01846	125	87
9	0.33738	100	22	0.08048	125	36
12	0.48454	14 103	14	0.24282	125	15
14	1.00000	14	1	1.00000	14	1
15	0.12844	103	66	0.04839	125	51
20	0.32461	103	25	0.08012	125	37
21	0.28615	103	31	0.10428	125	28
24	0.07796	100	88	0.01798	125	90
29	0.12851	100	65	0.02754	125	74
31	0.12455	100	69	0.02419	125	78
34	0.07962	100	87	0.02279	125	79
35	0.10131	100	78	0.02252	125	81
36	0.36283	100	19	0.15807	125	24
37	0.19572	100	43	0.04745	125	52
38	0.30177	100	28	0.09732	125	32
40	0.98297	100 103	9	0.29862	125	9
42	0.39169	100	17	0.18663	125	19
43	0.06698	100	93	0.01433	125	96
45	0.32857	100	23	0.07638	125	39
46	0.61736	100	11	0.26021	125	12
47	0.11071	100	72	0.02527	125	76
49	0.25735	100	36	0.10354	125	29
50	0.22569	100	38	0.12137	125	25
51	0.11605	100	71	0.03295	125	65
53	0.29695	100	30	0.17923	100	21
55	0.38660	103	18	0.10506	125	27
57	0.17904	100	47	0.03891	125	57
58	0.27552	103	34	1.00000	58	1
59	0.02049	125	102	0.01192	125	99
60	0.56232	100 125	13	0.54996	100 125	6
61	0.09254	100	80	0.02775	125	73
62	0.17945	100	46	0.04597	125	53
63	0.05501	100	97	0.01175	125	100
69	0.15998	100	51	0.08906	125	34
71	0.07392	100	90	0.01624	125	91
75	0.05768	100	96	0.01371	125	97
76	0.32722	100	24	0.25726	125	13
78	0.13623	100	60	0.03146	125	68
83	0.12697	100	68	0.02916	125	70
84	0.58563	103	12	0.19833	125	18
85	0.14745	103	55	0.04877	125	50
86	0.30006	100	29	0.07075	125	41
93	0.15319	100	54	0.04427	125	55
95	0.43432	100 103	16	0.10013	125	30
98	0.27062	100	35	0.06472	125	43
100	1.00000	100	1	1.00000	100	1
102	0.33771	100	21	0.44143	100	8
103	1.00000	103	1	0.46866	125	7
105	0.15480	100	53	0.17633	125	22
110	0.13768	100	59	0.03330	125	64
113	0.05241	100	99	0.01444	125	94
115	0.17871	100	48	0.08567	125	35
121	0.10587	103	75	0.02782	125	72
123	0.10904	100	74	0.03267	125	66
125	1.00000	125	1	1.00000	125	1
126	0.10139	100	77	0.03342	125	63
127	0.08865	100	84	0.02107	125	84
128	0.13552	103	61	0.03829	125	58
130	0.07611	100	89	0.01618	125	93
131	0.27757	100	32	0.16271	125	23

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132	0.13370	100	62	0.03155	125	67
133	0.25268	100	37	0.07176	125	40
137	0.30851	100	27	0.07655	125	38
139	0.48390	100	15	0.20724	125	17
140	0.08654	100	85	0.01843	125	88
142	0.08910	100	83	0.02212	125	83
152	0.34260	100	20	0.18371	125	20
154	0.05299	100	98	0.01063	125	101
156	0.10983	100	73	0.03368	125	62
159	0.09063	100	81	0.02972	125	69
160	0.18193	100	45	0.05281	125	48
161	0.09544	100	79	0.02213	125	82
162	0.16337	100	50	0.03661	125	60
163	0.05147	100	100	0.00971	125	102
164	1.00000	164	1	0.25196	125	14
168	0.16436	100	49	0.06011	125	45
173	0.07258	100	92	0.01809	125	89
174	0.31727	100	26	0.06407	125	44
176	0.07287	100	91	0.01619	125	92
177	0.12188	100	70	0.02655	125	75
182	0.10220	100	76	0.02256	125	80
185	0.13982	103	57	0.04944	125	49
190	0.08337	100	86	0.01897	125	86
191	0.15846	100	52	0.04188	125	56
192	0.12837	100	67	0.03725	125	59
193	0.06506	100	95	0.01441	125	95
194	0.18850	100	44	0.06658	125	42
195	0.21477	100	40	0.09957	125	31
196	0.85674	100 125 164	10	0.26524	125	11
197	0.13269	100	64	0.03607	125	61
198	0.13823	100	58	0.02490	125	77
200	1.00000	200	1	0.28714	125	10
202	0.22387	100	39	0.05817	125	46
203	1.00000	203	1	1.00000	203	1
204	0.14578	100	56	0.02887	125	71
205	1.00000	205	1	0.21208	100	16
206	0.19629	100	42	0.12050	125	26
207	0.06510	100	94	0.01944	125	85
210	0.27657	100	33	0.05779	125	47

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-E-3 Comparative CRS-SBM scores for 2007 with different measures of capital

Overall Efficiency						
Using shareholders' equity				Using furniture and equipment		
MFMC	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.04937	100	88	0.01199	125	90
4	0.10488	103	54	0.07647	125	23
6	0.12678	100	42	0.03340	125	46
7	0.06810	100	75	0.01625	125	81
9	0.25633	103	17	0.08144	125	20
12	0.24853	103	19	0.20661	103	9
14	0.29689	103	12	0.27076	125	6
15	0.05225	103	86	0.03515	125	45
20	0.21514	103	23	0.08143	125	21
21	0.11364	103	49	0.06098	125	29
24	0.05693	100	84	0.01379	125	87
29	0.09364	100	58	0.02440	125	63
31	0.08811	100	59	0.02157	125	67
34	0.07355	100	70	0.01884	125	74
35	0.06677	100	76	0.01659	125	80
36	0.34604	100	11	0.08785	125	17
37	0.11603	100	48	0.02817	125	54
38	0.21607	100	22	0.05455	125	32
40	0.24673	100	20	0.06247	125	27
43	0.08387	100	63	0.02174	125	66
45	0.26707	100	16	0.07078	125	24
46	0.54995	100	7	0.14776	125	12
47	0.10175	103	56	0.02613	125	61
49	0.18861	100	29	0.09966	125	15
50	0.19133	100	27	0.05790	125	30
51	0.10175	100	55	0.02543	125	62
53	0.22934	100	21	0.07039	125	25
55	0.17281	103	32	0.08568	125	18
57	0.16067	100	34	0.03802	125	38
58	0.20325	103	25	0.12627	125	14
59	0.02239	125	95	0.00926	125	94
61	0.06150	100	81	0.01545	125	85
62	0.13891	100	40	0.04330	125	36
63	0.02930	103	94	0.01271	125	88
69	0.12486	100	43	0.02845	125	52
71	0.14463	100	37	0.03718	125	40
75	0.05125	103	87	0.02011	125	68
76	0.21237	100	24	0.05328	125	33
78	0.08658	100	61	0.01943	125	70
83	0.08579	103	62	0.03078	125	48
84	0.40844	103	10	0.21293	125	7
85	0.07036	103	73	0.03688	125	41
86	0.13901	100	39	0.03606	125	43
93	0.15701	100	35	0.03957	125	37
95	0.27348	103	15	1.00000	95	1
98	0.27933	100	14	0.06841	125	26
100	1.00000	100	1	1.00000	100	1
102	1.00000	102	1	0.05561	125	31
103	1.00000	103	1	1.00000	103	1
105	0.28315	100	13	0.14629	125	13
110	0.12210	100	45	0.03074	125	49
113	0.04300	103	90	0.01127	125	91
115	0.14418	100	38	0.04481	125	35
121	0.08012	103	64	0.03084	125	47
125	1.00000	125	1	1.00000	125	1
126	0.07084	100	72	0.01814	125	77
127	0.07826	100	67	0.01932	125	72
128	0.05711	103	83	0.02771	125	57
130	0.06568	100	79	0.01591	125	83
131	0.18744	100	30	0.09602	125	16
132	0.10706	100	53	0.02830	125	53
133	0.20135	100	26	0.04851	125	34
137	0.25393	100	18	0.06135	125	28

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139	0.42997	100	9	0.18163	125	10
140	0.06866	100	74	0.01665	125	79
142	0.05337	103	85	0.01685	125	78
152	0.18678	100	31	0.07864	125	22
154	0.04190	100	91	0.01005	125	92
156	0.09471	100	57	0.02369	125	65
159	0.08677	100	60	0.02389	125	64
160	0.10742	100	52	0.02898	125	51
161	0.07592	100	68	0.01935	125	71
162	0.11256	100	50	0.02794	125	55
163	0.03624	100	93	0.00862	125	95
168	0.13868	100	41	0.08222	125	19
173	0.06195	100	80	0.01571	125	84
174	0.16349	100	33	0.03561	125	44
176	0.06578	100	78	0.01618	125	82
177	0.07277	100	71	0.01828	125	76
182	0.07833	100	65	0.01950	125	69
185	0.04871	103	89	0.02730	125	58
190	0.05825	100	82	0.01407	125	86
191	0.12380	100	44	0.02902	125	50
192	0.12072	100	46	0.03607	125	42
193	0.03958	100	92	0.00965	125	93
194	0.11228	100	51	0.02626	125	60
196	1.00000	196	1	0.17664	125	11
197	0.07830	100	66	0.01929	125	73
198	0.06653	100	77	0.01263	125	89
200	0.47176	103 125	8	0.21214	125	8
203	1.00000	203	1	1.00000	203	1
204	0.11611	100	47	0.02723	125	59
206	0.14614	100	36	0.03751	125	39
207	0.07377	100	69	0.01879	125	75
210	0.18937	100	28	0.02775	125	56

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-E-4 Comparative CRS-SBM scores for 2008 with different measures of capital

Overall Efficiency						
Using shareholders' equity				Using furniture and equipment		
MFMC	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.06936	100	71	0.01722	125	81
4	0.12356	103	45	0.06909	125	31
6	0.09404	46	59	0.03344	125	49
7	0.08422	100	65	0.02102	125	70
9	0.28667	46	13	0.09187	125	22
12	0.36512	103	10	0.16784	125	9
14	0.53273	46 103	7	0.21310	125	5
15	0.06101	46	82	0.01877	125	77
20	0.24916	46	23	0.08615	125	26
21	0.10910	103	52	0.03974	125	42
24	0.06890	100	72	0.01847	125	78
29	0.05422	100	86	0.01409	125	87
31	0.05397	100	87	0.01484	125	85
34	0.05629	46	84	0.01756	125	80
35	0.10524	100	56	0.02399	125	65
36	0.37877	46	9	0.12109	125	14
37	0.11621	100	48	0.03123	125	53
38	0.09978	46	58	0.03506	125	48
40	0.07030	46	70	0.02407	125	64
43	0.06407	100	79	0.01787	125	79
45	0.26292	100	19	0.19856	125	7
46	1.00000	46	1	0.36614	125	4
47	0.08838	46	62	0.02710	125	61
49	0.16023	100	35	0.10846	125	17
50	0.28164	100	17	0.08747	125	25
51	0.13157	100	43	0.03678	125	47
53	0.40558	100	8	0.10806	125	19
55	0.22664	46	26	0.07580	125	28
57	0.11501	100	49	0.03019	125	55
58	0.28277	103	16	0.11274	125	15
61	0.06024	46	83	0.02081	125	71
62	0.14484	100	38	0.03998	125	41
63	0.02524	46	91	0.00790	125	91
69	0.23610	100	25	0.06510	125	32
71	0.13102	46	44	0.03890	125	44
76	0.13805	100	41	0.15211	125	10
78	0.09133	100	61	0.02323	125	67
83	0.06880	46	73	0.02047	125	73
84	0.26264	46	20	0.09085	125	23
85	0.08376	46	66	0.02746	125	59
86	0.06671	46	77	0.01943	125	76
93	0.18065	46	31	0.06087	125	35
95	0.22117	46	27	0.06249	125	34
98	0.20722	46	29	0.06341	125	33
100	1.00000	100	1	1.00000	100	1
103	1.00000	103	1	1.00000	103	1
105	0.00449	46	93	0.00347	125	93
110	0.11121	46	51	0.03205	125	51
113	0.07830	46	67	0.02713	125	60
115	0.19021	100	30	0.11197	125	16
121	0.07646	46	68	0.02534	125	62
125	1.00000	125	1	1.00000	125	1
126	0.09270	100	60	0.02429	125	63
127	0.06516	100	78	0.12292	125	13
128	0.06187	46	81	0.01992	125	74
130	0.05462	100	85	0.01473	125	86
131	0.21404	100	28	0.12670	125	12
132	0.10632	46	55	0.03324	125	50
133	0.28295	100	15	0.07277	125	30
137	0.28322	100	14	0.07351	125	29
139	0.35653	46	11	0.19691	125	8
140	0.06238	100	80	0.01671	125	83
142	0.14721	46	37	0.04807	125	36

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152	0.24542	100	24	0.10526	125	20
154	0.05124	100	88	0.01323	125	89
156	0.13365	100	42	0.08957	125	24
159	0.10815	100	53	0.03058	125	54
160	0.16151	100	34	0.04409	125	38
161	0.07374	46	69	0.02077	125	72
162	0.10303	46	57	0.02931	125	57
163	0.05120	100	89	0.01256	125	90
168	0.16156	100	33	0.10839	125	18
173	0.04383	46	90	0.01347	125	88
174	0.25869	100	22	0.04636	125	37
176	0.08507	100	64	0.02256	125	69
177	0.10718	100	54	0.02757	125	58
182	0.06770	46	75	0.01975	125	75
185	0.01699	46	92	0.00561	125	92
190	0.06689	100	76	0.01716	125	82
191	0.17292	100	32	0.03918	125	43
192	0.12119	46	47	0.03743	125	45
193	0.06802	100	74	0.01493	125	84
194	0.08797	100	63	0.02352	125	66
196	0.56694	46	6	0.21053	125	6
197	0.13962	100	39	0.02973	125	56
198	0.13824	100	40	0.02285	125	68
200	0.34346	103	12	0.13114	125	11
203	0.26175	100	21	0.08139	125	27
204	0.11325	100	50	0.03138	125	52
206	0.15593	100	36	0.04029	125	40
207	0.12223	100	46	0.03733	125	46
210	0.26583	100	18	0.04244	125	39
217	1.00000	217	1	0.09226	125	21

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

Table I-E-5 Comparative CRS-SBM scores for 2009 with different measures of capital

Overall Efficiency						
Using shareholders' equity				Using furniture and equipment		
MFMC	SBM	Ref.	Rank	SBM	Ref.	Rank
2	0.06684	100	74	0.02964	125	83
4	0.07747	100	64	0.05435	125	57
6	0.09868	100	55	0.07137	125	46
7	0.08650	100	59	0.03606	125	78
9	0.23953	100	19	0.13264	125	23
12	0.00002	100	95	0.00002	125	95
14	0.24104	100	18	0.15410	125	19
15	0.05933	100	83	0.03893	125	70
20	0.23018	100	20	0.14711	125	21
21	0.11611	100	49	0.07845	125	40
24	0.06891	100	72	0.03643	125	77
29	0.07943	100	62	0.02987	125	82
31	0.07714	100	65	0.03996	125	69
34	0.08981	100	57	0.05520	125	56
35	0.10771	100	52	0.04227	125	65
36	0.31329	100	14	0.20167	125	14
37	0.11910	100	48	0.06094	125	53
38	0.06778	100	73	0.04835	125	61
40	0.29908	100	15	0.17298	125	16
43	0.07326	100	68	0.03872	125	71
45	1.00000	45	1	0.27930	125	7
46	0.60339	100	7	0.50343	100	3
47	0.08880	100	58	0.04872	125	60
49	0.14272	100	37	0.17812	125	15
50	0.26275	100	16	0.16986	125	17
51	0.13888	100	38	0.07283	125	45
53	0.35490	100	11	0.16249	100	18
55	0.12805	100	42	0.07477	125	42
57	0.18196	100	27	0.07367	125	43
58	0.16127	100	33	0.10611	125	30
61	0.00117	100	94	0.00077	125	94
62	0.17493	100	28	0.07133	125	47
63	0.01846	100	93	0.01111	125	93
69	0.32852	100	13	0.23460	100	10
71	0.12169	100	46	0.06453	125	51
76	0.12508	100	44	0.20537	125	13
78	0.14369	100	36	0.07108	125	48
83	0.04583	100	90	0.02682	125	87
84	0.17140	100	30	0.11114	125	29
85	0.07218	100	69	0.04625	125	62
86	0.06392	100	77	0.03720	125	73
93	0.19697	100	26	0.12886	125	24
95	0.06981	100	70	0.03573	125	79
98	0.09003	100	56	0.05198	125	59
100	1.00000	100	1	1.00000	100	1
103	1.00000	103	1	0.45490	125	4
105	0.04605	100	89	0.08273	125	36
110	0.07889	100	63	0.04400	125	64
113	0.04774	100	87	0.03168	125	80
115	0.17193	100	29	0.20953	125	12
121	0.06331	100	79	0.04149	125	66
125	1.00000	125	1	1.00000	125	1
126	0.12112	100	47	0.06260	125	52
127	0.06583	100	75	0.03676	125	76
128	0.02722	100	92	0.01534	125	92
130	0.05414	100	84	0.02593	125	88
131	0.83185	45 100 217	6	0.21834	125	11
132	0.06399	100	76	0.13910	125	22
133	0.25524	100	17	0.12715	125	25

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137	0.21064	100	23	0.10192	125	31
139	0.34998	100	12	0.29046	125	6
140	0.04534	100	91	0.02156	125	91
142	0.06106	100	82	0.04010	125	68
152	0.37313	100	10	0.24035	125	9
154	0.06351	100	78	0.03116	125	81
156	0.12481	100	45	0.06802	125	50
159	0.11241	100	50	0.05814	125	54
160	0.13288	100	41	0.07944	125	38
161	0.06268	100	80	0.03720	125	74
162	0.07697	100	66	0.03751	125	72
163	0.05326	100	85	0.02524	125	90
168	0.16093	100	34	0.08328	125	34
173	0.04681	100	88	0.02550	125	89
174	0.22081	100	22	0.07887	125	39
176	0.07694	100	67	0.04082	125	67
177	0.12704	100	43	0.07637	125	41
182	0.06908	100	71	0.03681	125	75
185	0.04822	100	86	0.02917	125	84
190	0.06186	100	81	0.02746	125	86
191	0.22495	100	21	0.10010	125	32
192	0.10456	100	54	0.09520	125	33
193	0.08055	100	61	0.02817	125	85
194	0.10909	100	51	0.04433	125	63
195	0.08512	100	60	0.11509	125	27
196	0.53158	100	8	0.38728	125	5
197	0.15332	100	35	0.05612	125	55
198	0.10700	100	53	0.05371	125	58
200	0.16554	100	32	0.11288	125	28
203	0.20074	100	25	0.11956	125	26
204	0.13595	100	39	0.07327	125	44
206	0.16866	100	31	0.08298	125	35
207	0.13347	100	40	0.07961	125	37
210	0.20425	100	24	0.06813	125	49
217	1.00000	217	1	0.15366	100	20
221	0.45852	100 103	9	0.26628	125	8

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). SBM corresponds to the efficiency score obtained by the Slacks-Based-Measure under CRS (Equation I.15). Ref. corresponds to the reference set for each company obtained by the SBM approach (Equation I.21). Rank represents the ranking of the SBM scores which have been sorted by value, where SBM scores with value of 1.00000 obtain the highest ranking 1.

**PART II: FURTHER EVALUATION OF
EFFICIENCY OF MUTUAL FUND
MANAGEMENT COMPANIES**

1 Introduction

In Part I, we proposed a new specific model suitable for the banking industry to evaluate for the first time the efficiency of MFMCs based on the unoriented methodology proposed by Holod and Lewis (2011), which aims to close the debate between production and asset approaches. We also applied different management stages to an MFMC based on the ideas by Berkowitz and Qiu (2003). Therefore, our specific model captures the interaction between the different requirements of the stakeholders of an MFMC and the diverse sub-decisions making units (DMUs) within a company to properly evaluate the efficiency of these financial institutions (see Figure I-5, Part I).

To appropriately evaluate the efficiency of the MFMC, it was necessary to develop adequate indicators to describe the main management activities of these firms. Previously published DEA banking and insurance literature were the sources of the variables selected and appropriately represent the inputs and outputs in our original model. In our first management stage (portfolio management), we defined inputs that captured the labour, capital and risk that are assumed in the management of mutual fund portfolios and outputs that represented the success of this management, such as the amount of assets, gross return records, diversification capacity and investment categories in the market. As outlined in our unoriented model, the results of this *Portfolio Management stage* should be appropriate indicators of the quality of the mutual funds that will be sold by the *Marketing and Service stage* of the company. Therefore, in our second management stage, we used intermediate inputs coming from outputs of the previous *Portfolio Management stage*, with the only difference being the nature of the returns (net return records). These inputs were considered to evaluate the success or failure of the marketing and distribution of the company, which was measured by the unitholder flows and the net money flows, as well as by the new management fees received. Finally, the *Overall Efficiency* of the MFMCs considered the final profits obtained by the company shareholders after the whole production process, i.e., after the *Portfolio Management* and Marketing stages of the mutual funds managed by the company.

The application of a non-oriented frontier approach using the slacks-based measure of the DEA model proposed by Tone (2001) was discussed in detail in Part I to justify this technique as an initial and appropriate tool to obtain the efficiency scores of the MFMCs. The consistency of the efficiency results obtained by using different measures of the most controversial variables included in our model proved the robustness of our findings. Although the returns-to-scale influence suggests that size seems to be an important variable to successfully sell mutual funds, we provide evidence that the best-managed funds are not the most-efficient sellers. This considerably reduces overall shareholder profits. This major finding of the study is consistent for all the years analysed in Part I of this thesis.

However, DEA and the following extensions of this method fail to identify the ‘best practice’ competitors that can serve as a benchmark for the companies analysed in light of the striking differences potentially found between the characteristics of the companies evaluated and those considered as ‘best practice’ competitors. This failure could limit the accuracy of some results in the literature. This problem is especially relevant in those industries where competitors show assorted characteristics, such as the Spanish fund industry. That is, the set of a DEA ‘best practice’ frontier formed by bank-owned large fund companies offering an assorted range of large funds might not be an appropriate benchmark for small-sized independent managers focused on a small number of funds, thereby offering misleading efficiency rankings of fund families.

In other words, the original SBM model evaluates the efficiency of the MFMCs, referring to the furthest frontier point within a range, but this evaluation approach may go to a remote point of the frontier and therefore to an inappropriate identification of the reference set, i.e., the ‘best practice’ benchmark for each MFMC analysed. This problem may be particularly important in the Spanish fund industry, where the selection of appropriate benchmarks is very important due to both the high concentration and assorted characteristics of the management companies.

In this part of the thesis, we overcome this limitation by using recent and unexplored variations of the original slacks-based measure (SBM) proposed by Tone (2010). These

new and innovative variations allow for the comparison of fund companies with their true ‘best practice’ competitors (benchmark) according to both management resources and objectives, thereby fitting fully to the assorted characteristics of the Spanish mutual funds industry. We will also analyse the effect of these variants on the efficiency rankings obtained by the original SBM as a first step to quantifying the bias because of the use of inappropriate reference sets of competitors. The application of these new SBM techniques will improve the accuracy of results and will complement the mere consideration of variable-returns-to-scale to evaluate the efficiency of DMUs with different scale features. These variations proposed by Tone (2010) are based on the hyperplanes rather than the vertices of the frontier, allowing the method to find more suitable facets with respect to the analysed DMU.

Finally, applying the clustering process originally proposed by Tone (2010), we achieve a refined evaluation of efficient companies focused only on those facets formed by competitors with similar characteristics. Our clustering proposal is based on the assets managed and labour effort by the companies because, as we found in Part I (Table I-3), a large number of small fund companies managing a residual market share and a reduced number of huge fund companies dominate the industry. Under these clustering criteria, we assume the hypothesis that fund companies with clustering-homogeneous size should have similar opportunities to reach efficiency at every management stage, thereby solving the potential scale effects in the distribution and marketing stage that were anticipated in Part I. The use of this clustering variation of Tone (2010) allows for the identification of locally efficient companies in relation to competitors with similar clustering characteristics.

Tone (2010) used different variations versus the original SBM (Tone, 2001) as a good approach to determine potential benchmark bias in this efficiency methodology. It is necessary to check in this Part II whether the patterns of efficiency obtained by the above techniques are persistent over time, i.e., if the efficiency results obtained by the different stages and companies correspond to stable patterns of the management process. Otherwise, if these efficiency results were subject to a considerable variability along time, that would question the conclusions obtained for specific dates and companies; that is, it would not be

possible to differentiate efficient management strategies from other temporary factors and results. This analysis should separate those companies that follow clearly efficient management patterns from other companies with much more erratic management results in relation to competitors. This persistence phenomenon has been extensively discussed in the literature on mutual funds but, as far as we know, never applied to fund management companies. To test this hypothesis, we will use a nonparametric approach based on clusters of efficiency instead of mere ranking quartiles or quintiles, allowing us to identify homogeneous groups of efficiency to provide greater reliability to our conclusions.

Finally, we aim to identify those major factors that seem mainly to drive the efficiency results of the fund management companies. This issue is particularly important in those financial markets with assorted competitors and characteristics, such as the Spanish fund industry. This analysis will be conducted based on the efficiency clusters designed in the previous persistence analysis, which will allow us to draw conclusions based on similar working groups and to further contrast those relevant factors in the efficiency scores over our time horizon.

This Part II unfolds as follows: Section 2 reviews the major concepts of the Variations to the original SBM approach and the empirical results of these variants in the Spanish mutual fund companies. Section 3 illustrates the persistence phenomenon in the efficiency scores and determines those relevant factors which may potentially drive the persistence results obtained by the companies along our horizon of study. Finally, Section 4 concludes and summarises the major findings of the study.

2 Variations of the Slacks-based measure (SBM) of efficiency

2.1 Basic concepts of the new measures

Part I has shown that there are many approaches to implement DEA methodology. Basic radial models such as CCR (Charnes et al., 1978) and BCC (Banker et al., 1984) adopt proportional changes of inputs or outputs and usually do not deal directly with the slacks for the calculation of efficiency scores. The additive DEA model (Charnes et al., 1985) has no scalar measure or ratio efficiency per se, but it can discriminate between efficient and inefficient DMUs by the existence of excesses in inputs and/or shortfalls in outputs (slacks). Tone (2001) states that this additive model has no means of gauging the depth of inefficiency in a form similar to radial efficiency scores and introduces a non-radial model, which addresses the slacks of each input and output individually and independently to integrate them into an efficiency measure (SBM). However, Tone (2010) states that the problem with this slacks-based measure is that it aims to minimise this score, and the appropriate referent point could be far from the DMU analysed.

In most DEA models, the production possibility set is a polyhedral convex set with vertices corresponding to the efficient DMUs found by the corresponding DEA method. Based on Simmonard (1966), Tone (2010) argues that a polyhedral convex set can be defined by its vertices or by its supporting hyperplanes (See Figure II-1, Figure II-2). Tone (2010) then proposes variants of the original SBM of efficiency (Tone, 2001), which are based on the hyperplanes instead of the vertices of the reference frontier. The first variation (Variation I) aims to obtain the minimum slacks-based measure point on the facet (supporting the hyperplane) that the SBM finds for the objective DMU – that is, to find the nearest referent point on the efficient frontier. This author extends this approach to consider all facets of the production possibility set (Variation II). Finally, there are two additional variants because the exhaustive enumeration of all facets required in Variation II might need huge computing resources: Variation III clusters all

facets based on common characteristics of their components, and Variation IV makes a random search of these facets.

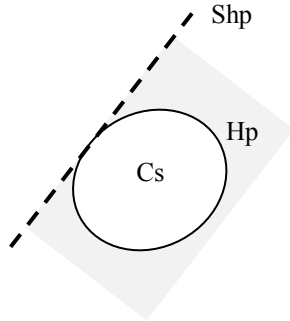


Figure II-1 Convex set

Cs = Convex set
 Shp = Supporting hyperplane
 Hp = the half-space delimited by the hyperplane that contains Cs

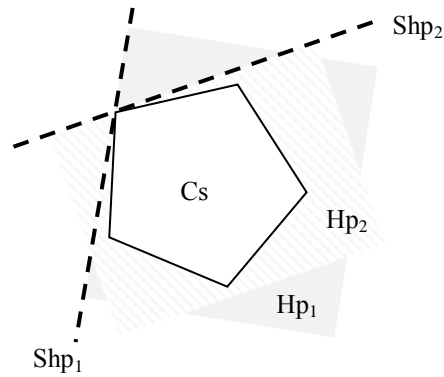


Figure II-2 Convex set with two Shp

Cs = Convex set
 Shp₁ = Supporting hyperplane 1
 Shp₂ = Supporting hyperplane 2
 Hp₁ = Hyperplane 1
 Hp₂ = Hyperplane 2

As mentioned above, a major contribution by Tone (2010) is that in his new proposal he considers supporting hyperplanes such as the facets of production possibility set P (see Figure II-1, Figure II-2). Let $P_0 = (\xi_j, \eta_j)$ ($j = 1, \dots, k$) be k DMUs in P . Tone (2010) makes a linear combination of these k DMUs with positive coefficients as

$$\begin{aligned} \xi_0 &= w_1 \xi_1 + \dots + w_k \xi_k, \\ \eta_0 &= w_1 \eta_1 + \dots + w_k \eta_k \\ \text{where } w_j &> 0 \quad (j = 1, \dots, k) \end{aligned} \tag{II.1}$$

Tone (2010) shows in his Theorem 2 that if (ξ_0, η_0) defined above is CRS-efficient, then (ξ_j, η_j) ($j = 1, \dots, k$) must be CRS-efficient, and in Theorem 3 that there exists a supporting hyperplane to P at (ξ_0, η_0) that also supports P at (ξ_j, η_j) ($j = 1, \dots, k$).⁶²

⁶² See section 2.3 of Tone (2010) for further details about the facets of the production possibility set.

As previously noted in Part I of this thesis, we consider a set of n DMUs, where each DMU $_j$ ($j = 1, 2, 3, \dots, n$) uses the same m inputs x_{ij} ($i = 1, 2, 3, \dots, m$), possibly in different positive amounts, and produces the same s outputs y_{rj} ($r = 1, 2, 3, \dots, s$), also possibly in different positive amounts. λ is a non-negative set of variables $(\lambda_1, \dots, \lambda_n)$ that represents the intensity vector;⁶³ s^- and s^+ are the non-negative sets of input excesses and output shortfalls, respectively.

Under the hypothesis of constant returns to scale, the production possibility set P was defined in Part I of this thesis as expression (I.7):

$$P = \left\{ (x, y) \mid x \geq \sum_j X_j \lambda_j, \quad y \leq \sum_j Y_j \lambda_j, \quad \lambda_j \geq 0 \right\} \quad (\text{I.7})$$

Let us review the SBM model (equation I.15) proposed by Tone (2001) to present further variations (Tone, 2010) to this original approach. In this model, an objective DMU (x_o, y_o) will be considered as efficient in terms of Pareto-Koopmans when it has no input excesses and no output shortfalls for any optimal solution, that is, when $p_o^{min} = 0$

$$\text{minimise} \quad p = \frac{1 - (1/m) \sum_{i=1}^m s_i^- / x_{io}}{1 + (1/s) \sum_{r=1}^s s_r^+ / y_{ro}}$$

s.t.

$$\sum_{j=1} x_j \lambda_j + s^- = x_o$$

$$\sum_{j=1} y_j \lambda_j - s^+ = y_o$$

$$\lambda_j, s_i^-, s_r^+ \geq 0 \quad (\text{I.15})$$

⁶³ We can impose some constraints on λ -weights, such as $\sum_j \lambda_j = 1$ (see expression (I.8), the BCC approach), to modify the constant-returns-to-scale production possibility set P .

The reference-set R_0 to the objective DMU being analysed in equation (I.15) was already presented in the expression (I.21) of Part I; it is defined as the set of DMUs corresponding to positive λ_j^*

$$R_0 = \{j | \lambda_j^* > 0, j = 1, \dots, n\} \quad (\text{I.21})$$

According to the following expressions (I.24) and (I.25), the objective DMU can be projected in terms of the reference-set R_0 , this being projection efficient (see Theorem 1 of Tone, 2010):

$$\bar{x}_0 = x_0 - s^{-*} = \sum_{j \in R_0} x_j \lambda_j^* \quad (\text{I.24})$$

$$\bar{y}_0 = y_0 + s^{+*} = \sum_{j \in R_0} y_j \lambda_j^* \quad (\text{I.25})$$

Tone (2010) states that the objective function expressed by the original SBM might project the objective DMU (\bar{x}_0, \bar{y}_0) onto a very remote point on the frontier because the basic SBM aims to find the worst efficiency score associated with the relatively maximum slacks under the constraints of the SBM model (equation I.15). These remote projections could sometimes be hard to interpret in terms of appropriate efficiency comparisons.

To overcome this limitation, as previously explained (see expression II.1), Tone (2010) explores facets of the production possibility set P to define the existence of a supporting hyperplane (Facet) to P which includes efficient linear combinations of the DMUs analysed.

For each inefficient DMU detected in the original SBM expression (I.15), the reference set R_0 is obtained according to expression (I.21), which includes only efficient DMUs (see Theorem 2 of Tone, 2010). Next, **SBM Variation I** looks for the nearest point on this reference set by minimising the slacks-based measure from the frontier.

Therefore, this variant modifies the basic SBM model, maximising the objective function rather than minimising it. That is, it evaluates the minimum slacks-based measure and hence the maximum score p_o^{max} on the efficient supporting hyperplane as follows:

$$\text{maximise} \quad p = \frac{1 - 1/m \sum_{i=1}^m s_i^- / x_{io}}{1 + 1/s \sum_{r=1}^s s_r^+ / y_{ro}}$$

s. t.

$$\sum_{j \in R_0} x_j \lambda_j + s^- = x_o$$

$$\sum_{j \in R_0} y_j \lambda_j - s^+ = y_o$$

$$\lambda_j, s_i^-, s_r^+ \geq 0 \quad (\text{II.2})$$

Therefore, Variation I requires only one easy-to-implement additional solution for each inefficient DMU detected in the original SBM. Because this variant works with the same facet as the original SBM model, the new scores will be at least similar to those obtained in the basic SBM:

$$p_o^{max} \geq p_o^{min} \quad (\text{II.3})$$

However, there might be other facets of production possibility set P apart from those defined by reference set R_o . All these facets should be considered to appropriately evaluate the efficiency of the objective DMU (x_o, y_o) . Tone (2010) proposes a method to enumerate all facets of P . First, this author defines that a subset of efficient DMUs in P is called *friends* if a linear combination of this subset is also efficient. Then, *maximal friends* are those *friends* when the result of any addition of an efficient DMU (not in *friends*) to *friends* is no more *friends*. Finally, a *friends* is dominated by other friends (*dominated friends*) if the set of efficient DMUs is a subset of others.

Let $P_j = (\xi_j, \eta_j)$ ($j = 1, \dots, K$) be the CRS-efficient DMUs in the production possibility set P , where K is the number of efficient DMUs. Tone (2010) defines the following algorithm to find the *maximal friends* facets.⁶⁴

<p>Begin</p> <p> For $k = 1$ to K</p> <p> Find <i>maximal friends</i> of P_k</p> <p> Next k</p> <p> Delete <i>dominated friends</i> from the set of <i>friends</i></p> <p> Obtain the set of facets from the final set of <i>friends</i></p> <p>End</p> <p> Subroutine Find <i>maximal friends</i> of P_k</p> <p> Excluded P_1, \dots, P_{k-1} from the candidates of <i>friends</i></p> <p> Enumerate all <i>friends</i> of P_k</p> <p> Remove <i>dominated friends</i> from the set of <i>friends</i></p> <p> Exit sub</p>	(II.4)
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SBM Variation II searches while minimising the SBM score obtained from all facets through three steps. First, this variant finds the set of efficient DMUs by solving the basic SBM expression (I.15) and finding the maximal friends of $P_j = (\xi_j, \eta_j)$ ($j = 1, \dots, K$). Then, this variant enumerates all facets and only selects those *maximal friends*. This second step requires the application of the aforementioned algorithm (II.4). Finally, for each inefficient DMU, Variation I is applied but only for those facets (h) selected in the previous step as *maximal friends*.

⁶⁴ Additional explanations of this algorithm can be found in section 4 of Tone (2010).

$$p_o^{(h)} = \max \frac{1 - 1/m \sum_{i=1}^m s_i^- / x_{io}}{1 + 1/s \sum_{r=1}^s s_r^+ / y_{ro}}$$

s.t.

$$x_o = \sum_{j \in R(h)} \xi_j \lambda_j + s^-$$

$$y_o = \sum_{j \in R(h)} \eta_j \lambda_j - s^+$$

$$\lambda_j, s_i^-, s_r^+ \geq 0 \quad (\text{II.5})$$

where $R(h)$ is the set of efficient DMUs that span each Facet (h) obtained in the second step of this variant. The efficiency score p_o^{all} of each objective DMU is obtained as the maximum p_o^h obtained for all *maximal friends* facets (h).

$$p_o^{all} = \max_h \{p_o^{(h)}\} \quad (\text{II.6})$$

Therefore, Tone (2010) finds the following inequalities among the three SBM scores:⁶⁵

$$p_o^{all} \geq p_o^{max} \geq p_o^{min} \quad (\text{II.7})$$

However, the enumeration of facets required by Variation II might need large computation resources for real and large-scale problems. To solve this potential problem, **SBM Variation III** modifies Variation II by using a clustering DMUs process. This variant again requires three steps. First, it is necessary to classify all DMUs in clusters. Second, this variant obtains the efficient DMUs according to the basic SBM model. Observe that this step is similar to the first step of Variation II. Therefore, this variant obtains efficiency scores $p_o^{(h)}$ (equation II.5) for each inefficient DMU with respect to

⁶⁵ See brief examples 1 and 2 in Tone's (2010) study for further details of these inequalities.

the *maximal friends* facets composed by efficient DMUs included in the same cluster that each inefficient DMU analysed.⁶⁶

If this model finds no feasible solution for the new within-cluster facets, the DMU analysed is considered to be efficient in its cluster, that is, globally inefficient but locally efficient in relation to the DMUs with common clustering characteristics. The contribution of this variation is quite relevant for homogeneous reference frontiers with regard to the characteristics of those DMUs analysed because it represents a more accurate analysis when working with heterogeneous industry samples.

According to Tone (2010), the merits of this modification are as follows:

- The enumeration of facets and the selection of *maximal friends* can be largely reduced by introducing a considerable number of clusters.
- In the case of the inefficient DMUs, the efficiency score is acquired in reference to the efficient DMUs in the same cluster. Thus, the results are more adequate and comprehensible because the DMUs are compared with competitors that show common clustering characteristics.

Tone (2010) proposed another modified version of Variation II that requires much less time and space in computing terms in the enumeration of facets. Thus, **SBM Variation IV** approximates a random search method for enumerating all facets of P . Based on the creation of random directions around efficient DMUs obtained from the basic SBM, this variant finds facets by repeating this random search until a sufficiently large number of facets is found. After that, the efficiency scores $p_o^{(h)}$ of Variation II are obtained for each inefficient DMU for those *maximal friends* facets randomly obtained.

⁶⁶ If none of the DMUs in the Cluster analysed is efficient, Tone (2010) proposes to pick up the efficient DMUs in the adjacent clusters to form the *maximal friends* facets.

The steps proposed by Tone (2010) are as follows (see the illustrative example in Figure II-3): (1) Let us use the set of K efficient DMUs $P_j = (\xi_j, \eta_j)$ ($j = 1, \dots, K$) to find the centre of gravity G for inputs and outputs.

$$X_G = (\xi_1 + \dots + \xi_K)/K, \quad (\text{II.8})$$

$$Y_G = (\eta_1 + \dots + \eta_K)/K \quad (\text{II.9})$$

(2) Create random directions around each efficient DMU, where for each efficient DMU included in set $P_j = (\xi_j, \eta_j)$, the direction from G to P_j is computed as $(\xi_j - X_G, \eta_j - Y_G)$; then, using random numbers, this direction is slightly disturbed (d_x, d_y) .

(3) Find a facet by solving the following linear program in $t \in R$ and $\lambda \in R^k$:

$$\begin{aligned} & \max t \\ & \text{s.t.} \\ & X_G + d_x t \geq \xi_1 \lambda_1 + \dots + \xi_k \lambda_k \\ & Y_G + d_y t \geq \eta_1 \lambda_1 + \dots + \eta_k \lambda_k \\ & t, \lambda \geq 0 \end{aligned} \quad (\text{II.10})$$

Let an optimal solution be (t^*, η^*) . Tone (2010) states that $t^* = 0$ implies that the centre G is efficient and all $P_j = (\xi_j, \eta_j)$ ($j = 1, \dots, K$) are *friends*. Otherwise, if $t^* > 0$, then the reference DMUs corresponding to positive λ_j^* s form a facet of P , because the optimal solution is obtained on a limit of P .

(4) Repeat the random search around the K efficient DMUs until a sufficient number of facets is found.

(5) Apply Variation II to determine the efficiency scores of the inefficient DMUs.⁶⁷

⁶⁷ Details of this random search procedure may be found in section 6.2 of Tone (2010).

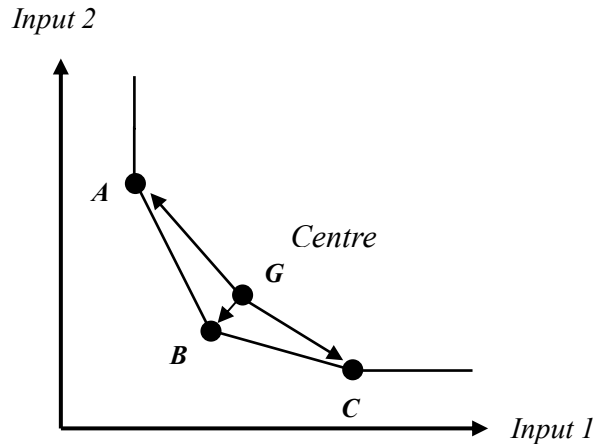


Figure II-3 Random search around efficient DMUs (Tone, 2010)

The rationale behind this random search is that the perturbed direction around vertices of the reference frontier allows for the probable finding of facets because several facets may be connected at the same vertex. Tone (2010) confirms the possibility of utilising any positive linear combination of efficient DMUs instead of the centre of gravity for Variation IV resolution.⁶⁸

To summarise, this section has reviewed the methodological variations proposed by Tone (2010) to find the nearest point on the efficiency frontier. First, the original SBM is modified to obtain the minimum slacks-based point on the supporting hyperplane (facet) that the SBM found for each DMU (Variation I). Then, this modification is extended to all facets to the production possibility set (Variation II). Finally, due to the massive enumeration of facets required by Variation II, Tone (2010) proposes two additional variations to save computing resources: the clustering process (Variation III) and the random exploration (Variation IV).

⁶⁸ Similarly to Part I, the empirical application of the SBM variations proposed by Tone (2010) has been run by the R project software because of adaptation of the R library package "nonparaeff" to the characteristics required by the new SBM variants. More details about this package are in Footnote 52 of Part I.

2.2 Comparative evaluation between the SBM original model and SBM Variation I

First, for each Stage considered in the conceptual model developed in Section 3 of Part I (see Figure I-5), we run the original SBM-efficiency model under constant-returns-to-scale (equation I.15). After that, we run SBM Variation I, thereby maximising the objective function for the reference set obtained for each SBM-inefficient company (equation II.2). The results obtained by our multi-management approach for the Stage 1 *Portfolio Management*, Stage 2 *Marketing and Service*, and Stage 3 *Overall Efficiency* are shown in Table II-F-1, Table II-F-2, Table II-F-3, Table II-F-4, and Table II-F-5, respectively (Appendix F).

According to Tone (2010), the scores are higher or equal in Variation I than in SBM ($p_o^{max} \geq p_o^{min}$); therefore, they show an improvement of the average efficiency score as a consequence of the new approach for every stage. Table II-1 shows this same finding in aggregate terms. Table II-1 shows that the Spearman rank correlations⁶⁹ for both SBM and Variation I are nearly equal to 1 for every stage and year analysed, which raises questions regarding the relevance of this variant in the empirical results of our study.

These rank correlation patterns are also quite visible when we plot the scores obtained by the original SBM and Variation I. In Figure II-G-1, Figure II-G-2, Figure II-G-3, Figure II-G-4, and Figure II-G-5 (Appendix G), the comparative plots appear to be around a straight line, almost perfect for *Marketing* and *Overall Efficiency stages*, which provides evidence of the high degree of similarity between the SBM and Variation I scores obtained in our sample. As in Table II-1, the portfolio management is the stage where graphs show more changes in Variation I scores with respect to those obtained by the original SBM. Nonetheless, the effect is not relevant in the new efficiency rankings as the Spearman rank correlation demonstrates by finding significant and positive correlation coefficients quite close to 1. On the other hand, both *Marketing stage* and *Overall Efficiency* show comparative plots that form almost perfect straight lines, which

⁶⁹ The conclusions provided by the Kendall rank correlation coefficients are quite similar. Detailed results are available upon request.

provide evidence of the residual effect of Variation I on the original SBM scores previously obtained.

Table II-1 Comparative findings between SBM and Variation I scores

Year	MMS	Average CRS-SBM Scores	Eff.	Average CRS-VAR I Scores	Eff.	Spearman Rank Corr.
2005	Stage 1	0.53359	28	0.61504	28	0.949**
	Stage 2	0.26553	7	0.26975	7	0.999**
	Overall Eff.	0.25697	10	0.26012	10	0.999**
2006	Stage 1	0.48751	22	0.55256	22	0.961**
	Stage 2	0.32233	8	0.33223	8	0.999**
	Overall Eff.	0.26854	8	0.27322	8	0.999**
2007	Stage 1	0.51105	21	0.57980	21	0.956**
	Stage 2	0.34870	6	0.36163	6	0.997**
	Overall Eff.	0.19564	6	0.19665	6	0.999**
2008	Stage 1	0.37089	13	0.41167	13	0.976**
	Stage 2	0.13860	2	0.13860	2	1.000**
	Overall Eff.	0.19781	5	0.19785	5	0.999**
2009	Stage 1	0.45114	20	0.53649	20	0.929**
	Stage 2	0.26990	7	0.27195	7	0.999**
	Overall Eff.	0.19333	5	0.19650	5	0.999**

MMS = Multi-Management Stages

Eff. = Number of Efficient Companies

** Correlation is significant at the 0.01 level (2-tailed)

Table II-1 supports all the major findings already obtained by the original SBM in Part I (Subsection 4.2). The new approach also highlights the fact that companies seem to be worse mutual fund sellers than portfolio managers, thereby reducing considerably the overall profits of the company shareholders. Similarly to previous SBM findings, the Variation I' scores also show a decreasing efficiency pattern in the Spanish fund industry. Table II-1 also reports that lower efficiency scores involve mostly higher rank correlations. For example, the *Marketing and Service stage* in 2008 shows the lowest average efficiency for the entire sample while the correlation is perfect and positive. On the other hand, in the *Portfolio Management stage* where we find the highest levels of efficiency, the Spearman rank correlation is slightly lower than the other results, but still quite significant.

Furthermore, Table II-2 supports evidence obtained by Variation I comparable to that found by the original SBM in Part I (see Table I-5), where the efficiency scores in the *Portfolio Management stage* are the highest for any year analysed. The rank correlation coefficient obtained by Variation I also provides a negative relationship between the *Portfolio Management* and the *Marketing and Service stage* of the companies from 2005 to 2009. We also find similar evidence but less significance between *Portfolio Management stage* and *Overall Efficiency* of the companies, and a strongly significant and positive rank relationship between the commercial skills and the overall efficiency for the entire time horizon. Table II-2 therefore provides robustness to the major conclusions in the original SBM results in terms of the relationship among the different stages of the model raised in Part I.

Table II-2 Average CRS-Variation I scores and Spearman's rank correlation

Year	MMS	Average CRS-VAR I Scores	Eff.	Spearman Rank Corr. (Scores)		
				Stage 1	Stage 2	Overall Eff.
2005	Stage 1	0.61504	28	1	-0.281**	-0.147
	Stage 2	0.26975	7		1	0.560**
	Overall Eff.	0.26012	10			1
2006	Stage 1	0.55256	22	1	-0.451**	-0.287**
	Stage 2	0.33223	8		1	0.419**
	Overall Eff.	0.27322	8			1
2007	Stage 1	0.57980	21	1	-0.254**	-0.187
	Stage 2	0.36163	6		1	0.359**
	Overall Eff.	0.19665	6			1
2008	Stage 1	0.41167	13	1	-0.643**	-0.220*
	Stage 2	0.13860	2		1	0.525**
	Overall Eff.	0.19785	5			1
2009	Stage 1	0.53649	20	1	-0.243*	-0.055
	Stage 2	0.27195	7		1	0.708**
	Overall Eff.	0.19650	5			1

MMS = Multi-Management Stages

Eff. = Number of Efficient Companies

* Correlation is significant at the 0.05 level (2-tailed)

** Correlation is significant at the 0.01 level (2-tailed)

2.3 Random searches for facets (Variation IV)

As previously discussed, Tone (2010) suggests two modified versions to compute SBM Variation II that reduce the massive enumeration of the required facets to obtain the efficiency scores, i.e., a random search for facets (Variation IV) and clustering facets (Variation III). This subsection details the empirical application of the random search approach to find the facets required for the Variation II efficiency scores. After that, the Variation II efficiency scores will be obtained and compared with the conclusions drawn by the original SBM rankings in Part I of this thesis for the different years and management stages analysed.

According to Tone (2010), the SBM Variation II could be directly applied (equation II.5) if we found a few efficient MFMCs, which would not require huge computation resources to obtain the *maximal friends* solution of the algorithm (II.4). By contrast, the application of the random search Variation IV approach would be necessary if we found a large number of efficient MFMCs and the *maximal friends* facets were difficult to obtain in terms of computing resources.

The SBM efficiency results obtained in our model (Table II-1) show an important number of efficient companies for Stage 1, which implies a massive enumeration of facets in that stage to obtain the *maximal friends* required to apply Variation II. For instance, consider the *Portfolio Management stage* in 2005, where we have $m = 3$ inputs, $s = 4$ outputs, and $k = 28$ efficient companies, which would suppose a total enumeration of 268,435,455 facets.⁷⁰ This problem of the massive enumeration of facets does not appear to be important for the *Marketing* and *Overall Efficiency stages*, which provide far fewer efficient companies. Nevertheless, we apply random search-Variation IV to the whole model to provide uniformity to our methodology and results, although the enumeration of facets is not a computing resources problem in both *Marketing* and *Overall Efficiency stages*. In any case, the facets obtained are similar.

⁷⁰ In the worst case, we should enumerate ${}_{28}C_{14} = 40,116,600$ combinations of 14 MFMCs. Logically, most of these are not efficient cases.

Therefore, the creation of random directions around the centre of gravity of each efficient company allows for the enumeration of facets using much less computing resources and time, this being an issue especially relevant for the *Portfolio Management stage* (Stage 1) of our mutual fund company model.

In our case, we conduct 50,000 simulations inside $\pm 40\%$ bounds around the gravity centre for each SBM-efficient company obtained in the different stages of our model.⁷¹ Therefore, we identify those *maximal friends* facets based on the *friends* obtained in the above-mentioned simulations. Table II-3 details the number of *friends* and *maximal friends* found in our simulations for each different stage and year analysed (see Appendix H where the *maximal friends* facets are detailed for each period). Although the total number of simulations amounts to 8,400,000, the advantage in terms of computing is huge, considering that, just for Stage 1 in 2005, we would have worked with more than 268 million possible combinations.

⁷¹ To test the robustness of this random search, we further compute 25,000 simulations for each efficient company with random limits to 20% and 30%, respectively. Detailed results are available upon request.

Table II-3 Friends and Maximal Friends (Variation IV)

Year	MMS	Friends	Maximal friends	# simulations (perturbed direction $\pm 40\%$)
2	Stage 1	391	55	1,400,000
0	Stage 2	16	6	350,000
5	Overall Eff.	27	8	500,000
2	Stage 1	436	81	1,100,000
0	Stage 2	33	4	400,000
6	Overall Eff.	23	6	400,000
2	Stage 1	323	40	1,050,000
0	Stage 2	11	4	300,000
7	Overall Eff.	14	5	300,000
2	Stage 1	131	15	650,000
0	Stage 2	4	1	100,000
8	Overall Eff.	12	3	250,000
2	Stage 1	280	47	1,000,000
0	Stage 2	18	4	350,000
9	Overall Eff.	13	2	250,000

MMS = Multi-Management Stages

Once we have identified the *maximal friends* facets randomly obtained, we apply Variation II (equation II.5) to obtain the efficiency scores $p_o^{(h)}$ for each inefficient MFMC. According to Tone (2010), the new Variation II results provided in Appendix I confirm the inequalities (II.7) among the three SBM scores, i.e., $p_o^{all} \geq p_o^{max} \geq p_o^{min}$, and therefore display an additional improvement of the average efficiency score in the new approach for every stage (see Table II-4).

As with the previous comparison between the original SBM and Variation I scores (Table II-1), we also checked for Spearman rank correlation between both efficiency rankings, and again we found evidence of high and significantly positive correlations. This result raises questions regarding the relevance of this modification in practical terms for our study. However, the correlation between Variation I and Variation II scores is still high but lower than that existing between SBM and Variation I. This is an expected result given that the modification proposed by Tone (2010) in Variation II is

greater than in Variation I. Finally, although these results call into question the relevance of Variation II in relation to the conclusions obtained by the original SBM in terms of our study, it seems that Variation II provides the most reliable scores according to Tone (2010).

Table II-4 Comparative findings between Variation I and Variation II scores

Year	MMS	Average CRS-VAR I Scores	Eff.	Average CRS-VAR II Scores	Eff.	Spearman Rank Corr.
2005	Stage 1	0.61504	28	0.76477	28	0.930**
	Stage 2	0.26975	7	0.45864	7	0.959**
	Overall Eff.	0.26012	10	0.60882	10	0.853**
2006	Stage 1	0.55256	22	0.71020	22	0.942**
	Stage 2	0.33223	8	0.53522	8	0.913**
	Overall Eff.	0.27322	8	0.55799	8	0.887**
2007	Stage 1	0.57980	21	0.71353	21	0.945**
	Stage 2	0.36163	6	0.61103	6	0.778**
	Overall Eff.	0.19665	6	0.46422	6	0.899**
2008	Stage 1	0.41167	13	0.57478	13	0.902**
	Stage 2	0.13860	2	0.25275	2	0.973**
	Overall Eff.	0.19785	5	0.42443	5	0.877**
2009	Stage 1	0.53649	20	0.73151	20	0.851**
	Stage 2	0.27195	7	0.66851	7	0.822**
	Overall Eff.	0.19650	5	0.49584	5	0.911**

MMS = Multi-Management Stages

Eff. = Number of Efficient Companies

Average CRS-VAR II Scores are based on *maximal friends* facets randomly obtained by Variation IV

** Correlation is significant at the 0.01 level (2-tailed)

Finally, Table II-5 also shows evidence similar to that found both in Part I (Table I-5) and in the previous subsection (Table II-2), where the highest efficiency scores are obtained by the *Portfolio Management stage*. This stage proves to be quite independent of the efficiency of the marketing and distribution of the mutual funds, and therefore of the overall efficiency of the company. Moreover, Table II-5 shows the positive and significant relationship between the marketing and distribution skills of the company and overall efficiency for the whole time horizon. That is, the most refined measure

proposed by Tone (2010) confirms the major findings obtained in Part I by using the original SBM approach. This result verifies the robustness of our main conclusion, such as the apparent dichotomy between the efficiency of portfolio management and the distribution and marketing skills in the Spanish mutual fund industry.

Table II-5 Average CRS-Variation II scores and Spearman's rank correlation

Year	MMS	Average CRS-VAR II Scores	Eff.	Spearman Rank Corr. (Scores)		
				Stage 1	Stage 2	Overall Eff.
2	Stage 1	0.76477	28	1	-0.294**	-0.089
0	Stage 2	0.45864	7		1	0.467**
0	Overall Eff.	0.60882	10			1
5	Stage 1	0.71020	22	1	-0.470**	-0.316**
0	Stage 2	0.53522	8		1	0.499**
0	Overall Eff.	0.55799	8			1
6	Stage 1	0.71353	21	1	-0.121	-0.156
0	Stage 2	0.61103	6		1	0.426**
0	Overall Eff.	0.46422	6			1
7	Stage 1	0.57478	13	1	-0.659**	-0.424**
0	Stage 2	0.25275	2		1	0.603**
0	Overall Eff.	0.42443	5			1
8	Stage 1	0.73151	20	1	-0.049	0.051
0	Stage 2	0.66851	7		1	0.601**
0	Overall Eff.	0.49584	5			1
9						

MMS = Multi-Management Stages

Eff. = Number of Efficient Companies

Average CRS-VAR II Scores are based on *maximal friends* facets randomly obtained by Variation IV

** Correlation is significant at the 0.01 level (2-tailed)

2.4 The search for locally efficient companies applying Variation III

As in the previous section, Variation III allows for a refined evaluation of efficient companies because the target company is going to be referred to the best practice frontier formed by fund companies with more homogeneous characteristics than the evaluated fund company. That is, this variant compares competitors with quite similar resources and therefore opportunities to gain efficiency.

Variation III is in fact a refinement of SBM Variation II (equation II.5) proposed by Tone (2010), which enumerates all the facets of the efficient frontier (*maximal friends*). However, Variation III will only enumerate those facets formed by companies with similar characteristics based on specific clustering standards. Therefore, the scores obtained by this variant may be extremely sensitive to the clustering process, which highlights the relevance of a proper clustering methodology.

A cluster process consists of finding groups in data. This idea has been applied in many areas including astronomy, archaeology, medicine, chemistry, education, psychology, linguistics and sociology. The objective of our study is to set groups in such a way that MFMCs in the same group have similar management resources. The need to classify cases in line with the objectivity standards of modern science has given rise to automatic and sophisticated classification procedures. The present study is based on the well-known ideas of Kaufman and Rousseeuw (2005).

Several clustering techniques may be applicable, and a priori arguments may not suffice to narrow down the choice of a single method. According to Kaufman and Rousseeuw (2005), it is necessary to run more than one method and to carefully analyse and compare the resulting classifications, making use of their graphical displays and other statistical measures, i.e., review what the data are trying to tell us before selecting the clustering technique. Studies in other scientific fields (e.g., Ahlquist and Breuning, 2012) have already argued that a priori generation of equal-size clusters may be a serious shortcoming. Therefore, it is necessary to identify the main data attributes as a prerequisite step in clustering analysis because the lack of this previous work may lead to important limitations of the clustering results.

We can find two major types of clustering techniques, depending on both the type of data available and the particular purpose of the research: 1) partitioning and 2) hierarchical techniques. A brief explanation of both techniques is necessary to determine which best fits our analysis.

A partitioning technique constructs K clusters (groups) that together satisfy the requirements of a partition:

Each group must contain at least one object;

Each object must belong to exactly one group

These conditions imply that there are equal or fewer clusters K than there are objects n :

$$K \leq n \quad (\text{II.11})$$

According to the clustering standards, two different clusters cannot have any object in common, and the K groups together must add up to the full dataset.

Partitioning techniques are applied if we want to classify the objects into K clusters, where K is fixed. The algorithms used here aim to find a “good” partition in the sense that objects of the same cluster should be very close or related to each other. The purpose is to uncover a structure already present in the data. Let us briefly examine each technique:

The PAM technique (Partitioning Around Medoids) can be applied when the data are a dissimilarity matrix, i.e., objects within a cluster show a high degree of similarity, while objects belonging to different clusters are as dissimilar as possible. This technique is an improvement of the well-known K -means clustering process where each observation belongs to the cluster with the nearest mean, serving as a prototype of the cluster. In other words, PAM is based on the search for K representative objects among the objects of the dataset. The algorithm used by PAM represents objects that are often called centrotypes or medoids of the clusters. The K -medoid method tries to find “spherical” clusters, that is, clusters that have roughly ball-shaped forms. To be accurate, the average distance or dissimilarity (using K -medoid technique) of the representative object (K) to all the other objects of the same cluster is minimised to obtain accurate “spherical” clusters. The PAM technique is especially recommended if a researcher is

also interested in the representative objects themselves, which may be very useful for data reduction or characterisation purposes.

An extension of PAM is the technique called CLARA (Clustering Large Applications), which was developed to analyse large datasets. Its clustering objective is the same as in the above-mentioned K -medoid. That is, CLARA tries to find K representative objects that are centrally located in the clusters previously determined.

Finally, fuzzy analysis may be considered in these partitioning techniques. It can be applied to the same datasets as the PAM approach, but the nature of its algorithm is different. For example, instead of confirming that “object n belongs to cluster 1”, fuzzy analysis could state that “object n belongs for 90% to cluster 1, for 5% to cluster 2, and for 5% to cluster 3”. This implies that this object is mostly assigned to cluster 1, but that there is still a glimpse of doubt in favour of clusters 2 and 3. Thus, intermediate objects that are not clearly assigned can be described by a means called membership coefficients (Kaufman and Rousseeuw, 2005).⁷²

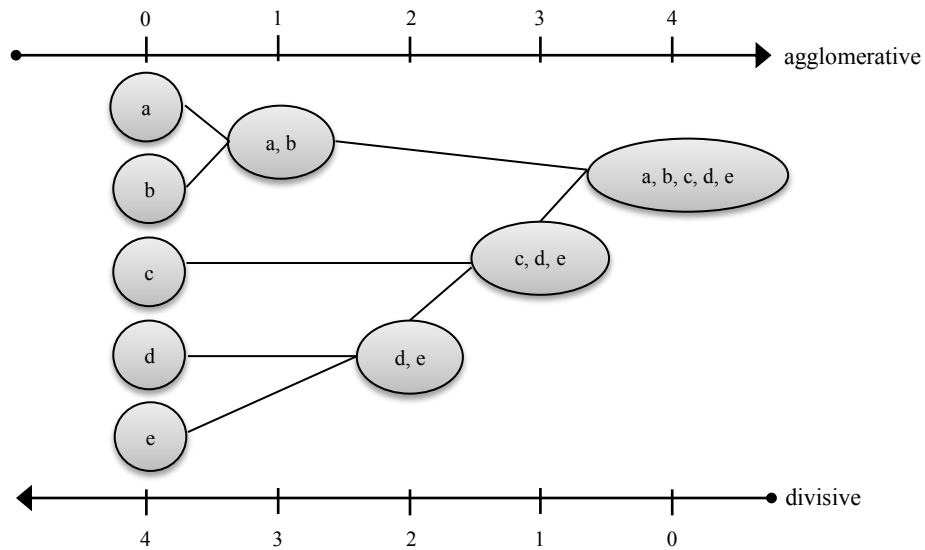
On the other hand, hierarchical techniques do not construct a single partition with K clusters, as they address all values of K in the same run. That is, in the partition with $K = n$, each object forms a separate cluster with only a single element. All values of $K = 2, 3, \dots, n - 1$ are covered in a type of gradual transition. The only difference between $K = r$ and $K = r + 1$ is that one (two) of the K clusters splits up (combine) to obtain $r + 1$ (r) clusters. A hierarchical technique suffers from the defect that it can never repair what was done in previous steps, i.e., once a hierarchical technique has joined/split up two or more objects, the clustering process can no longer be reversed.

There are two types of hierarchical techniques: agglomerative and divisive. They construct their hierarchies in opposite directions, potentially yielding different results. The agglomerative approach starts when all objects are separate, and the divisive

⁷² The PAM, CLARA and FANNY partitioning techniques are available in the Cluster package (Version 1.14.4, August 2013) available in the R project library (URL <http://www.r-project.org>). Additional explanations of these partitioning techniques are included in Kaufman and Rousseeuw (2005).

techniques start when all objects are together. Figure II-4 shows an example process with a dataset of $n = 5$ objects. In the upper arrow, the agglomerative technique starts when all objects are separate; then in the first step, two clusters are merged, and so on until only one cluster is left. Otherwise, the divisive technique starts with all objects together in one group; in each following step, a cluster is split up until there are n clusters. The agglomerative and divisive hierarchies coincide in the illustration of Figure II-4, but they usually are different.

Figure II-4 Distinction between agglomerative and divisive techniques



Note: The scheme is a replication of Figure 11 from Kaufman and Rousseeuw (2005, p.45)

The agglomerative nesting technique (AGNES) accepts the same data as the aforementioned partitioning techniques (Partitioning around medoids, and Fuzzy Analysis). However, the number of clusters is no longer requested because of the hierarchical nature of this agglomerative technique. Therefore, this approach executes a complete sequence of cluster combinations. In a first step, those closest objects are joined in r clusters, and in all succeeding steps, the r closest clusters are merged. All these mergers are based on the dissimilarities among their objects. There exist different

agglomerative nesting algorithms that differ only in their definition of among-cluster dissimilarity.⁷³

On the other hand, the divisive approach of the hierarchical clustering methods is covered by the DIANA technique.⁷⁴ The qualitative advantage of this divisive analysis consists of the fact that most users are interested in the major structure of their data, which is drawn by a few large clusters in the first steps, rather than in a detailed description of the individual objects. According to Kaufman and Rousseeuw (2005, p.49), the agglomerative techniques start with the details and then works its way up to large clusters, which may however be affected by unfortunate clustering decisions in the first steps of the process. Agglomerative techniques suffer from the defect that once two or more objects have been joined in a new cluster, they can no longer be separated. Instead, the divisive techniques start with the main large clusters. That is, the first step consists of splitting up all the objects included in the dataset into two major groups, and then the divisive method goes on to divide them further into smaller clusters. Based on observations made by Kaufman and Rousseeuw (2005), the large clusters are determined first and thereby are less likely to suffer from earlier steps.

Therefore, it is often useful to compare the clusters provided by the agglomerative nesting technique (AGNES) with those obtained by the divisive analysis (DIANA) to identify potential clustering bias resulting from the different natures of the two hierarchical methods. Such a comparison should be based on the results of both techniques, which are displayed in a similar banner (dendrogram⁷⁵).

⁷³ AGNES technique allows calculating dissimilarities between observations by using sum-of-squares of differences (Euclidean distance), and sum of absolute differences (Manhattan distance). This technique also allows the definition of clustering rules to group a set of objects, such as the unweighted pair-group average, single linkage (nearest neighbour method), complete linkage (furthest neighbour method), ward's method, weighted average linkage, and flexible method.

⁷⁴ The AGNES and DIANA techniques are available in the Cluster package (Version 1.14.4, August 2013) available in the R project library (URL <http://www.r-project.org>). Additional explanations of these hierarchical techniques are included in Kaufman and Rousseeuw (2005).

⁷⁵ A dendrogram is the graphical representation of the clusters produced by a hierarchical clustering process. The bottom row usually represents individual objects displaying a 0 similarity. The distances among a merged group increases with the level of the merger; i.e., the height of each cluster displayed in the dendrogram is proportional to the value of the dissimilarity between the clusters at that height. In past years, dendrograms were frequently produced with printers without graphic capability; therefore, the lines were obtained by

Our clustering proposal is based on two dimensions of the sample: the assets under management and the number of employees of each fund company. Either variable may be considered an appropriate proxy for the management resources of a mutual fund company to determine the most suitable number of clusters for our study. As we have previously discussed along these lines, the Spanish fund industry is extremely concentrated, thereby generating a competition map wherein a large number of small companies manage a residual market share and a reduced number of huge companies dominate the industry. Because of this concentration, opportunities and resources to gain efficiency seem to be quite different for both groups. In this sense, fund companies will employ mainly skilled labour, which is a major determinant of the production function of this industry.

Consequently, we assume the hypothesis that fund companies with both clustering-homogeneous assets under management and labour size should have similar opportunities to reach efficiency at every management stage proposed in our conceptual model (Figure I-5). Therefore, our intention is to set clusters formed by fund companies with similar typologies based on management resources, which are determined by both assets and labour variables.

An a priori determination of Z clusters using partitioning methods could provide an inappropriate number of clusters. According to examples provided by Kaufman and Rousseeuw (2005, chapter 6), divisive techniques (DIANA) obtain very similar results to those yielded by partitioning methods (PAM and FANNY) and agglomerative nesting techniques (AGNES). Apparently, DIANA is sufficiently clear-cut to uncover “right” clusters in the first steps of the process.⁷⁶ According to the usual standards in the clustering process (Kaufman and Rousseeuw (2005, p.37), developing hierarchical techniques and comparing the resulting classifications is recommended, making use of the graphical displays before forcing the number of clusters. That is, our clustering

means of the usual alphabetic characters. Currently, dendrograms are much better developed to provide more illustrative displays of hierarchical results (see further details in Forina et al., 2002).

⁷⁶ It obtains a similar picture using AGNES, although with a reversed process.

objective is not to force Z clusters, but rather that the clusters obtained are well made and reasonable.⁷⁷

Our clustering procedure starts by running both agglomerative nesting (AGNES) and divisive (DIANA) techniques to compare results and thereby identify the most appropriate Z clusters. As previously discussed, the divisive analysis run by the DIANA technique seems to be more useful in our concentrated industry because it starts by finding a few large clusters rather than with an exhaustive description of the individual companies. In our case, the sample presents two well-identified major groups: the largest companies and the remaining competitors in the fund industry. According to Kaufman and Rousseeuw (2005), these first clusters are less likely to suffer from earlier wrong decisions than agglomerative techniques. That is, DIANA primarily splits our sample into two major groups and then goes on to divide them further into smaller parts. In other words, the largest and best-identified clusters are found first, which makes it less susceptible to being affected in the initial steps.⁷⁸ Despite the above statement, we also run both hierarchical techniques to compare the dendrograms and to analyse the robustness of the clusters obtained. The MFMCs sorted by both clustering AGNES (Agglomerative Nesting) and DIANA (Divisive Analysis) show similar classifications in approximately 90% of cases, which provides robustness to the results provided by both hierarchical techniques.⁷⁹

⁷⁷ In the past, divisive clustering techniques were not generally available and rarely were applied due to huge computation time requirements for the first steps. Because of the good results provided by Mcnaughton-Smith et al. (1964), divisive clustering techniques have become more accessible. More recently, Bradley et al. (1998) presented a fast, scalable and single-pass version of K-means with fewer data requirements; Scholkopf et al. (1998) proposed a Kernel K-means to detect arbitrarily-shaped clusters with an appropriate choice of kernel similarity function; and Steinbach et al. (2000) proposed a hierarchical divisive version of K-means, called bisecting K-means, which recursively partitions the data into two clusters at each step.

⁷⁸ To divide the selected cluster, the algorithm in DIANA first looks for its most disparate observation (i.e., which has the largest average dissimilarity to the other observations of the selected cluster). This observation initiates the "splinter group". In subsequent steps, the algorithm reassigns observations that are closer to the "splinter group" than to the "old party". The result is a division of the selected cluster into two new clusters.

⁷⁹ AGNES technique was applied by using a Manhattan metric and the unweighted pair-group average method (by default in the Cluster package of the R program). Using the unweighted average linkage method (see Figure II-4), when two clusters a and b are joined into new cluster (a, b) , the position of (a, b) is between the positions of a and b , weighted for the number of objects in the two joined clusters (roughly ball-shaped). In evaluating the position of (a, b) , all the original objects have the same weight (this is the reason for the name "unweighted"). DIANA technique use the inverse approach; it starts when all objects are together by splitting up and forming approximately similar ball-shaped clusters with the objects, such as the unweighted average method implemented by AGNES which allows their banners to be highly comparable.

The aim of Variation III proposed by Tone (2010) is to solve the problem of massive enumeration of facets potentially present in Variation II because of the combination of several efficient companies. Therefore, if we consider only a few clusters, this analysis would make little sense to solve the aforementioned problem because a relevant number of efficient companies might still be present in the same cluster, and the decrease of the enumerated facets could be residual. On the other hand, if we consider many clusters, we could find many locally efficient MFMCs in our analysis, thereby distorting the appropriate identification of locally efficient companies because of mere excessive clustering (see subsection 2.1 for details).

To address this problematic issue, choice of the number of clusters is based on the dendrograms found by the divisive hierarchical approach included in the divisive technique (DIANA).⁸⁰ After drawing a line at the same height (approximately 5) for each dendrogram and period, we could generally identify four major clusters for each year based on the mix of assets and labour variables. That is, we usually identified four major groups at the same proportional value of the dissimilarity between the clusters. The presence of this stable clustering pattern along our time horizon seems to provide robustness to our clustering proposal.⁸¹ Finally, we named each cluster according to the characteristics found in the group.

However, MFMCs in 2005 and 2006 were divided into additional clusters for an appropriate search for locally efficient companies.⁸² Five clusters were initially obtained in 2005 at the same dendrogram's height = 5. However, the number of Stage1-efficient

⁸⁰ Dendrograms are shown in Appendix J. To calculate dissimilarities among observations, we used the sum of absolute differences; i.e., we applied the “Manhattan metric” rather than the sum-of-squares of differences “Euclidean distances”. Roughly speaking, the latter is similar to considering a straight line between the objects, whereas Manhattan distance defines proximity like a city block distance. In our case, we obtained quite similar cluster results for both measures (more than 97% of cases). Details about the DIANA program are available in Cluster package (Version 1.14.4, August 2013) available in the R project library (URL <http://www.r-project.org>).

⁸¹ Cluster 2 (corresponding to large firms) seems to be divided into two sub-clusters, but nevertheless we decided to keep it as a whole due to its small number of components (generally five companies) to avoid any bias in the search for locally efficient companies as a consequence of few components in the cluster.

⁸² This additional sub-division in 2005 and 2006 was only considered for stage 1 due to the incidence of the high number of efficient companies in the original clusters, which involved computing problems to enumerate all the facets required by Variation II.

companies included in Cluster #5 was still quite important, thereby involving a problem of computing resources to enumerate all the required facets to run Variation III. This is the reason for sub-dividing Cluster #5 in the subsequent three dendrogram branches. This additional split solved the massive enumeration problem of the facets required by Variation III without forcing the appropriate clusters drawn by the 2005 dendrogram (see details in Appendix J).⁸³

Four clusters were initially obtained in 2006 at the same dendrogram height. In 2005, we found the same problem: a large number of Stage1-efficient companies present in a cluster enumerated all the facets required to search for locally efficient companies. In this case, we split up Cluster #4 into the following two branches to solve the massive enumeration of facets (see details in Appendix J).

For the remaining years (2007 to 2009), the number of efficient companies was low enough to enumerate all the facets required by Variation III without computing resource problems. Therefore, no additional division was required of the clusters initially obtained at height 5 by the dendrograms in that period. The following Table II-6 shows the entire cluster classification.

Table II-6 Cluster classification by year

2005	2006	2007 to 2009
Cluster of the largest firms (C1)	Cluster of the largest firms (C1)	Cluster of the largest firms (C1)
Cluster of large firms (C2)	Cluster of large firms (C2)	Cluster of large firms (C2)
Cluster of large-midsize firms (C3)	Cluster of midsize firms (C3)	Cluster of midsize firms (C3)
Cluster of small-midsize firms (C4)		
Cluster of large-small sized firms (C5.1)	Cluster of large-small sized firms (C4.1)	Cluster of small firms (C4)
Cluster of mid-small sized firms (C5.2)		
Cluster of small-small sized firms (C5.3)	Cluster of small-small sized firms (C4.2)	

⁸³ The dendrograms displayed in Appendix J show the hierarchy of clusters, which is quite similar to a tree. The banner plots the diameter of each cluster being split. The observations are listed in the order found by the DIANA algorithm, and the numbers in the height vector are represented as bars between the observations. The leaves of the clustering tree are the original observations. A branch splits up at the diameter of the cluster being split.

The overall findings after applying the divisive clustering procedure previously described (without considering additional sub-clusters) show that Cluster #1 includes the two largest fund companies within the sample in 2005, 2006 and 2007, and the three largest companies in 2008 and 2009. Cluster #2 includes five large companies within the sample in 2005 and 2006, and four large companies in 2007, 2008 and 2009. Cluster #3 of midsize companies (C3 and C4 were joined in 2005) includes a range from 14 to 19 companies for our whole time horizon. Finally, to differentiate the extremely small companies existing in Spain, Cluster 4 (named C5.1, C5.2, C5.3 in 2005) integrates a range of from 69 to 76 companies in all the years analysed.

Table II-7 shows some descriptive statistics of these clusters, thereby highlighting the assorted characteristics of these major groups by their main management variables. The data sample figures show the extreme concentration of the Spanish fund industry. Note both average assets and unitholders, where the proportional difference between the largest and the smallest companies is more than 100 to 1, although this difference is lower after 2008. We also find mean differences of approximately 20 to 1 in variables such as shareholder equity and the number of funds managed by the companies. In the case of profits, the proportional difference is approximately 60 to 1; in number of employees, the proportion is approximately 12 to 1.

Under our clustering hypothesis, the extreme differences between the assets managed by each cluster included in Table II-7 should confirm the different resources available to each company to gain efficiency in the three stages considered in our model (see Figure I-5, Part I). Likewise, the clustering hypothesis also applies to the number of employees (named labour above) because this factor of production is one of the most important for this financial industry, as was previously said, due to the high qualifications required of the employees.

Table II-7 Descriptive Statistics of Clusters

Year	Cluster	Num. of MFMCs	Num. of Employees (L _k)	Shareholders' Equity (1)** (SE _k)	Assets Managed** (AM _k)	Num. of Funds (NF _k)	Num. of Unitholders
2005	1	2	144	77,507	53,319,085	229	1,740,660
2005	2	5	82	54,869	13,145,941	127	416,610
2005	3 and 4	19	34	15,797	2,753,956	38	86,906
2005	5	76	9	3,628	491,361	12	16,197
2006	1	2	140	84,806	53,145,598	244	1,766,859
2006	2	5	87	59,163	13,679,317	146	444,456
2006	3	18	34	18,671	3,378,712	39	98,641
2006	4	77	11	3,830	453,269	12	14,367
2007	1	2	133	82,014	47,743,697	263	1,560,868
2007	2	4	110	42,492	13,820,432	175	419,612
2007	3	19	36	29,387	3,789,362	42	115,841
2007	4	70	11	4,624	457,521	13	15,000
2008	1	3	128	72,848	27,361,780	245	874,034
2008	2	4	92	87,217	8,206,037	145	313,397
2008	3	17	36	26,132	2,187,013	40	71,855
2008	4	69	12	5,326	340,199	13	11,929
2009	1	3	123	93,625	26,021,083	202	787,505
2009	2	4	88	94,090	7,342,766	113	269,509
2009	3	14	37	29,239	2,545,975	41	84,597
2009	4	74	12	5,575	370,762	12	11,486

Figures are shown as the mean of the cluster.

(1) At beginning of the year; ** Thousand Euros

After the exhaustive clustering process previously defined, we continue to identify locally efficient companies, i.e., those companies that cannot be referred to the efficient frontier formed by companies belonging to the same cluster as the target company. Therefore, it is necessary to identify the efficient companies within each cluster. Then, according to Variation III, we must enumerate all the facets to be potentially compared with each target company analysed within each cluster. After that, we should select those efficient combinations (*friends*) for each cluster that are not dominated by any other efficient combination (*maximal friends*). That is, we will select those *maximal friends* for every cluster to run Variation III instead of searching for all the *maximal friends* combinations of all SBM-efficient companies (Variation II, equation II.5).

The search for the *maximal friend* facets has been quite different for each stage. For the case of the marketing and distribution unit (Stage 2) and the overall unit (Stage 3), the *maximal friends* were easily found due to the reduced number of SBM-efficient companies included in the reference sets for each stage. Thus, the algorithm proposed by Tone (2010) to find the *maximal friend facets* within each cluster to run expression (II.5) was easily developed. For the case of the portfolio management unit (stage 1), the existence of a much higher number of efficient companies involved huge computational resources. Indeed, the consideration of four clusters aims to reduce largely the enumeration of facets and the identification of *maximal friends*. As stated above, this is an important contribution because it has allowed finding locally efficient companies with regard to similar competitors that otherwise would not have been possible to identify.

The results found by Variation III in the different management stages are displayed in Appendix K. To clarify the major findings, Table II-8 summarises globally efficient, globally inefficient, and locally efficient companies for each management stage. For example, reports that there were 74 globally inefficient companies but 26 locally efficient companies in the *Portfolio Management stage* in 2005. That is, the analysis of efficiency restricted to homogeneous competitors reveals that 26 out of 74 initially considered inefficient companies are efficient in their *Portfolio Management stage*, which might be considered a relevant percentage. The average ratio of locally efficient companies to globally inefficient companies in Stage 1 is 26.19% for the entire time horizon. This might be considered measurement bias due to inappropriate reference frontiers with different characteristics from the evaluated company. This potential bias seems quite important in the *Portfolio Management stage* with so many efficient companies previously found with the different SBM variations.

Table II-8 also indicates that this potential measurement bias is not relevant in the overall efficiency analysis. For example, other than 10 companies overall efficient in 2005, there was just one other company which was locally overall efficient, according to homogeneous reference sets. The aforementioned measurement bias provided by the average ratio of globally overall-inefficient companies that become locally overall efficient is approximately 2.42% for the whole time horizon.

Regarding the second management stage (*Marketing and Service stage*), the concentration of all globally efficient companies in one cluster from 2005 to 2008 does not permit finding locally efficient companies in an accurate way (see details in Appendix L). This unique efficient cluster was Cluster #3 in 2005, and Cluster #4 for the remaining years. According to Tone (2010), if none of the DMUs in a cluster is efficient, we should take the efficient DMUs in the adjacent clusters as appropriate reference sets. Because of this approach, the results should be similar to those previously obtained by Variation II because the efficient reference set is unique for the whole dataset (see Table II-L-1, Appendix L). Only for 2009 was it possible to properly follow this approach. We take the adjacent Cluster #1 (with one efficient company) to search for locally efficient companies in Cluster #2, whilst Cluster #4 (with 6 efficient companies) was considered an adjacent cluster to search for locally efficient companies in Cluster #3 (see Table II-L-1, Appendix L).

In our opinion, the aforementioned adjacent-cluster process in Stage 2 could not be very suitable because of the excessive differences previously found in the clusters. We should at least be cautious with the results in those adjacent clusters with striking differences. That is, to find locally efficient companies based on reference frontiers with extremely different characteristics from the target companies' does not seem to be the best solution when a cluster does not have any efficient company to form part of the frontier. Therefore, we note the measurement bias obtained in the 2009 sample, where we find that the average ratio of locally efficient (Stage 2) companies to globally inefficient companies was approximately 0.68%.

Table II-8 Efficient Companies by Stage

Year	MMS	Total Companies	Globally Efficient	Globally Inefficient	Locally Efficient
2005	Stage 1		28	74	26
	Stage 2	102	7	95	0
	Overall Eff.		10	92	1
2006	Stage 1		22	80	22
	Stage 2	102	8	94	0
	Overall Eff.		8	94	4
2007	Stage 1		21	74	19
	Stage 2	95	6	89	0
	Overall Eff.		6	89	2
2008	Stage 1		13	80	16
	Stage 2	93	2	91	0
	Overall Eff.		5	88	3
2009	Stage 1		20	75	17
	Stage 2	95	7	88	3
	Overall Eff.		5	90	1

Previous findings seem to provide evidence that the efficiency measurement of the *Portfolio Management stage* is quite sensitive to inappropriate reference frontiers. That is, an appropriate SBM evaluation of the *Portfolio Management stage* should consider homogeneous sets of companies with similar management resources that may be proxied by the assets under management. Therefore, in the case that the reference companies included in the efficient frontier do not belong to the same cluster as the target company, traditional SBM scores could be biased in favour of efficient companies in this *Portfolio Management stage*. Otherwise, this bias appears to be less important to both the *Marketing and Service stage* and the *Overall stage*. However, these results may be affected in the *Marketing and Service stage* by the high concentration of efficient companies in few clusters, which could be problematic if we consider adjacent clusters with striking differences to search for locally efficient companies.

3 Persistence in funds management companies: Do the best winners and losers usually repeat?

Performance persistence in mutual funds has been extensively analysed by researchers since the first studies by Sharpe (1966) and Jensen (1968). This phenomenon refers to the hypothesis that mutual funds with better (worse) performance records than competitors will keep these better (worse) performance records over time. The importance of this hypothesis is very relevant in the mutual fund industry in terms of prediction of future performance based on mutual funds' past reputation, i.e., the accepted wisdom in the fund industry that winners repeat (Kahn and Rudd, 1995). This abovementioned importance is highlighted by financial marketing because past results are one of the major factors that drive investors to choose a mutual fund (Ippolito, 1992; Capon et al., 1996; and Sirri and Tufano, 1998).

Later studies in the 1980s and 1990s found contradictory results in this field, some supporting the persistence hypothesis (e.g., Lehman and Modest, 1987; Grinblatt and Titman, 1992; Brown and Goetzmann, 1995; Malkiel, 1995; Elton et al., 1996; and Gruber, 1996), but some also rejecting this “hot hands phenomenon”⁸⁴ (e.g., Bogle, 1992; and Carhart, 1997).⁸⁵ More recently, Busse et al. (2010) and Fama and French (2010) applied Carhart's four-factor model (1997) to find little evidence of persistence. However, Cohen et al. (2005) found that fund returns exhibit significant persistence, even after adjusting for momentum in stock returns. Lynch and Musto (2003) provided evidence of more sensitive persistence in winner funds than in losers. Kosowski et al. (2006) found that growth-oriented funds exhibit strong persistence, while income-oriented funds exhibit little evidence of this phenomenon.

⁸⁴ “Hot hands phenomenon” was first used in the Hendricks and Zeckhauser' (1993) study.

⁸⁵ According to Carhart (1997), the persistence could be caused by managers' costs and the momentum effect, rather than managers' ability. Recently, Gottesman and Morey (2007) and Fama and French (2010) identify costs as the source of persistence. Andreu et al. (2007) also found similar evidence for Spanish money market funds.

The assorted conclusions can be explained by the different characteristics of the portfolios examined, the different evaluation periods, and the survivorship bias influence. The incidence of terminated mutual funds being excluded from the sample (survivorship bias) on persistence results has been addressed by different authors including Grinblatt and Titman (1992); Brown et al. (1992); Malkiel (1995); Blake and Timmermann (1998); and Hallahan and Faff (2001); and more recently Rohleder, et al. (2011). The magnitude of this bias in persistence findings is still a controversial topic in this field.

The analysis of this persistence phenomenon has also been extensively analysed in Europe in recent decades. Studies include Ter Horst et al. (1998) for the Dutch market, Cortez et al. (1999) and Cortez and Silva (2002) for the Portuguese market, Dahlquist et al. (2000) for the Swedish market, Otten and Bams (2001) for European industry considering different countries, Giles et al. (2002) for the U.K. market, Casarin et al. (2005) for the Italian market, Christensen (2005) for Danish industry, and Ciriaco and Santamaria (2005) and Vicente and Ferruz (2005) for the Spanish market.

Although most studies have focused on persistence equity mutual funds, there are also works on bond mutual funds and money market funds (e.g., Blake et al., 1993; Domian and Reichenstein, 1998; Philpot et al., 2000; Christoffersen and Musto, 2002; and Polwittoon and Tawatnuntachai, 2006).

The literature provides evidence that previous persistence research focused on mutual funds but not on the fund companies' efficiency over time. Taking into account our efficiency model (Figure I-5), the persistence analysis of fund companies should test for this persistence hypothesis in the efficiency achieved by the portfolios managed by the company (*Portfolio Management* stage, Stage 1) together with the efficiency in the company's abilities to sell these funds (Stage 2). In addition, our analysis should also consider the overall persistence hypothesis for the company as a whole (*Overall Efficiency* stage). That is, our innovative approach can more completely approximate this phenomenon by including in some way the traditional persistence idea related to portfolio

management ability and also by considering the remaining management stages that also add value to a fund company.

Therefore, the aim of this section is to test for whether mutual funds management companies maintain their relative efficiency rankings over time for the different management stages considered in our model. If the persistence hypothesis is accepted, then those companies with high efficiency scores in a specific management stage will keep their high efficiency rankings over time.

Our study fills this gap in the literature by analysing the persistence phenomenon based on the efficiency scores obtained by the SBM variants for the different management stages considered in our model. This analysis aims to be an original and further contribution to mutual fund literature because evidence on the compelling topic of performance persistence is inconclusive, and this phenomenon has never been explored for mutual fund companies.

3.1 Methodology of performance persistence

The metrics employed to measure the persistence phenomenon in the performance mutual funds industry have been determined by parametric and non-parametric techniques. Parametric techniques have focused on the use of recursive portfolios to test this phenomenon. Recursive portfolios proposed by Grinblatt and Titman (1993) and Hendricks et al. (1993), then improved by Carhart (1997) evaluate persistence by estimating their future performance according to the past performance of the mutual funds. These techniques have been applied in different markets in recent years by Bollen and Busse (2004), Cohen et al. (2005), Kacperczyk et al. (2008), Busse et al. (2010), Fama and French (2010), and Benos and Jochev (2011).

Recent studies distinguish among different management-style groups and consider the cross-sectional significance of recursive portfolios to measure the persistence hypothesis (Matallín et al., 2014a). Matallín et al. (2014b) also apply recursive portfolios to

evaluate the persistence of mutual fund performance obtained by partial frontiers based in DEA and FDH methodology.

On the other hand, the use of non-parametric techniques based on contingency tables widely used in other scientific disciplines has also been very frequent in mutual funds literature (Brown et al., 1992; Brown and Goetzmann, 1995; Allen and Tan, 1999; Silva et al., 2005; and Elyasiani and Jia, 2011, among others). However, a potential weak point of contingency tables is the a priori determination of groups to be compared to other groups in subsequent years. Therefore, those groups based on medians, quartiles, quintiles and other breakpoints may be discretionary elements that could affect the robustness of the persistence results in the sense that there may be miniscule differences in the performance records among some funds included in adjacent groups. Cortez et al. (1999) already noted how results from contingency tables for small mutual fund samples should be interpreted with caution. Furthermore, there are studies which question the application of some non-parametric tests to aggregated tables because this aggregated information could lead to the existence of this phenomenon solely due to the consideration of a high number of portfolios (Andreu et al., 2007; Cuthbertson et al., 2010).

Taken into account the aforementioned limitations, our approach to persistence is based on analysis of contingency tables, but efficiency groups are consistent in terms of their efficiency scores instead of the definition of predetermined groups through median or quartile breakpoints. In our approach, we propose cluster techniques as a first step to design robust efficiency groups in contingency tables rather than mere upper and lower median-groups.

These cluster techniques have been applied to the SBM Variation II efficiency scores obtained for each management stage included in our model.⁸⁶ We run divisive clustering

⁸⁶ The scores obtained by SBM Variation II were used to obtain the efficiency clusters, specifically those obtained from the random search approach (Variation IV). According to Tone (2010), this variant evaluates each MFMC from all facets, thereby finding the most accurate SBM efficiency scores.

algorithms (DIANA) to identify the most suitable efficiency groups.⁸⁷ Note that this divisive technique is based on the natural search for Z representative objects among all the objects of the dataset initially included in one group (see details in section 2.4). The DIANA algorithm seems again to be more useful than agglomerative techniques in our concentrated efficiency results because it starts by finding a few large clusters rather than an exhaustive description of the individual efficiency results.

Our clustering approach first finds two major efficiency groups for each management stage, the *winner*s companies and the *loser*s companies. Then, the divisive technique splits up these groups to obtain four efficiency groups for each management stage (*Top Winners*, *Winners*, *Losers*, and *Bottom Losers*). The main advantage of this clustering process is that we obtain consistent efficiency groups rather than the mere consideration of median or quartile groups with the same number of companies. Appendix M shows the dendrograms of the efficiency clusters obtained by the DIANA technique for each management stage and year.

Non-parametric statistics applied to *winner*s and *loser*s efficiency clusters in 2x2 contingency tables include the following:

- The Z-test applied by Malkiel (1995):

$$Z = \frac{(Y - np)}{\sqrt{np(1 - p)}}$$

$$Z \approx N(0,1) \tag{II.12}$$

If we want to contrast the winners' persistence, then Y is the number of *winner* companies in two consecutive periods, n is the total number of *winner*s in the first period, and p is the probability of a *winner* mutual fund company being a *winner* in the

⁸⁷ The DIANA algorithm was applied in the Cluster package of R program (Version 1.14.4, August 2013) under the same conditions as the aforementioned clustering process to find the groups of companies required by Variation III in section 2.4 of this chapter.

next period under the null hypothesis of no persistence. In the case of *losers'* persistence, Y is the number of *loser* companies in two consecutive periods, n is the total number of *loser* in the first period, and p is the probability of a *loser* mutual fund company being a *loser* in the next period under the null hypothesis of no persistence.⁸⁸

- The OR ratio applied by Brown and Goetzmann (1995):

$$OR = \frac{WW * LL}{WL * LW} \quad (\text{II. 13})$$

where WW (LL) represents the number of fund companies that are *winner* (*losers*) in two consecutive periods and WL (LW) represents the number of mutual fund companies which were *winner* (*losers*) in the first period but then were *losers* (*winner*) in the next year.

This expression is the ratio of mutual fund companies that show persistence relative to those that do not. If the ratio equals one, each category is implied to have half the number of companies included in the winners and losers in the first year, and no persistence is therefore found. That is, if there are n (n') companies in the winners' (losers') cluster in period t , the lack of persistence will involve $n/2$ ($n'/2$) companies in both WL (LW) and WW (LL) groups. In this case, the null hypothesis (no persistence) cannot be rejected. Brown and Goetzmann (1995) use the subsequent Z-statistic based on OR ratio to develop an accurate persistence test as follows:

$$Z = \frac{\ln(OR)}{\sigma_{\log OR}}$$

$$Z \approx N(0,1) \quad (\text{II.14})$$

⁸⁸ In our equations, p takes different values for each compared period, because the number of winners and losers is different due to our efficiency cluster approach; i.e., both the probability of a winner company being a winner in the following period and the probability of a loser company being a loser in the following period were calculated for each period.

Additionally, to test persistence based on 2x2 contingency tables is possible by using chi-square as follows:

- The chi-square applied by Kahn and Rudd (1995):

$$X^2 = \sum_{i=1}^n \sum_{j=1}^n \frac{(O_{ij} - E_{ij})^2}{E_{ij}}$$

$$X^2 \approx X_{(r-1)(c-1)}^2 \quad (\text{II.15})$$

where O_{ij} (E_{ij}) is the actual (expected) frequency in the i^{th} row and j^{th} column in the contingency table, r is the number of rows and c is the number of columns. Thus, in the case of a 2x2 contingency table, this distribution presents one degree of freedom.⁸⁹

Application of the previous tests provides information about the existence of persistence for two consecutive periods, but it does not provide an overall result for longer time horizons. To overcome this limitation, aggregating results provided by the contingency tables in different periods is a common practice employed in the financial literature to produce an overall finding that allows us to conclude whether efficiency persistence exists for the entire period. As previously addressed in the literature, conclusions obtained from this practice should be interpreted with caution because the aggregation of information could lead to confirming the existence of this phenomenon solely due to the consideration of a high number of mutual fund companies.

In this sense, although Cochran's test (1954) also uses aggregate information which could provide biased evidence of the persistence hypothesis, it is a comprehensive measure that provides an overall result for the entire time period. This statistic is calculated as follows:

⁸⁹ Carpenter and Lynch (1999) studied the power of several tests used in persistence studies. They found that the chi-square test is powerful and more robust to the presence of survivorship bias compared to other tests.

- Overall results using Cochran's test (1954):

$$Y = \frac{\sum_{i=1}^g w_i d_i}{[\sum_{i=1}^g w_i P_i Q_i]^{1/2}}$$

$$Y \approx N(0,1) \quad (\text{II.16})$$

where:

$$P_i = \frac{n_{i1}P_{i1} + n_{i2}P_{i2}}{(n_{i1} + n_{i2})} \quad (\text{II.17})$$

$$Q_i = (1 - P_i) \quad (\text{II.18})$$

$$d_i = (P_{i1} - P_{i2}) \quad (\text{II.19})$$

$$w_i = \frac{n_{i1}n_{i2}}{(n_{i1} + n_{i2})} \quad (\text{II.20})$$

g is the number of 2x2 contingency tables, n_{i1} and n_{i2} are the sample sizes in the two groups ($WW + WL$) and ($LL + LW$), respectively, and p_{i1} (p_{i2}) is the relationship between WW (WL) and n_{i1} (n_{i2}).

To further refine the persistence conclusions, we develop 4x4 contingency tables. A 4x4 contingency table splits the mutual fund companies of each period into four subsets by considering DIANA clustering methodology. Therefore, Q_i is the subset of companies' scores included in the i^{th} groups of the cluster studied (previously called *Top Winners, Winners, Losers and Bottom Losers*), where i takes values from 1 to 4. In this case, two non-parametric tests are applied: the chi-square test and the residual analysis proposal by Haberman (1973).

The chi-square test involves comparing expected and observed frequencies similarly to 2x2 contingency tables. In the case of a 4x4 contingency table, this distribution presents nine degree of freedom (see expression II.15).

On the other hand, Haberman's (1973) residual analysis identifies the categories responsible for a significant chi-square value, i.e., this analysis makes it easier to identify those groups most responsible for the phenomenon of efficiency persistence. One adjusted residual d_{ij} can be calculated for each cell in the 4x4 contingency tables as follows:

- Residual analysis applying Haberman's test (1973):

$$d_{ij} = \frac{e_{ij}}{\sqrt{v_{ij}}}$$

$$d_{ij} \approx N(0,1) \quad (\text{II.21})$$

where:

$$e_{ij} = \frac{n_{ij} - E_{ij}}{\sqrt{E_{ij}}} \quad (\text{II.22})$$

$$v_{ij} = \left(1 - \frac{n_i}{N}\right) \left(1 - \frac{n_j}{N}\right) \quad (\text{II.23})$$

n_{ij} (E_{ij}) is the observed (expected) frequency in the i^{th} row and j^{th} column of the contingency table, and n_i (n_j) is the total number of observations in the i^{th} row and j^{th} column.

3.2 Empirical results of efficiency persistence

After proper identification of *winner*s and *loser*s in clusters from 2005 to 2009 (see Appendix M), we first apply the previous non-parametric tests for 2x2 contingency tables individually for each management stage. Table II-9, Table II-10, and Table II-11 show the results of these metrics to test for the persistence phenomenon.

The first column of the following 2x2 tables shows the consecutive annual periods analysed. The following four columns report the number of companies included in each category of the 2x2 contingency tables. That is, WW reports the number of *winner*s companies in year t and $t+1$. WL reports the number of *winner*s companies in year t but *loser*s in $t+1$. LW reports the number of *loser*s in year t but *winner*s in $t+1$. LL reports the number of *loser*s companies in year t and $t+1$. Then, columns 6 and 7 report the Z-test of Malkiel for both *winner*s and *loser*s. Columns 8 and 9 report both the Brown and Goetzmann statistic as the Kahn and Rudd chi-square test. Finally, the result for the entire period (2005-2009) is reported in the last column of the table using the Y-test of Cochran.

Table II-9 and Table II-11 show a significant persistence phenomenon in efficiency scores clustered, i.e., both the *Portfolio Management stage* (Stage 1) and *Overall Efficiency*, although less significance is observed between 2008 and 2009. Cochran's test also confirms this evidence for the whole period from 2005 to 2009. These findings support the proposition that those companies that manage their mutual funds better than their competitors are usually the same during the time horizon analysed. A similar conclusion could be drawn for the efficiency of a mutual fund company as a whole. On the other hand, the aforementioned conclusions could also be extended to those companies with worse efficiency records than their competitors.

Table II-10 shows significant persistence in the marketing and distribution stage (Stage 2) for the first two comparative periods (2005-2006, 2006-2007). However, there is a lack of this persistence phenomenon in 2007-2008 and 2008-2009, which might be

driven by the consequences of the financial crisis affecting the successful distribution of mutual funds managed by the company.⁹⁰ This finding on the persistence phenomenon is achieved for the entire period when Cochran's test is used. However, this result must be taken with caution, considering the limitations previously indicated for the Cochran's measure; i.e., aggregate consideration of information tends to skew towards persistence.

Table II-9 Efficiency persistence for stage 1 based on 2x2 contingency tables

Stage 1	WW	WL	LW	LL	Malkiel	Malkiel	B&G	K&R	Cochran
					Z-test Winners	Z-test Losers			
2005-2006	37	33	1	27	2.418*	3.823**	3.260**	20.464**	9.146**
2006-2007	36	3	18	38	4.472**	3.732**	4.856**	33.922**	
2007-2008	25	27	0	40	3.388**	3.863**	N.A.	26.406**	
2008-2009	14	12	17	50	2.219*	1.382	2.551*	6.834**	

Table II-10 Efficiency persistence for stage 2 based on 2x2 contingency tables

Stage 2	WW	WL	LW	LL	Malkiel	Malkiel	B&G	K&R	Cochran
					Z-test Winners	Z-test Losers			
2005-2006	27	2	8	61	6.450**	4.181**	5.626**	59.085**	8.160**
2006-2007	30	2	35	28	3.082**	2.197*	3.214**	14.328**	
2007-2008	3	60	0	29	0.671	0.989	N.A.	1.428	
2008-2009	3	0	87	3	0.316	0.058	N.A.	0.103	

Table II-11 Efficiency persistence for overall efficiency based on 2x2 contingency tables

Overall eff.	WW	WL	LW	LL	Malkiel	Malkiel	B&G	K&R	Cochran
					Z-test Winners	Z-test Losers			
2005-2006	13	4	1	80	7.327**	3.357**	4.805**	64.954**	14.804**
2006-2007	9	3	1	82	7.278**	2.767**	4.561**	60.621**	
2007-2008	7	2	2	81	6.866**	2.261*	4.609**	52.259**	
2008-2009	10	0	8	75	6.455**	2.241*	N.A.	46.687**	

* Statistically significant at 0.05 level (2-tailed)

** Statistically significant at 0.01 level (2-tailed)

N.A. in the Brown and Goetzmann test represents unrealisable calculations as a consequence of a zero value in any category considered in the OR ratio.

⁹⁰ Further research for longer time horizons will be necessary to test this hypothesis more appropriately.

To look more deeply at the persistence effect using non-parametric measures, we use 4x4 contingency tables. The results are presented in, Table II-12, Table II-13 and Table II-14. Columns 2 to 5 contain the 4x4 contingency tables that result when comparing the four efficiency clusters (*Top Winners* or “*Top W*”, *Winners* or “*W*”, *Losers* or “*L*”, *Bottom Losers* or “*Bott L*”) in the two consecutive annual periods analysed (t , $t-1$). The analysis of residuals of Haberman is shown in parentheses. These latter statistics identify the categories responsible for a significant chi-square value, which is presented in the last column of the table.

First, we find a significant value of the chi-square test for all comparative periods and for each fund management company. This result provides evidence of a significant persistence in efficiency records obtained by the companies in the different management stages. The analysis of residuals provides some relevant findings to identify those efficiency clusters responsible for these significant results. We find significant persistence of those companies included in the “*Top W*” clusters in two consecutive years. This finding indicates that the best-managed companies persist over time, and that this result is robust for portfolio management and distribution of mutual funds as well as the overall efficiency of the company as a whole. That is, the significant persistence results previously obtained might be strongly caused by the best mutual fund companies in efficiency terms. However, the significant results found for the “*Bott L*” clusters in two consecutive years also support the significant role of the worst-managed companies in this persistence phenomenon (with the exception of Stage 2).

Table II-12 Performance persistence for stage 1 based on 4x4 contingency tables

2005-2006	Top W	W	L	Bott L	χ^2
Top W	22 (6.176)**	4 (0.200)	4 (-4.249)**	3 (-1.837)	72.704**
W	4 (-2.889)**	7 (1.879)	24 (3.599)**	2 (-2.727)**	
L	0 (-3.285)**	0 (-1.894)	13 (1.863)	9 (2.899)**	
Bott L	1 (-0.616)	0 (-0.899)	0 (-2.144)*	5 (4.089)**	
2006-2007	Top W	W	L	Bott L	χ^2
Top W	19 (5.924)**	6 (-0.977)	1 (-4.439)**	1 (-0.155)	72.036**
W	1 (-1.582)	10 (4.378)**	1 (-2.327)*	0 (-0.777)	
L	4 (-3.006)**	12 (0.367)	22 (3.092)**	0 (-1.668)	
Bott L	2 (-1.718)	0 (-3.046)**	13 (3.216)**	3 (2.923)**	
2007-2008	Top W	W	L	Bott L	χ^2
Top W	11 (5.386)**	4 (0.314)	6 (-2.670)**	4 (-1.346)	57.880**
W	1 (-1.714)	9 (3.408)**	15 (1.092)	2 (-2.630)**	
L	0 (-3.047)**	0 (-3.191)**	22 (2.006)*	15 (2.590)**	
Bott L	0 (-0.682)	0 (0.714)	0 (-1.650)	3 (2.964)**	
2008-2009	Top W	W	L	Bott L	χ^2
Top W	7 (3.060)**	1 (-0.498)	5 (-1.193)	0 (-1.496)	31.460**
W	3 (0.149)	3 (1.354)	7 (0.006)	0 (-1.496)	
L	5 (-2.150)*	7 (1.233)	28 (2.036)*	3 (-1.581)	
Bott L	5 (-0.093)	0 (-2.083)*	10 (-1.380)	9 (4.173)**	

* Statistically significant at 0.05 level (2-tailed)

** Statistically significant at 0.01 level (2-tailed)

Table II-13 Performance persistence for stage 2 based on 4x4 contingency tables

2005-2006	Top W	W	L	Bott L	χ^2
Top W	11 (6.952)**	0 (-1.496)	0 (-1.782)	0 (-3.112)**	131.312**
W	9 (3.448)**	7 (3.076)**	2 (-1.083)	0 (-4.152)**	
L	0 (-2.214)*	8 (4.214)**	8 (3.211)**	0 (-3.867)**	
Bott L	0 (-5.440)**	0 (-4.567)**	10 (-0.411)	43 (8.066)**	
2006-2007	Top W	W	L	Bott L	χ^2
Top W	7 (3.910)**	10 (0.278)	0 (-2.942)**	0 (-0.667)	46.918**
W	2 (0.089)	11 (1.491)	2 (-1.494)	0 (-0.619)	
L	3 (0.359)	16 (2.454)*	1 (-2.702)**	0 (-0.738)	
Bott L	0 (-3.370)**	16 (-3.316)**	25 (5.573)**	2 (1.572)	
2007-2008	Top W	W	L	Bott L	χ^2
Top W	2 (3.692)**	1 (2.596)**	3 (0.192)	6 (-2.023)*	27.505**
W	0 (-1.595)	0 (-1.121)	16 (2.178)*	35 (-1.288)	
L	0 (-0.922)	0 (-0.648)	2 (-2.271)*	25 (2.630)**	
Bott L	0 (-0.213)	0 (-0.150)	0 (-0.778)	2 (0.849)	
2008-2009	Top W	W	L	Bott L	χ^2
Top W	2 (4.119)**	0 (-3.547)**	0 (-0.149)	0 (-0.212)	33.242**
W	1 (2.897)**	0 (-2.494)*	0 (-0.105)	0 (-0.149)	
L	5 (2.075)*	17 (-1.354)	0 (-0.560)	0 (-0.796)	
Bott L	2 (-4.011)**	63 (3.039)**	1 (0.610)	2 (0.867)	

* Statistically significant at 0.05 level (2-tailed)

** Statistically significant at 0.01 level (2-tailed)

Table II-14 Performance persistence for overall efficiency based on 4x4 contingency tables

2005-2006	Top W	W	L	Bott L	χ^2
Top W	9 (8.328)**	1 (0.891)	1 (-3.418)**	0 (-2.226)*	119.873**
W	1 (0.540)	2 (3.737)**	2 (-1.216)	1 (-0.666)	
L	0 (-4.906)**	1 (-1.904)	52 (6.020)**	14 (-2.473)*	
Bott L	0 (-1.362)	0 (-0.834)	1 (-4.083)**	13 (5.751)**	
2006-2007	Top W	W	L	Bott L	χ^2
Top W	6 (7.157)**	1 (1.434)	2 (-1.081)	0 (-3.186)**	95.425**
W	1 (1.749)	1 (3.037)**	1 (-0.203)	0 (-1.779)	
L	0 (-3.155)**	1 (-0.835)	32 (4.659)**	21 (-2.604)**	
Bott L	0 (-1.822)	0 (-1.167)	2 (-4.247)**	27 (5.502)**	
2007-2008	Top W	W	L	Bott L	χ^2
Top W	5 (8.705)**	1 (1.530)	0 (-1.895)	0 (-2.764)**	115.598**
W	0 (-0.422)	1 (2.503)*	2 (1.131)	0 (-1.921)	
L	0 (-1.886)	2 (0.408)	24 (4.756)**	11 (-3.888)**	
Bott L	0 (-2.299)*	0 (-2.045)*	7 (-4.130)**	39 (5.861)**	
2008-2009	Top W	W	L	Bott L	χ^2
Top W	6 (8.255)**	0 (-0.879)	0 (-2.013)*	0 (-2.152)*	116.221**
W	1 (1.196)	3 (4.240)**	0 (-1.625)	0 (-1.737)	
L	1 (-1.421)	7 (2.415)*	20 (3.215)**	5 (-3.882)**	
Bott L	0 (-3.190)**	0 (-3.610)**	16 (-1.432)	34 (5.493)**	

* Statistically significant at 0.05 level (2-tailed)

** Statistically significant at 0.01 level (2-tailed)

A more detailed analysis of the residuals of the main diagonals (from “*Top W*” to “*Bott L*”) of the 4x4 contingency tables should provide evidence about the probability of a mutual fund company remaining in the same efficiency cluster in the next year. That is, the significance of positive values in the main diagonals of the contingency tables shows that it is quite difficult for a company to change its efficiency category in relation to competitors. Table II-14 shows that the main diagonal of the *Overall Efficiency stage* provides positive and significant values in all years, indicating that the efficiency clusters remains quite similar without many changes. This result is somewhat less evident for the *Portfolio Management stage* (stage 1).

Supporting the robustness of these previous results, negative and significant values are obtained in some cases when the opposite diagonal (from “*Bott L*” to “*Bott L*”) is traced. This finding indicates that efficiency scores are unlikely to change to a very different efficiency cluster. Similarly, when we compare two opposite categories (“*Top W*” to “*Bott L*”) the results show negative and significant values in most figures.

Furthermore, it will be interesting to analyse the effect of the companies’ resources in the different categories of persistence studied. These management resources were approximated by different variables such as number of employees, shareholders’ equity, assets managed and number of funds.⁹¹ These proxy variables make sense because all of them are very close to the company’s size; that is, the larger a company is, the more resources the company has to manage and sell their mutual funds. The following tables (Table II-15, Table II-16, Table II-17, and Table II-18) show the aggregate mean values of these proxy variables after considering whole-year frequencies of each 4x4 persistence cell for each stage. After that, a two-tailed t-statistic was used to check for the null hypothesis that the aggregate mean of any cell is different from the aggregate mean of the remaining 15 persistence categories.

⁹¹ Other potential variables such as number of unitholders are not shown here, because the results are very similar to those obtained from assets managed and number of funds. Detailed results are available upon request.

Table II-15 Aggregate mean and deviation value by number of employees

Stage 1	Top W	W	L	Bott L	Mean(sd)
Top W	33 (48)**	20 (18)	28 (46)	7 (5)**	19 (20)
W	25 (38)	31 (28)*	20 (20)	8 (4)**	
L	12 (13)**	15 (16)*	21 (21)	23 (15)	
Bott L	7 (7)**	N.A.	17 (12)**	20 (12)	
Stage 2	Top W	W	L	Bott L	Mean(sd)
Top W	6 (5)**	10 (7)**	7 (4)**	8 (5)**	21 (17)
W	7 (6)**	9 (4)**	11 (7)**	15 (10)**	
L	10 (7)**	11 (8)**	17 (11)*	43 (33)**	
Bott L	76 (75)	21 (18)	36 (32)**	46 (48)**	
Overall eff.	Top W	W	L	Bott L	Mean(sd)
Top W	15 (17)**	72 (96)	33 (42)	N.A.	24 (26)
W	6 (1)**	12 (10)**	50 (73)	29 (7)	
L	6 (0)**	7 (6)**	18 (24)**	21 (27)	
Bott L	N.A.	N.A.	12 (10)**	32 (29)**	

Significance is the result of a test cell average of aggregated data from other cells; N.A. represents unrealisable calculations because of not any company in the category; values are rounded.

* Statistically significant at 0.05 level (2-tailed)

** Statistically significant at 0.01 level (2-tailed)

Table II-16 Aggregate mean and deviation value by shareholder equity

Stage 1	Top W	W	L	Bott L	Mean(sd)
Top W	17,600 (27,190)*	7,620 (7,030)**	14,040 (28,301)	3,633 (5,638)**	10,657 (17,261)
W	9,024 (12,161)	14,306 (15,499)	8,587 (13,602)**	2,956 (2,750)**	
L	22,860 (60,471)	8,418 (14,198)	17,315 (31,819)*	9,880 (11,827)	
Bott L	1,848 (2,705)**	N.A.	7,017 (8,918)**	14,756 (16,804)	
Stage 2	Top W	W	L	Bott L	Mean(sd)
Top W	1,536 (2,167)**	6,449 (11,846)*	1,759 (882)**	2,424 (1,679)**	10,845 (14,773)
W	3,533 (7,541)**	1,922 (1,234)**	6,581 (12,039)**	5,412 (5,366)**	
L	7,769 (15,800)	5,453 (8,282)	4,985 (5,573)**	30,254 (35,335)**	
Bott L	32,238 (41,794)	13,530 (25,560)	21,414 (27,393)**	28,259 (33,880)**	
Overall eff.	Top W	W	L	Bott L	Mean(sd)
Top W	14,442 (22,255)	29,386 (35,770)	30,699 (45,648)	N.A.	13,890 (18,894)
W	705 (80)**	15,555 (23,311)	32,497 (33,190)	6,735 (3,812)	
L	2,736 (6)**	1,440 (626)	9,330 (20,089)**	15,190 (29,842)	
Bott L	N.A.	N.A.	4,594 (5,111)**	17,262 (25,885)**	

Significance is the result of a test cell average of aggregated data from other cells; N.A. represents unrealisable calculations because of not any company in the category; Figures at beginning of the year and in Thousand Euros.

* Statistically significant at 0.05 level (2-tailed)

** Statistically significant at 0.01 level (2-tailed)

Table II-17 Aggregate mean and deviation value by assets managed

Stage 1	Top W	W	L	Bott L	Mean(sd)
Top W	7,539,816 (14,975,718)**	2,499,459 (3,201,010)	3,325,931 (8,192,804)	378,302 (953,669)**	
W	2,823,376 (5,207,405)	3,515,910 (4,994,322)	1,424,367 (2,727,004)**	179,167 (212,571)**	1,785,673 (3,343,212)
L	976,676 (2,281,639)*	1,375,273 (2,723,223)	1,287,945 (2,382,732)**	631,778 (822,021)**	
Bott L	25,296 (25,754)**	N.A.	597,174 (997,307)**	204,622 (451,003)**	
Stage 2	Top W	W	L	Bott L	Mean(sd)
Top W	25,310 (33,572)**	39,984 (25,595)**	45,585 (46,153)**	184,752 (111,506)**	
W	62,322 (42,010)**	149,904 (134,241)**	276,866 (653,622)**	431,404 (322,675)**	1,889,355 (2,397,924)
L	140,571 (169,060)**	344,061 (616,185)	589,123 (807,103)**	4,400,304 (4,105,803)**	
Bott L	8,098,920 (8,931,617)	1,181,748 (1,648,450)**	4,097,127 (4,499,935)**	10,161,691 (16,219,252)**	
Overall eff.	Top W	W	L	Bott L	Mean(sd)
Top W	2,190,088 (8,648,521)	20,835,636 (31,708,106)	14,560,510 (22,447,224)	N.A.	
W	11,987 (4,073)**	1,682,461 (2,725,110)	12,635,747 (23,743,689)	2,025,700 (414,881)	4,717,816 (8,088,991)
L	265,705 (30,428)**	43,118 (36,207)	1,929,439 (6,241,025)	2,147,219 (4,760,439)	
Bott L	N.A.	N.A.	272,582 (348,713)**	2,731,413 (4,048,466)	

Significance is the result of a test cell average of aggregated data from other cells; N.A. represents unrealisable calculations because of not any company in the category; Figures in Thousand Euros.

* Statistically significant at 0.05 level (2-tailed)

** Statistically significant at 0.01 level (2-tailed)

Table II-18 Aggregate mean and deviation value by number of funds

Stage 1	Top W	W	L	Bott L	Mean(sd)
Top W	71 (93)**	34 (27)	45 (75)	3 (2)**	
W	45 (62)	46 (39)**	24 (28)	9 (7)**	24 (29)
L	16 (19)**	23 (28)	18 (19)**	13 (12)**	
Bott L	3 (3)**	N.A.	10 (10)**	6 (8)**	
Stage 2	Top W	W	L	Bott L	Mean(sd)
Top W	2 (1)**	3 (2)**	4 (2)**	8 (2)**	
W	4 (2)**	6 (3)**	5 (2)**	16 (7)**	26 (24)
L	6 (3)**	8 (4)	10 (2)**	66 (62)**	
Bott L	113 (118)	26 (21)	56 (62)**	78 (87)**	
Overall eff.	Top W	W	L	Bott L	Mean(sd)
Top W	9 (33)**	103 (153)	66 (97)	N.A.	
W	2 (1)**	18 (25)	79 (128)	9 (1)**	31 (45)
L	8 (1)**	4 (2)	21 (41)**	30 (40)	
Bott L	N.A.	N.A.	12 (7)**	47 (56)**	

Significance is the result of a test cell average of aggregated data from other cells; N.A. represents unrealisable calculations because of not any company in the category; the values are rounded.

* Statistically significant at 0.05 level (2-tailed)

** Statistically significant at 0.01 level (2-tailed)

The above aggregate tables show a high persistence for the biggest companies located in the extreme cell *Top Winners–Top Winners* for the *Portfolio Management stage* (Stage 1). This result is robust in terms of employees, assets, equity and number of funds. That is, the size, as a proxy for portfolio management resources, seems to play an important role in the portfolio management persistence of the best fund companies. On the other hand, the aggregate concentration in the extreme cell *Bottom Losers–Bottom Losers* shows not-significant differences in terms of labour and capital in contrast with the remaining variables (number of funds and assets), where the aggregate values show significantly lower values than the overall average.

The opposite pattern is found in the *Marketing and Service stage* (Stage 2), where size, as a proxy for resources, seems to play a significant and negative role in the persistence of this stage. Therefore, the bigger a company is, the more problems it will have in efficiently distributing mutual funds. This result is robust in terms of employees, assets, equity and number of funds. Note that these findings are very consistent with those found in Table I-4 of this thesis.

The aforementioned pattern is also found in the *Overall Efficiency stage*, although the results are not so significantly clear for all persistence cells and variables. This situation could suggest that the marketing and service of the fund companies also affect considerably the persistence of the overall efficiency of the companies. This result is consistent with those obtained in Table I-5 of Part I.

Looking in more detail at the results from persistently extreme winners and losers, the null hypothesis that the aggregate mean of the characteristics of *Top Winners–Top Winners* is different from the aggregate mean of *Bottom Losers–Bottom Losers* is shown in Table II-19. The table is constructed by subtracting mean values of both extreme persistence groups, then determining the statistical significance for each difference. The results are robust with those provided by the previous comparisons and suggest that companies with more resources are better at managing than at selling their portfolios,

and therefore impacting in the overall efficiency levels. The differences in these extreme persistence categories are significantly important for the *Marketing and Service stage*.

Table II-19 Comparison of mean values between “*Top W - Top W*” and “*Bott L – Bott L*” companies

	Employees	Equity	Assets	Funds	Unitholders
Stage 1	14	2,844**	7,335,194	65**	247,956**
Stage 2	-40**	-26,723**	-10,136,381**	-76**	-332,479**
Overall eff.	-17**	-2,820	-541,325	-38**	-11,320

* Statistically significant at 0.05 level (2-tailed)

** Statistically significant at 0.01 level (2-tailed)

Figures of equity and assets are in thousand Euros.

In general, the results of this sub-section provide evidence of persistence in efficiency records obtained by the companies in the different management stages. Residual analysis of the main diagonals (from “*Top W*” to “*Bott L*”) also shows evidence about the significant probability of a mutual fund company to remain in the same efficiency cluster in the next year. Likewise, negative and significant values are obtained in some cases when the opposite diagonal (from “*Bott L*” to “*Bott L*”) is traced, indicating that those efficiency scores are unlikely to change to a very different efficiency cluster. Further research involving parametric tools could be useful to test for the robustness of these major findings.

4 Conclusions and Summaries

Based on the sub-DMUs approach of Holod and Lewis (2011) and considering the interaction between the different management subsystems within a mutual fund company proposed by Berkowitz and Qiu (2003), Part I of this thesis evaluated the efficiency of Spanish Mutual Fund Companies by applying the slack-based measure (SBM) approach proposed by Tone (2001). In Part II, we go a step further by developing a study that overcomes the limitation of an appropriate reference set identification. This problem is particularly important in the Spanish fund industry due to both the high concentration and assorted characteristics of the management companies. We use for this purpose the recent and unexplored variations of SBM proposed by Tone (2010), which allow comparing fund companies with their true ‘best practice’ competitors by similar management resources.

We analysed the effect of these variations (based on the hyperplanes instead of the vertices of the frontier) on efficiency rankings previously obtained by the original SBM. First, Variation I improves the average efficiency score as a consequence of the new approach for every stage, although the Spearman rank correlation for both SBM and Variation I are nearly to 1 for every stage and year analysed. On the other hand, the comparative plots for *Marketing* and *Overall Efficiency stages* provided evidence of the high degree of similarity between the SBM and Variation I scores obtained in our sample; i.e., we found evidence of the residual effect of Variation I on the original SBM scores previously obtained. The aforementioned finding raises questions regarding the relevance of this variant in the empirical results of our study.

Similarly to previous SBM findings, the Variation I scores also show a decreasing efficiency pattern in the Spanish fund industry. In the same line, this variant highlights the fact that companies seem to be worse mutual fund sellers than portfolio managers, thereby reducing considerably the overall profits of company shareholders.

Continuing with the SBM variants proposed by Tone (2010), we applied SBM Variation II, which minimises the SBM score obtained from all facets found. SBM Variation II was computed using both *maximal friend* facets randomly obtained (Variation IV) and clustering facets search (Variation III) as suggested in the aforementioned model. Variation IV again improves the average efficiency scores for every stage, and the Spearman rank correlation between efficiency rankings (SBM original and Variation IV), again showed evidence of both high and significantly positive correlations. However, the correlation between Variation I and Variation IV scores, while still high, is lower than that between SBM original and Variation I.

Furthermore, Variation IV also showed a positive and significant relationship between the marketing and distribution skills of the company and overall efficiency for the entire time horizon. Therefore, this new measure proposed by Tone (2010) confirms the major findings obtained in Part I by using the SBM original. This result verifies the robustness of our previous conclusions, such as the apparent dichotomy between the efficiency of portfolio management and the distribution and marketing skills in the Spanish mutual fund industry.

On the other hand, we conducted a refined search for locally efficient companies by applying a clustering process (Variation III) based on a divisive hierarchical technique and different measurement dimensions (assets under management and number of employees of each fund company). This approach allowed us to find evidence that an appropriate evaluation of the *Portfolio Management stage* should consider homogeneous sets of companies with similar management resources. Therefore, when the reference companies included in the efficient frontier do not belong to the same cluster as the target company, traditional SBM scores could be biased. Otherwise, this bias appeared to be less important to both the *Marketing* and *Overall stages*.

In addition to the use of the different variations (Tone, 2010) versus the SBM original (Tone, 2001), we also analysed whether the patterns of efficiency obtained are persistent over time. Our persistence analysis is an innovative and complete approach due to the

inclusion in certain respects of the traditional performance persistence idea related to portfolio management abilities and also due to the consideration of the remaining management stages, which also add value to a fund company. The main advantage of our clustering approach has been that we obtained consistent efficiency groups rather than the mere consideration of median or quartile groups with the same number of companies.

First, using 2x2 contingency tables, we found a significant phenomenon of persistence in the efficiency clusters, both in the *Portfolio Management* and in *Overall Efficiency stage*, although less significance is observed between 2008 and 2009. Cochran's test also confirmed this evidence for the entire period from 2005 to 2009. These findings supported the underlying hypothesis that those companies that manage their mutual funds better than their competitors are usually the same during the time horizon analysed. These conclusions could also be extended to those companies with worse efficiency records than their competitors.

Regarding persistence of the marketing and distribution stage, results showed significant persistence for the first two comparative periods (2005-2006, 2006-2007). However, there is a lack of this persistence phenomenon in 2007-2008 and 2008-2009, which might be driven by the consequences of the financial crisis in the successful distribution of mutual funds managed by the company.

Moreover, 4x4 contingency tables provided evidence of a significant persistence in the efficiency records obtained by the companies in the different management stages. We found a significant value of the chi-square test for all comparative periods and for each fund management company. Additionally, we found significant persistence of both the best-and worst-managed companies in two consecutive years. This finding is robust for portfolio management and distribution of mutual funds as well as the overall efficiency of the company as a whole. A more detailed analysis of the residuals of these 4x4 contingency tables indicates that the efficiency scores are unlikely to change to a very different efficiency cluster.

Furthermore, we found that size, as a proxy for portfolio management resources, seems to play an important and positive role in the portfolio management persistence of the best fund companies. This result was robust in terms of employees, assets, equity and number of funds. On the other hand, we found that companies with fewer management resources are consistently the worst fund companies, especially in terms of number of funds and assets managed.

An opposite pattern was found in the *Marketing and Service stage*, where size, as a proxy for resources, seems to play a significant and negative role in the persistence of this stage. Therefore, the bigger a company is, the more problems it has efficiently distributing their mutual funds. This result is robust in terms of employees, assets, equity and number of funds. Noteworthy is that the same pattern was also found in the *Overall Efficiency stage*, although the results are not so significantly clear for every persistence cell and variable. This situation suggests that the marketing and service of fund companies is also considerably affecting the persistence of the overall efficiency of the companies.

Finally, looking in more detail at the difference in management resources from extreme persistence groups, we found robust results with those provided by the previous comparisons, thereby suggesting that companies with more resources are better at managing than at selling their portfolios, therefore impacting in the overall efficiency levels. The differences in these extreme persistence categories are significantly important for the *Marketing and Service stage*.

Appendix F

Table II-F-1 Variation I-efficiency under CRS for 2005

Stage 1 Portfolio Management				Stage 2 Marketing and Service			Overall Efficiency		
MFMC	VARI	Ref.	Rank	VARI	Ref.	Rank	VARI	Ref.	Rank
1	0.38665	161	79	0.08966	59	91	0.04431	196	99
2	0.50263	177	66	0.11600	59	69	0.04597	100	97
4	0.47083	14 161	68	0.11224	59	74	0.14556	125	55
6	1.00000	6	1	0.12635	59	61	0.09447	196	71
7	1.00000	7	1	0.11275	59	72	0.07625	100	82
9	0.26032	12 177	88	0.18803	59	47	0.27668	100	25
12	1.00000	12	1	0.01733	59	102	1.00000	12	1
14	1.00000	14	1	0.12632	59	62	1.00000	14	1
15	0.66942	12 161	39	0.08926	59	92	0.13830	200	57
20	0.53726	12 161	57	0.08596	59	95	0.26639	196	28
21	0.21632	161	94	0.12651	59	60	0.20771	125	36
24	0.44598	12 177	72	0.10505	59	78	0.06403	100	89
29	0.55348	43	53	0.10109	59	83	0.08061	100	79
31	0.66068	12 177	40	0.14563	59	54	0.07928	100	80
34	0.23310	177	93	0.15740	59	49	0.10354	196	68
35	0.94340	40 160 176 177	29	0.14150	59	55	0.08897	100	75
36	0.31692	174	83	0.48623	46	14	0.34542	100	19
37	0.62130	40 174 177	44	0.26371	46	30	0.17881	100	43
38	0.35231	160 176	81	0.11694	59	67	0.19097	100	40
40	1.00000	40	1	1.00000	40	1	1.00000	40	1
42	0.20996	174	95	0.48345	46	15	0.33274	100	20
43	1.00000	43	1	0.08761	59	93	0.05890	100	90
45	0.65554	12 174 177	41	0.15939	59	48	0.21755	100	35
46	0.25689	152	90	1.00000	46	1	0.58714	100	14
47	0.53955	12 176	56	0.12083	59	64	0.07059	100	84
49	0.57863	40 177	49	0.26299	46	31	0.19546	100	39
50	0.30691	177	85	0.84876	59 205	9	0.22464	100	33
51	0.65158	40 160 161	43	0.21977	59	39	0.13051	100	59
53	1.00000	53	1	0.61620	46	11	0.26866	100	27
55	1.00000	55	1	0.04260	59	101	0.30758	196	21
57	0.42220	174	76	0.20929	46	42	0.15238	100	51
58	0.74463	12 161	35	0.11267	59	73	0.18955	196	41
59	0.08443	161	101	1.00000	59	1	0.14996	125	53
60	0.12200	40 160	99	0.58960	46	13	0.35570	196	18
61	0.33483	161	82	0.09937	59	84	0.03392	196	100
62	0.74987	160 161 176	34	0.15399	59	51	0.15798	100	49
63	0.83527	12 43 154	31	0.10850	59	75	0.04542	100	98
69	0.31002	174 177	84	0.41929	46	18	0.12470	100	61
71	1.00000	71	1	0.10392	59	79	0.10305	100	70
73	0.71078	12 174 177	37	0.20019	59	44	0.17174	100	45
75	0.53706	12 161	58	0.07912	59	97	0.05309	100	94
76	1.00000	76	1	0.09364	59	86	0.26174	100	29
78	0.53648	174 177	59	0.23072	59	37	0.12188	100	62
83	0.65236	12 55 161	42	0.09589	59	85	0.07843	100	81
84	0.52326	14 161	63	0.09259	59	87	0.52573	125 196	15
85	0.55414	12 161	52	0.06204	59	100	0.15520	196	50
86	0.50307	161	64	0.09105	59	88	0.23166	100	32
93	0.20083	161	96	0.26485	59	29	0.13984	196	56
95	0.77913	55 161	33	0.11970	59	66	0.28886	100	24
98	0.52763	12 43 160	61	0.08995	59	90	0.18462	100	42
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
102	0.15566	152	97	0.81978	40 196	10	0.70115	100 196	13
103	1.00000	103	1	0.15074	59	52	1.00000	103	1
105	0.54799	163 177	54	0.08666	59	94	0.01924	100	101
110	0.57282	161	51	0.10239	59	81	0.08687	100	77
113	0.23546	161	92	0.10707	59	76	0.01282	196	102
115	0.52512	100 174	62	0.32630	46	24	0.17119	100	46
121	0.27680	161	87	0.09003	59	89	0.08832	196	76
123	0.38691	160	78	0.19068	59	46	0.11286	100	63
125	0.02147	161	102	0.44341	46	16	1.00000	125	1

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126	0.45016	177	71	0.19487	59	45	0.09324	100	72
127	1.00000	127	1	0.07950	59	96	0.10429	100	67
128	1.00000	128	1	0.10130	59	82	0.10884	196	65
130	1.00000	130	1	0.10245	59	80	0.06501	100	86
131	0.28311	152	86	0.32630	46	25	0.29562	100	22
132	0.45702	177	69	0.14965	59	53	0.08999	100	74
133	0.61604	40 174	45	0.38862	46	22	0.24787	100	31
137	0.43701	12 174	73	0.21974	59	40	0.20631	100	37
138	0.39042	152	77	1.00000	138	1	0.78334	100 152	11
139	0.24469	177	91	0.28047	59	28	0.41251	196	17
140	1.00000	140	1	0.12058	59	65	0.05620	100	92
142	0.35601	161	80	0.07777	59	98	0.08564	196	78
152	1.00000	152	1	0.41096	196	20	1.00000	152	1
154	1.00000	154	1	0.13565	59	58	0.04904	100	96
156	0.54752	55 174	55	0.24016	59	35	0.11235	100	64
159	0.50301	174 177	65	0.25823	59	33	0.10316	100	69
160	1.00000	160	1	0.25883	46	32	0.15886	100	48
161	1.00000	161	1	0.07408	59	99	0.07363	100	83
162	0.72785	12 55 154 177	36	0.13594	59	57	0.12509	100	60
163	1.00000	163	1	0.13043	59	59	0.05362	100	93
164	0.12391	174	98	0.94237	40 196	8	0.75246	125 196	12
168	0.48103	12 174	67	0.23932	59	36	0.15203	100	52
173	0.45081	177	70	0.11520	59	71	0.05016	100	95
174	1.00000	174	1	0.35947	46	23	0.50134	205	16
176	1.00000	176	1	0.10591	59	77	0.06601	100	85
177	1.00000	177	1	0.15581	59	50	0.09050	100	73
182	0.60649	176 177	47	0.11549	59	70	0.06486	100	87
185	0.43585	12 43	74	0.20259	59	43	0.27353	100	26
189	0.25766	152	89	0.60172	46 196	12	0.29159	100	23
190	0.67208	12 176 177	38	0.12094	59	63	0.06482	100	88
191	0.57749	174 177	50	0.28520	59	27	0.16560	100	47
192	1.00000	192	1	0.14081	59	56	0.10601	100	66
193	1.00000	193	1	0.11656	59	68	0.05747	100	91
194	0.83731	12 43 174 198	30	0.21982	59	38	0.17578	100	44
195	0.78890	174 198	32	0.43738	59	17	0.24808	100	30
196	0.11880	40 152	100	1.00000	196	1	1.00000	196	1
197	0.60959	174 177	46	0.25072	46	34	0.13386	100	58
198	1.00000	198	1	0.30823	59	26	0.14766	100	54
200	0.42880	55 174 177	75	0.21853	59	41	1.00000	200	1
202	0.58329	40 177	48	0.41174	46	19	0.22371	100	34
205	1.00000	205	1	1.00000	205	1	1.00000	205	1
206	0.52945	152 174	60	0.39816	196	21	0.20326	100	38

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). VAR I corresponds to the efficiency score obtained by the Variation I maximizing the objective function for the reference set obtained for each SBM-inefficient company (equation II.2). Ref. corresponds to the reference set for each company obtained by the VAR I approach. Rank represents the ranking of the VAR I scores which have been sorted by value, where VAR I scores with value of 1.00000 obtain the highest ranking 1.

Table II-F-2 Variation I-efficiency under CRS for 2006

Stage 1 Portfolio Management				Stage 2 Marketing and Service			Overall Efficiency		
MFMC	VARI	Ref.	Rank	VARI	Ref.	Rank	VARI	Ref.	Rank
2	0.55640	177	40	0.16883	196	80	0.05102	100	101
4	0.40949	14 51	61	0.20362	196	63	0.20237	103	41
6	0.33720	51 161	69	0.19315	196	67	0.13330	100	63
7	0.89402	12 53 163	25	0.18790	196	68	0.08920	100	82
9	0.23473	140 174	89	0.28269	196	35	0.33738	100	22
12	1.00000	12	1	0.00577	202	102	0.74887	14	11
14	1.00000	14	1	0.17761	202	74	1.00000	14	1
15	0.72103	12 127	31	0.00899	202	101	0.12844	103	66
20	0.61056	12 127	37	0.14657	196	93	0.32461	103	25
21	0.24047	127	87	0.17893	196	72	0.28615	103	31
24	0.45168	12 177	56	0.15599	196	88	0.07796	100	88
29	0.41895	128 205	59	0.16855	196	81	0.12851	100	65
31	0.91759	12 53 163	23	0.28779	196	34	0.12455	100	69
34	0.21397	177	93	0.53787	202	17	0.07962	100	87
35	0.53292	12 174	41	0.19937	196	64	0.10131	100	78
36	0.21919	174	92	0.45964	203	21	0.36283	100	19
37	0.49610	12 174 205	53	0.25101	196	45	0.19572	100	43
38	0.28681	51 163	84	0.20477	196	60	0.30177	100	28
40	0.12219	12 174	99	1.00000	40	1	0.98315	100 103	9
42	0.17388	174	96	0.41011	203	24	0.39169	100	17
43	0.85881	128 198 205	28	0.15153	196	91	0.06698	100	93
45	0.51323	12 174	46	0.21997	196	54	0.32857	100	23
46	0.33416	205	71	0.89877	196 203	10	0.61736	100	13
47	0.45723	12 51	55	0.19877	196	66	0.11071	100	72
49	0.35110	174	67	0.26393	196	42	0.25735	100	36
50	0.38948	174	63	0.61073	196	15	0.22569	100	38
51	1.00000	51	1	0.16411	196	83	0.11605	100	71
53	1.00000	53	1	0.62849	196 203	14	0.29695	100	30
55	1.00000	55	1	0.10510	196	99	0.38660	103	18
57	0.34766	12 174	68	0.22792	196	50	0.17904	100	47
58	0.61071	12 51 128	36	0.17201	202	78	0.27552	103	34
59	1.00000	59	1	0.74001	196	12	0.02049	125	102
60	0.09444	205	100	1.00000	60	1	0.63776	100 125	12
61	0.28675	161	85	0.16816	196	82	0.09254	100	80
62	1.00000	62	1	0.22136	196	53	0.17945	100	46
63	0.82739	12 128 163	29	0.14954	202	92	0.05501	100	97
69	0.52203	174 205	43	0.29226	203	33	0.15998	100	51
71	1.00000	71	1	0.16032	196	86	0.07392	100	90
75	0.44193	12 51 128	57	0.15477	196	90	0.05768	100	96
76	0.67039	55 174	34	0.16160	196	85	0.32722	100	24
78	0.51700	12 174	45	0.23471	196	47	0.13623	100	60
83	0.88700	12 55 127 140	26	0.22539	196	51	0.12697	100	68
84	0.55934	14 127	39	0.18098	196	71	0.58563	103	14
85	0.50734	12 51 128	49	0.08541	202	100	0.14745	103	55
86	0.50325	12 51	51	0.12824	196	97	0.30006	100	29
93	0.31691	51	74	0.27776	196	36	0.15319	100	54
95	0.49968	12 14 51	52	0.19907	196	65	0.46012	103	16
98	0.38824	12 51	64	0.15588	196	89	0.27062	100	35
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
102	0.29598	205	80	0.53768	203	18	0.33771	100	21
103	0.19099	55 127	95	0.22883	196	49	1.00000	103	1
105	0.16744	12 174	98	0.21509	196	55	0.15480	100	53
110	0.51800	12 51	44	0.16956	196	79	0.13768	100	59
113	0.23504	12 51	88	0.20468	196	61	0.05241	100	99
115	0.35917	174	66	0.26007	203	43	0.17871	100	48
121	0.29824	14 51	77	0.20370	196	62	0.10587	103	75
123	0.31644	12 205	75	0.21301	196	56	0.10904	100	74
125	0.06359	51	102	0.46693	196	20	1.00000	125	1
126	0.29205	12 174	81	0.22891	196	48	0.10139	100	77
127	1.00000	127	1	0.13583	196	95	0.08865	100	84
128	1.00000	128	1	0.13471	202	96	0.13552	103	61
130	1.00000	130	1	0.17482	196	75	0.07611	100	89
131	0.23051	174 205	90	0.32372	196	31	0.27757	100	32
132	0.51123	12 51 177	48	0.26697	196	39	0.13370	100	62

PART II: FURTHER EVALUATION OF EFFICIENCY OF MUTUAL FUND MANAGEMENT COMPANIES

133	0.33064	174	72	0.39194	196	25	0.25268	100	37
137	0.29792	174	78	0.27432	196	38	0.30851	100	27
139	0.28955	51	82	0.36027	196	28	0.48390	100	15
140	1.00000	140	1	0.18767	196	69	0.08654	100	85
142	0.29763	12 51	79	0.18135	196	70	0.08910	100	83
152	0.36580	174	65	1.00000	152	1	0.34260	100	20
154	1.00000	154	1	0.13960	196	94	0.05299	100	98
156	0.43289	174	58	0.27650	196	37	0.10983	100	73
159	0.32332	12 174	73	0.33236	196	30	0.09063	100	81
160	1.00000	160	1	0.26484	196	40	0.18193	100	45
161	1.00000	161	1	0.11123	196	98	0.09544	100	79
162	0.67058	55 140 177	33	0.21295	196	57	0.16337	100	50
163	1.00000	163	1	0.16022	196	87	0.05147	100	100
164	0.22963	205	91	0.76309	196	11	1.00000	164	1
168	0.26852	174	86	0.70646	196 202	13	0.16436	100	49
173	0.51246	12 177	47	0.16336	196	84	0.07258	100	92
174	1.00000	174	1	0.38481	196	26	0.31727	100	26
176	0.87968	140 177	27	0.17342	196	76	0.07287	100	91
177	1.00000	177	1	0.17241	196	77	0.12188	100	70
182	0.70413	12 51 163 177	32	0.20785	196	59	0.10220	100	76
185	0.28944	51 128	83	0.46887	202	19	0.13982	103	57
190	0.46243	12 51 205	54	0.22225	196	52	0.08337	100	86
191	0.52271	174	42	0.25457	196	44	0.15846	100	52
192	0.63720	161 174	35	0.17847	196	73	0.12837	100	67
193	0.56544	174 177	38	0.24157	196	46	0.06506	100	95
194	1.00000	194	1	0.31866	196	32	0.18850	100	44
195	0.39708	174 198	62	0.90278	152 196 202	9	0.21477	100	40
196	0.17049	205	97	1.00000	196	1	0.96884	100 125 164	10
197	0.50619	174	50	0.26409	196	41	0.13269	100	64
198	1.00000	198	1	0.60641	202	16	0.13823	100	58
200	0.33604	161	70	0.34650	196	29	1.00000	200	1
202	0.41224	177 205	60	1.00000	202	1	0.22387	100	39
203	0.07730	205	101	1.00000	203	1	1.00000	203	1
204	0.90701	14 174 177 205	24	0.21243	196	58	0.14578	100	56
205	1.00000	205	1	0.42132	203	23	1.00000	205	1
206	0.31628	174	76	0.43377	196	22	0.19629	100	42
207	0.19626	205	94	1.00000	207	1	0.06510	100	94
210	0.76209	174 205	30	0.38090	203	27	0.27657	100	33

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). VAR I corresponds to the efficiency score obtained by the Variation I maximizing the objective function for the reference set obtained for each SBM-inefficient company (equation II.2). Ref. corresponds to the reference set for each company obtained by the VAR I approach. Rank represents the ranking of the VAR I scores which have been sorted by value, where VAR I scores with value of 1.00000 obtain the highest ranking 1.

Table II-F-3 Variation I-efficiency under CRS for 2007

Stage 1 Portfolio Management				Stage 2 Marketing and Service			Overall Efficiency		
MFMC	VAR I	Ref.	Rank	VAR I	Ref.	Rank	VAR I	Ref.	Rank
2	0.63460	7 50 198	37	0.27159	50	57	0.04937	100	88
4	0.55677	12 14 127	44	0.23952	50	74	0.10488	103	54
6	0.54725	174 177	46	0.26111	46	61	0.12678	100	42
7	1.00000	7	1	0.30253	50	37	0.06810	100	75
9	0.27374	12 177	82	0.28129	46	45	0.25633	103	17
12	1.00000	12	1	0.02373	50	95	0.24853	103	19
14	1.00000	14	1	0.09046	50	94	0.29689	103	12
15	0.77309	12 127	27	0.18102	50	92	0.05225	103	86
20	0.29695	12 51	77	0.19869	50	90	0.21514	103	23
21	0.32596	12 51	72	0.27917	46	50	0.11364	103	49
24	0.39582	12 152	67	0.27726	46	53	0.05693	100	84
29	0.42171	12 50 198	65	0.26095	50	62	0.09364	100	58
31	0.50438	12 152	51	0.29044	46	42	0.08811	100	59
34	0.30280	12 50	76	0.43667	50	22	0.07355	100	70
35	0.82942	12 174 177	23	0.25775	46	66	0.06677	100	76
36	0.29045	174	81	0.46345	46	18	0.34604	100	11
37	0.51333	12 152 198	49	0.27020	46	58	0.11603	100	48
38	0.24290	51	85	0.27542	46	56	0.21607	100	22
40	0.29544	174	79	0.66181	50	8	0.24673	100	20
43	0.73357	50 128 198	30	0.25369	50	67	0.08387	100	63
45	0.72216	12 174 177	31	0.22965	46	79	0.26707	100	16
46	0.19931	174	89	1.00000	46	1	0.54995	100	8
47	0.46521	12 51	60	0.26043	50	64	0.10175	103	56
49	0.34824	198	70	0.50501	50	16	0.18861	100	29
50	1.00000	50	1	1.00000	50	1	0.19133	100	27
51	1.00000	51	1	0.22873	46	80	0.10175	100	55
53	1.00000	53	1	0.65008	46 100	9	0.22934	100	21
55	1.00000	55	1	0.18614	50	91	0.17281	103	32
57	0.34003	12 198	71	0.28414	46	44	0.16067	100	34
58	0.54674	12 127	47	0.21110	50	88	0.20325	103	25
59	0.08603	51	92	0.75383	100	7	0.02239	125	95
61	0.29518	14 51	80	0.26365	50	59	0.06150	100	81
62	0.71958	12 50 177	33	0.41194	50	24	0.13891	100	40
63	0.47037	12 51	57	0.22016	50	85	0.02930	103	94
69	0.47715	198	55	0.63894	50	10	0.12486	100	43
71	0.71594	14 177	34	0.27921	50	49	0.14463	100	37
75	0.46734	12 51 128	58	0.22553	50	83	0.05125	103	87
76	0.70046	14 174 177	35	0.26010	46	65	0.21237	100	24
78	1.00000	78	1	0.33522	50	31	0.08658	100	61
83	1.00000	83	1	0.23689	50	76	0.08579	103	62
84	0.59960	14 51 127	41	0.21743	50	86	0.40844	103	10
85	0.48836	12 51 128	54	0.17989	50	93	0.07036	103	73
86	0.72143	12 50 127	32	0.24629	50	70	0.13901	100	39
93	0.31287	51	73	0.62909	50	11	0.15701	100	35
95	0.58787	12 51	42	0.20359	50	89	0.27348	103	15
98	0.30854	12 50	75	0.27866	46	51	0.27933	100	14
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
102	1.00000	102	1	0.53741	46	14	1.00000	102	1
103	0.23013	12 51	86	0.33876	46	30	1.00000	103	1
105	0.08895	51	91	0.38316	46	28	0.28315	100	13
110	0.46244	12 51	61	0.27668	50	55	0.12210	100	45
113	0.22104	14 51	87	0.22404	50	84	0.04300	103	90
115	0.37316	174	68	0.40079	46	26	0.14418	100	38
121	0.29673	12 51	78	0.24177	50	72	0.08012	103	64
125	0.01458	51	94	0.52661	100	15	1.00000	125	1
126	0.61450	152 177 198	40	0.23754	46	75	0.07084	100	72
127	1.00000	127	1	0.24432	50	71	0.07826	100	67
128	1.00000	128	1	0.22772	50	81	0.05711	103	83
130	0.63996	12 50 128	36	0.23219	50	77	0.06568	100	79
131	0.20083	198	88	0.30184	46	38	0.18744	100	30
132	0.57854	12 51 177	43	0.21567	46	87	0.10706	100	53
133	0.47351	174	56	0.41140	46	25	0.20135	100	26
137	0.46543	12 174	59	0.27974	46	47	0.25393	100	18
139	0.35828	12 51	69	0.38813	46	27	0.42997	100	9

PART II: FURTHER EVALUATION OF EFFICIENCY OF MUTUAL FUND MANAGEMENT COMPANIES

140	1.00000	140	1	0.23008	50	78	0.06866	100	74
142	0.31110	12 51	74	0.24925	50	69	0.05337	103	85
152	1.00000	152	1	0.45273	46	20	0.18678	100	31
154	0.97396	12 51 128 140 198	22	0.22616	50	82	0.04190	100	91
156	0.52434	174 177	48	0.44584	50	21	0.09471	100	57
159	0.39990	152 198	66	0.42791	50	23	0.08677	100	60
160	1.00000	160	1	0.46731	50	17	0.10742	100	52
161	0.77077	14 55 127 177	28	0.32144	50	34	0.07592	100	68
162	0.61857	51 127	39	0.29792	50	39	0.11256	100	50
163	1.00000	163	1	0.24070	50	73	0.03624	100	93
168	0.50159	174 177	52	0.32660	46	33	0.13868	100	41
173	0.42885	12 50	64	0.27699	50	54	0.06195	100	80
174	1.00000	174	1	0.58505	50	12	0.16349	100	33
176	0.80657	12 51 140 177	25	0.30405	50	35	0.06578	100	78
177	1.00000	177	1	0.27741	46	52	0.07277	100	71
182	0.44739	12 51	62	0.28875	50	43	0.07833	100	65
185	0.27039	51 128	84	0.29669	46	40	0.04871	103	89
190	0.62888	12 50 198	38	0.27991	50	46	0.05825	100	82
191	0.73583	174 198	29	0.25283	46	68	0.12380	100	44
192	0.43269	177	63	1.00000	192	1	0.12072	100	46
193	0.77926	152 177 198	26	0.26127	50	60	0.03958	100	92
194	0.49623	12 198	53	0.27968	46	48	0.11228	100	51
196	0.06738	198	93	1.00000	196	1	1.00000	196	1
197	0.50973	198	50	0.29052	46	41	0.07830	100	66
198	1.00000	198	1	0.26090	46	63	0.06653	100	77
200	0.18292	51 127	90	0.30270	46	36	0.56729	103 125	7
203	0.01218	174	95	1.00000	203	1	1.00000	203	1
204	0.82784	12 50 177 198	24	0.45445	50	19	0.11611	100	47
206	0.55269	152	45	0.33323	46	32	0.14614	100	36
207	0.27342	174	83	0.38133	46	29	0.07377	100	69
210	1.00000	210	1	0.56271	50	13	0.18937	100	28

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). VAR I corresponds to the efficiency score obtained by the Variation I maximizing the objective function for the reference set obtained for each SBM-inefficient company (equation II.2). Ref. corresponds to the reference set for each company obtained by the VAR I approach. Rank represents the ranking of the VAR I scores which have been sorted by value, where VAR I scores with value of 1.00000 obtain the highest ranking 1.

Table II-F-4 Variation I-efficiency under CRS for 2008

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency		
	VAR I	Ref.	Rank	VAR I	Ref.	Rank	VAR I	Ref.	Rank
2	0.34102	14 174	43	0.07661	100	70	0.06936	100	71
4	0.50953	14 127	27	0.05026	100	88	0.12356	103	45
6	0.30824	174	48	0.09354	100	50	0.09404	46	59
7	0.65044	154 174	17	0.08171	100	64	0.08422	100	65
9	0.27604	14 174	53	0.17328	100	17	0.28667	46	13
12	1.00000	12	1	0.00580	100	93	0.36512	103	10
14	1.00000	14	1	0.02121	100	92	0.53627	46 103	7
15	1.00000	15	1	0.04369	100	91	0.06101	46	82
20	0.32022	14 127 174	47	0.07434	100	73	0.24916	46	23
21	0.30457	127	49	0.08726	100	56	0.10910	103	52
24	0.18848	174	72	0.09683	100	47	0.06890	100	72
29	0.16027	174	78	0.08021	100	67	0.05422	100	86
31	0.23317	174	61	0.10037	100	46	0.05397	100	87
34	0.17439	174	75	0.11094	100	39	0.05629	46	84
35	0.51037	14 174	26	0.10075	100	43	0.10524	100	56
36	0.12706	174	81	0.29986	100	5	0.37877	46	9
37	0.21857	174	67	0.10292	100	42	0.11621	100	48
38	0.19608	174	69	0.10059	100	45	0.09978	46	58
40	0.08362	174	88	0.22784	100	8	0.07030	46	70
43	0.25822	14 174	54	0.08061	100	66	0.06407	100	79
45	0.33274	14 174	46	0.11144	100	37	0.26292	100	19
46	0.03121	210	91	1.00000	46	1	1.00000	46	1
47	0.46085	14 154	36	0.08297	100	62	0.08838	46	62
49	0.17135	174	76	0.13015	100	32	0.16023	100	35
50	0.20987	174	68	0.21096	100	14	0.28164	100	17
51	0.56217	127 174	22	0.10495	100	40	0.13157	100	43
53	0.21879	210 217	66	0.22162	100	10	0.40558	100	8
55	1.00000	55	1	0.05008	100	89	0.22664	46	26
57	0.24432	174	56	0.13549	100	29	0.11501	100	49
58	0.48977	12 127	31	0.04599	100	90	0.28277	103	16
61	0.24133	14 174	57	0.08615	100	59	0.06024	46	83
62	0.44015	14 174	37	0.11135	100	38	0.14484	100	38
63	0.61609	14 154	21	0.06001	100	82	0.02524	46	91
69	0.19360	174	71	0.21138	100	12	0.23610	100	25
71	0.46351	14 174	34	0.07507	100	71	0.13102	46	44
76	0.76308	14 55 127 174	14	0.07948	100	69	0.13805	100	41
78	0.35198	174	42	0.09341	100	51	0.09133	100	61
83	1.00000	83	1	0.05749	100	84	0.06880	46	73
84	0.47955	14 127	32	0.05633	100	85	0.26264	46	20
85	0.68399	12 127	16	0.05201	100	86	0.08376	46	66
86	0.47427	14 154	33	0.06724	100	77	0.06671	46	77
93	0.18364	174	74	0.18624	100	15	0.18065	46	31
95	0.64537	14 154	18	0.06332	100	80	0.22117	46	27
98	0.10369	174	87	0.08690	100	58	0.20722	46	29
100	0.01982	210	92	1.00000	100	1	1.00000	100	1
103	0.11734	14 174	82	0.14642	100	24	1.00000	103	1
105	0.08180	174	89	0.15219	100	22	0.00449	46	93
110	0.50199	154	28	0.07954	100	68	0.11121	46	51
113	0.22532	14 174	63	0.07449	100	72	0.07830	46	67
115	0.16674	174	77	0.16853	100	19	0.19021	100	30
121	0.24507	14 174	55	0.08205	100	63	0.07646	46	68
125	0.01161	174	93	0.32678	100	4	1.00000	125	1
126	0.28950	174	50	0.09261	100	52	0.09270	100	60
127	1.00000	127	1	0.05888	100	83	0.06516	100	78
128	1.00000	128	1	0.05170	100	87	0.06187	46	81
130	0.64198	14 154	19	0.06347	100	79	0.05462	100	85
131	0.11549	174	83	0.17022	100	18	0.21404	100	28
132	0.15380	174	79	0.10070	100	44	0.10632	46	55
133	0.23824	174	59	0.22745	100	9	0.28295	100	15
137	0.23982	174	58	0.16323	100	20	0.28322	100	14
139	0.19445	154 174	70	0.21125	100	13	0.35653	46	11
140	0.69896	154	15	0.06216	100	81	0.06238	100	80
142	0.42401	14 154 174	38	0.07049	100	75	0.14721	46	37
152	0.23478	174 217	60	0.21544	100	11	0.24542	100	24

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154	1.00000	154	1	0.07043	100	76	0.05124	100	88
156	0.27965	174	52	0.13949	100	27	0.13365	100	42
159	0.18658	174	73	0.12133	100	33	0.10815	100	53
160	1.00000	160	1	0.14901	100	23	0.16151	100	34
161	0.49258	14 55 174	30	0.08691	100	57	0.07374	46	69
162	0.35985	14 174	41	0.08763	100	55	0.10303	46	57
163	0.49559	154 174	29	0.06445	100	78	0.05120	100	89
168	0.22249	174	65	0.13937	100	28	0.16156	100	33
173	0.10768	174	85	0.08971	100	54	0.04383	46	90
174	1.00000	174	1	0.14526	100	25	0.25869	100	22
176	0.63727	154 174	20	0.08379	100	61	0.08507	100	64
177	0.46210	174	35	0.09069	100	53	0.10718	100	54
182	0.33530	14 174	44	0.08063	100	65	0.06770	46	75
185	0.55629	12 127 160	23	0.09539	100	49	0.01699	46	92
190	0.53469	154 174	25	0.07405	100	74	0.06689	100	76
191	0.39324	174	39	0.13382	100	30	0.17292	100	32
192	0.38279	174	40	0.09595	100	48	0.12119	46	47
193	0.28315	174	51	0.08454	100	60	0.06802	100	74
194	0.23248	174	62	0.11563	100	35	0.08797	100	63
196	0.05126	174	90	0.56400	100	3	0.56694	46	6
197	0.33478	174	45	0.11451	100	36	0.13962	100	39
198	1.00000	198	1	0.10335	100	41	0.13824	100	40
200	0.11153	14 174	84	0.18564	100	16	0.34346	103	12
203	0.10584	174	86	0.22949	100	7	0.26175	100	21
204	0.53757	14 174	24	0.12076	100	34	0.11325	100	50
206	0.22317	174	64	0.13368	100	31	0.15593	100	36
207	0.13731	174	80	0.16322	100	21	0.12223	100	46
210	1.00000	210	1	0.14436	100	26	0.26583	100	18
217	1.00000	217	1	0.27604	100	6	1.00000	217	1

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). VAR I corresponds to the efficiency score obtained by the Variation I maximizing the objective function for the reference set obtained for each SBM-inefficient company (equation II.2). Ref. corresponds to the reference set for each company obtained by the VAR I approach. Rank represents the ranking of the VAR I scores which have been sorted by value, where VAR I scores with value of 1.00000 obtain the highest ranking 1.

Table II-F-5 Variation I-efficiency under CRS for 2009

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency		
	VAR I	Ref.	Rank	VAR I	Ref.	Rank	VAR I	Ref.	Rank
2	0.41886	7 76	56	0.14137	196	76	0.06684	100	74
4	1.00000	4	1	0.09835	196	90	0.07747	100	64
6	0.41433	50 76	59	0.21464	196	34	0.09868	100	55
7	1.00000	7	1	0.15481	196	66	0.08650	100	59
9	0.23876	76	85	0.24172	196	23	0.23953	100	19
12	0.66097	14 46	29	0.00009	196	94	0.00002	100	95
14	1.00000	14	1	0.00003	196	95	0.24104	100	18
15	0.45053	14	50	1.00000	15	1	0.05933	100	83
20	0.30511	14 51	74	0.18649	196	49	0.23018	100	20
21	0.24252	14 46	83	0.19100	196	45	0.11611	100	49
24	0.34371	46 76	64	0.14132	196	77	0.06891	100	72
29	0.42025	45 76 131	54	0.14545	196	71	0.07943	100	62
31	0.66870	7 51	27	0.29884	196	18	0.07714	100	65
34	0.15321	76	89	0.15801	196	64	0.08981	100	57
35	0.75657	76 210	25	0.15662	196	65	0.10771	100	52
36	0.15772	46 76	87	0.44315	196	11	0.31329	100	14
37	0.36394	46 76	62	0.17962	196	51	0.11910	100	48
38	0.12349	14 50	93	0.14600	196	70	0.06778	100	73
40	0.56056	45 46 76 100	34	0.34411	196	15	0.29908	100	15
43	0.64809	7 14 128	31	0.18516	196	50	0.07326	100	68
45	1.00000	45	1	0.19426	196	43	1.00000	45	1
46	1.00000	46	1	1.00000	46	1	0.60339	100	8
47	0.44254	76	52	0.17026	196	53	0.08880	100	58
49	0.38321	46 76	61	0.21968	196	30	0.14272	100	37
50	1.00000	50	1	0.40832	196	13	0.26275	100	16
51	1.00000	51	1	0.23329	196	26	0.13888	100	38
53	0.05305	45	95	1.00000	53	1	0.35490	100	11
55	1.00000	55	1	0.11516	196	89	0.12805	100	42
57	0.55684	46 76 210	35	0.21488	196	33	0.18196	100	27
58	0.41861	14 51	58	0.13294	196	84	0.16127	100	33
61	0.32884	50 76	66	0.13415	196	82	0.00117	100	94
62	1.00000	62	1	0.21576	196	32	0.17493	100	28
63	0.50707	14 51	43	0.09070	196	91	0.01846	100	93
69	0.27015	45 46	78	0.91835	46 53 196	10	0.32852	100	13
71	0.87395	7 14 50 76	21	0.15199	196	68	0.12169	100	46
76	1.00000	76	1	0.16795	196	54	0.12508	100	44
78	0.48283	46 76	46	0.19213	196	44	0.14369	100	36
83	0.41918	14 76	55	0.12291	196	87	0.04583	100	90
84	0.48720	14 51	44	0.19749	196	40	0.17140	100	30
85	0.54846	14 51	37	0.08841	196	92	0.07218	100	69
86	0.51972	14 51	40	0.17899	196	52	0.06392	100	77
93	0.32194	14 50	68	0.29418	196	20	0.19697	100	26
95	0.81564	7 14 140	23	0.14857	196	69	0.06981	100	70
98	0.24244	7	84	0.22425	196	29	0.09003	100	56
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
103	0.18388	76	86	0.15821	196	63	1.00000	103	1
105	0.14898	14 46	90	0.21197	196	35	0.04605	100	89
110	0.58006	7	32	0.14326	196	73	0.07889	100	63
113	0.31340	14 51	71	0.13521	196	81	0.04774	100	87
115	0.25544	50 76 100	81	0.32520	196	16	0.17193	100	29
121	0.25098	14 50	82	0.14466	196	72	0.06331	100	79
125	0.13837	46	91	0.97340	46 53 196	8	1.00000	125	1
126	0.33415	46 76	65	0.16061	196	61	0.12112	100	47
127	1.00000	127	1	0.14041	196	78	0.06583	100	75
128	1.00000	128	1	0.02592	196	93	0.02722	100	92
130	0.86584	7 14 128 140	22	0.15991	196	62	0.05414	100	84
131	1.00000	131	1	0.18828	196	48	0.84278	45 100 217	6
132	0.25683	76	80	0.16460	196	59	0.06399	100	76
133	0.45421	45 46 76	49	0.30729	196	17	0.25524	100	17
137	0.28537	46 76	77	0.20871	196	37	0.21064	100	23
139	0.48051	46 76 210	47	0.24574	196	22	0.34998	100	12

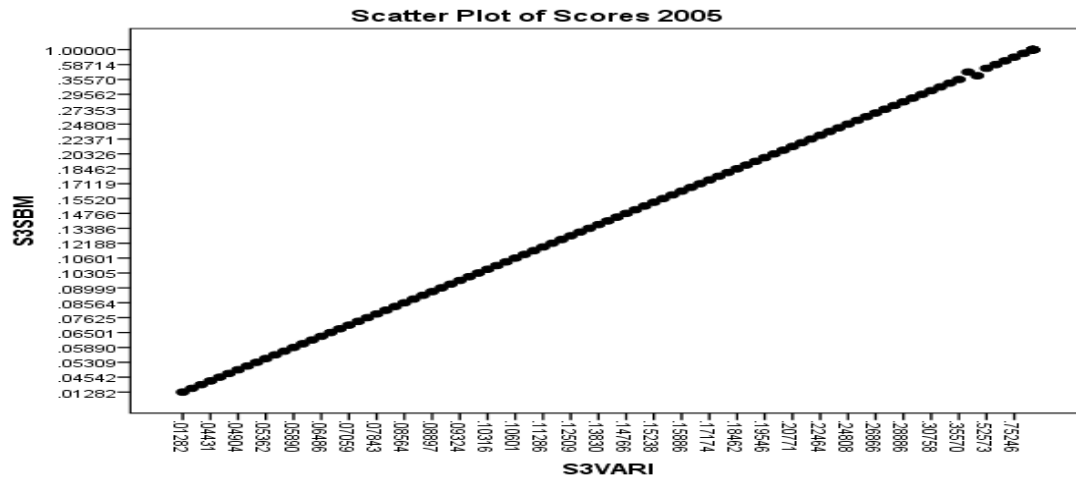
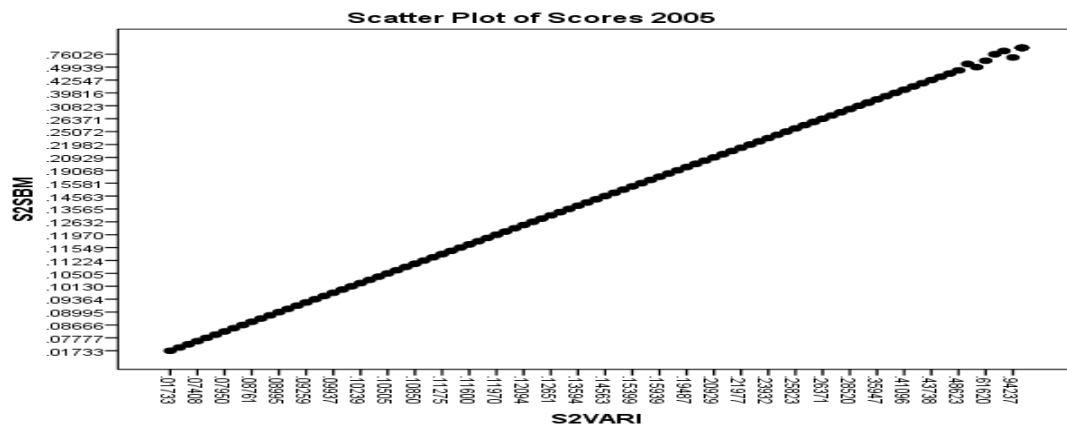
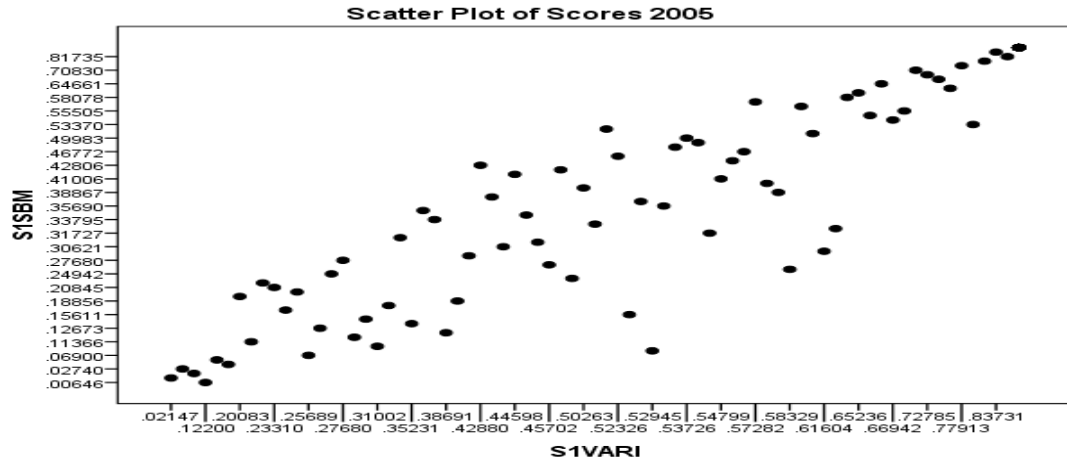
PART II: FURTHER EVALUATION OF EFFICIENCY OF MUTUAL FUND MANAGEMENT COMPANIES

140	1.00000	140	1	0.13995	196	79	0.04534	100	91
142	0.41864	14 51	57	0.12292	196	86	0.06106	100	82
152	1.00000	152	1	0.43377	196	12	0.37313	100	10
154	0.66777	14 46 76	28	0.15444	196	67	0.06351	100	78
156	0.54375	50 76	38	0.18926	196	46	0.12481	100	45
159	0.32656	46 76	67	0.18860	196	47	0.11241	100	50
160	0.31841	46 76	69	0.21135	196	36	0.13288	100	41
161	0.56409	76	33	0.14168	196	75	0.06268	100	80
162	0.51520	46 76	41	0.12123	196	88	0.07697	100	66
163	0.47683	7 14	48	0.13196	196	85	0.05326	100	85
168	0.26591	46 76	79	0.20162	196	38	0.16093	100	34
173	0.52658	14 51	39	0.16598	196	57	0.04681	100	88
174	1.00000	174	1	0.28618	196	21	0.22081	100	22
176	0.76119	14 51 76	24	0.16460	196	58	0.07694	100	67
177	0.31780	50 76	70	0.23625	196	25	0.12704	100	43
182	0.42454	50 76	53	0.14183	196	74	0.06908	100	71
185	0.44913	14 51	51	0.13941	196	80	0.04822	100	86
190	0.69864	7	26	0.13383	196	83	0.06186	100	81
191	0.65853	46 76 174	30	0.21636	196	31	0.22495	100	21
192	0.38878	50 76	60	0.19708	196	41	0.10456	100	54
193	1.00000	193	1	0.16678	196	56	0.08055	100	61
194	0.54980	76 210	36	0.16364	196	60	0.10909	100	51
195	0.48317	76 210	45	0.16712	196	55	0.08512	100	60
196	0.29905	46 76	75	1.00000	196	1	0.53158	100	9
197	0.50977	174	42	0.24011	196	24	0.15332	100	35
198	0.30795	14 50 76	72	0.20008	196	39	0.10700	100	53
200	0.11656	14 46	94	0.29435	196	19	0.16554	100	32
203	0.12525	45 46 76	92	0.37516	196	14	0.20074	100	25
204	0.34667	46 76	63	0.22747	196	28	0.13595	100	39
206	0.29661	45 46 76	76	0.23200	196	27	0.16866	100	31
207	0.15729	46 76	88	1.00000	207	1	0.13347	100	40
210	1.00000	210	1	0.19581	196	42	0.20425	100	24
217	1.00000	217	1	0.92002	46 53	9	1.00000	217	1
221	0.30515	50 76	73	1.00000	221	1	0.74804	100 103	7

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). VAR I corresponds to the efficiency score obtained by the Variation I maximizing the objective function for the reference set obtained for each SBM-inefficient company (equation II.2). Ref. corresponds to the reference set for each company obtained by the VAR I approach. Rank represents the ranking of the VAR I scores which have been sorted by value, where VAR I scores with value of 1.00000 obtain the highest ranking 1.

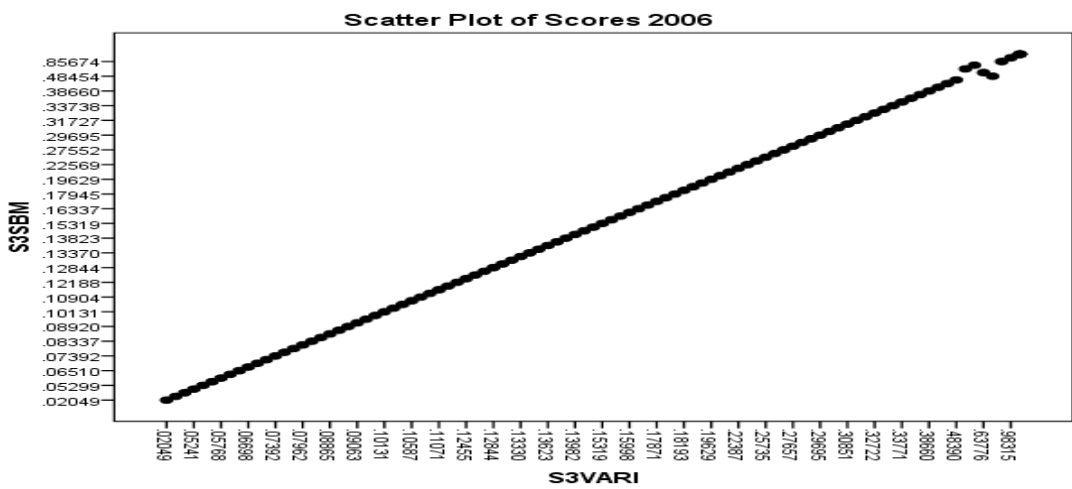
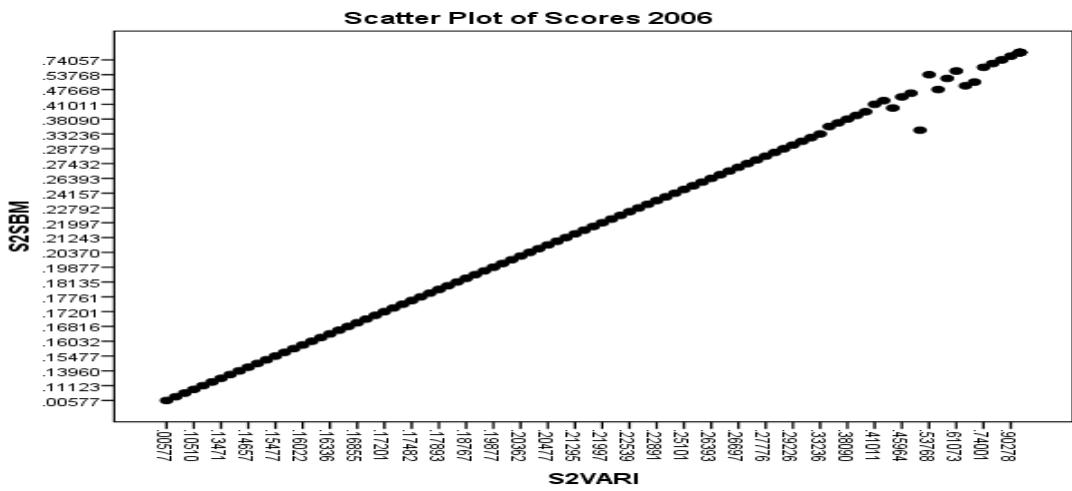
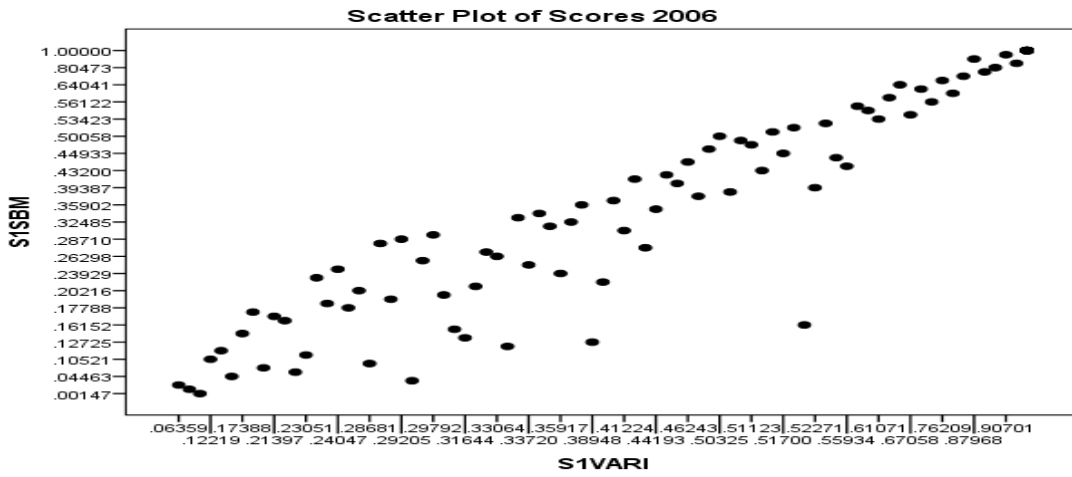
Appendix G

Figure II-G-1 Comparative graphs between SBM and Variation I scores for 2005



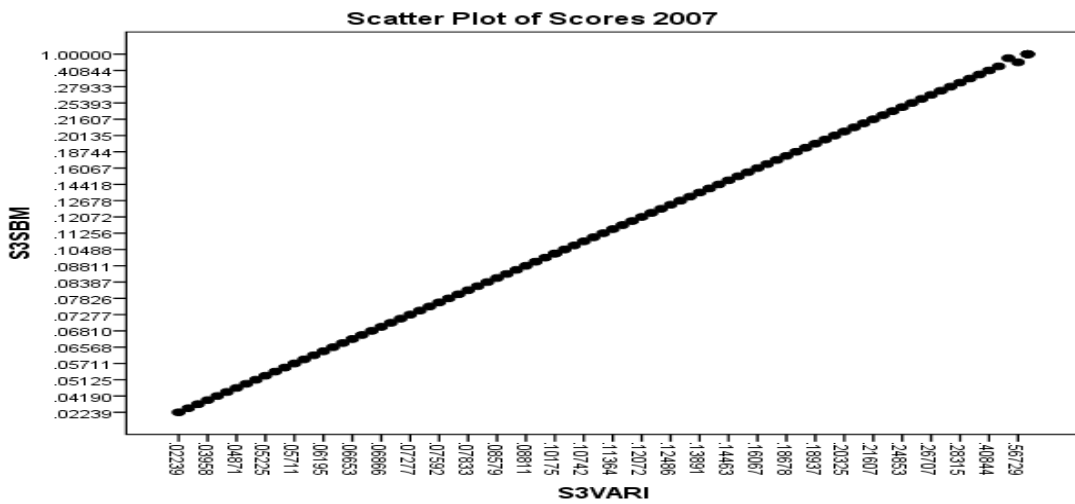
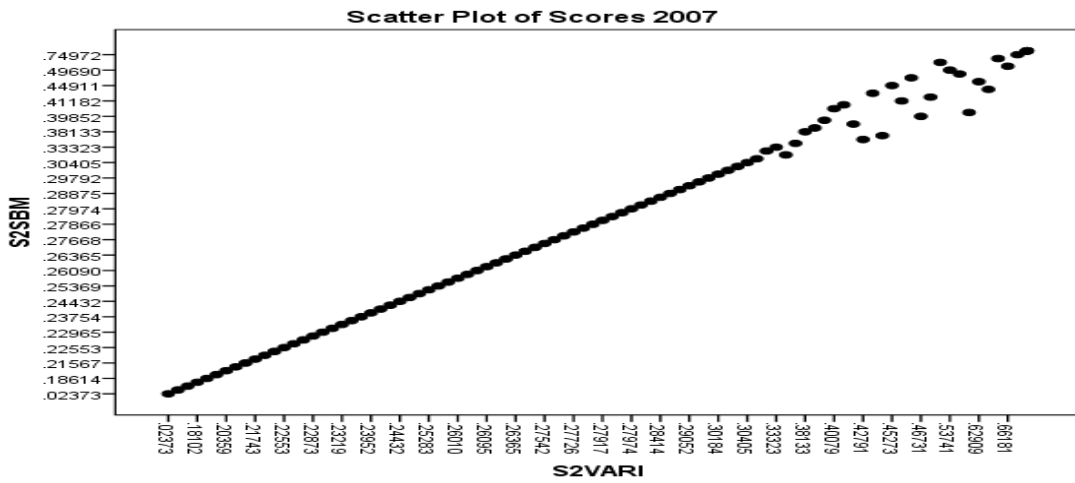
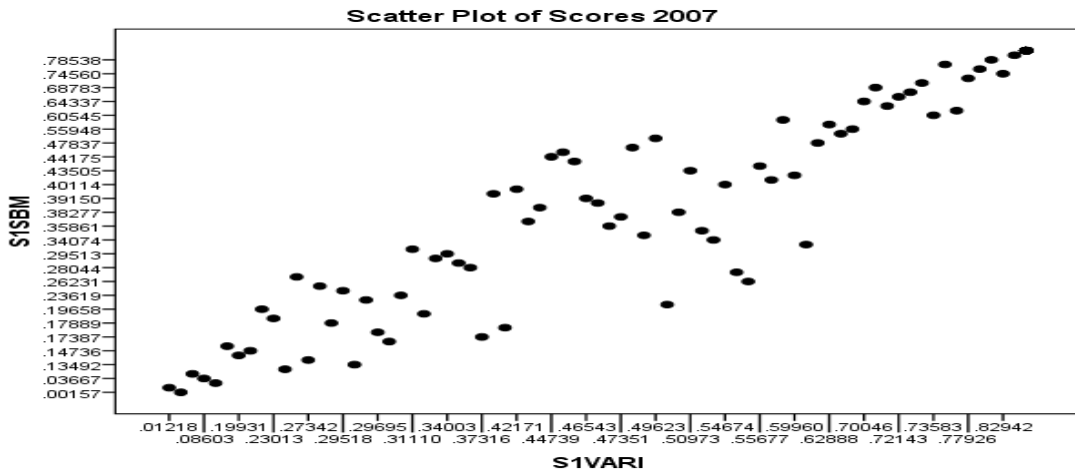
S#SBM = efficiency scores obtained by the CRS-SBM for Stage # of the multi-manag. stage model
 S#VARI = efficiency scores obtained by the CRS-Var. I for Stage # of the multi-manag. stage model

Figure II-G-2 Comparative graphs between SBM and Variation I scores for 2006



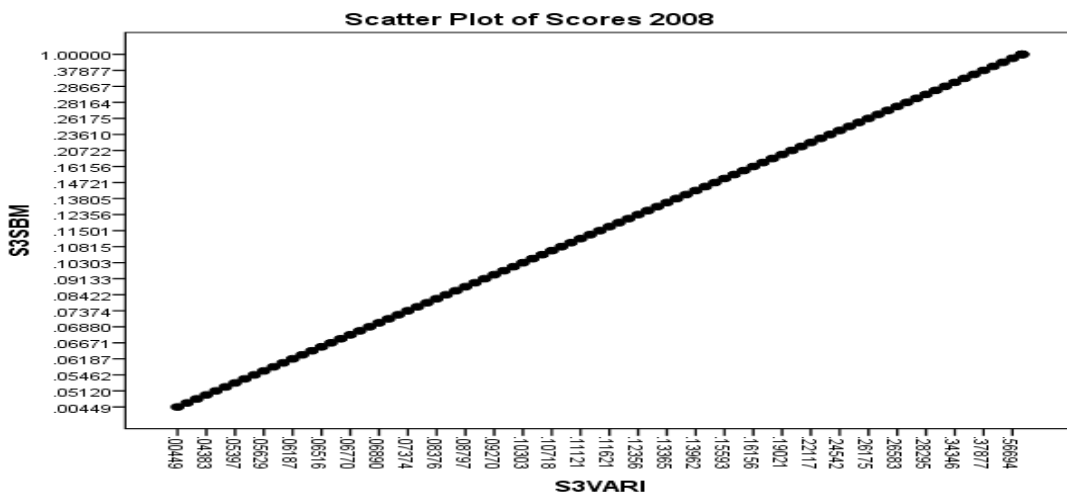
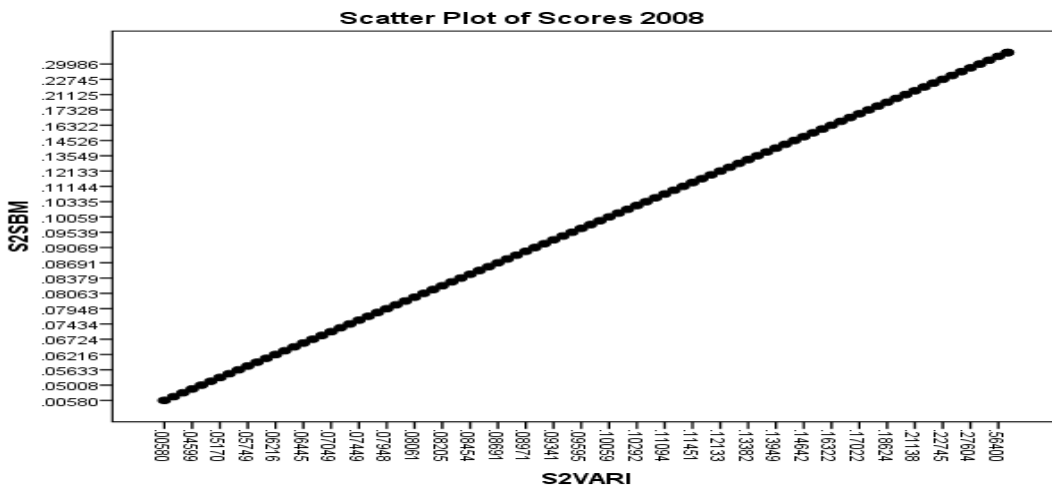
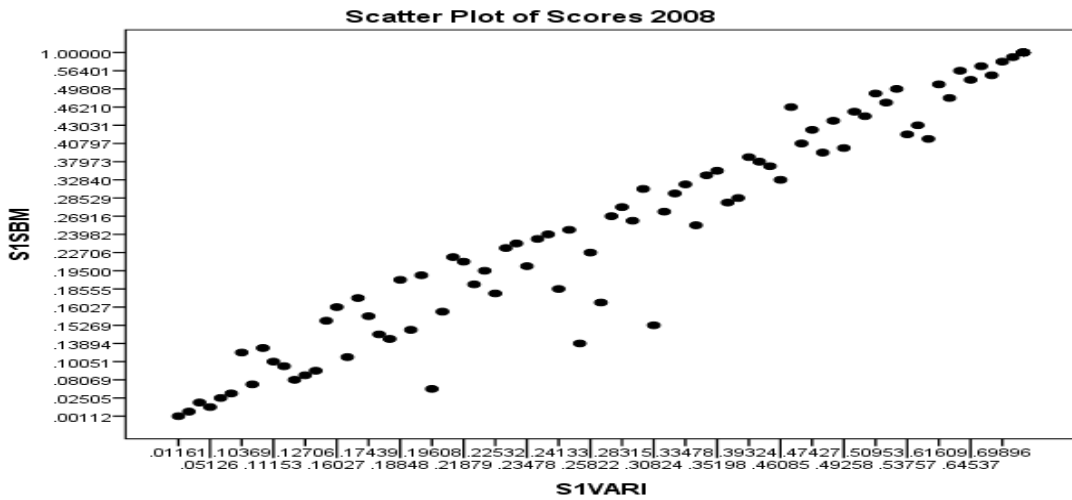
S#SBM = efficiency scores obtained by the CRS-SBM for Stage # of the multi-manag. stage model
 S#VARI = efficiency scores obtained by the CRS-Var. I for Stage # of the multi-manag. stage model

Figure II-G-3 Comparative graphs between SBM and Variation I scores for 2007



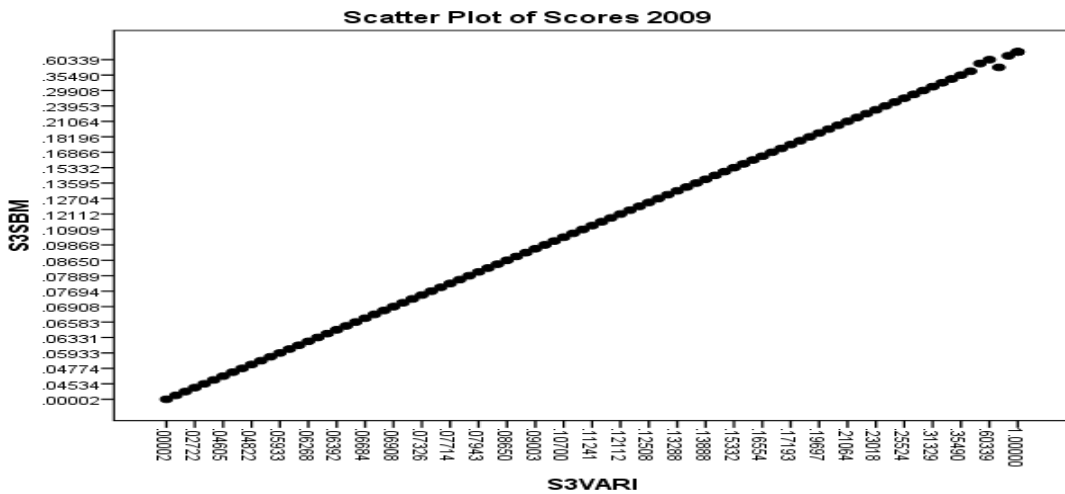
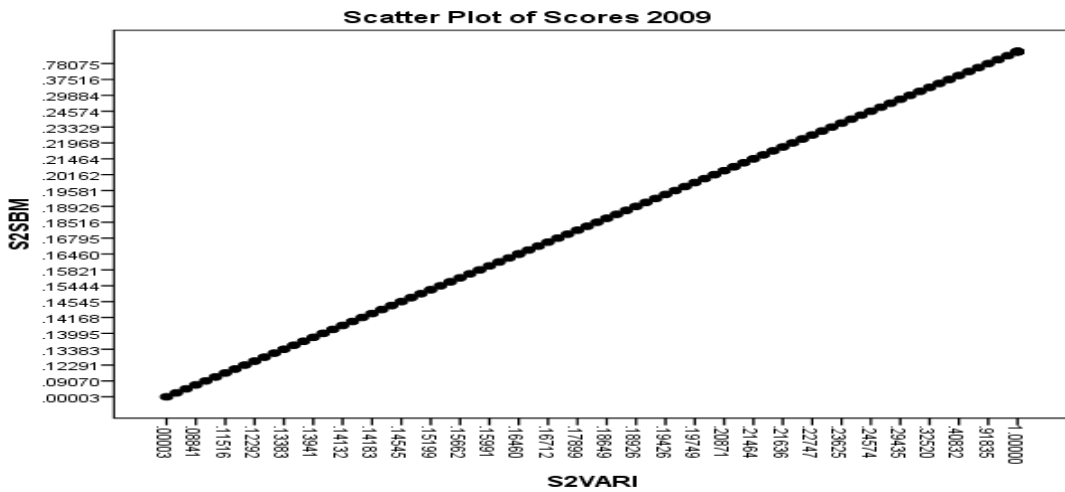
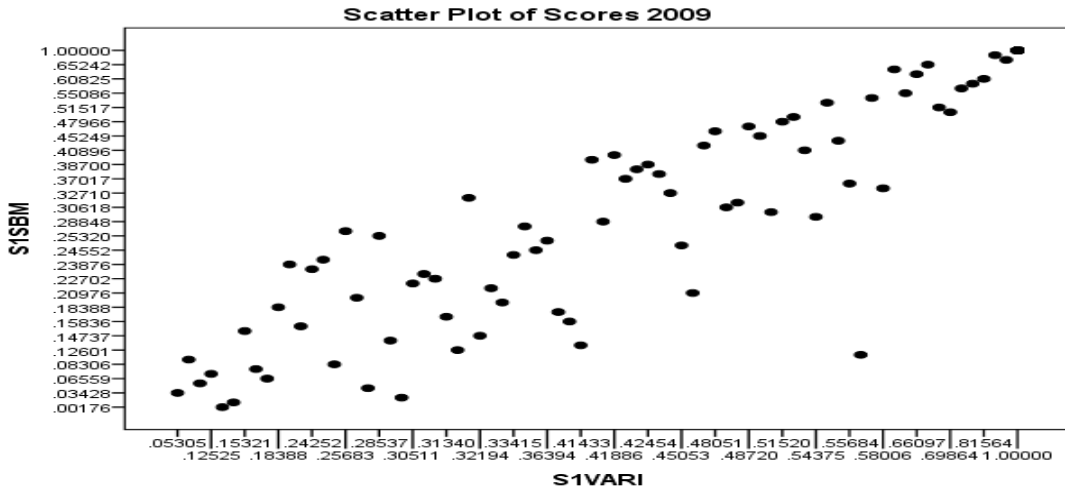
S#SBM = efficiency scores obtained by the CRS-SBM for Stage # of the multi-manag. stage model
 S#VARI = efficiency scores obtained by the CRS-Var. I for Stage # of the multi-manag. stage model

Figure II-G-4 Comparative graphs between SBM and Variation I scores for 2008



S#SBM = efficiency scores obtained by the CRS-SBM for Stage # of the multi-manag. stage model
 S#VARI = efficiency scores obtained by the CRS-Var. I for Stage # of the multi-manag. stage model

Figure II-G-5 Comparative graphs between SBM and Variation I scores for 2009



S#SBM = efficiency scores obtained by the CRS-SBM for Stage # of the multi-manag. stage model
 S#VARI = efficiency scores obtained by the CRS-Var. I for Stage # of the multi-manag. stage model

Appendix H

Code inside are the identifying number of the mutual fund management company (Official registers, CNMV).

Table II-H-1 Maximal Friends found for 2005

Stage 1	Stage 2	Overall Eff.
6 40 160 161	40 46 59	12 14 200
53 160 174 177	40 100 196	12 14 125
55 76 100 174	40 46 196	14 125 200
12 100 152 174	46 59 205	100 152 205
7 154 160 163 176	46 138 205	40 100 200
7 160 161 163 176	46 138 196	100 196 200
7 154 160 161 176		40 103 125 200
7 154 161 163 176		40 125 196 200
12 71 161 176 177		
12 55 128 154 161		
12 55 154 176 177		
12 128 130 154 161		
12 40 43 174 177		
12 55 154 161 176		
12 55 161 176 177		
12 40 43 160 177		
12 40 160 174 177		
12 40 43 174 205		
6 40 161 174 192		
43 163 174 177 205		
40 161 174 177 192		
55 127 128 154 161		
12 55 71 161 177		
55 140 154 176 177		
55 76 161 177 193		
55 76 161 174 193		
55 76 174 177 193		
55 161 174 177 193		
6 161 174 177 192		
55 76 161 174 177		
12 14 55 100 174		
12 152 174 198 205		
12 14 103 160 161		
12 128 154 160 161		
12 43 128 154 160		
12 40 152 174 205		
76 161 174 177 193		
43 163 177 198 205		
40 43 160 174 177		
163 174 177 198 205		
12 43 160 163 176 177		
12 154 160 161 163 176		
12 43 154 160 163 176		
7 12 154 160 161 163		
12 43 154 163 176 177		
12 43 163 174 177 198		
12 14 55 161 174 177		
12 14 40 55 161 174		
12 40 55 161 174 177		
12 14 55 127 128 161		
12 43 163 174 198 205		
14 40 55 76 161 174		
55 127 140 154 161 176		
7 12 130 154 161 163		
12 40 160 161 176 177		

Table II-H-2 Maximal Friends found for 2006

Stage 1	Stage 2	Overall Eff.
12 100 174 205	60 196 203	100 203
12 14 100 174	60 152 196	100 205
14 55 100 174	152 196 202	14 103 125
12 14 71 174 177	40 100 196 203 207	100 125 164
12 14 55 127 128		103 125 164 200
12 14 174 177 205		100 103 164 200
12 14 51 127 128		
12 51 59 127 128		
12 51 59 128 160		
51 62 130 160 163		
12 14 51 59 127		
53 163 177 198 205		
128 154 177 198 205		
12 174 194 198 205		
51 62 140 161 177		
51 127 130 140 163		
51 62 140 163 177		
14 71 161 174 177		
62 127 130 140 163		
51 62 127 140 163		
51 62 127 130 163		
53 174 177 198 205		
128 174 177 198 205		
53 163 174 198 205		
12 51 55 127 128 140		
12 14 55 140 174 177		
12 128 140 154 174 198		
12 128 140 174 177 205		
12 163 174 177 198 205		
12 128 154 174 198 205		
12 140 154 174 177 198		
12 128 154 163 198 205		
12 154 174 177 198 205		
12 14 51 55 127 140		
12 14 51 55 140 177		
12 51 55 140 177 205		
12 51 53 154 163 177		
12 51 53 128 154 163		
12 51 128 140 154 205		
12 128 140 154 177 205		
12 51 53 154 177 205		
12 14 51 55 177 205		
14 51 55 161 174 177		
12 51 53 128 160 163		
12 53 163 174 177 205		
12 51 140 154 177 205		
12 53 128 160 163 205		
12 14 51 140 161 177		
12 14 51 127 140 161		
12 51 53 128 154 205		
12 51 128 130 154 163		
12 51 71 140 163 177		
12 53 128 154 163 205		
12 51 140 154 163 177		
12 51 128 130 140 154		
12 51 128 130 160 163		
12 51 71 140 161 177		
12 154 163 177 198 205		
12 53 154 163 177 205		
12 51 127 128 130 140		
12 128 140 154 174 177		
14 51 55 140 161 177		
14 51 55 174 177 205		
12 128 154 174 177 205		
14 51 55 127 140 161		
14 55 140 161 174 177		
12 51 71 140 161 163		

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12 14 71 140 161 177
 12 140 154 163 177 198
 12 51 130 140 154 163
 51 55 127 140 161 177
 12 51 127 130 140 161
 12 128 140 154 163 198
 12 128 130 140 154 163
 12 51 130 140 161 163
 51 128 130 140 154 163
 51 128 140 154 177 205
 128 140 154 174 177 198
 12 51 53 128 160 205
 51 62 127 130 140 161
 51 62 130 140 161 163

Table II-H-3 Maximal Friends found for 2007

Stage 1	Stage 2	Overall Eff.
12 14 100 174	46 50	100 102
50 51 128 160	50 192	103 125 196
50 128 160 198	46 100 196	100 103 196
12 100 174 210	100 196 203	100 125 203
100 102 174 210		100 125 196
100 174 198 210		
51 127 128 140 198		
7 12 128 163 198		
7 50 78 163 198		
7 12 78 163 198		
12 50 53 152 198		
12 53 102 152 198		
12 102 152 174 198		
12 102 174 198 210		
102 152 174 198 210		
53 102 152 198 210		
12 51 127 140 174 177		
7 12 50 128 140 198		
7 12 50 140 177 198		
12 50 51 128 140 198		
12 51 140 174 177 198		
7 12 50 78 177 198		
12 51 128 140 174 198		
12 14 51 127 174 177		
12 50 51 140 177 198		
7 12 50 127 128 140		
12 14 50 51 127 177		
12 50 51 127 140 177		
12 50 51 127 128 140		
12 14 83 127 174 177		
12 14 51 83 127 174		
14 51 55 127 174 177		
14 51 55 83 127 174		
12 14 51 152 174 177		
12 50 51 152 177 198		
12 51 152 174 177 198		
12 51 55 83 127 128		
14 55 83 127 174 177		
12 14 51 55 83 127		
51 127 140 174 177 198		

Table II-H-4 Maximal Friends found for 2008

Stage 1	Stage 2	Overall Eff.
14 55 127 174	46 100	100 217
15 198 210 217		46 103 125
55 83 127 174		46 100 125
15 174 198 210		
15 174 210 217		
15 174 198 217		
14 15 154 160 174		
15 128 154 174 198		
15 128 154 160 174		
12 14 127 154 160		
14 127 154 160 174		
127 128 154 160 174		
12 14 15 128 154 160		
12 15 127 128 154 160		
12 14 15 127 128 154		

Table II-H-5 Maximal Friends found for 2009

Stage 1	Stage 2	Overall Eff.
4 14 127	15 196 207	45 100 217
14 45 46 76	46 53 196	100 103 125
14 45 76 100	100 196 221	
45 174 193 217	46 100 196	
45 100 152 174		
46 50 100 152		
46 50 152 174		
46 100 152 174		
50 100 152 174		
174 193 210 217		
7 46 76 128 210		
7 62 76 193 210		
7 76 128 193 210		
14 51 127 128 140		
7 14 51 76 193		
7 76 128 140 193		
14 46 50 76 100		
14 50 76 100 174		
45 46 76 131 210		
45 76 131 193 210		
46 76 128 131 210		
45 46 76 174 210		
45 46 152 174 217		
7 46 50 51 62		
7 46 50 76 210		
50 62 76 193 210		
45 76 174 193 210		
7 50 62 76 193		
7 50 51 62 193		
7 50 76 193 210		
45 46 131 210 217		
45 131 193 210 217		
50 76 174 193 210		
7 50 62 193 210		
45 46 100 152 217		
50 51 76 174 193		
7 46 50 62 210		
45 46 174 210 217		
46 50 76 174 210		
7 14 46 76 128 140		
7 14 46 51 128 140		
7 14 50 51 76 140		
7 14 46 50 76 140		
7 14 46 50 51 140		
14 46 50 55 76 140		
7 50 51 76 140 193		
45 46 50 76 100 174		

Appendix I

Table II-I-1 Variation II-efficiency under CRS for 2005

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency		
	VAR II	Ref.	Rank	VAR II	Ref.	Rank	VAR II	Ref.	Rank
1	0.62660	7 12 130 163	83	0.27981	40 59	86	0.35942	12 125	96
2	0.75563	43 163 198 205	52	0.31795	40 59	63	0.52468	12 125	77
4	0.66861	12 161 163 176	74	0.22070	138 205	97	0.53497	12 125	72
6	1.00000	6	1	0.34996	40 59	53	0.41680	12 125	94
7	1.00000	7	1	0.30571	40 59	68	0.56110	14 125 200	53
9	0.63778	14 40 76	79	0.42290	138	43	0.72773	14 125 200	19
12	1.00000	12	1	0.03048	59 205	102	1.00000	12	1
14	1.00000	14	1	0.15673	59 205	100	1.00000	14	1
15	0.82141	14 55 128	41	0.26393	138 196	91	0.36761	12	95
20	0.72579	127 128	64	0.27083	40 59	88	0.67038	14 125 200	27
21	0.50003	103	89	0.32631	40 59	60	0.65387	12 14 125	30
24	0.75395	12 163 198 205	54	0.29171	40 59	78	0.50810	12 125	81
29	0.71638	43 163 198 205	66	0.31670	40 59	64	0.53732	14 125 200	71
31	0.86055	12 43 163 174 205	36	0.33585	40 59	58	0.55487	14 125 200	56
34	0.55701	6 40 161 192	85	0.35654	40 59	51	0.51384	12 125	78
35	0.94340	40 160 176	30	0.34986	40 59	54	0.55032	14 125 200	61
36	0.70789	6 40 174	68	0.74505	40 46 59	14	0.71421	125 196 200	21
37	0.75725	40 160 174	51	0.54733	40 59	31	0.58279	125 196 200	43
38	0.44163	6 161 177	96	0.33099	40 59	59	0.58539	40 103 125	41
40	1.00000	40	1	1.00000	40	1	1.00000	40	1
42	0.55395	6 40 174	86	0.76168	59 205	13	0.70030	125 196 200	23
43	1.00000	43	1	0.30158	40 59	71	0.53175	14 125 200	73
45	0.83581	40 161 174 177 192	38	0.37312	40 59	48	0.61285	100 196 200	36
46	0.40978	76 100 174	97	1.00000	46	1	0.89171	100 196 200	12
47	0.72792	12 163 177	62	0.30867	138 196	67	0.48598	12 125	85
49	0.82668	40 160 174	40	0.56928	40 59	29	0.60595	125 196 200	39
50	0.47064	53 160 177	92	0.84876	59 205	11	0.62303	125 196 200	33
51	0.73582	40 160 176	61	0.41680	40 59	46	0.56549	14 125 200	51
53	1.00000	53	1	0.85623	46 138 205	10	0.75751	100 152 205	18
55	1.00000	55	1	0.14995	100 196	101	0.71366	14 125 200	22
57	0.76376	40 160 174	49	0.57301	40 59	28	0.57366	125 196 200	49
58	0.91842	14 127 128	32	0.20126	59 205	99	0.61374	12 125	34
59	0.30026	163 177 198 205	99	1.00000	59	1	0.51300	100 152	79
60	0.26563	100 152 174	101	0.83982	138 205	12	0.59208	100 200	40
61	0.63330	12 43 163 177	81	0.28864	40 59	82	0.33033	12 125	99
62	0.75193	7 160 176	55	0.36214	40 59	50	0.60902	14 125 200	38
63	0.90147	12 154 163	33	0.28119	138 196	85	0.19893	12	101
69	0.46557	53 174 177	93	0.58979	40 46 196	26	0.55364	125 196 200	58
71	1.00000	71	1	0.28994	138 196	80	0.66885	12 125	28
73	0.92704	12 40 43 174 177	31	0.43503	40 59	42	0.44003	100 200	91
75	0.74529	127 128	56	0.25971	40 59	93	0.30949	12	100
76	1.00000	76	1	0.30367	40 59	70	0.68595	12 14 200	25
78	0.75539	12 40 160 174	53	0.46455	40 59	39	0.55339	125 196 200	60
83	0.85938	127 128	37	0.25938	138 196	94	0.44032	12 125	90
84	0.68312	12 55 154 177	72	0.26262	138 196	92	0.79814	14 125 200	15
85	0.73626	14 55 128	60	0.22043	100 196	98	0.52740	12 125	75
86	0.71626	12 43 163 205	67	0.28257	40 59	84	0.65912	14 125 200	29
93	0.79252	6 40 161	46	0.51670	138	32	0.50436	152 205	82
95	0.89153	12 55 71 161	34	0.29656	138 205	74	0.71739	14 125 200	20
98	0.70620	6 161 174 177 192	70	0.29607	40 59	75	0.61333	14 125 200	35
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
102	0.31407	100 152	98	0.87433	40 100 196	9	0.85804	100 196 200	14
103	1.00000	103	1	0.36319	138	49	1.00000	103	1
105	0.57865	43 163 177	84	0.29383	40 59	77	0.33605	12 125	98
110	0.82896	6 40 161 192	39	0.29750	40 59	73	0.54501	12 125	64
113	0.47324	43 163 198	91	0.28964	40 59	81	0.15111	12	102
115	0.63595	152 174 198	80	0.59846	40 46 196	25	0.58144	40 125 196	44
121	0.63208	12 163 198 205	82	0.29020	40 59	79	0.45007	12 125	89
123	0.65598	6 40 161 174	76	0.42226	40 59	44	0.52713	12 125	76
125	0.29971	40 152	100	0.74046	40 46 59	15	1.00000	125	1

PART II: FURTHER EVALUATION OF EFFICIENCY OF MUTUAL FUND MANAGEMENT COMPANIES

126	0.73891	12 152 174 198	59	0.40895	40 59	47	0.53913	14 125 200	70
127	1.00000	127	1	0.27067	40 59	89	0.54432	12 125	66
128	1.00000	128	1	0.23907	138 205	96	0.34918	12	97
130	1.00000	130	1	0.27649	138 196	87	0.50369	12 125	84
131	0.49723	6 40 174	90	0.63725	40 46 59	21	0.67312	125 196 200	26
132	0.70784	6 161 174 177 192	69	0.34403	40 59	56	0.54441	14 125 200	65
133	0.77377	6 40 174	47	0.66087	40 46 59	19	0.63329	125 196 200	31
137	0.76210	6 40 161 174 192	50	0.45070	40 59	41	0.61074	125 196 200	37
138	0.52947	100 174	88	1.00000	138	1	0.87403	100 152 205	13
139	0.65153	6 40 161 192	77	0.57697	138 196	27	0.75887	125 196 200	17
140	1.00000	140	1	0.28707	138 205	83	0.48565	12 125	86
142	0.63993	12 43 163 205	78	0.25835	40 59	95	0.46239	12 125	88
152	1.00000	152	1	0.67312	138 205	18	1.00000	152	1
154	1.00000	154	1	0.31909	138 205	62	0.43217	12 125	92
156	0.87372	6 40 161 174	35	0.47301	40 59	38	0.55355	14 125 200	59
159	0.55228	6 174 177	87	0.48575	40 59	35	0.54168	14 125 200	68
160	1.00000	160	1	0.60223	40 46 59	24	0.57909	125 196 200	45
161	1.00000	161	1	0.26421	40 59	90	0.51171	12 125	80
162	0.72785	12 55 154 177	63	0.31282	138 205	66	0.57846	14 125 200	46
163	1.00000	163	1	0.30099	138 205	72	0.55365	12 14 125	57
164	0.45298	6 40	95	0.99062	40 46 196	8	0.96576	125 196 200	11
168	0.79636	6 40 174 192	45	0.46296	40 59	40	0.57214	14 125 200	50
173	0.71892	43 163 198 205	65	0.31924	40 59	61	0.46823	12 125	87
174	1.00000	174	1	0.61053	40 46 59	23	0.50420	152	83
176	1.00000	176	1	0.30501	40 59	69	0.55930	12 14 125	54
177	1.00000	177	1	0.35193	40 59	52	0.54031	14 125 200	69
182	0.79826	163 177 198 205	44	0.29527	138 196	76	0.54738	14 125 200	63
185	0.67524	12 14 40 161	73	0.48411	138	36	0.78413	14 125 200	16
189	0.46387	100 152 174	94	0.70898	138 196	16	0.57592	100 196 200	47
190	0.80388	160 163	43	0.31498	40 59	65	0.54755	12 125	62
191	0.76970	6 40 161 174 192	48	0.50057	40 59	33	0.57575	40 103 125	48
192	1.00000	192	1	0.34489	40 59	55	0.56362	14 125 200	52
193	1.00000	193	1	0.33926	40 59	57	0.52845	14 125 200	74
194	0.96417	12 43 163 205	29	0.42058	40 59	45	0.43186	100 200	93
195	0.81140	12 152 174 198 205	42	0.62417	40 46 59	22	0.58317	100 200	42
196	0.17975	53	102	1.00000	196	1	1.00000	196	1
197	0.74443	53 174 177	57	0.56574	40 46 59	30	0.55629	125 196 200	55
198	1.00000	198	1	0.48037	40 59	37	0.54342	100 200	67
200	0.74171	14 40 76 174	58	0.48832	138	34	1.00000	200	1
202	0.70264	40 43 160 174	71	0.68120	40 46 59	17	0.62341	40 125 196	32
205	1.00000	205	1	1.00000	205	1	1.00000	205	1
206	0.66200	53 174	75	0.65272	59 205	20	0.69738	152 205	24

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). VAR II corresponds to the efficiency score obtained by the Variation II were obtained for each inefficient MFMC for those maximal friends facets randomly obtained (equation II.5). Ref. corresponds to the reference set for each company obtained by the VAR II approach. Rank represents the ranking of the VAR II scores which have been sorted by value, where VAR II scores with value of 1.00000 obtain the highest ranking 1.

Table II-I-2 Variation II-efficiency under CRS for 2006

Stage 1 Portfolio Management				Stage 2 Marketing and Service			Overall Efficiency		
MFMC	VAR II	Ref.	Rank	VAR II	Ref.	Rank	VAR II	Ref.	Rank
2	0.76302	12 194 198 205	39	0.38246	40 207	62	0.45674	14 125	74
4	0.65677	12 71 140 163	61	0.28594	60 152	89	0.54439	14 125	52
6	0.62620	12 174 194 198	64	0.43680	40 207	55	0.43127	205	77
7	0.92460	51 62 130 161 163	25	0.35240	40 207	64	0.47971	14 125	70
9	0.47173	12 51 59 160	87	0.64083	152	35	0.66103	14 103 125	21
12	1.00000	12	1	0.01947	100	102	0.75389	14 125	12
14	1.00000	14	1	0.17987	152 202	99	1.00000	14	1
15	0.79851	12 163 198	33	0.02534	40 207	101	0.29632	100 205	98
20	0.72182	12 128 130 160 163	45	0.26938	40 207	94	0.64227	14 103 125	24
21	0.55499	12 194 198 205	81	0.32579	40 207	74	0.64913	14 125	22
24	0.67410	12 174 194 198	55	0.33449	40 207	72	0.45021	14 125	76
29	0.71728	12 194 198	46	0.34337	40 207	66	0.52976	14 103 125	58
31	0.92942	12 128 163 198 205	24	0.46282	152	53	0.55008	14 125	50
34	0.41117	194	92	0.55128	152 202	44	0.40819	14 125	82
35	0.72666	51 53 163 177	44	0.43514	40 207	56	0.53537	14 103 125	56
36	0.61475	12 100 174 205	68	0.87299	40 100 203 207	14	0.70620	100 164 200	16
37	0.66009	51 128 160 205	60	0.55942	40 207	42	0.55590	100 103 164	47
38	0.47039	59 128 160	88	0.42218	40 207	58	0.59579	103 125 164	30
40	0.32999	12 100 205	100	1.00000	40	1	0.98433	100 103 164	9
42	0.61532	12 100 174 205	67	0.87276	40 100 203 207	16	0.72530	125 164 200	15
43	0.85881	128 198 205	29	0.33644	40 207	71	0.46221	14 125	73
45	0.63289	55 100 174	63	0.47303	40 207	51	0.55619	100 205	46
46	0.44947	12 100 205	91	0.99975	40 100 196 203 207	9	0.92683	100 164 200	11
47	0.70298	12 174 194 198	50	0.32133	40 207	78	0.37733	14	85
49	0.66715	12 53 160 205	57	0.69934	40 207	29	0.58445	100 103	35
50	0.58216	53 174	73	0.94882	152 196 202	12	0.61515	100 164 200	28
51	1.00000	51	1	0.50674	40	48	0.52860	14 103 125	59
53	1.00000	53	1	0.84560	60 152 196	19	0.64044	100 164 200	26
55	1.00000	55	1	0.25899	40 207	95	0.68509	14 103 125	17
57	0.62520	160 163 205	65	0.57878	40 207	39	0.49194	100 103	67
58	0.85071	12 163 198 205	30	0.23559	60 152	96	0.58768	14 125	34
59	1.00000	59	1	0.99682	40 100 203 207	11	0.29920	14 125	97
60	0.24630	100 205	102	1.00000	60	1	0.64109	100 125 164	25
61	0.59269	12 163 198 205	72	0.33426	40 207	73	0.37193	14 125	86
62	1.00000	62	1	0.47263	40 207	52	0.56125	14 103 125	42
63	0.91481	12 128 163 198 205	28	0.21810	60 203	98	0.17987	125	101
69	0.66627	100 174 205	59	0.79356	152 202	24	0.56488	100 164 200	39
71	1.00000	71	1	0.32165	40 207	77	0.30128	14 125	96
75	0.75271	12 128 163 198 205	40	0.27509	40 207	92	0.16342	205	102
76	0.78042	55 100 174	36	0.33718	40 207	70	0.51453	203	66
78	0.67956	53 163 198	53	0.50595	40 207	49	0.52076	103 125 200	64
83	0.91684	55 127 128 140	27	0.31079	60 152	81	0.34769	14	87
84	0.70903	12 130 140 161 163	48	0.27195	40 207	93	0.73148	14 103 125	14
85	0.73183	12 128 130 163	43	0.17243	40 207	100	0.33784	14	89
86	0.76417	130 163	38	0.28051	40 207	90	0.47574	203	71
93	0.57992	14 51 59	74	0.65739	152 196 202	33	0.48836	100 205	68
95	0.76530	12 140 163 198	37	0.29491	40 207	85	0.66212	14 103 125	20
98	0.55810	12 59 128 160	80	0.34177	40 207	68	0.57258	14 103 125	38
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
102	0.51440	100 205	83	0.87296	60 196 203	15	0.73669	100 205	13
103	0.38528	14 51 59	94	0.50826	152	47	1.00000	103 200	1
105	0.36626	198	97	0.43115	40 207	57	0.56422	103 125 164	41
110	0.73521	12 163 174 198 205	41	0.34326	40 207	67	0.54138	14 125	55
113	0.52875	12 128 163 198 205	82	0.31169	40 207	79	0.23904	14	100
115	0.66654	100 174 205	58	0.73514	152 202	25	0.57794	100 164 200	36
121	0.50667	12 194	85	0.31118	40 207	80	0.33765	205	90
123	0.60074	12 59 128 160	71	0.54124	40 207	45	0.48363	14 125	69
125	0.36668	59	96	0.87219	40 100 203 207	17	1.00000	125	1
126	0.65378	53 198	62	0.55886	40 207	43	0.52118	103 125 200	62
127	1.00000	127	1	0.29025	40 207	87	0.32800	14	92
128	1.00000	128	1	0.22138	40 207	97	0.31909	14 125	94
130	1.00000	130	1	0.28851	40 207	88	0.32748	14 125	93
131	0.41081	59 160	93	0.72486	152 202	27	0.64591	100 103 164	23

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132	0.78740	12 194 198 205	35	0.39001	60 152	60	0.54244	14 103 125	54
133	0.67243	12 51 53 160	56	0.83003	40 100 203 207	22	0.61416	100 103 200	29
137	0.69427	51 62 130 160 163	52	0.63603	202	36	0.62216	100 103 200	27
139	0.61652	12 51 59 160	66	0.67829	152 196	31	0.67693	103 125 200	18
140	1.00000	140	1	0.29186	40 207	86	0.41418	14 125	80
142	0.56461	12 194 198	79	0.29926	40 207	84	0.31907	14	95
152	0.50657	100 174 205	86	1.00000	152	1	0.67646	100 205	19
154	1.00000	154	1	0.30893	40 207	82	0.27667	14 125	99
156	0.57239	51 62 177	77	0.59897	40 207	38	0.52733	14 103 125	60
159	0.57673	194 198 205	75	0.56534	202	40	0.51875	14 103 125	65
160	1.00000	160	1	0.71811	152 202	28	0.55753	100 103 200	45
161	1.00000	161	1	0.27822	40 207	91	0.38533	14	84
162	0.79632	62 127 130 140 163	34	0.32401	40 207	75	0.56472	14 103 125	40
163	1.00000	163	1	0.30353	40 207	83	0.33355	14 125	91
164	0.36081	12 100 205	99	0.99912	40 100 203 207	10	1.00000	164	1
168	0.45616	12 174 194 198	90	0.72728	152 202	26	0.55156	103 125 200	49
173	0.69672	12 194 198 205	51	0.34740	40 207	65	0.45029	14 125	75
174	1.00000	174	1	0.82887	152 202	23	0.59514	205	31
176	0.92048	127 140 177	26	0.33870	40 207	69	0.41192	14 125	81
177	1.00000	177	1	0.45706	40 207	54	0.52306	14 103 125	61
182	0.83536	62 127 140 163	31	0.32289	40 207	76	0.55880	14 103 125	43
185	0.50748	14 51 161 174	84	0.55976	60 152 196	41	0.41422	100 205	79
190	0.73438	12 194 198 205	42	0.36147	40 207	63	0.46594	14 125	72
191	0.70352	12 100 174 205	49	0.65664	40 100 207	34	0.54423	100 103 200	53
192	0.71115	51 161 174 177	47	0.41637	40 207	59	0.54855	14 103 125	51
193	0.56809	51 161 174 177	78	0.38680	40 207	61	0.52093	14 103 125	63
194	1.00000	194	1	0.52539	152	46	0.39306	205	83
195	0.46293	53 198	89	0.90278	152 196 202	13	0.59030	100 103 164	32
196	0.37136	12 100 174 205	95	1.00000	196	1	0.96884	100 125 164	10
197	0.67727	53 163 177 198	54	0.66382	40 100 207	32	0.53292	100 103 200	57
198	1.00000	198	1	0.60641	202	37	0.42886	100 103	78
200	0.57644	12 100 174 205	76	0.67928	60 152	30	1.00000	200	1
202	0.60647	194 205	70	1.00000	202	1	0.55534	103 125 164	48
203	0.31533	100 205	101	1.00000	203	1	1.00000	203	1
204	0.93688	14 51 174 205	23	0.49157	40 207	50	0.55792	14 103 125	44
205	1.00000	205	1	0.85146	60 196 203	18	1.00000	205	1
206	0.61435	51 53 128 160	69	0.84098	152 202	20	0.58777	100 164 200	33
207	0.36325	59 128 160	98	1.00000	207	1	0.34119	14 125	88
210	0.80531	12 100 174 205	32	0.83242	40 100 196 207	21	0.57738	100 205	37

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). VAR II corresponds to the efficiency score obtained by the Variation II were obtained for each inefficient MFMC for those maximal friends facets randomly obtained (equation II.5). Ref. corresponds to the reference set for each company obtained by the VAR II approach. Rank represents the ranking of the VAR II scores which have been sorted by value, where VAR II scores with value of 1.00000 obtain the highest ranking 1.

Table II-I-3 Variation II-efficiency under CRS for 2007

Stage 1 Portfolio Management				Stage 2 Marketing and Service			Overall Efficiency		
MFCM	VAR II	Ref.	Rank	VAR II	Ref.	Rank	VAR II	Ref.	Rank
2	0.73765	50 128 160 198	45	0.71470	50 192	29	0.32951	103 125	68
4	0.57473	12 55 127 128	66	0.34536	192	87	0.32190	203	71
6	0.62460	12 174 177 198	63	0.75677	50 192	22	0.44004	102	45
7	1.00000	7	1	0.59117	50 192	52	0.30129	125 203	75
9	0.50099	12 100 174 210	78	0.62241	50 192	45	0.53460	125 203	31
12	1.00000	12	1	0.05120	203	95	0.53552	103 125 196	30
14	1.00000	14	1	0.16538	100 203	94	0.55875	103 125 196	24
15	0.81201	12 51 55 127 128	32	0.27005	192	92	0.29333	100 102	81
20	0.49159	12 128 163	83	0.33632	192	88	0.52517	125 203	33
21	0.49170	7 12 50 78	82	0.54681	50 192	66	0.49030	100 102	38
24	0.55148	78 163	71	0.58997	50 192	53	0.28080	103 125	85
29	0.58382	128 160 198	65	0.58586	192	56	0.37154	103 125	61
31	0.64518	12 53 102 152 198	60	0.62189	50 192	46	0.30853	103 125	74
34	0.42337	12 53 102 198	89	0.77819	50 192	15	0.25152	125	89
35	0.88065	51 127 174 177 198	26	0.71950	50 192	27	0.34709	103 125	63
36	0.43001	100 102 174 210	88	0.71030	46 50	30	0.70994	100 103 196	8
37	0.70727	50 51 128 160	52	0.76821	50 192	18	0.47679	103 125	40
38	0.43082	128 160	87	0.57132	50 192	61	0.40318	100 102	53
40	0.52373	12 53 102 198	73	0.66697	50 192	37	0.61079	103 125 196	15
43	0.83502	128 163 198	29	0.58218	50 192	59	0.33799	103 125	66
45	0.80771	14 51 152 174 177	33	0.77842	50 192	14	0.58336	100 103 196	20
46	0.46034	12 50 53	85	1.00000	46	1	0.94814	100 103 196	7
47	0.72840	55 174 177	48	0.52426	192	68	0.32168	102	72
49	0.56269	50 160 198	69	0.85213	50 192	7	0.55088	103 125 196	26
50	1.00000	50	1	1.00000	50	1	0.63371	100 103 196	13
51	1.00000	51	1	0.67947	50 192	35	0.41695	103 125	49
53	1.00000	53	1	0.66656	46 100 196	38	0.64932	100 103 196	12
55	1.00000	55	1	0.29344	192	91	0.51499	125 203	35
57	0.56745	12 53 102	68	0.66007	50 192	40	0.43161	100 102	47
58	0.59534	12 14 127	64	0.29991	192	90	0.55795	103 125 196	25
59	0.20928	12 50 53	94	0.76049	100 196 203	21	0.13339	102	94
61	0.65556	78 163	58	0.58806	192	55	0.19836	102	92
62	0.74013	14 50 177	44	0.82177	50 192	10	0.40030	100 102	54
63	0.77048	12 78 163	39	0.36025	192	85	0.15150	125	93
69	0.84198	12 53 102 152 198	27	0.78601	50 192	13	0.56487	100 103 196	23
71	0.73057	14 55 174 177	47	0.52268	192	69	0.44667	100 102	43
75	0.77490	7 12 128 163 198	36	0.38798	192	82	0.22771	102	90
76	0.74769	14 55 83 174 177	41	0.55710	50 192	64	0.50166	100 102	37
78	1.00000	78	1	0.83755	50 192	9	0.38602	103 125	56
83	1.00000	83	1	0.37920	192	83	0.34001	102	65
84	0.69572	12 127 140 177	53	0.35162	192	86	0.67157	103 125 196	10
85	0.74094	51 55 128	43	0.26762	192	93	0.31229	100 102	73
86	0.76915	7 127 128 140	40	0.43898	192	77	0.40687	102	52
93	0.64192	12 53 102	61	0.62965	50 192	44	0.44328	100 102	44
95	0.82524	14 50 127 177	30	0.37658	192	84	0.54674	125 203	28
98	0.50222	55 174 177	77	0.61237	50 192	49	0.48859	125 203	39
100	1.00000	100	1	1.00000	100	1	1.00000	100	1
102	1.00000	102	1	0.74881	46 50	23	1.00000	102	1
103	0.38590	14 152 174	91	0.60408	50 192	50	1.00000	103	1
105	0.42317	12 53 102	90	0.77192	50 192	16	0.60386	103 125 196	16
110	0.73395	7 12 50 78 198	46	0.58880	192	54	0.38347	102	57
113	0.49210	78 163	81	0.46282	192	75	0.12981	102	95
115	0.57373	50 51 152 198	67	0.74013	50 192	25	0.59204	100 103 196	18
121	0.49686	12 78 163	79	0.45278	192	76	0.37729	102	58
125	0.31214	100 102	92	0.74850	100 196 203	24	1.00000	125	1
126	0.71590	50 51 152	50	0.61982	50 192	47	0.47491	103 125	41
127	1.00000	127	1	0.41057	192	79	0.29344	125 203	79
128	1.00000	128	1	0.32087	192	89	0.29224	125 203	82
130	0.84006	7 128 140 198	28	0.40548	192	81	0.28628	125 203	84
131	0.43938	12 102 198	86	0.49739	46 50	72	0.59029	103 125 196	19
132	0.69070	7 12 50 78 198	54	0.55510	50 192	65	0.32925	102	69
133	0.67784	12 50 53 152	55	0.70848	46 50	32	0.59726	100 103 196	17
137	0.67767	12 102 152 174	56	0.63246	50 192	42	0.56702	103 125 196	22
139	0.53612	12 53 102	72	0.65211	46 50	41	0.65452	125 203	11

PART II: FURTHER EVALUATION OF EFFICIENCY OF MUTUAL FUND MANAGEMENT COMPANIES

140	1.00000	140	1	0.40860	192	80	0.29458	125 203	78
142	0.56049	7 12 128 163 198	70	0.48064	192	73	0.20124	102	91
152	1.00000	152	1	0.72640	46 50	26	0.61575	100 103 196	14
154	0.97683	12 51 128 140 174 198	22	0.46331	192	74	0.26783	125 203	88
156	0.63676	50 51 127 177	62	0.76939	50 192	17	0.41831	103 125	48
159	0.49528	102 174 198	80	0.71629	50 192	28	0.52105	103 125 196	34
160	1.00000	160	1	0.63034	50 192	43	0.44869	103 125	42
161	0.77271	14 55 83 177	37	0.54232	192	67	0.29672	125 203	76
162	0.71521	140 174 177 198	51	0.58285	192	58	0.33519	102	67
163	1.00000	163	1	0.42417	192	78	0.27186	125 203	86
168	0.66091	12 174 198 210	57	0.79909	50 192	12	0.53806	103 125 196	29
173	0.64626	50 78 163	59	0.56997	50 192	62	0.27104	103 125	87
174	1.00000	174	1	0.82051	50 192	11	0.55003	100 103	27
176	0.90774	50 127 140 177	24	0.60172	50 192	51	0.29180	125 203	83
177	1.00000	177	1	0.70883	50 192	31	0.41100	103 125	50
182	0.74126	7 50 78 163	42	0.58327	50 192	57	0.29466	125 203	77
185	0.51114	14 51 55 174	74	0.50878	50 192	70	0.34318	102	64
190	0.79102	7 78 177	35	0.61930	192	48	0.29341	103 125	80
191	0.89363	12 51 152 174 198	25	0.76480	50 192	20	0.53023	103 125 196	32
192	0.48852	12 78 177	84	1.00000	192	1	0.39958	100 102	55
193	0.82010	78 177 198	31	0.66960	50 192	36	0.32915	103 125	70
194	0.72245	12 127 140 174 177	49	0.58009	50 192	60	0.37199	103 125	60
196	0.25515	12 53 102	93	1.00000	196	1	1.00000	196	1
197	0.79756	12 50 53 152 198	34	0.69763	50 192	33	0.51077	103 125	36
198	1.00000	198	1	0.66168	50 192	39	0.40700	100 103	51
200	0.50988	12 53 102	75	0.50841	50 192	71	0.70358	103 125 196	9
203	0.09050	100 102	95	1.00000	203	1	1.00000	203	1
204	0.90917	12 50 78 177	23	0.68795	50 192	34	0.37510	102	59
206	0.77102	12 50 53	38	0.76548	50 192	19	0.58287	100 103 196	21
207	0.50417	12 102 198	76	0.56816	46 50	63	0.36365	103 125	62
210	1.00000	210	1	0.84159	50 192	8	0.43365	102	46

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). VAR II corresponds to the efficiency score obtained by the Variation II were obtained for each inefficient MFMC for those maximal friends facets randomly obtained (equation II.5). Ref. corresponds to the reference set for each company obtained by the VAR II approach. Rank represents the ranking of the VAR II scores which have been sorted by value, where VAR II scores with value of 1.00000 obtain the highest ranking 1.

Table II-I-4 Variation II-efficiency under CRS for 2008

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency		
	VAR II	Ref.	Rank	VAR II	Ref.	Rank	VAR II	Ref.	Rank
2	0.59550	15 198 217	35	0.18218	46	71	0.35560	46 103	53
4	0.59201	14 15 127 154	37	0.13894	46	88	0.33284	100 217	62
6	0.52650	174 198 217	51	0.19828	46	60	0.34609	103 125	56
7	0.80925	128 154 198	14	0.18782	46	67	0.30599	46 103	70
9	0.36174	15 174 217	82	0.29425	46	21	0.49467	100 125	24
12	1.00000	12	1	0.01660	46	93	0.52390	46 103 125	20
14	1.00000	14	1	0.06160	46	92	0.54483	46 103 125	17
15	1.00000	15	1	0.12520	46	91	0.27193	100 125	84
20	0.50616	55 83	56	0.19250	46	63	0.40915	100 125	40
21	0.46699	15 154 198	65	0.18760	46	68	0.31530	217	67
24	0.47246	15 174 198 217	64	0.23225	46	37	0.32907	103 125	64
29	0.39322	15 198 217	76	0.20744	46	53	0.33364	103 125	60
31	0.56887	15 174 198 217	44	0.21456	46	45	0.28657	103 125	76
34	0.48287	15 198 217	61	0.21542	46	44	0.27285	103 125	83
35	0.65590	127 154 160 174	27	0.21600	46	43	0.33305	46 103	61
36	0.37063	15 174 217	81	0.42683	46	6	0.74420	46 103 125	7
37	0.54192	15 198 217	49	0.21401	46	46	0.38167	46 103	47
38	0.40115	15 198 217	73	0.21218	46	50	0.26958	103	85
40	0.38246	15 217	77	0.38420	46	8	0.40095	103 125	41
43	0.51454	127 160	54	0.19313	46	62	0.30974	103 125	69
45	0.43452	83 174	68	0.23430	46	34	0.46843	100 217	31
46	0.13920	217	91	1.00000	46	1	1.00000	46	1
47	0.69560	83 174	25	0.20112	46	57	0.30029	100 125	71
49	0.41917	15 198 217	71	0.24778	46	30	0.47155	46 103	29
50	0.33344	174 217	83	0.32421	46	16	0.63336	46 103	10
51	0.59562	127 160 174	34	0.20762	46	52	0.47268	46 103	28
53	0.27972	174 210 217	87	0.36566	46	11	0.73957	100 217	8
55	1.00000	55	1	0.13828	46	89	0.41880	100 125	37
57	0.51792	15 198 217	53	0.24767	46	31	0.38191	46 103	46
58	0.54024	12 15 127	50	0.12830	46	90	0.47641	100 125	27
61	0.51851	15 198 217	52	0.20001	46	58	0.16347	217	90
62	0.56954	127 160 174	43	0.21331	46	49	0.41870	46 103	38
63	0.69772	12 127 128 160	24	0.15966	46	82	0.11021	125	91
69	0.37146	174 217	80	0.32570	46	15	0.55941	46 103	14
71	0.59620	14 55 127 174	33	0.18551	46	70	0.34417	100 125	57
76	0.76308	14 55 127 174	17	0.17985	46	73	0.34996	100 125	55
78	0.55830	15 174 198 217	46	0.20161	46	56	0.37182	46 103	50
83	1.00000	83	1	0.15958	46	83	0.28480	100 125	79
84	0.71522	55 83	19	0.15451	46	85	0.42505	100 125	34
85	0.78333	14 127 128	15	0.14807	46	87	0.27668	125	82
86	0.70258	83 127 174	21	0.17407	46	76	0.28630	103 125	77
93	0.55008	15 210 217	48	0.28808	46	23	0.48933	100 217	26
95	0.70181	14 15 127 154	22	0.16317	46	81	0.42574	100 125	33
98	0.29633	15 198 217	86	0.19082	46	64	0.52056	46 103	21
100	0.10899	217	93	1.00000	100	1	1.00000	100	1
103	0.26652	15 217	88	0.29630	46	20	1.00000	103	1
105	0.37281	198 217	79	0.29949	46	19	0.02662	103	93
110	0.69063	83 127 174	26	0.18667	46	69	0.33768	103 125	59
113	0.45911	15 198 217	67	0.18797	46	66	0.20438	217	89
115	0.31358	174 217	85	0.33163	46	12	0.60425	46 103 125	11
121	0.43192	55 83	69	0.22563	46	40	0.23755	125	86
125	0.15207	217	90	0.52552	46	4	1.00000	125	1
126	0.59300	174 198 217	36	0.19841	46	59	0.42103	46 103	35
127	1.00000	127	1	0.15643	46	84	0.28702	103 125	75
128	1.00000	128	1	0.14927	46	86	0.22909	125	87
130	0.75219	128 154 198	18	0.16639	46	79	0.27833	100 125	80
131	0.32688	198 217	84	0.32618	46	14	0.60183	46 103	12
132	0.42043	127 160	70	0.23296	46	36	0.36365	103 125	52
133	0.56980	15 174 217	42	0.38054	46	9	0.49075	46 103	25
137	0.58753	15 198 217	38	0.29120	46	22	0.51351	100 217	23
139	0.40092	15 210 217	74	0.32356	46	17	0.56555	100 125	13
140	0.77594	127 128 154 160	16	0.16461	46	80	0.28761	103 125	74
142	0.61804	55 83 127 174	30	0.17190	46	78	0.36644	217	51
152	0.23478	174 217	89	0.40889	46 100	7	0.55702	217	15

PART II: FURTHER EVALUATION OF EFFICIENCY OF MUTUAL FUND MANAGEMENT COMPANIES

154	1.00000	154	1	0.17462	46	75	0.27780	100 125	81
156	0.49718	15 174 217	58	0.28576	46	24	0.43458	46 103	32
159	0.48231	198 217	62	0.25591	46	26	0.53258	46 103	18
160	1.00000	160	1	0.24419	46	33	0.39589	46 103	42
161	0.50774	14 55 127 174	55	0.23221	46	38	0.29603	103 125	72
162	0.61621	83 174	31	0.22271	46	41	0.33100	103 125	63
163	0.62352	128 154 198	29	0.17247	46	77	0.29260	103 125	73
168	0.49062	15 174 217	59	0.27420	46	25	0.51716	46 103	22
173	0.37833	127 160	78	0.20908	46	51	0.22665	125	88
174	1.00000	174	1	0.24848	46	29	0.38556	217	44
176	0.70863	127 154 160 174	20	0.19609	46	61	0.31188	103 125	68
177	0.46210	174	66	0.20414	46	55	0.41989	46 103	36
182	0.57534	15 198 217	41	0.17979	46	74	0.28550	100 125	78
185	0.55629	12 127 160	47	0.19053	46	65	0.05271	125	92
190	0.65080	154 174 198	28	0.18093	46	72	0.31696	103 125	66
191	0.55974	127 160 174	45	0.25448	46	27	0.38501	46 103	45
192	0.48496	174 198 217	60	0.20662	46	54	0.52659	103 125	19
193	0.47632	174 198 217	63	0.21399	46	47	0.35046	46 103	54
194	0.59765	127 160 174	32	0.23202	46	39	0.34000	103 125	58
196	0.11883	174 217	92	0.73057	46	3	0.90450	46 103 125	6
197	0.57689	174 198 217	39	0.23425	46	35	0.37717	46 103	48
198	1.00000	198	1	0.21346	46	48	0.32454	46 103	65
200	0.39916	15 198 217	75	0.31903	46	18	0.41811	100 125	39
203	0.41864	15 198 217	72	0.36829	46	10	0.65514	46 103 125	9
204	0.69810	14 160 174	23	0.22154	46	42	0.39463	46 103	43
206	0.57541	15 174 198 217	40	0.25430	46	28	0.46986	46 103	30
207	0.50317	15 198 217	57	0.32682	46	13	0.54926	46 103 125	16
210	1.00000	210	1	0.24506	46	32	0.37691	100 125	49
217	1.00000	217	1	0.45084	46	5	1.00000	217	1

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). VAR II corresponds to the efficiency score obtained by the Variation II were obtained for each inefficient MFMC for those maximal friends facets randomly obtained (equation II.5). Ref. corresponds to the reference set for each company obtained by the VAR II approach. Rank represents the ranking of the VAR II scores which have been sorted by value, where VAR II scores with value of 1.00000 obtain the highest ranking 1.

Table II-I-5 Variation II-efficiency under CRS for 2009

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency		
	VAR II	Ref.	Rank	VAR II	Ref.	Rank	VAR II	Ref.	Rank
2	0.80165	51 174 193	30	0.53660	15 207	83	0.46496	103 125	50
4	1.00000	4	1	0.45781	15 207	89	0.25551	45	83
6	0.68003	45 131 193 210 217	55	0.74818	15 207	22	0.55521	45 217	44
7	1.00000	7	1	0.64147	15 207	61	0.36724	103 125	62
9	0.48553	14 50 100	87	0.56584	46 53	79	0.58148	103 125	29
12	0.71522	4 14 127	45	0.00027	15	94	0.00008	217	95
14	1.00000	14	1	0.00018	100	95	0.42524	125	56
15	0.67725	14 128	56	1.00000	15	1	0.18579	217	89
20	0.58430	76 128 131 210	78	0.75497	15 207	19	0.44927	103 125	52
21	0.59559	4 14 127	76	0.58665	15 207	73	0.38989	45	58
24	0.71157	45 76 131 193	47	0.61761	15 207	65	0.34035	103 125	68
29	0.44229	45 76 131 193	89	0.52118	15 207	85	0.44744	103 125	53
31	0.70822	7 51 193	49	0.82770	15 196 207	11	0.34294	103 125	66
34	0.64106	45 76 131 193 210	67	0.68499	15 196 207	47	0.44694	103 125	54
35	0.91060	50 51 76 174 193	21	0.63656	15 207	62	0.54157	103 125	46
36	0.61199	14 46 50 100	72	0.79022	46 53 196	13	0.77096	45 100 217	15
37	0.60574	128 131 210	74	0.70827	15 207	38	0.56360	103 125	39
38	0.41354	50 51 193	90	0.54374	15 207	82	0.45210	45 217	51
40	0.73725	46 50 76 100	40	0.76303	15 196 207	17	0.74270	45 100 217	16
43	0.76939	7 51 140 193	34	0.59379	15 207	70	0.36933	103 125	61
45	1.00000	45	1	0.70201	15 196 207	40	1.00000	45	1
46	1.00000	46	1	1.00000	46	1	0.95419	103 125	6
47	0.70415	76 128 131 210	51	0.71053	15 207	37	0.33056	103 125	71
49	0.57460	76 131 193 210	79	0.73296	15 196 207	29	0.58896	45	26
50	1.00000	50	1	0.66915	100 196 221	48	0.80871	45 100 217	12
51	1.00000	51	1	0.80063	15 207	12	0.58029	103 125	31
53	0.60637	45 46 100 217	73	1.00000	53	1	0.87864	100 217	8
55	1.00000	55	1	0.49118	15 207	87	0.35383	103 125	64
57	0.74001	45 76 131 193 210	37	0.71742	15 196 207	34	0.59213	45 217	25
58	0.65633	4 14 127	61	0.58506	15 207	74	0.32319	45 100	73
61	0.55714	128 193 210	83	0.59357	15 207	71	0.00520	45	94
62	1.00000	62	1	0.71535	15 196 207	35	0.58407	103 125	28
63	0.75616	14 127 128	35	0.45738	15 207	90	0.05511	125	93
69	0.67247	45 46 100 217	58	0.91835	46 53 196	10	0.82493	45 100 217	10
71	0.89938	7 14 51 76 193	22	0.68735	15 196 207	46	0.38014	103 125	59
76	1.00000	76	1	0.69840	15 207	44	0.37976	103 125	60
78	0.77159	45 46 76 131 210	33	0.75114	15 196 207	20	0.57683	103 125	34
83	0.69211	128 140 193	54	0.49672	15 207	86	0.16020	125	91
84	0.72170	4 14 127	43	0.71102	15 207	36	0.34288	125	67
85	0.63040	4 14 127	68	0.41993	15 207	92	0.21711	217	84
86	0.71752	127 128	44	0.66287	15 207	52	0.26866	125	82
93	0.60112	45 46 76 131	75	0.77985	15 196 207	14	0.55912	45 100	41
95	0.88045	7 14 76 128 140	24	0.63096	15 207	63	0.30047	103 125	76
98	0.40837	45 76 193	91	0.72137	15 207	32	0.44385	103 125	55
100	1.00000	100	1	1.00000	100	1	1.00000		1
103	0.39396	14 45 46 76	94	0.44139	196 207	91	1.00000	103	1
105	0.50767	46 131 210 217	86	0.74032	15 207	26	0.39596	45 217	57
110	0.70307	7 62 193	53	0.58784	15 207	72	0.34908	103 125	65
113	0.54074	128 193 210	84	0.55092	15 207	81	0.18844	45	88
115	0.62486	152 174 217	71	0.60497	100 196 221	68	0.63983	45 217	19
121	0.55843	4 127	82	0.65608	15 207	55	0.21266	45	85
125	0.35004	45 100 152	95	0.97340	46 53 196	8	1.00000	125	1
126	0.72582	7 14 46 50 51	42	0.70387	15 207	39	0.56625	103 125	36
127	1.00000	127	1	0.64348	15 207	59	0.28922	103 125	78
128	1.00000	128	1	0.13037	15	93	0.10484	125	92
130	0.86584	7 14 128 140	25	0.57746	15 207	76	0.28600	103 125	79
131	1.00000	131	1	0.69936	15 207	41	0.84278	45 100 217	9
132	0.58555	51 127 128 140	77	0.61431	15 207	66	0.31978	103 125	74
133	0.78873	45 46 50 76 100	31	0.71854	15 196 207	33	0.77101	45 100 217	14
137	0.70913	45 46 50 76 100	48	0.69888	15 196 207	43	0.72495	45 217	17
139	0.64264	14 45 46 76	66	0.69907	221	42	0.62223	103	20

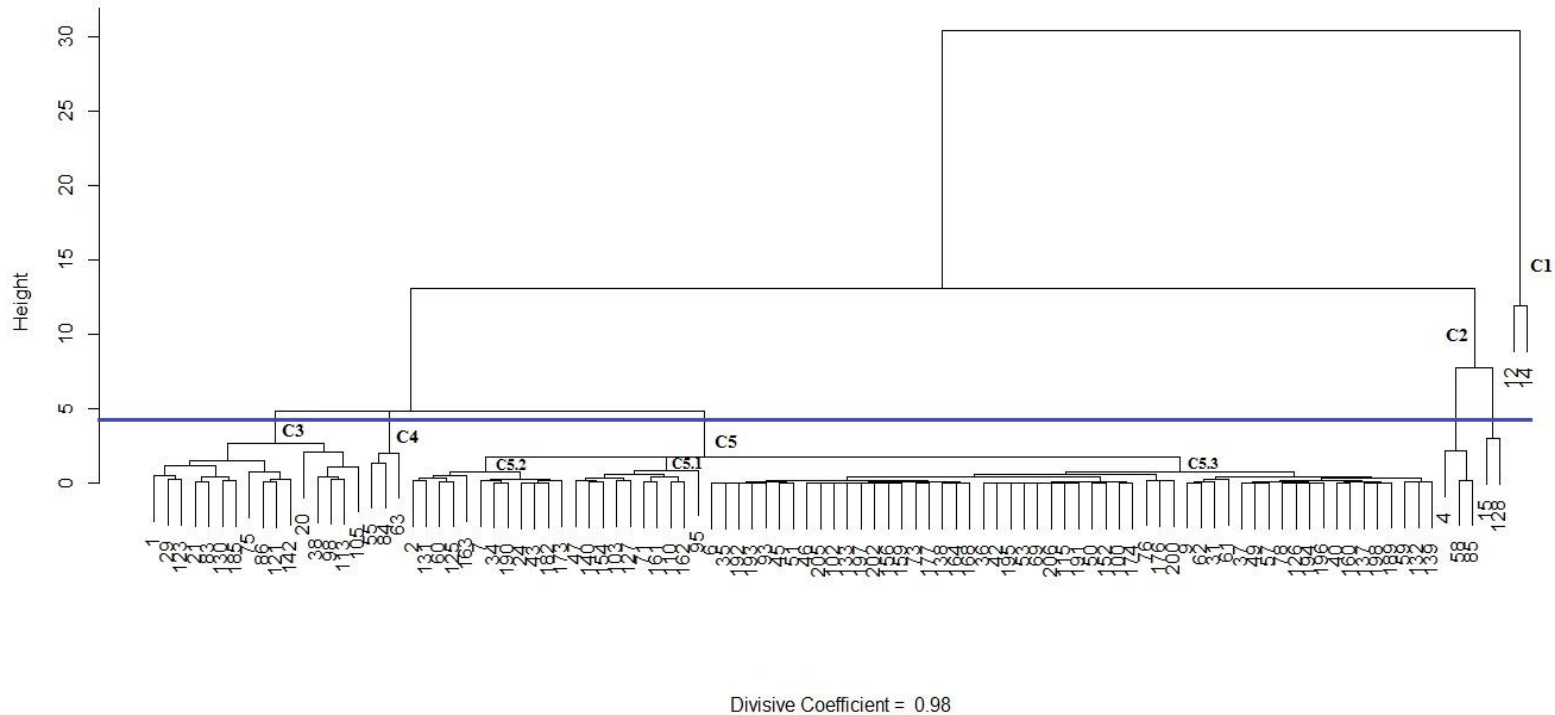
PART II: FURTHER EVALUATION OF EFFICIENCY OF MUTUAL FUND MANAGEMENT COMPANIES

140	1.00000	140	1	0.52274	15 207	84	0.27316	103 125	81
142	0.55940	4 127	81	0.65658	15 196 207	54	0.20576	45	87
152	1.00000	152	1	0.62595	100 196 221	64	0.82462	45 217	11
154	0.88375	76 128 131 210	23	0.65029	15 207	57	0.29836	103 125	77
156	0.73884	45 174 193 210	39	0.73722	15 196 207	27	0.56776	103 125	35
159	0.66175	51 174 193	60	0.64653	15 207	58	0.55737	103 125	42
160	0.73983	51 62 193	38	0.75677	15 196 207	18	0.56508	103 125	37
161	0.70324	45 76 193	52	0.60253	15 207	69	0.28188	125	80
162	0.81092	14 51 193	29	0.57762	15 207	75	0.36112	103 125	63
163	0.65303	51 140 193	62	0.46335	15 207	88	0.32834	103 125	72
168	0.70472	7 50 51 62 193	50	0.75065	15 196 207	21	0.62107	45 217	22
173	0.62974	7 128 140 193	69	0.64152	15 207	60	0.20992	125	86
174	1.00000	174	1	0.77893	15 196 207	15	0.61036	45 217	24
176	0.85227	14 51 76 193	26	0.66611	15 207	50	0.33760	103 125	69
177	0.57134	45 152 174 217	80	0.74047	15 207	25	0.57753	103 125	33
182	0.64577	128 193 210	65	0.66032	15 207	53	0.30782	103 125	75
185	0.53409	4 127	85	0.56639	15 196	78	0.16341	45	90
190	0.81579	14 51 193	28	0.57629	15 207	77	0.33619	103 125	70
191	0.82468	46 50 76 100	27	0.76499	15 196 207	16	0.61675	103 125	23
192	0.66503	45 174 193 217	59	0.72330	15 207	30	0.54223	45 217	45
193	1.00000	193	1	0.55944	15 207	80	0.52702	103 125	47
194	0.73075	7 62 76 193	41	0.65390	15 207	56	0.51501	103 125	48
195	0.62920	76 131 193 210	70	0.66586	15 207	51	0.49744	103 125	49
196	0.40002	14 50 100 174	92	1.00000	196	1	0.94074	103 125	7
197	0.64604	174 193 217	64	0.73452	15 196 207	28	0.56290	103 125	40
198	0.74111	51 174 193	36	0.66901	15 207	49	0.55700	103 125	43
200	0.39727	7 46 128 210	93	0.74663	15 196	23	0.56407	45 100	38
203	0.64785	46 131 217	63	0.61395	100 196 221	67	0.66650	45 217	18
204	0.71249	45 76 131 193 210	46	0.74594	15 196 207	24	0.58105	45	30
206	0.67473	45 46 50 100 174	57	0.69600	15 196 207	45	0.62200	45 217	21
207	0.45493	131 193	88	1.00000	207	1	0.58658	45 217	27
210	1.00000	210	1	0.72176	15 196 207	31	0.57897	103 125	32
217	1.00000	217	1	0.92002	46 53	9	1.00000	217	1
221	0.78730	14 46 50 100	32	1.00000	221	1	0.80444	103 125	13

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). VAR II corresponds to the efficiency score obtained by the Variation II were obtained for each inefficient MFMC for those maximal friends facets randomly obtained (equation II.5). Ref. corresponds to the reference set for each company obtained by the VAR II approach. Rank represents the ranking of the VAR II scores which have been sorted by value, where VAR II scores with value of 1.00000 obtain the highest ranking 1.

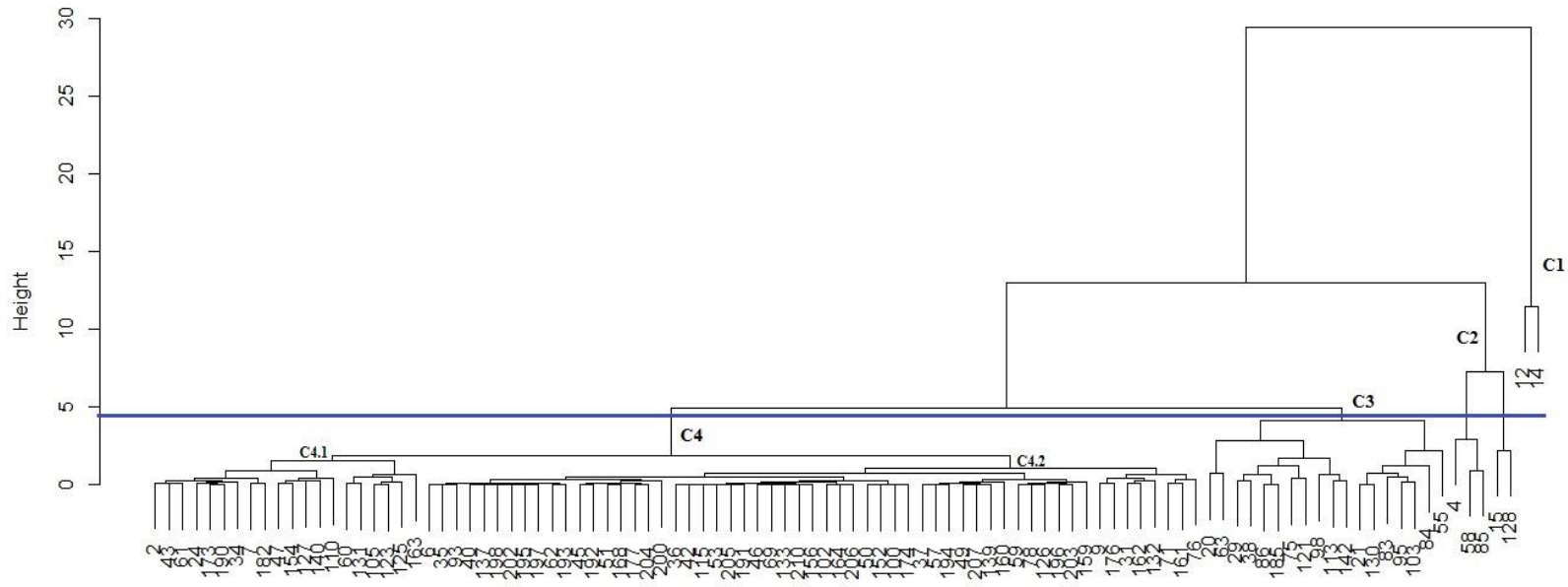
Appendix J

Figure II-J-1 Dendrogram (Divisive Analysis) DIANA for 2005



The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). C1 = Cluster of the largest firms, C2 = Cluster of large firms, C3 = Cluster of large-midsize firms, C4 = Cluster of small-midsize firms, C5 (small firms) where C5.1 Cluster of large-small sized firms, C5.2 = Cluster of mid-small sized firms, C5.3 = Cluster of small-small sized firms.. The values consider the assets and labour amounts. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

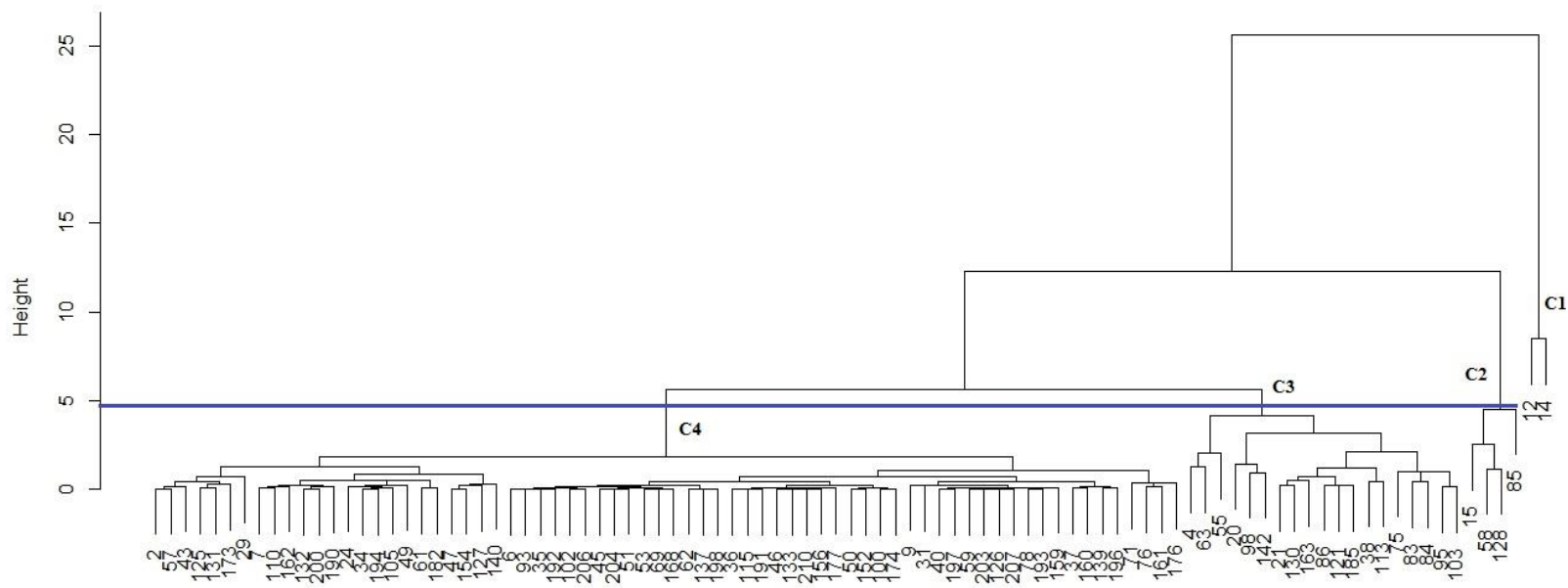
Figure II-J-2 Dendrogram (Divisive Analysis) DIANA for 2006



Divisive Coefficient = 0.98

The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). C1 = Cluster of the largest firms, C2 = Cluster of large firms, C3 = Cluster of large-midsize firms, C4 = Cluster of small-midsize firms, C5 (small firms) where C5.1 Cluster of large-small sized firms, C5.2 = Cluster of mid-small sized firms, C5.3 = Cluster of small-small sized firms.. The values consider the assets and labour amounts. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

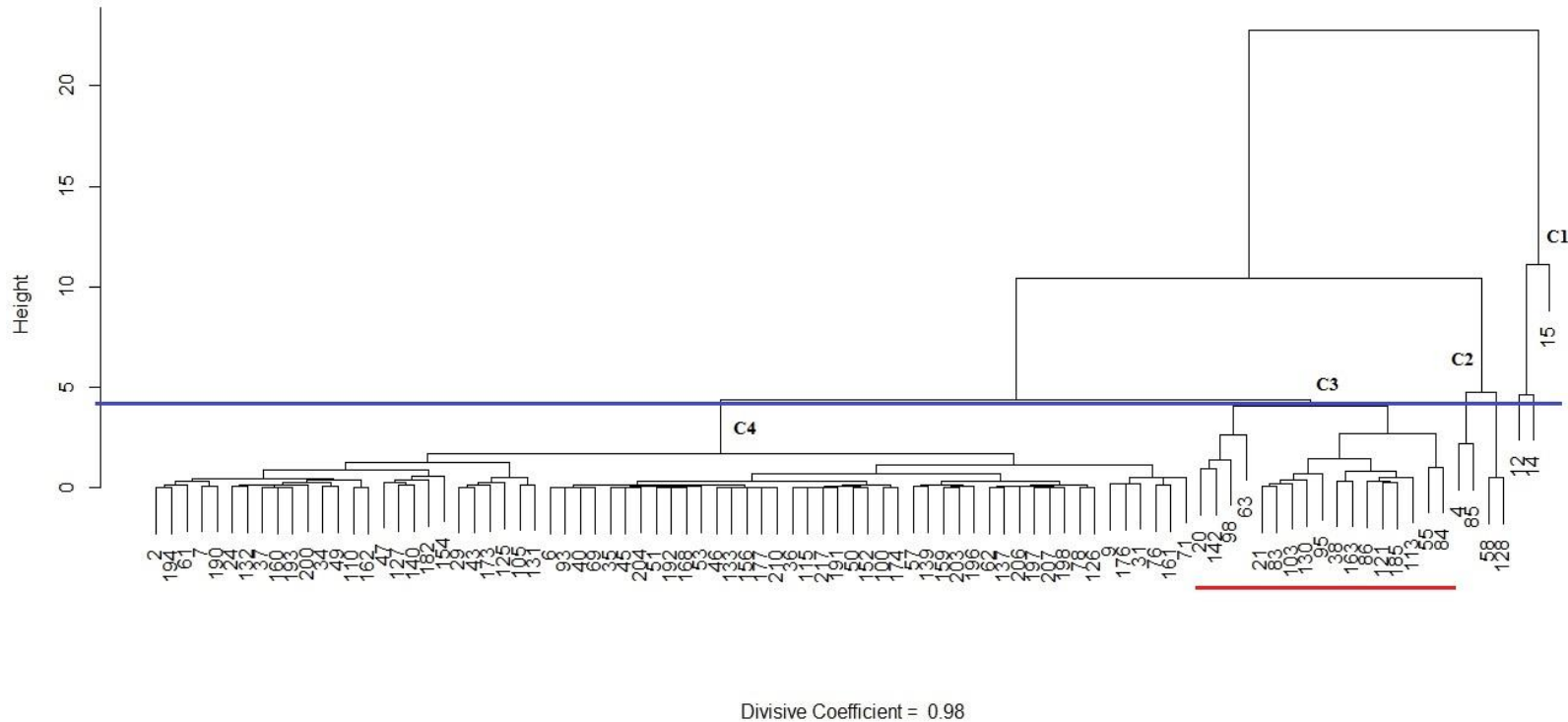
Figure II-J-3 Dendrogram (Divisive Analysis) DIANA for 2007



Divisive Coefficient = 0.98

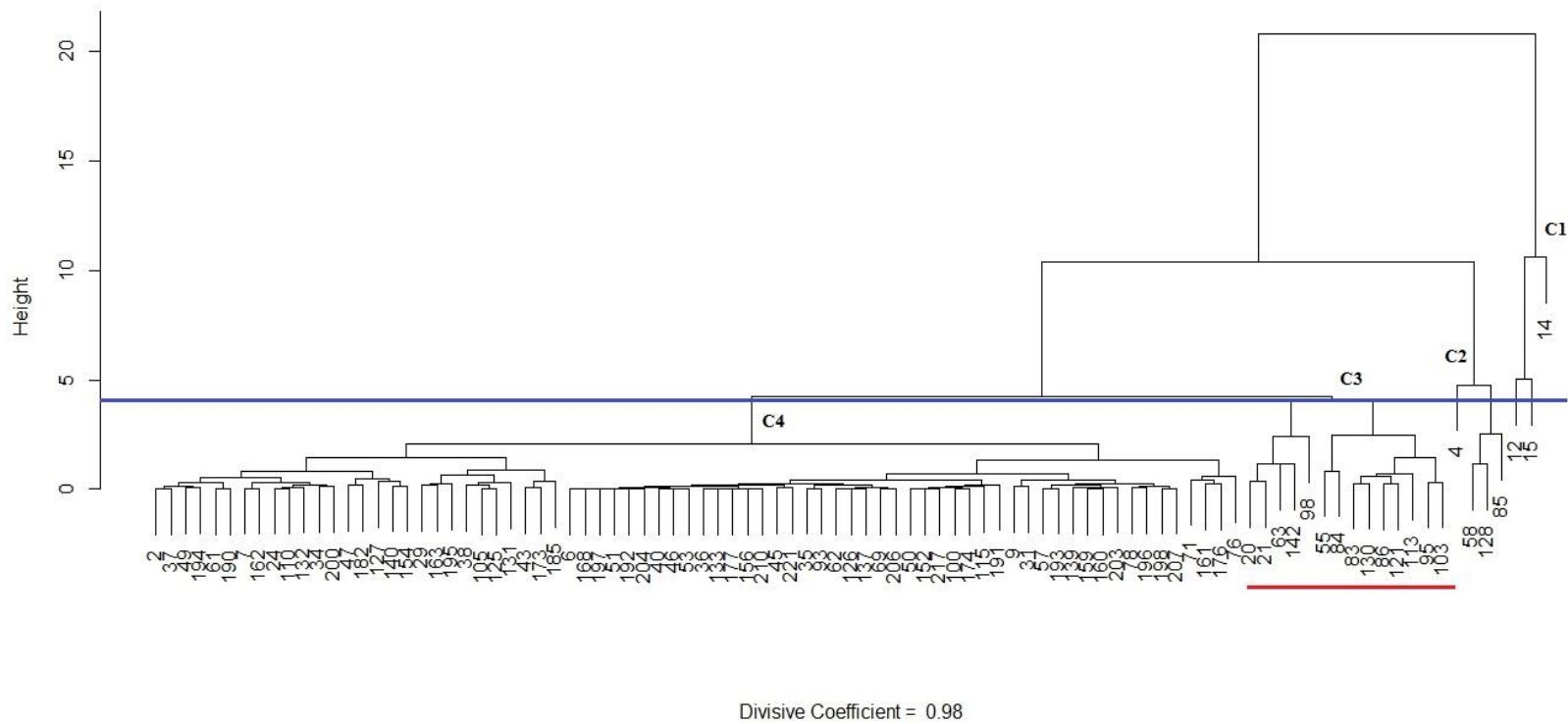
The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). C1 = Cluster of the largest firms, C2 = Cluster of large firms, C3 = Cluster of large-midsize firms, C4 = Cluster of small-midsize firms, C5 (small firms) where C5.1 Cluster of large-small sized firms, C5.2 = Cluster of mid-small sized firms, C5.3 = Cluster of small-small sized firms.. The values consider the assets and labour amounts. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-J-4 Dendrogram (Divisive Analysis) DIANA for 2008



The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). C1 = Cluster of the largest firms, C2 = Cluster of large firms, C3 = Cluster of large-midsize firms, C4 = Cluster of small-midsize firms, C5 (small firms) where C5.1 Cluster of large-small sized firms, C5.2 = Cluster of mid-small sized firms, C5.3 = Cluster of small-small sized firms.. The values consider the assets and labour amounts. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-J-5 Dendrogram (Divisive Analysis) DIANA for 2009



The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). C1 = Cluster of the largest firms, C2 = Cluster of large firms, C3 = Cluster of large-midsize firms, C4 = Cluster of small-midsize firms, C5 (small firms) where C5.1 Cluster of large-small sized firms, C5.2 = Cluster of mid-small sized firms, C5.3 = Cluster of small-small sized firms.. The values consider the assets and labour amounts. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Appendix K

Table II-K-1 Variation III-efficiency under CRS for 2005

Stage 1 Portfolio Management				Stage 2 Marketing and Service			Overall Efficiency			
MFMC	VAR III	Ref.	Remark	VAR III	Ref.	Remark	VAR III	Ref.	Remark	Cluster
1	1.00000	1	locally eff.	0.27981	40 59		0.15860	152		3
2	0.53841	43 163		0.31795	40 59		0.29616	103 125		5.2
4	1.00000	4	locally eff.	0.22070	138 205		1.00000	4	locally eff.	2
6	1.00000	6		0.34996	40 59		0.36169	152		5.3
7	1.00000	7		0.30571	40 59		0.33222	103 125		5.2
9	0.62045	40 76		0.42290	138		0.55360	103 125		5.3
12	1.00000	12		0.03048	59 205		1.00000	12		1
14	1.00000	14		0.15673	59 205		1.00000	14		1
15	1.00000	15	locally eff.	0.26393	138 196		0.36761	12		2
20	1.00000	20	locally eff.	0.27083	40 59		0.48653	103 125		3
21	1.00000	21	locally eff.	0.32631	40 59		0.38557	100 205		3
24	0.60829	7		0.29171	40 59		0.27904	103 125		5.2
29	1.00000	29	locally eff.	0.31670	40 59		0.44023	103 125		3
31	1.00000	31	locally eff.	0.33585	40 59		0.36068	103 125		5.3
34	0.29414	43 163		0.35654	40 59		0.34552	152		5.2
35	0.94340	40 160 176 177		0.34986	40 59		0.42367	103 125		5.3
36	0.70789	6 40 174		0.74505	40 46 59		0.71421	125 196 200		5.3
37	0.75725	40 160 174		0.54733	40 59		0.58279	125 196 200		5.3
38	1.00000	38	locally eff.	0.33099	40 59		0.58539	40 103 125		3
40	1.00000	40		1.00000	40		1.00000	40		5.3
42	0.55395	6 40 174		0.76168	59 205		0.70030	125 196 200		5.3
43	1.00000	43		0.30158	40 59		0.33802	103 125		5.2
45	0.76627	40 174 177 192		0.37312	40 59		0.61285	100 196 200		5.3
46	0.40978	76 100 174		1.00000	46		0.89171	100 196 200		5.3
47	0.63152	127 140		0.30867	138 196		0.25011	125		5.1
49	0.82668	40 160 174		0.56928	40 59		0.60595	125 196 200		5.3
50	0.47064	53 160 177		0.84876	59 205		0.62303	125 196 200		5.3
51	0.73582	40 160 176 177		0.41680	40 59		0.50581	103 125		5.3
53	1.00000	53		0.85623	46 138 205		0.75751	100 152 205		5.3
55	1.00000	55		0.14995	100 196		0.48570	103 125		4
57	0.76376	40 160 174		0.57301	40 59		0.57366	125 196 200		5.3
58	1.00000	58	locally eff.	0.20126	59 205		0.59002045	12 14		2
59	0.25762	6 174		1.00000	59		0.51300	100 152		5.3
60	0.03124	43		0.83982	138 205		0.59208	100 200		5.2
61	0.53174	40 160 176 177		0.28864	40 59		0.12951	152		5.3
62	0.74709	160 176		0.36214	40 59		0.51844	103 125		5.3
63	1.00000	63	locally eff.	0.28119	138 196		0.16068	125		4
69	0.46557	53 174 177		0.58979	40 46 196		0.55364	125 196 200		5.3
71	1.00000	71		0.28994	138 196		0.32593	103 125		5.1
73	1.00000	73	locally eff.	0.43503	40 59		0.44003	100 200		5.3
75	1.00000	75	locally eff.	0.25971	40 59		0.17286	152		3
76	1.00000	76		0.30367	40 59		0.54259	103 200		5.3
78	0.75348	40 160 174 177		0.46455	40 59		0.55339	125 196 200		5.3
83	1.00000	83	locally eff.	0.25938	138 196		0.25339	152		3
84	1.00000	84	locally eff.	0.26262	138 196		0.67584	103		4
85	1.00000	85	locally eff.	0.22043	100 196		0.51800	12 14		2
86	0.61898	130		0.28257	40 59		0.52340	103 200		3
93	0.70410	6 40 192		0.51670	138		0.50436	152 205		5.3
95	1.00000	95	locally eff.	0.29656	138 205		0.53526	40 103		5.1
98	1.00000	98	locally eff.	0.29607	40 59		0.56935	103 125		3
100	1.00000	100		1.00000	100		1.00000	100		5.3
102	0.31407	100 152		0.87433	40 100 196		0.85804	100 196 200		5.3
103	1.00000	103		0.36319	138		1.00000	103		5.1
105	1.00000	105	locally eff.	0.29383	40 59		0.12083	125		3
110	0.70524	71 161		0.29750	40 59		0.28489	103 125		5.1
113	0.34055	130		0.28964	40 59		0.05145	152		3
115	0.63595	152 174 198		0.59846	40 46 196		0.58144	40 125 196		5.3
121	1.00000	121	locally eff.	0.29020	40 59		0.32436	152		3
123	1.00000	123	locally eff.	0.42226	40 59		0.41587	103 125		3
125	0.05083	43		0.74046	40 46 59		1.00000	125		5.2

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126	0.72349	174 177 198 205		0.40895	40 59	0.53698	40 103 125	5.3
127	1.00000	127		0.27067	40 59	0.32729	152 205	5.1
128	1.00000	128		0.23907	138 205	0.34918	12	2
130	1.00000	130		0.27649	138 196	0.28981	103 125	3
131	1.00000	131	locally eff.	0.63725	40 46 59	0.67312	125 196 200	5.2
132	0.69072	6 174 177 192		0.34403	40 59	0.39513	103	5.3
133	0.77377	6 40 174		0.66087	40 46 59	0.63329	125 196 200	5.3
137	0.75743	6 40 174 192		0.45070	40 59	0.61074	125 196 200	5.3
138	0.52947	100 174		1.00000	138	0.87403	100 152 205	5.3
139	0.51429	6 40 192		0.57697	138 196	0.75887	125 196 200	5.3
140	1.00000	140		0.28707	138 205	0.27982	103 125	5.1
142	0.48450	130		0.25835	40 59	0.28830	152	3
152	1.00000	152		0.67312	138 205	1.00000	152	5.3
154	1.00000	154		0.31909	138 205	0.26866	103 125	5.1
156	0.85809	6 40 174 192		0.47301	40 59	0.54651	40 103 125	5.3
159	0.55228	6 174 177		0.48575	40 59	0.54150	40 103 125	5.3
160	1.00000	160		0.60223	40 46 59	0.57909	125 196 200	5.3
161	1.00000	161		0.26421	40 59	0.27909	103 125	5.1
162	1.00000	162	locally eff.	0.31282	138 205	0.40292	103 125	5.1
163	1.00000	163		0.30099	138 205	0.30286	103 125	5.2
164	0.45298	6 40		0.99062	40 46 196	0.96576	125 196 200	5.3
168	0.79636	6 40 174 192		0.46296	40 59	0.57055	40 103 125	5.3
173	0.50520	43 163		0.31924	40 59	0.24158	125	5.2
174	1.00000	174		0.61053	40 46 59	0.50420	152	5.3
176	1.00000	176		0.30501	40 59	0.30895	103 125	5.3
177	1.00000	177		0.35193	40 59	0.46910	103 200	5.3
182	1.00000	182	locally eff.	0.29527	138 196	0.31234	103 125	5.2
185	1.00000	185	locally eff.	0.48411	138	0.46112	103 125	3
189	0.46387	100 152 174		0.70898	138 196	0.57592	100 196 200	5.3
190	1.00000	190	locally eff.	0.31498	40 59	0.30322	103 125	5.2
191	0.76134	6 40 174 192		0.50057	40 59	0.57575	40 103 125	5.3
192	1.00000	192		0.34489	40 59	0.44654	103 125	5.3
193	1.00000	193		0.33926	40 59	0.39558	103 125	5.3
194	1.00000	194	locally eff.	0.42058	40 59	0.43186	100 200	5.3
195	0.80340	174 177 198 205		0.62417	40 46 59	0.58317	100 200	5.3
196	0.17975	53		1.00000	196	1.00000	196	5.3
197	0.74443	53 174 177		0.56574	40 46 59	0.55629	125 196 200	5.3
198	1.00000	198		0.48037	40 59	0.54342	100 200	5.3
200	0.63774	40 76		0.48832	138	1.00000	200	5.3
202	0.69287	40 160 174 177		0.68120	40 46 59	0.62341	40 125 196	5.3
205	1.00000	205		1.00000	205	1.00000	205	5.3
206	0.66200	53 174		0.65272	59 205	0.69738	152 205	5.3

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). VAR III corresponds to the efficiency score obtained by the Variation II. But this VAR III was obtained for each inefficient MFMC referred to those maximal friends facets formed by efficient companies belonging to the same cluster (adjacent clusters when no efficient MFMC is found in the objective cluster). Ref. corresponds to the reference set for each company obtained by the VAR III approach. Remark represents the locally efficient companies found in each cluster. Cluster is the group where each company was classified.

Table II-K-2 Variation III-efficiency under CRS for 2006

Stage 1 Portfolio Management				Stage 2 Marketing and Service			Overall Efficiency			
MFMC	VAR III	Ref.	Remark	VAR III	Ref.	Remark	VAR III	Ref.	Remark	Cluster
2	1.00000	2	locally eff.	0.38246	40 207		0.28124	125 200		4.1
4	1.00000	4	locally eff.	0.28594	60 152		1.00000	4	locally eff.	2
6	0.51667	71 174 177		0.43680	40 207		0.43127	205		4.2
7	1.00000	7	locally eff.	0.35240	40 207		0.29915	125 200		4.1
9	0.39342	62 161		0.64083	152		0.49644	100 205		4.2
12	1.00000	12		0.01947	100		0.74887	14		1
14	1.00000	14		0.17987	152 202		1.00000	14		1
15	1.00000	15	locally eff.	0.02534	40 207		0.25515	14		2
20	1.00000	20	locally eff.	0.26938	40 207		0.32461	103		3
21	0.27350	55		0.32579	40 207		0.28615	103		3
24	0.54664	154 163		0.33449	40 207		0.28839	125 200		4.1
29	1.00000	29	locally eff.	0.34337	40 207		0.38344	103		3
31	1.00000	31	locally eff.	0.46282	152		0.31676	125 200		4.2
34	0.25657	163		0.55128	152 202		0.24943	205		4.1
35	0.63503	62 161 177		0.43514	40 207		0.33791	125 200		4.2
36	0.42176	51 160 205		0.87299	40 100 203 207		0.70620	100 164 200		4.2
37	0.63828	194 205		0.55942	40 207		0.48986	100 200		4.2
38	0.20847	130		0.42218	40 207		1.00000	38	locally eff.	3
40	0.14162	51 177 205		1.00000	40		1.00000	40	locally eff.	4.2
42	0.41864	100 174 205		0.87276	40 100 203 207		0.72530	125 164 200		4.2
43	1.00000	43	locally eff.	0.33644	40 207		0.29143	125 200		4.1
45	0.59881	51 161 174		0.47303	40 207		0.55619	100 205		4.2
46	0.39304	53 160 205		0.99975	40 100 196 203 207		0.92683	100 164 200		4.2
47	0.59045	127 140 163		0.32133	40 207		0.31518	205		4.1
49	0.45957	194 205		0.69934	40 207		0.51549	100 205		4.2
50	0.58216	53 174		0.94882	152 196 202		0.61515	100 164 200		4.2
51	1.00000	51		0.50674	40		0.39651	125 200		4.2
53	1.00000	53		0.84560	60 152 196		0.64044	100 164 200		4.2
55	1.00000	55		0.25899	40 207		0.38660	103		3
57	0.51530	194 205		0.57878	40 207		0.45669	205		4.2
58	1.00000	58	locally eff.	0.23559	60 152		0.57232	14		2
59	1.00000	59		0.99682	40 100 203 207		0.12718	205		4.2
60	0.00543	163		1.00000	60		0.64109	100 125 164		4.1
61	0.47405	163		0.33426	40 207		0.28923	205		4.1
62	1.00000	62		0.47263	40 207		0.46243	100 205		4.2
63	1.00000	63	locally eff.	0.21810	60 203		0.06122	103		3
69	0.66627	100 174 205		0.79356	152 202		0.56488	100 164 200		4.2
71	1.00000	71		0.32165	40 207		0.27198	100 125		4.2
75	1.00000	75	locally eff.	0.27509	40 207		0.05953	103		3
76	0.69875	51 71 161 177		0.33718	40 207		0.51453	203		4.2
78	0.67742	174 194 198		0.50595	40 207		0.43966	100 200		4.2
83	1.00000	83	locally eff.	0.31079	60 152		0.14432	103		3
84	1.00000	84	locally eff.	0.27195	40 207		0.58563	103		3
85	1.00000	85	locally eff.	0.17243	40 207		0.33784	14		2
86	1.00000	86	locally eff.	0.28051	40 207		1.00000	86	locally eff.	3
93	0.41423	51 62 160		0.65739	152 196 202		0.48836	100 205		4.2
95	1.00000	95	locally eff.	0.29491	40 207		0.46012	103		3
98	1.00000	98	locally eff.	0.34177	40 207		0.47299	103		3
100	1.00000	100		1.00000	100		1.00000	100		4.2
102	0.51440	100 205		0.87296	60 196 203		0.73669	100 205		4.2
103	1.00000	103	locally eff.	0.50826	152		1.00000	103		3
105	0.19031	127 163		0.43115	40 207		0.56404	125 164 200		4.1
110	0.58597	140 154 163		0.34326	40 207		0.40015	205		4.1
113	0.36528	130		0.31169	40 207		0.06916	103		3
115	0.66654	100 174 205		0.73514	152 202		0.57794	100 164 200		4.2
121	0.37387	130		0.31118	40 207		0.10587	103		3
123	0.20823	127 163		0.54124	40 207		0.33986	205		4.1
125	0.05088	163		0.87219	40 100 203 207		1.00000	125		4.1
126	0.65378	53 198		0.55886	40 207		0.47035	125 200		4.2
127	1.00000	127		0.29025	40 207		0.26964	125		4.1
128	1.00000	128		0.22138	40 207		0.25396	14		2
130	1.00000	130		0.28851	40 207		0.12687	103		3
131	1.00000	131	locally eff.	0.72486	152 202		0.64583	100 164 200		4.1
132	0.56536	62 161 177		0.39001	60 152		0.36749	125 200		4.2

PART II: FURTHER EVALUATION OF EFFICIENCY OF MUTUAL FUND MANAGEMENT COMPANIES

133	0.54528	51 53 177 205		0.83003	40 100 203 207	0.60502	100 200	4.2
137	0.47453	51 62 177		0.63603	202	0.56537	100 200	4.2
139	0.43850	62 160		0.67829	152 196	0.66774	125 200	4.2
140	1.00000	140		0.29186	40 207	0.29023	125 200	4.1
142	0.42262	130		0.29926	40 207	0.10846	103	3
152	0.50657	100 174 205		1.00000	152	0.67646	100 205	4.2
154	1.00000	154		0.30893	40 207	0.26309	125 200	4.1
156	0.57239	51 62 177		0.59897	40 207	0.39479	125 200	4.2
159	0.57673	194 198 205		0.56534	202	0.39265	125 200	4.2
160	1.00000	160		0.71811	152 202	0.54804	100 200	4.2
161	1.00000	161		0.27822	40 207	0.28819	100 125	4.2
162	0.58751	62 161 177		0.32401	40 207	0.40770	100 205	4.2
163	1.00000	163		0.30353	40 207	0.27019	125 200	4.1
164	0.27403	51 160 205		0.99912	40 100 203 207	1.00000	164	4.2
168	0.45032	174 194 198		0.72728	152 202	0.54093	125 200	4.2
173	0.54079	154 163		0.34740	40 207	0.28408	125 200	4.1
174	1.00000	174		0.82887	152 202	0.59514	205	4.2
176	1.00000	176	locally eff.	0.33870	40 207	0.28461	125 200	4.2
177	1.00000	177		0.45706	40 207	0.40245	125 200	4.2
182	0.81727	127 140 163		0.32289	40 207	0.31414	125 200	4.1
185	1.00000	185	locally eff.	0.55976	60 152 196	0.13982	103	3
190	0.63175	127 163		0.36147	40 207	0.29285	125 200	4.1
191	0.69915	51 174 177 205		0.65664	40 100 207	0.51609	100 200	4.2
192	0.71115	51 161 174 177		0.41637	40 207	0.37779	205	4.2
193	0.56809	51 161 174 177		0.38680	40 207	0.32269	125 200	4.2
194	1.00000	194		0.52539	152	0.39306	205	4.2
195	0.46293	53 198		0.90278	152 196 202	0.58970	100 164 200	4.2
196	0.33785	59 160		1.00000	196	0.96884	100 125 164	4.2
197	0.67523	53 177 198		0.66382	40 100 207	0.50211	100 200	4.2
198	1.00000	198		0.60641	202	0.37287	100 200	4.2
200	0.44056	71 174		0.67928	60 152	1.00000	200	4.2
202	0.60647	194 205		1.00000	202	0.50382	125 200	4.2
203	0.31533	100 205		1.00000	203	1.00000	203	4.2
204	1.00000	204	locally eff.	0.49157	40 207	0.37110	205	4.2
205	1.00000	205		0.85146	60 196 203	1.00000	205	4.2
206	0.61428	51 53 160 205		0.84098	152 202	0.58777	100 164 200	4.2
207	0.35137	51 59 160		1.00000	207	0.18664	205	4.2
210	0.78713	174 194 205		0.83242	40 100 196 207	0.57738	100 205	4.2

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). VAR III corresponds to the efficiency score obtained by the Variation II. But this VAR III was obtained for each inefficient MFMC referred to those maximal friends facets formed by efficient companies belonging to the same cluster (adjacent clusters when no efficient MFMC is found in the objective cluster). Ref. corresponds to the reference set for each company obtained by the VAR III approach. Remark represents the locally efficient companies found in each cluster. Cluster is the group where each company was classified.

Table II-K-3 Variation III-efficiency under CRS for 2007

Stage 1 Portfolio Management				Stage 2 Marketing and Service			Overall Efficiency			
MFMC	VAR III	Ref.	Remark	VAR III	Ref.	Remark	VAR III	Ref.	Remark	Cluster
2	0.65477	50 140 177 198		0.71470	50 192		0.28714	125 196		4
4	1.00000	4	locally eff.	0.34536	192		0.10488	103		3
6	0.62132	140 174 177 198		0.75677	50 192		0.44004	102		4
7	1.00000	7		0.59117	50 192		0.30129	125 203		4
9	0.32111	50 51 127 140		0.62241	50 192		0.53460	125 203		4
12	1.00000	12		0.05120	203		0.24853	103		1
14	1.00000	14		0.16538	100 203		0.29689	103		1
15	1.00000	15	locally eff.	0.27005	192		0.05225	103		2
20	0.45726	55 83		0.33632	192		0.21514	103		3
21	1.00000	21	locally eff.	0.54681	50 192		0.11364	103		3
24	0.52280	7 78		0.58997	50 192		0.27964	125 203		4
29	0.55646	127 174 198		0.58586	192		0.32535	125 196		4
31	0.54201	7 78 177		0.62189	50 192		0.30370	125 203		4
34	0.40615	78		0.77819	50 192		0.25152	125		4
35	0.88065	51 127 174 177 198		0.71950	50 192		0.30511	125 196		4
36	0.43001	100 102 174 210		0.71030	46 50		0.58785	100 196		4
37	0.63106	51 127 174		0.76821	50 192		0.38249	102		4
38	1.00000	38	locally eff.	0.57132	50 192		1.00000	38	locally eff.	3
40	0.42644	50 160		0.66697	50 192		0.49023	100 102		4
43	0.78265	127 140 198		0.58218	50 192		0.31926	125 203		4
45	0.78175	51 127 140 174 177		0.77842	50 192		0.49991	100 203		4
46	0.45504	50 53 198		1.00000	46		0.79692	100 196		4
47	0.66235	140 174		0.52426	192		0.32168	102		4
49	0.56269	50 160 198		0.85213	50 192		0.42513	100 102		4
50	1.00000	50		1.00000	50		0.50655	100 102		4
51	1.00000	51		0.67947	50 192		0.33298	125 196		4
53	1.00000	53		0.66656	46 100 196		0.64187	100 102		4
55	1.00000	55		0.29344	192		0.17281	103		3
57	0.49022	127 140 174		0.66007	50 192		0.43161	100 102		4
58	1.00000	58	locally eff.	0.29991	192		0.20325	103		2
59	0.17313	50 53 198		0.76049	100 196 203		0.13339	102		4
61	0.60290	7 78		0.58806	192		0.19836	102		4
62	0.72003	7 50 140 177		0.82177	50 192		0.40030	100 102		4
63	1.00000	63	locally eff.	0.36025	192		0.02930	103		3
69	0.82772	50 53 152 198		0.78601	50 192		0.41821	102		4
71	1.00000	71	locally eff.	0.52268	192		0.44667	100 102		4
75	0.65472	55 83		0.38798	192		0.05125	103		3
76	1.00000	76	locally eff.	0.55710	50 192		0.50166	100 102		4
78	1.00000	78		0.83755	50 192		0.32337	100 203		4
83	1.00000	83		0.37920	192		0.08579	103		3
84	0.62812	55		0.35162	192		0.40844	103		3
85	1.00000	85	locally eff.	0.26762	192		0.07036	103		2
86	1.00000	86	locally eff.	0.43898	192		0.15163	103		3
93	0.44678	7 50 78 177		0.62965	50 192		0.44328	100 102		4
95	1.00000	95	locally eff.	0.37658	192		0.27348	103		3
98	1.00000	98	locally eff.	0.61237	50 192		0.34052	103		3
100	1.00000	100		1.00000	100		1.00000	100		4
102	1.00000	102		0.74881	46 50		1.00000	102		4
103	1.00000	103	locally eff.	0.60408	50 192		1.00000	103		3
105	0.41569	53 102 210		0.77192	50 192		0.49546	100 102		4
110	0.64117	140 174 198		0.58880	192		0.38347	102		4
113	0.47913	163		0.46282	192		0.04300	103		3
115	0.57373	50 51 152 198		0.74013	50 192		0.54797	102		4
121	0.36385	55 83		0.45278	192		0.08012	103		3
125	0.31214	100 102		0.74850	100 196 203		1.00000	125		4
126	0.71590	50 51 152		0.61982	50 192		0.32017	125 196		4
127	1.00000	127		0.41057	192		0.29344	125 203		4
128	1.00000	128		0.32087	192		0.05711	103		2
130	1.00000	130	locally eff.	0.40548	192		0.07581	103		3
131	0.39274	100 102 210		0.49739	46 50		0.45212	100 102		4
132	0.68160	7 50 78 177		0.55510	50 192		0.32925	102		4
133	0.59676	51 127 174		0.70848	46 50		0.44049	100 102		4
137	0.55419	51 127 174		0.63246	50 192		0.48915	100 203		4
139	0.41263	51 127 174		0.65211	46 50		0.65452	125 203		4

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140	1.00000	140		0.40860	192	0.29458	125 203	4
142	1.00000	142	locally eff.	0.48064	192	0.05337	103	3
152	1.00000	152		0.72640	46 50	0.58499	100 102	4
154	1.00000	154	locally eff.	0.46331	192	0.26783	125 203	4
156	0.63676	50 51 127 177		0.76939	50 192	0.32770	125 196	4
159	0.49528	102 174 198		0.71629	50 192	0.34083	125 196	4
160	1.00000	160		0.63034	50 192	0.34623	125 196	4
161	1.00000	161	locally eff.	0.54232	192	0.29672	125 203	4
162	0.71521	140 174 177 198		0.58285	192	0.33519	102	4
163	1.00000	163		0.42417	192	0.07466	103	3
168	0.58851	100 174 198		0.79909	50 192	0.47483	100 102	4
173	0.59978	127 174 177		0.56997	50 192	0.27086	125 203	4
174	1.00000	174		0.82051	50 192	0.50130	102	4
176	0.90774	50 127 140 177		0.60172	50 192	0.29180	125 203	4
177	1.00000	177		0.70883	50 192	0.31705	125 196	4
182	0.68433	7 78		0.58327	50 192	0.29466	125 203	4
185	1.00000	185	locally eff.	0.50878	50 192	0.04871	103	3
190	0.79102	7 78 177		0.61930	192	0.28609	125 203	4
191	0.86966	51 140 174 198		0.76480	50 192	0.40536	102	4
192	0.47765	7 78 177		1.00000	192	0.39958	100 102	4
193	0.82010	78 177 198		0.66960	50 192	0.28237	125 196	4
194	0.70433	127 140 174		0.58009	50 192	0.34829	125 203	4
196	0.24555	53 102 210		1.00000	196	1.00000	196	4
197	0.71777	51 152 174 177		0.69763	50 192	0.32651	100 196	4
198	1.00000	198		0.66168	50 192	0.29946	100 196	4
200	0.26175	51 140 174		0.50841	50 192	1.00000	200	locally eff. 4
203	0.09050	100 102		1.00000	203	1.00000	203	4
204	1.00000	204	locally eff.	0.68795	50 192	0.37510	102	4
206	0.70215	50 53 152 198		0.76548	50 192	0.49278	100 102	4
207	0.48356	50 160		0.56816	46 50	0.29845	125 196	4
210	1.00000	210		0.84159	50 192	0.43365	102	4

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). VAR III corresponds to the efficiency score obtained by the Variation II. But this VAR III was obtained for each inefficient MFMC referred to those maximal friends facets formed by efficient companies belonging to the same cluster (adjacent clusters when no efficient MFMC is found in the objective cluster). Ref. corresponds to the reference set for each company obtained by the VAR III approach. Remark represents the locally efficient companies found in each cluster. Cluster is the group where each company was classified.

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Table II-K-4 Variation III-efficiency under CRS for 2008

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency			Cluster
	VAR III	Ref.	Remark	VAR III	Ref.	Remark	VAR III	Ref.	Remark	
2	0.55475	198 210 217		0.18218	46		0.29286	100 125		4
4	1.00000	4	locally eff.	0.13894	46		0.12356	103		2
6	0.52650	174 198 217		0.19828	46		0.27098	217		4
7	0.74892	154 198		0.18782	46		0.30478	100 125		4
9	0.28569	154 160 174		0.29425	46		0.49467	100 125		4
12	1.00000	12		0.01660	46		0.36512	103		1
14	1.00000	14		0.06160	46		1.00000	14	locally eff.	1
15	1.00000	15		0.12520	46		0.10475	103		1
20	0.50616	55 83		0.19250	46		0.25325	103		3
21	1.00000	21	locally eff.	0.18760	46		0.10910	103		3
24	0.43788	127 160 174		0.23225	46		0.29213	100 125		4
29	0.37074	198 210 217		0.20744	46		0.28128	100 125		4
31	0.52841	127 160 174		0.21456	46		0.27733	100 125		4
34	0.39259	198 210 217		0.21542	46		0.26941	100 125		4
35	0.65590	127 154 160 174		0.21600	46		0.31635	100 125		4
36	0.36531	174 210 217		0.42683	46		0.57524	100 125		4
37	0.48326	127 160 174		0.21401	46		0.33161	100 125		4
38	1.00000	38	locally eff.	0.21218	46		0.26958	103		3
40	0.36682	210 217		0.38420	46		0.27632	100 125		4
43	0.51454	127 160		0.19313	46		0.28628	100 125		4
45	0.42642	127 174		0.23430	46		0.46843	100 217		4
46	0.13920	217		1.00000	46		1.00000	46		4
47	0.54379	127 160		0.20112	46		0.30029	100 125		4
49	0.39555	210 217		0.24778	46		0.38283	217		4
50	0.33344	174 217		0.32421	46		0.55068	100 217		4
51	0.59562	127 160 174		0.20762	46		0.34480	100 125		4
53	0.27972	174 210 217		0.36566	46		0.73957	100 217		4
55	1.00000	55		0.13828	46		0.28478	103		3
57	0.51225	127 160 174		0.24767	46		0.33010	100 125		4
58	1.00000	58	locally eff.	0.12830	46		0.28277	103		2
61	0.39352	154 198		0.20001	46		0.16347	217		4
62	0.56954	127 160 174		0.21331	46		0.35607	100 125		4
63	1.00000	63	locally eff.	0.15966	46		0.04286	103		3
69	0.37146	174 217		0.32570	46		0.53336	100 217		4
71	1.00000	71	locally eff.	0.18551	46		0.34417	100 125		4
76	1.00000	76	locally eff.	0.17985	46		0.34996	100 125		4
78	0.54594	174 198 210 217		0.20161	46		0.31020	100 125		4
83	1.00000	83		0.15958	46		0.16334	103		3
84	0.71522	55 83		0.15451	46		0.27710	103		3
85	1.00000	85	locally eff.	0.14807	46		0.11254	103		2
86	1.00000	86	locally eff.	0.17407	46		0.17864	103		3
93	0.35103	154 160 174		0.28808	46		0.48933	100 217		4
95	1.00000	95	locally eff.	0.16317	46		1.00000	95	locally eff.	3
98	1.00000	98	locally eff.	0.19082	46		1.00000	98	locally eff.	3
100	0.10899	217		1.00000	100		1.00000	100		4
103	0.12726	55		0.29630	46		1.00000	103		3
105	0.37281	198 217		0.29949	46		0.01290	217		4
110	0.60674	154 198		0.18667	46		0.32598	100 125		4
113	0.42956	55 83		0.18797	46		0.11492	103		3
115	0.31358	174 217		0.33163	46		0.48571	217		4
121	0.43192	55 83		0.22563	46		0.11745	103		3
125	0.15207	217		0.52552	46		1.00000	125		4
126	0.59300	174 198 217		0.19841	46		0.31147	100 125		4
127	1.00000	127		0.15643	46		0.28638	100 125		4
128	1.00000	128		0.14927	46		0.08928	103		2
130	1.00000	130	locally eff.	0.16639	46		0.16835	103		3
131	0.32688	198 217		0.32618	46		0.41567	100 125		4
132	0.42043	127 160		0.23296	46		0.31888	100 125		4
133	0.52102	210 217		0.38054	46		0.49028	100 217		4
137	0.52200	127 160 174		0.29120	46		0.51351	100 217		4
139	0.37404	154 160 174		0.32356	46		0.56555	100 125		4
140	0.75356	127 154 160		0.16461	46		0.28568	100 125		4
142	1.00000	142	locally eff.	0.17190	46		0.24264	103		3
152	0.23478	174 217		0.40889	46 100		0.55702	217		4

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154	1.00000	154		0.17462	46	0.27780	100 125	4
156	0.43426	127 160 174		0.28576	46	0.34638	100 125	4
159	0.48231	198 217		0.25591	46	0.32472	100 125	4
160	1.00000	160		0.24419	46	0.36959	100 125	4
161	1.00000	161	locally eff.	0.23221	46	0.29284	100 125	4
162	0.59302	127 160		0.22271	46	0.31872	100 125	4
163	1.00000	163	locally eff.	0.17247	46	0.24790	103	3
168	0.45170	174 210 217		0.27420	46	0.38043	217	4
173	0.37833	127 160		0.20908	46	0.22665	125	4
174	1.00000	174		0.24848	46	0.38556	217	4
176	0.70863	127 154 160 174		0.19609	46	0.30540	100 125	4
177	0.46210	174		0.20414	46	0.32165	100 125	4
182	0.53107	127 160		0.17979	46	0.28550	100 125	4
185	1.00000	185	locally eff.	0.19053	46	0.02283	103	3
190	0.65080	154 174 198		0.18093	46	0.29108	100 125	4
191	0.55974	127 160 174		0.25448	46	0.36215	100 125	4
192	0.48496	174 198 217		0.20662	46	0.33380	100 125	4
193	0.47632	174 198 217		0.21399	46	0.28834	100 125	4
194	0.59765	127 160 174		0.23202	46	0.30790	100 125	4
196	0.11883	174 217		0.73057	46	0.74770	100 125	4
197	0.57689	174 198 217		0.23425	46	0.33082	100 125	4
198	1.00000	198		0.21346	46	0.31557	100 125	4
200	0.24084	154 160		0.31903	46	0.41811	100 125	4
203	0.41566	198 210 217		0.36829	46	0.50574	100 217	4
204	0.69701	127 160 174		0.22154	46	0.32886	100 125	4
206	0.56438	174 210 217		0.25430	46	0.35660	100 125	4
207	0.48295	198 217		0.32682	46	0.33694	100 125	4
210	1.00000	210		0.24506	46	0.37691	100 125	4
217	1.00000	217		0.45084	46	1.00000	217	4

MFMC is the identifying number of the mutual fund management company (Official registers, CNMV). VAR III corresponds to the efficiency score obtained by the Variation II. But this VAR III was obtained for each inefficient MFMC referred to those maximal friends facets formed by efficient companies belonging to the same cluster (adjacent clusters when no efficient MFMC is found in the objective cluster). Ref. corresponds to the reference set for each company obtained by the VAR III approach. Remark represents the locally efficient companies found in each cluster. Cluster is the group where each company was classified.

Table II-K-5 Variation III-efficiency under CRS for 2009

MFMC	Stage 1 Portfolio Management			Stage 2 Marketing and Service			Overall Efficiency			Cluster
	VAR III	Ref.	Remark	VAR III	Ref.	Remark	VAR III	Ref.	Remark	
2	0.80165	51 174 193		0.45045	196 221		0.28627	45		4
4	1.00000	4		1.00000	4	locally eff.	0.10309	103		2
6	0.68003	45 131 193 210 217		0.59040	100 196 221		0.55521	45 217		4
7	1.00000	7		0.35864	46 53		0.29391	100 125		4
9	0.45885	45 76 131		0.56584	46 53		0.54143	100 217		4
12	1.00000	12	locally eff.	0.00027	15		0.00002	103		1
14	1.00000	14		0.00014	15		0.26384	103		1
15	0.45053	14		1.00000	15		0.06753	103		1
20	1.00000	20	locally eff.	0.36956	46 53		0.36525	103		3
21	1.00000	21	locally eff.	0.38580	46 53		0.16846	103		3
24	0.71157	45 76 131 193		0.41924	196 221		0.28916	100 125		4
29	0.44229	45 76 131 193		0.39140	196 221		0.28534	100 125		4
31	0.70822	7 51 193		0.58216	207		0.29484	100 125		4
34	0.64106	45 76 131 193 210		0.59045	100 196 221		0.42434	45		4
35	0.91060	50 51 76 174 193		0.47228	196 221		0.32496	45		4
36	0.60312	45 46 50 100		0.79022	46 53 196		0.77096	45 100 217		4
37	0.60235	51 62		0.51933	196 221		0.47989	45		4
38	0.41354	50 51 193		0.54099	100 196 221		0.45210	45 217		4
40	0.73725	46 50 76 100		0.71830	100 196 221		0.74270	45 100 217		4
43	0.76939	7 51 140 193		0.34648	46 53		0.28897	100 125		4
45	1.00000	45		0.53343	196 207		1.00000	45		4
46	1.00000	46		1.00000	46		0.90408	100 125		4
47	0.62314	51 127 140		0.38054	46 53		0.30395	100 125		4
49	0.57460	76 131 193 210		0.61111	196 221		0.58896	45		4
50	1.00000	50		0.66915	100 196 221		0.80871	45 100 217		4
51	1.00000	51		0.60758	207		0.57433	45 217		4
53	0.60637	45 46 100 217		1.00000	53		0.87864	100 217		4
55	1.00000	55		0.27712	46 53		0.22358	103		3
57	0.74001	45 76 131 193 210		0.56961	196 221		0.59213	45 217		4
58	1.00000	58	locally eff.	1.00000	58	locally eff.	0.21441	103		2
61	0.55229	76 131 193		0.35523	196 221		0.00520	45		4
62	1.00000	62		0.56781	196 221		0.58257	45 217		4
63	1.00000	63	locally eff.	0.23863	46 53		0.03139	103		3
69	0.67247	45 46 100 217		0.91835	46 53 196		0.82493	45 100 217		4
71	1.00000	71	locally eff.	0.35200	46 53		0.31971	100 125		4
76	1.00000	76		0.36153	46 53		0.31078	100 125		4
78	0.77159	45 46 76 131 210		0.61132	100 196 221		0.52772	45		4
83	1.00000	83	locally eff.	0.30583	46 53		0.12077	103		3
84	1.00000	84	locally eff.	0.27599	207		0.26888	103		3
85	1.00000	85	locally eff.	1.00000	85	locally eff.	0.10838	103		2
86	1.00000	86	locally eff.	0.33126	46 53		0.16316	103		3
93	0.60112	45 46 76 131		0.71798	100 196 221		0.55912	45 100		4
95	1.00000	95	locally eff.	0.31710	46 53		0.15341	103		3
98	1.00000	98	locally eff.	0.44361	207		1.00000	98	locally eff.	3
100	1.00000	100		1.00000	100		1.00000	100		4
103	1.00000	103	locally eff.	0.44139	196 207		1.00000	103		3
105	0.50767	46 131 210 217		0.65357	207		0.39596	45 217		4
110	0.70307	7 62 193		0.36270	196 221		0.30384	45		4
113	1.00000	113	locally eff.	0.32354	46 53		0.14210	103		3
115	0.62486	152 174 217		0.60497	100 196 221		0.63983	45 217		4
121	1.00000	121	locally eff.	0.35047	46 53		0.11456	103		3
125	0.35004	45 100 152		0.97340	46 53 196		1.00000	125		4
126	0.71705	7 46 50 51 140		0.59395	100 196 221		0.52596	45 217		4
127	1.00000	127		0.33133	46 53		0.28610	100 125		4
128	1.00000	128		0.13037	15		0.04659	103		2
130	1.00000	130	locally eff.	0.32431	46 53		0.13988	103		3
131	1.00000	131		0.59530	100 196 221		0.84278	45 100 217		4
132	0.57406	51 127 140		0.36288	46 53		0.28538	100 125		4
133	0.78873	45 46 50 76 100		0.67852	100 196 221		0.77101	45 100 217		4
137	0.70913	45 46 50 76 100		0.63224	100 196 221		0.72495	45 217		4
139	0.64247	45 46 76 131		0.69907	221		0.61289	45 100		4
140	1.00000	140		0.31064	46 53		0.27153	100 125		4
142	1.00000	142	locally eff.	0.32615	46 53		0.13536	103		3
152	1.00000	152		0.62595	100 196 221		0.82462	45 217		4

PART II: FURTHER EVALUATION OF EFFICIENCY OF MUTUAL FUND MANAGEMENT COMPANIES

154	0.78907	7 76 140 193	0.35833	46 53	0.28402	100 125	4
156	0.74473	45 174 193 210 217	0.59156	100 196 221	0.52874	45 217	4
159	0.66175	51 174 193	0.57533	100 196 221	0.48137	45 217	4
160	0.73983	51 62 193	0.54640	207	0.50979	45	4
161	0.70324	45 76 193	0.34344	46 53	0.28188	125	4
162	0.78837	51 76 140 193	0.36431	196 221	0.29290	100 125	4
163	0.65303	51 140 193	0.31209	196 221	0.27689	100 125	4
168	0.70472	7 50 51 62 193	0.61646	196 207	0.62107	45 217	4
173	0.62145	7 140 193	0.34721	46 53	0.20992	125	4
174	1.00000	174	0.67633	196 207	0.61036	45 217	4
176	0.80912	51 76 140 193	0.35737	46 53	0.29385	100 125	4
177	0.57134	45 152 174 217	0.62044	207	0.56870	45 217	4
182	0.61293	51 127 140	0.35425	46 53	0.28889	100 125	4
185	0.48127	127	0.34498	46 53	0.16341	45	4
190	0.81471	7 51 140 193	0.35306	196 221	0.28141	100 125	4
191	0.82468	46 50 76 100	0.63640	196 221	0.58315	45 100	4
192	0.66503	45 174 193 217	0.58396	100 196 221	0.54223	45 217	4
193	1.00000	193	0.39889	196 221	0.28332	100 125	4
194	0.73075	7 62 76 193	0.45032	196 221	0.31870	45	4
195	0.62920	76 131 193 210	0.50625	196 221	0.29604	45	4
196	0.39988	45 50 100 174	1.00000	196	0.70665	100 125	4
197	0.64604	174 193 217	0.60156	196 207	0.50871	45 217	4
198	0.74111	51 174 193	0.57329	100 196 221	0.48466	45 217	4
200	0.39610	7 46 62	0.62489	46 53	0.56407	45 100	4
203	0.64785	46 131 217	0.61395	100 196 221	0.66650	45 217	4
204	0.71249	45 76 131 193 210	0.59055	196 221	0.58105	45	4
206	0.67473	45 46 50 100 174	0.60713	100 196 221	0.62200	45 217	4
207	0.45493	131 193	1.00000	207	0.58658	45 217	4
210	1.00000	210	0.59377	100 196 221	0.56867	45 217	4
217	1.00000	217	0.92002	46 53	1.00000	217	4
221	0.51774	46 76 100	1.00000	221	0.60895	100 125	4

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Appendix L

Table II-L-1 Efficient Companies by Clusters

Year	MMS	Clusters	Total Companies	Globally Efficient	Globally Inefficient	Locally Efficient
2005	Stage 1	1	2	2	0	0
		2	5	1	4	4
		3	16	1	15	12
		4	3	1	2	2
		5.1	10	6	4	2
		5.2	12	3	9	3
	5.3	54	14	40	3	
	Stage 2	1	2	0	2	0
		2	5	0	5	0
		3 and 4	19	0	19	0
Overall Eff.	5	76	7	69	0	
	1	2	2	0	0	
	2	5	0	5	1	
Overall Eff.	3 and 4	19	0	19	0	
	5	76	8	68	0	
	2006	Stage 1	1	2	2	0
2			5	1	4	4
3			18	2	16	11
4.1			20	4	16	4
4.2			57	13	44	3
Stage 2		1	2	0	2	0
		2	5	0	5	0
		3	18	0	18	0
		4	77	8	69	0
Overall Eff.		1	2	1	1	0
	2	5	0	5	1	
	3	18	1	17	2	
	4	77	6	71	1	
2007	Stage 1	1	2	2	0	0
		2	4	1	3	3
		3	19	3	16	11
		4	70	15	55	5
	Stage 2	1	2	0	2	0
		2	4	0	4	0
		3	19	0	19	0
		4	70	6	64	0
	Overall Eff.	1	2	0	2	0
		2	4	0	4	0
3		19	1	18	1	
4		70	5	65	1	
2008	Stage 1	1	3	3	0	0
		2	4	1	3	3
		3	17	2	15	10
		4	69	7	62	3
	Stage 2	1	3	0	3	0
		2	4	0	4	0
		3	17	0	17	0
		4	69	2	67	0

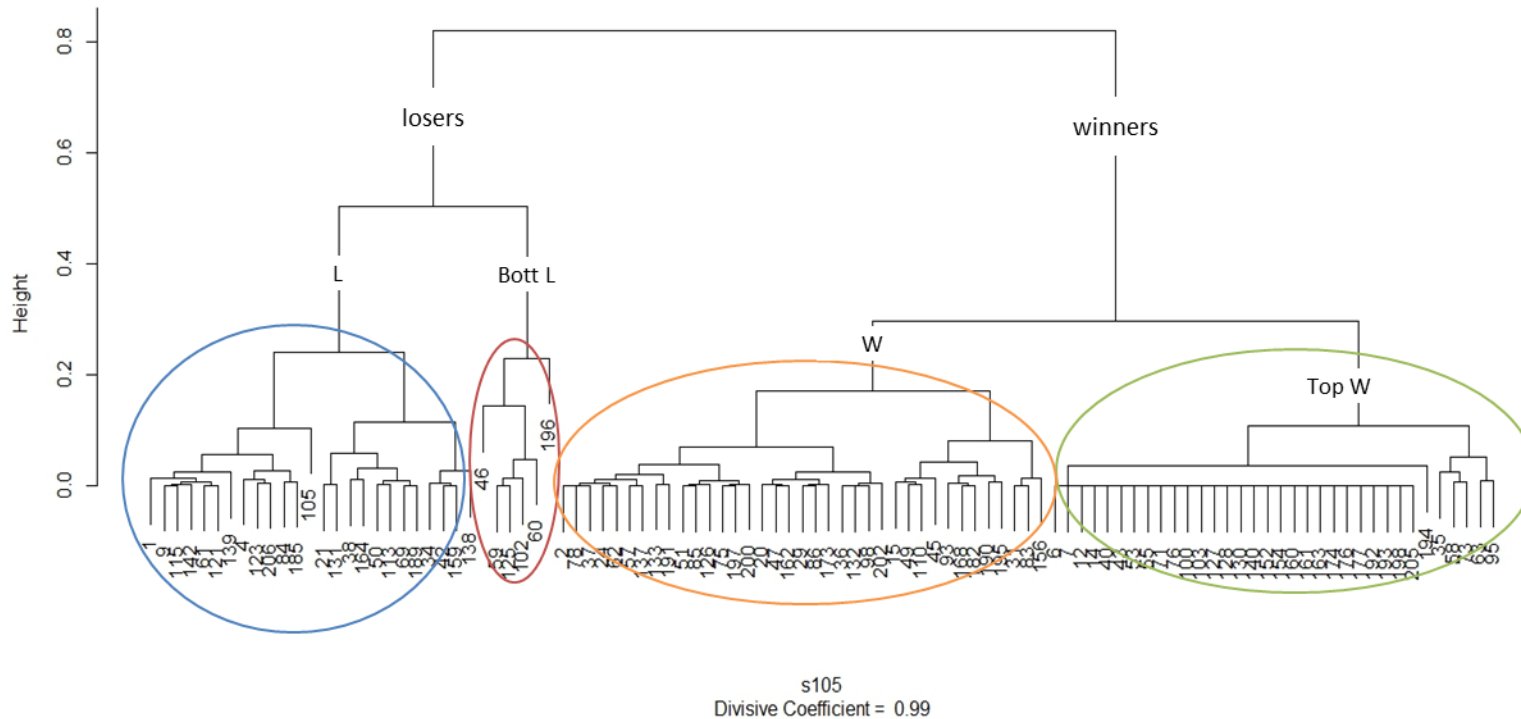
PART II: FURTHER EVALUATION OF EFFICIENCY OF MUTUAL FUND MANAGEMENT COMPANIES

		1	3	0	3	1
	Overall	2	4	0	4	0
	Eff.	3	17	1	16	2
		4	69	4	65	0
		1	3	1	2	1
	Stage 1	2	4	2	2	2
		3	14	1	13	13
		4	74	16	58	1
2		1	3	1	2	0
0	Stage 2	2	4	0	4	3
0		3	14	0	14	0
9		4	74	6	68	0
		1	3	0	3	0
	Overall	2	4	0	4	0
	Eff.	3	14	1	13	1
		4	74	4	70	0

Highlighted in grey we can see that whole the sample of globally inefficient companies become in locally efficient.

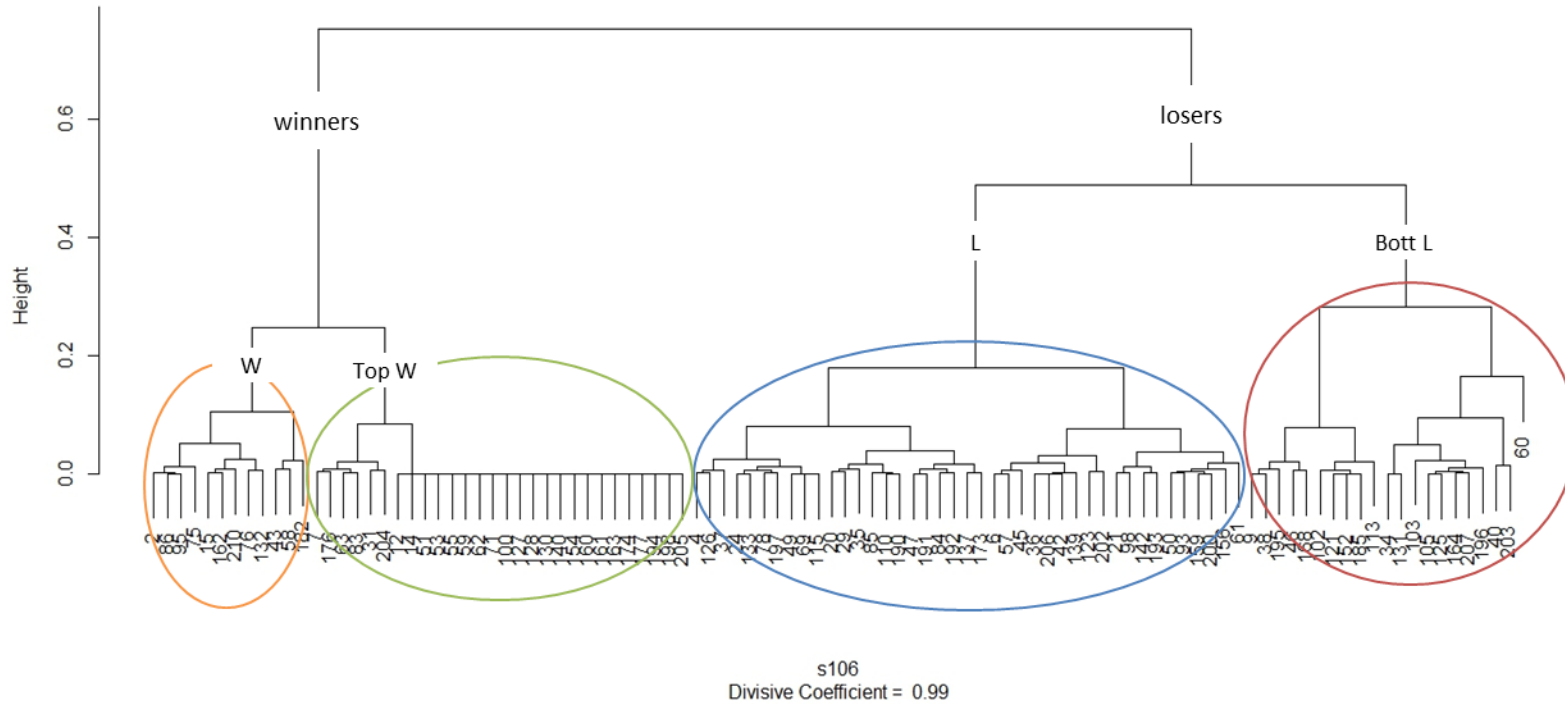
Appendix M

Figure II-M-1 Dendrogram DIANA of efficiency scores for stage 1 of 2005



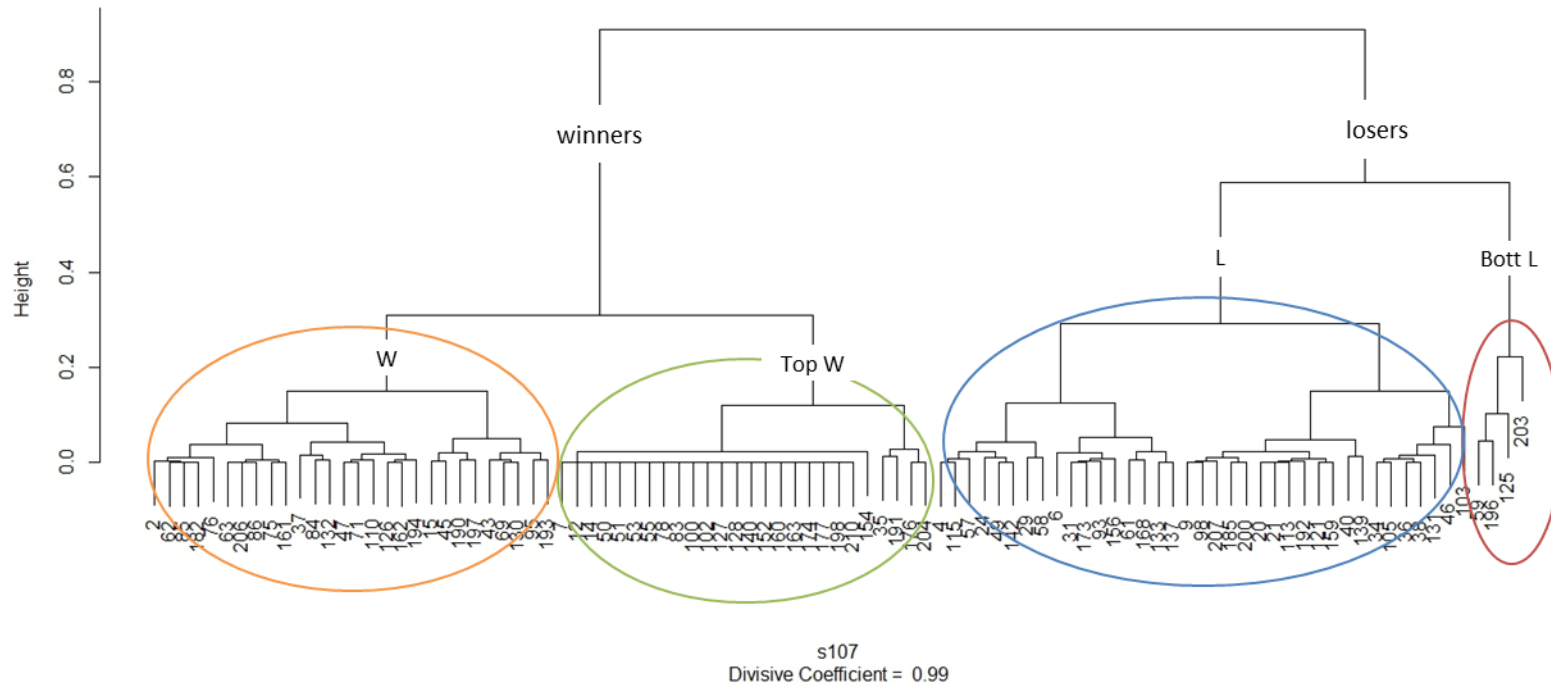
The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-M-2 Dendrogram DIANA of efficiency scores for stage 1 of 2006



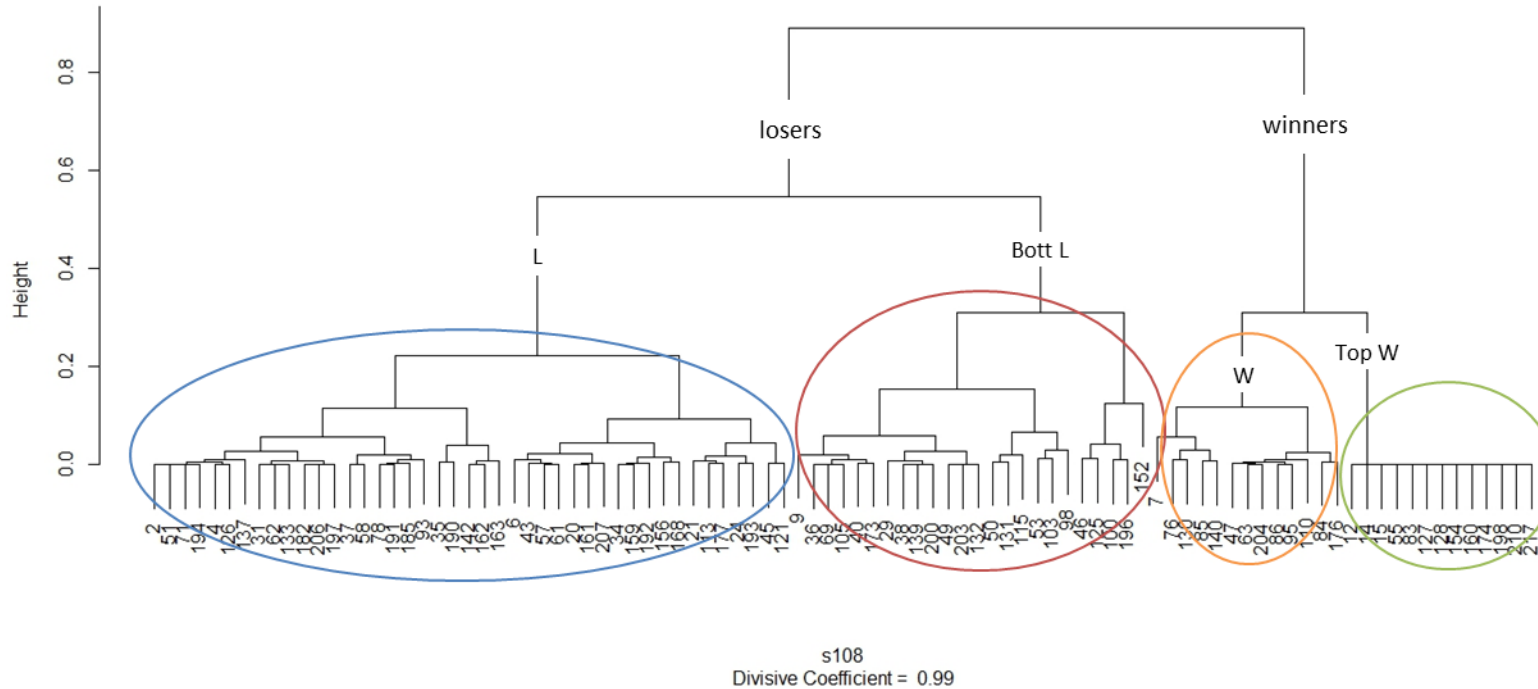
The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-M-3 Dendrogram DIANA of efficiency scores for stage 1 of 2007



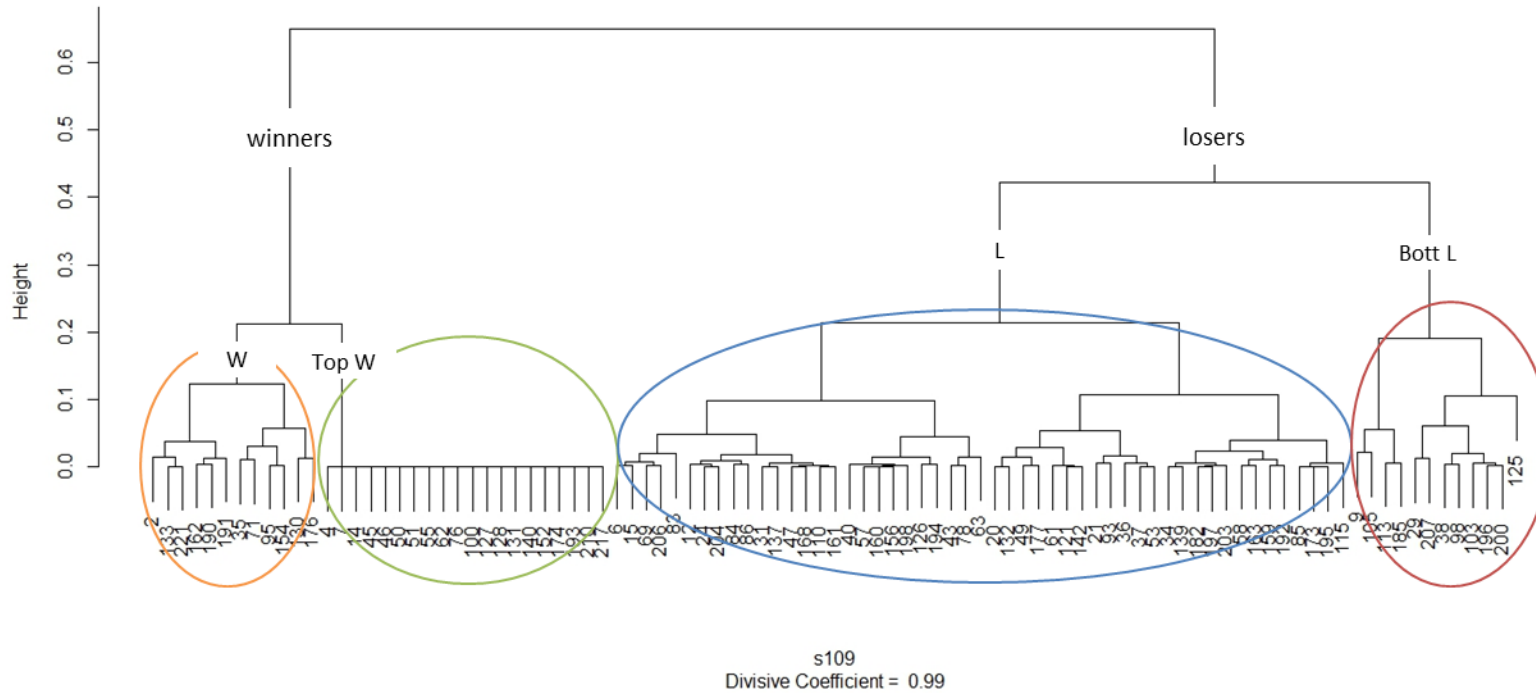
The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-M-4 Dendrogram DIANA of efficiency scores for stage 1 of 2008



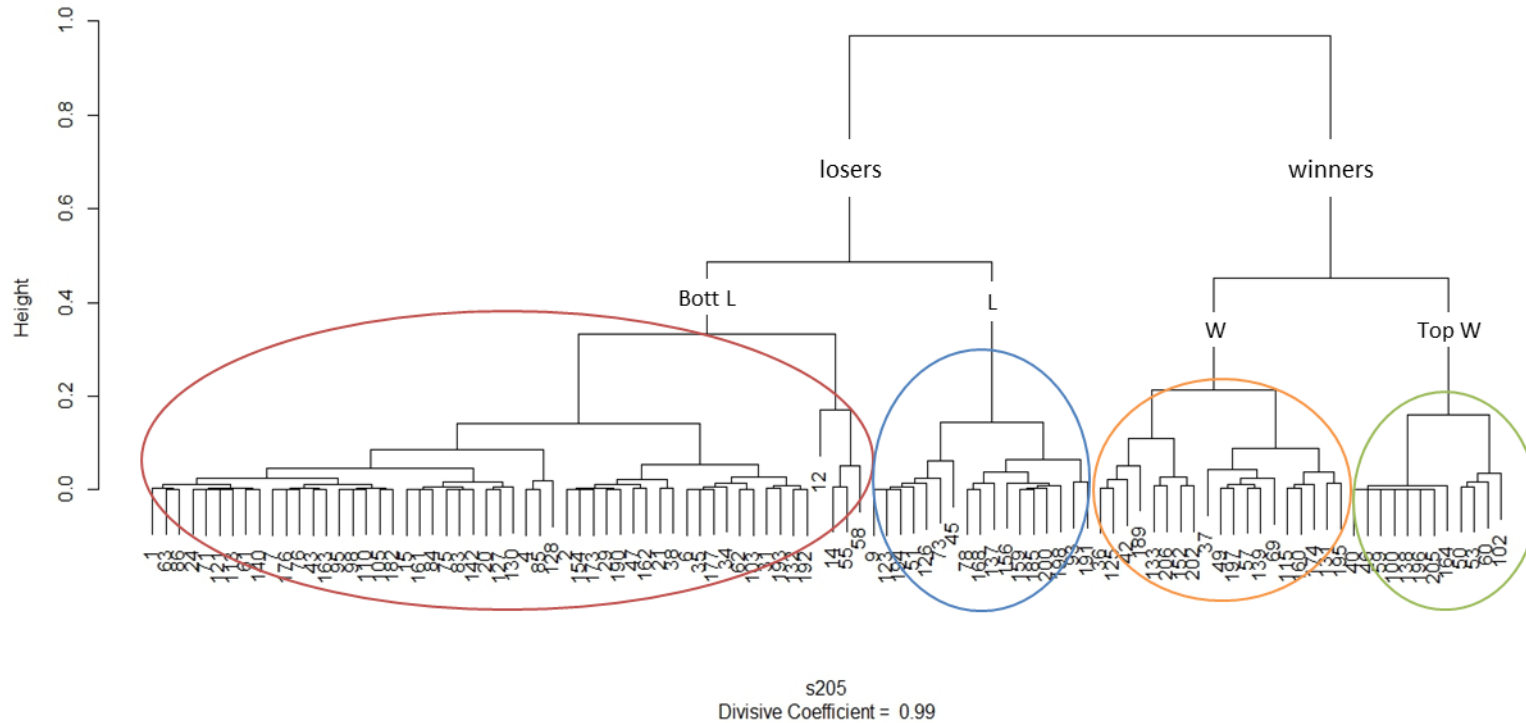
The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-M-5 Dendrogram DIANA of efficiency scores for stage 1 of 2009



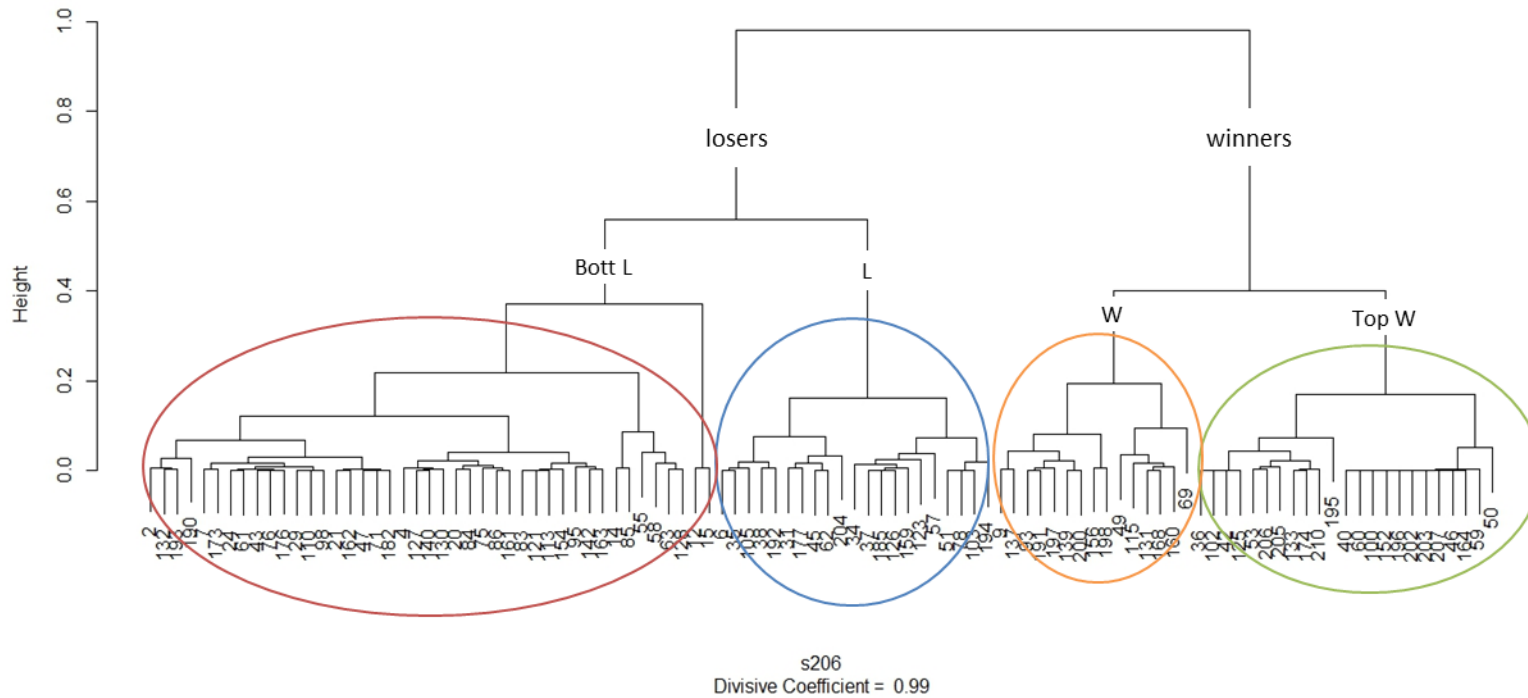
The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-M-6 Dendrogram DIANA of efficiency scores for stage 2 of 2005



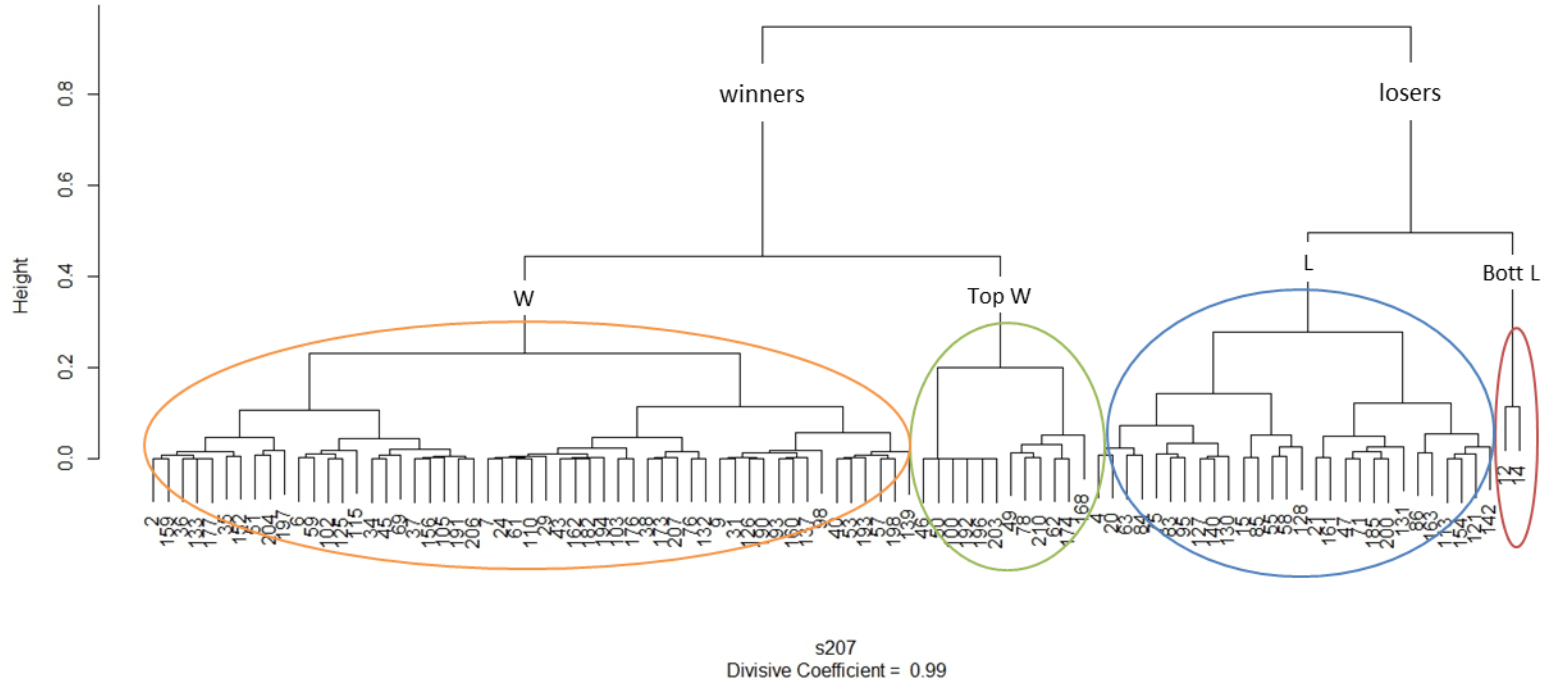
The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-M-7 Dendrogram DIANA of efficiency scores for stage 2 of 2006



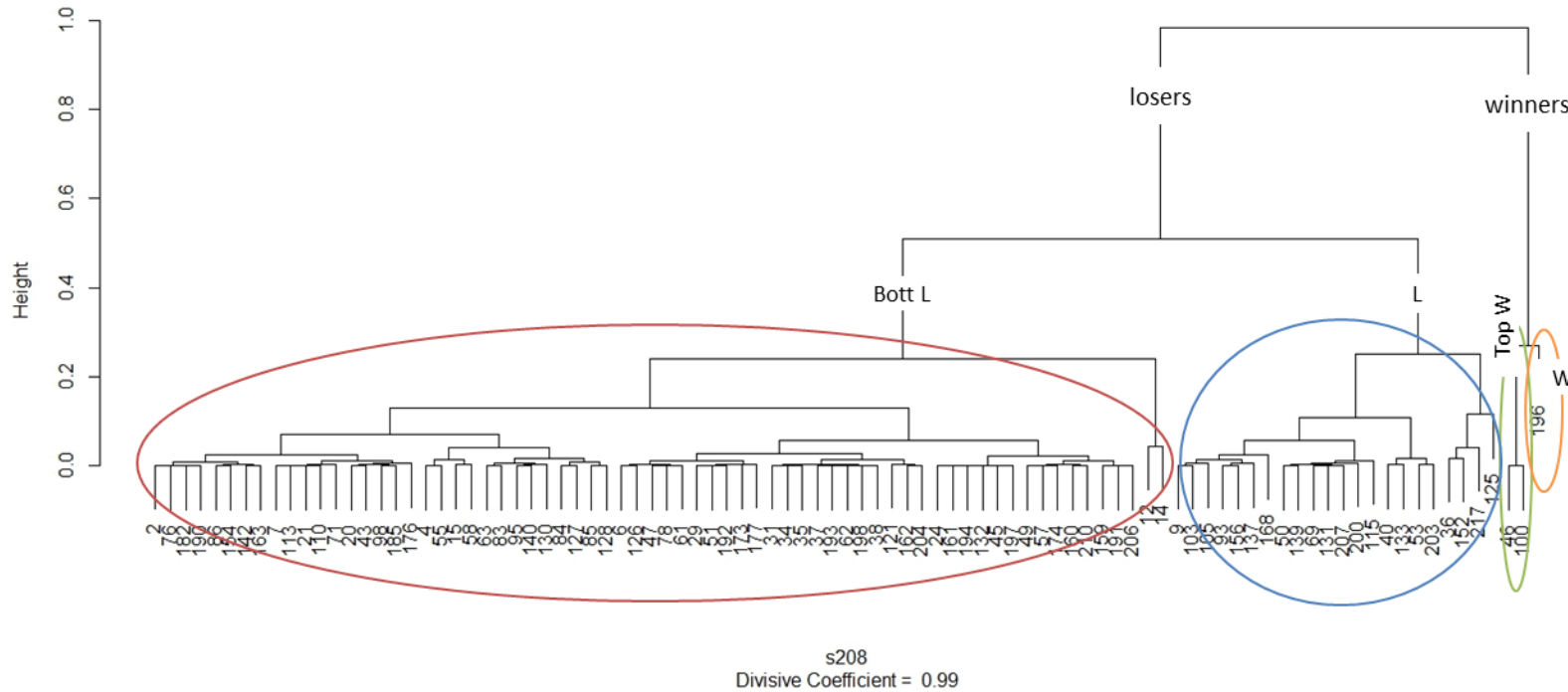
The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-M-8 Dendrogram DIANA of efficiency scores for stage 2 of 2007



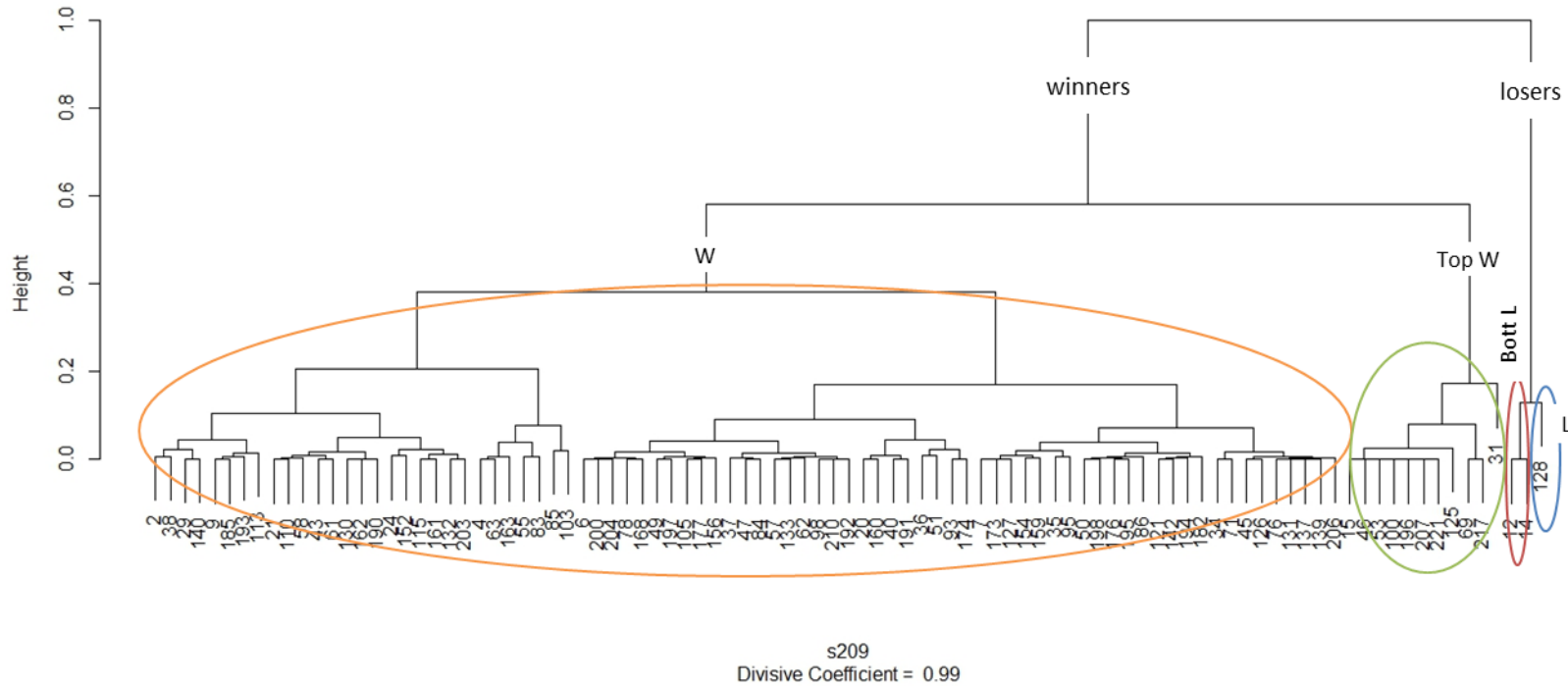
The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-M-9 Dendrogram DIANA of efficiency scores for stage 2 of 2008



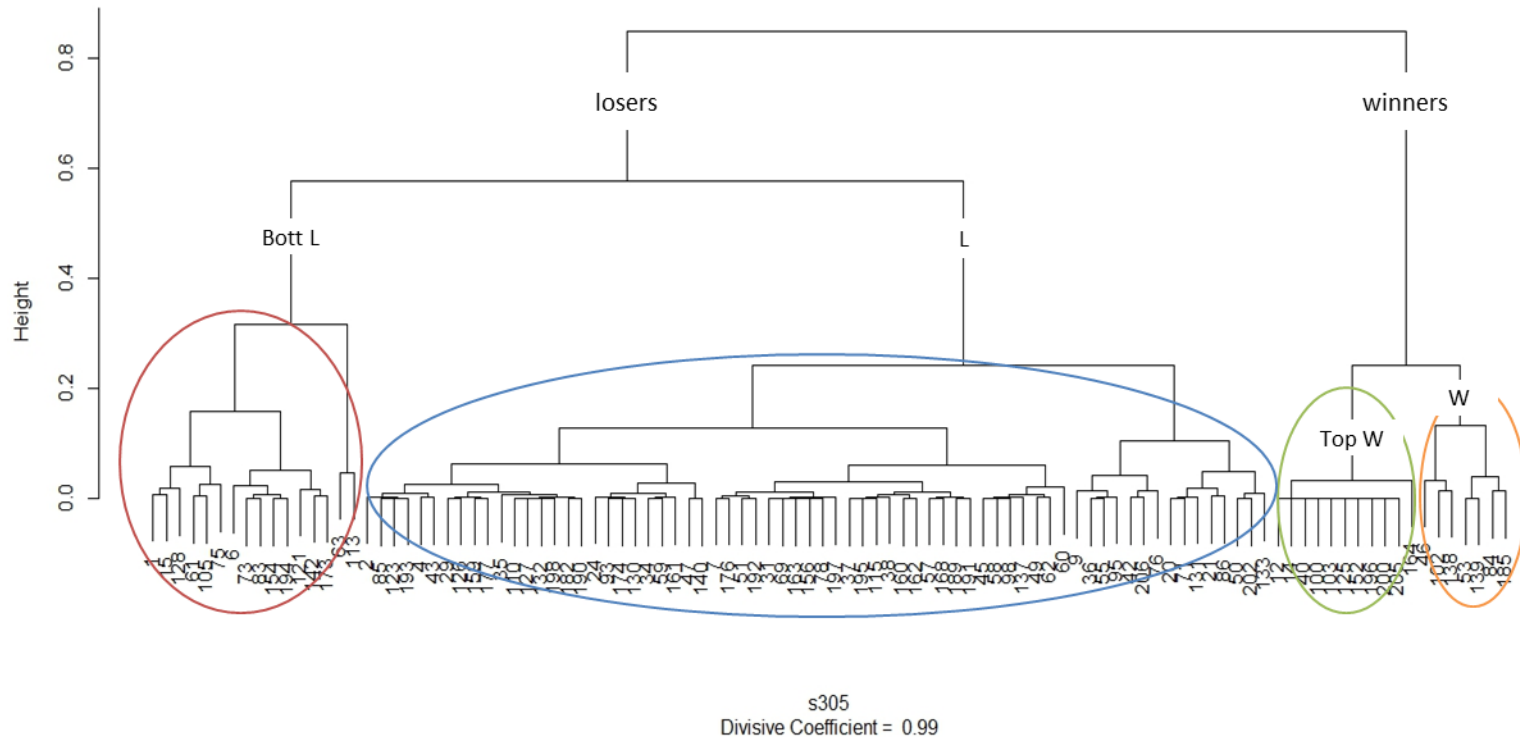
The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-M-10 Dendrogram DIANA of efficiency scores for stage 2 of 2009



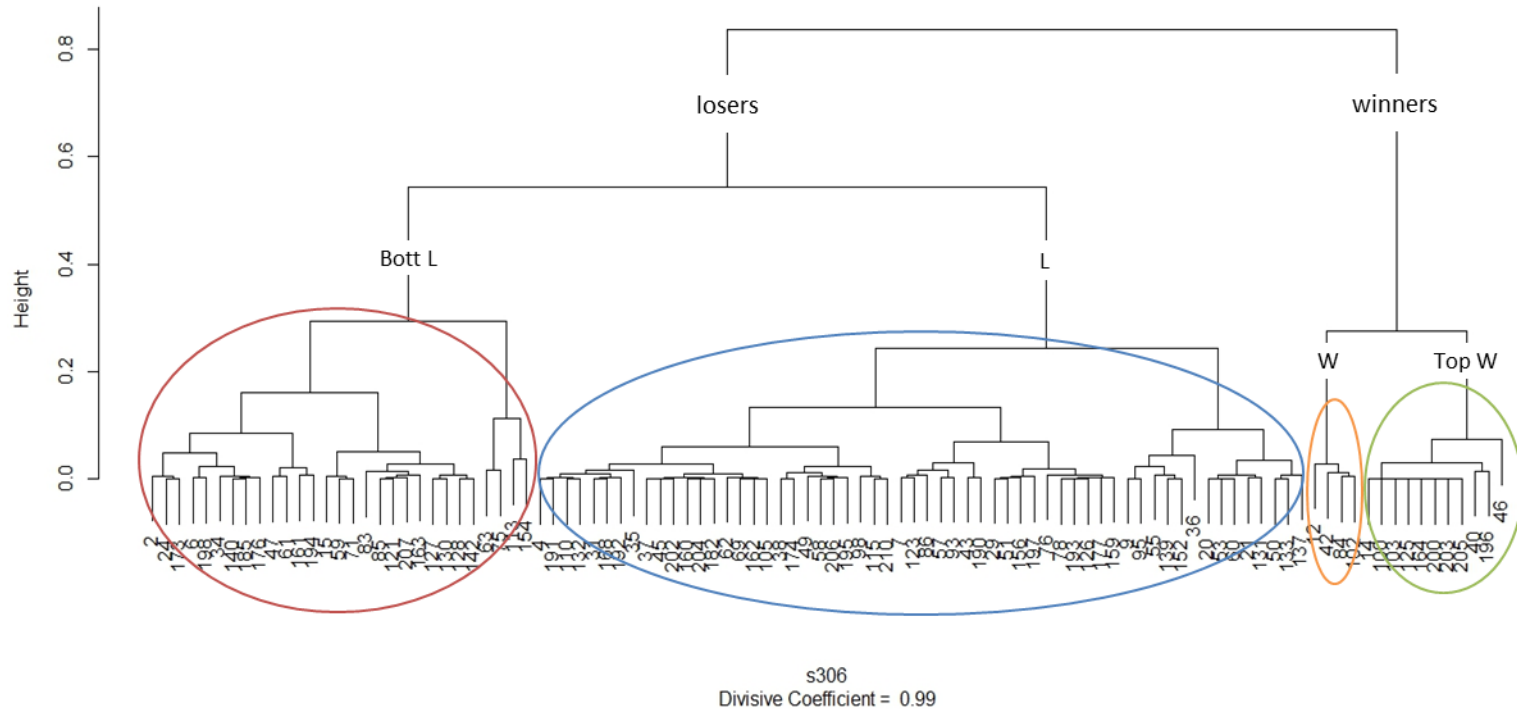
The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-M-11 Dendrogram DIANA of efficiency scores for overall efficiency of 2005



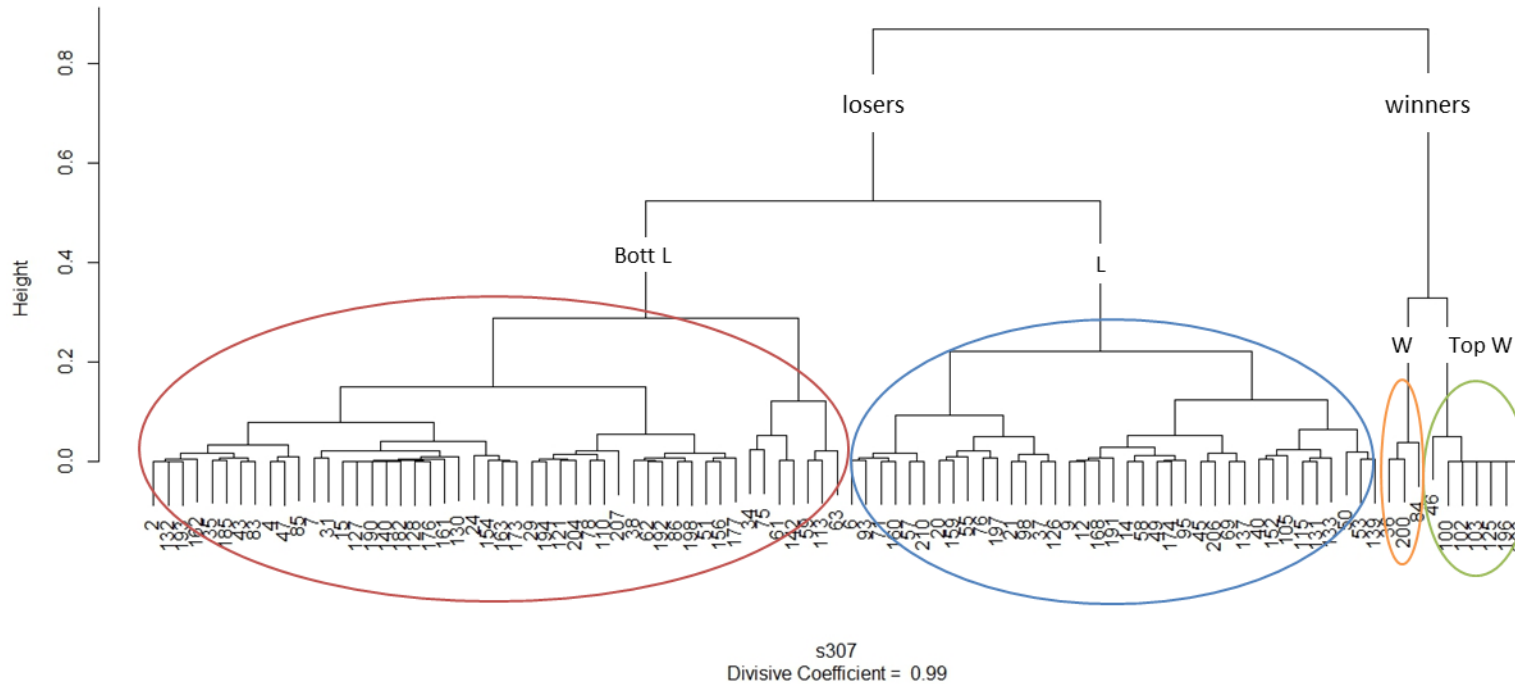
The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-M-12 Dendrogram DIANA of efficiency scores for overall efficiency of 2006



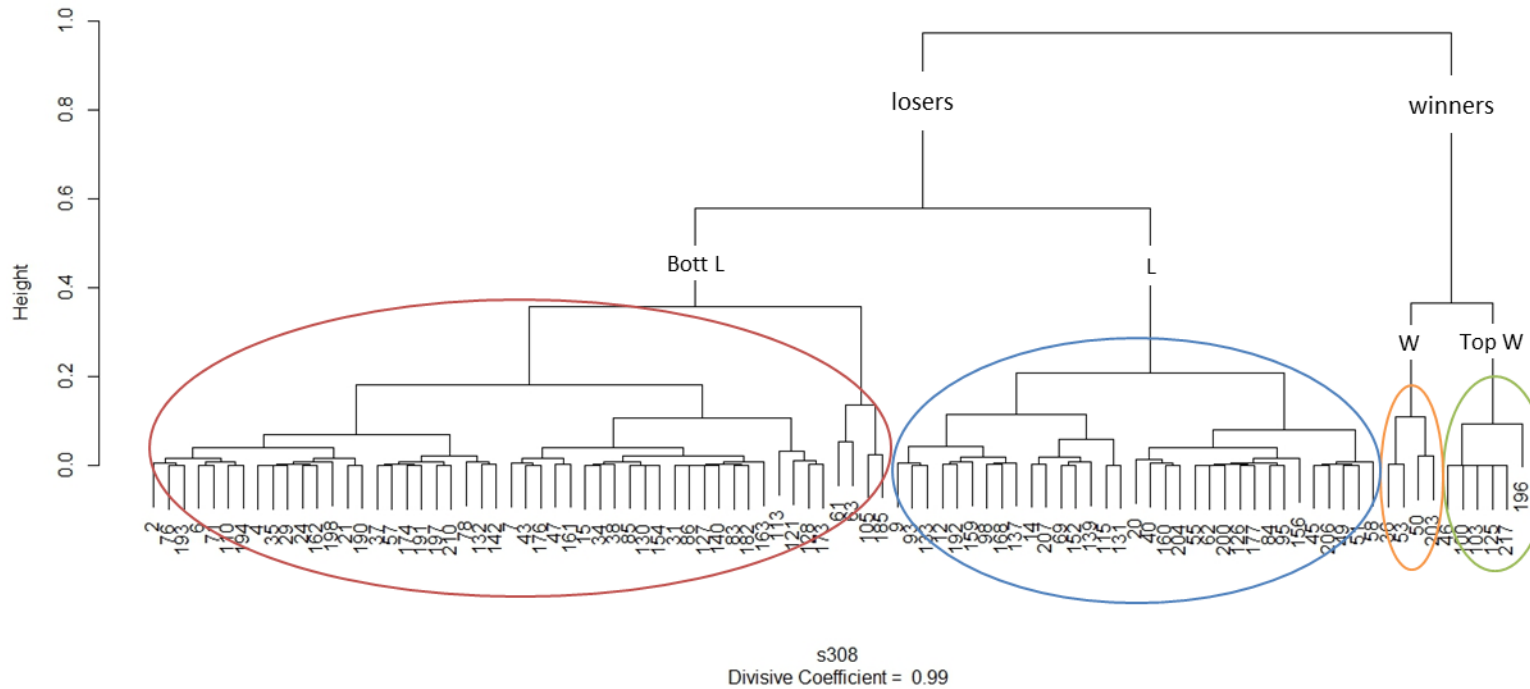
The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-M-13 Dendrogram DIANA of efficiency scores for overall efficiency of 2007



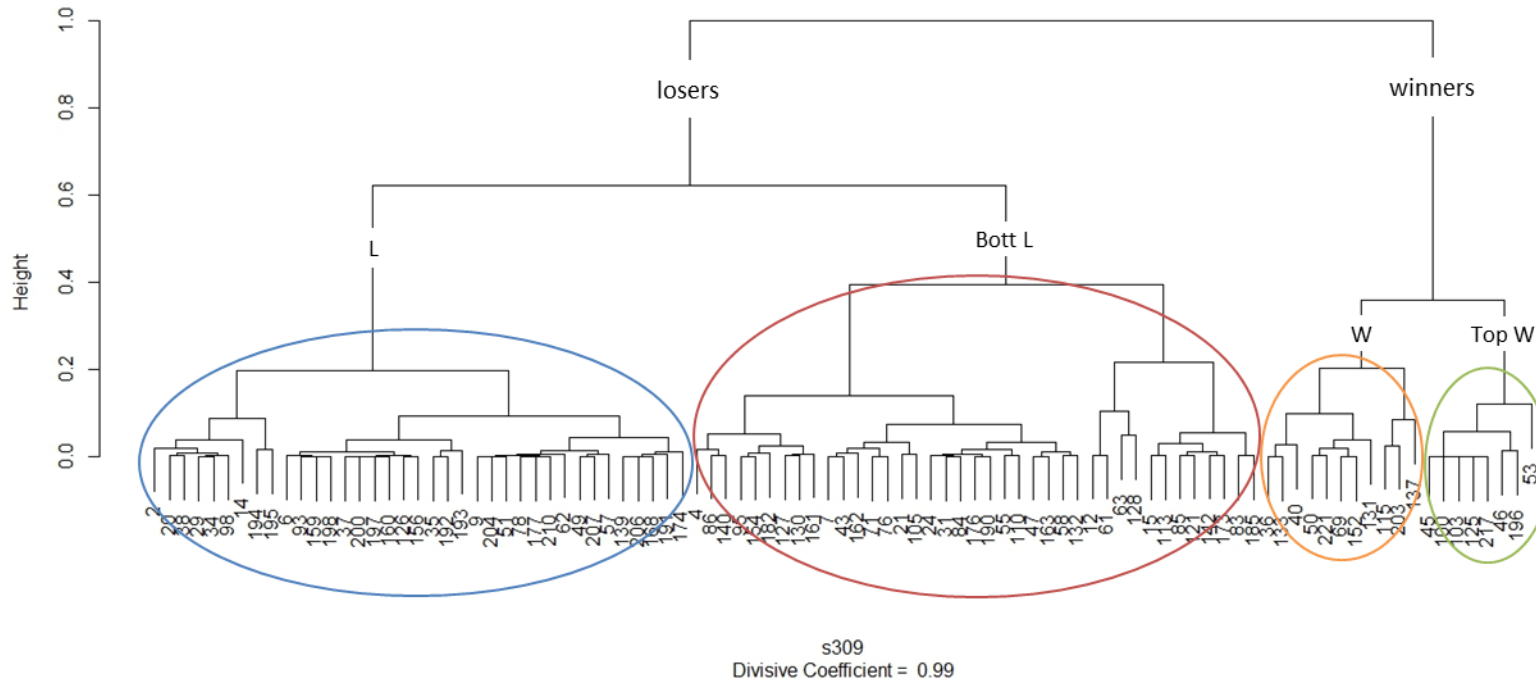
The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-M-14 Dendrogram DIANA of efficiency scores for overall efficiency of 2008



The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Figure II-M-15 Dendrogram DIANA of efficiency scores for overall efficiency of 2009



The code classified is the identifying number of the mutual fund management company (Official registers, CNMV). “winners” = cluster of the companies with best efficiency scores, “losers” = cluster of the companies with poor efficiency scores, this last for 2x2 contingency tables; “Top W” = cluster of the companies with top efficiency scores, “W” = cluster of the companies with high efficiency scores, “Bott L” = cluster of the companies with bottom efficiency scores, “L” = cluster of the companies with low efficiency scores, this last for 4x4 contingency tables. Metric used was Manhattan. The Divisive Coefficient (DC) is defined by $DC = \frac{1}{n} \sum_{i=1}^n \iota(i)$. For each object i it measures the length $\iota(i)$ of its line in the banner, with respect to normalized scale, i.e. all $\iota(i)$ lie between 0 and 1.

Final conclusions and further research

The main conclusions and contributions of this dissertation are summarised in this section. The main objective of the thesis was to fill a gap in the financial literature and to shed additional light by analysing the efficiency of mutual fund companies in Spain, one of the most relevant fund industries in the Euro market. The important market concentration of Spanish mutual fund companies was a challenging feature in the effort to obtain an appropriate evaluation for this industry. In this industry, a few, very large and well-diversified Mutual Fund Management Companies coexist together with a huge number of small managers specialising in fund strategy by sector and/or geographical area. Therefore, the question arose as to how to select an accurate methodology and management variables to analyse appropriately so heterogeneous a set of Spanish mutual fund companies.

To address the above issues, the first part of the thesis presented a review of the early DEA efficiency literature, a brief explanation of the basic efficiency models and a review of major contributions for the financial industry, finishing with a discussion of the two most traditional approaches to measuring the efficiency of financial institutions. The study then incorporated an innovative model that includes three interacting management subsystems within a mutual fund company: *Portfolio Management*, *Marketing and Service*, and *Overall Efficiency*. This model provides a successful interaction between the inputs and outputs selected for the different units, thereby attempting to overcome the traditional debate between the production and intermediation approaches in banking and insurance companies. The application of a non-oriented frontier approach based in the slacks proposed by Tone (2001) has been discussed in detail and tested in the efficiency scores found in our sample based on the Mutual Fund Management Companies. The consistency of the efficiency results obtained using additional measures proved the robustness of the results.

The results provided evidence for the low impact of portfolio management abilities on the efficiency of the marketing and sales process of the mutual funds managed by the

company. Furthermore, the results supported evidence that the best-managed funds are not the most efficient sellers of their products, thereby considerably reducing overall profits for shareholders. Analysing the influence of returns-to-scale, we found that size appears to be an important variable for successfully selling funds, which takes on more relevance in a concentrated market with heavy bank participation such as the Spanish fund industry.

As was explained, the DEA methodology and the following extensions fail to identify ‘best practice’ competitors who can serve as benchmarks of the companies analysed. This limitation could therefore limit the accuracy of some results in the previous literature. This is a relevant problem in industries where competitors show assorted characteristics, such as the Spanish fund industry. To provide a solution to the above problem, the second part of the thesis tries to overcome this limitation by using recent and unexplored variations of the original slacks-based measure (SBM) proposed by Tone (2010).

Thus, the second part of the thesis starts with a review of the major concepts of the SBM Variations. These variants were developed as a complementary methodology. Finally, we made a comparative analysis of the results of these variants with the original model. We then developed a study of persistence phenomena based on the previously found efficiency scores. Finally, those factors that might drive persistence results were analysed.

We do not find significant differences between the efficiency scores of the original SBM found in the first part and the SBM variations of the second part. Additionally, the results obtained using the more refined evaluation proposed by Tone (2010) in the SBM Variations model, i.e., based in the clustering process approach, showed evidence that an appropriate SBM evaluation of the *Portfolio Management stage* should consider homogeneous sets of companies with similar management resources. However, this issue appeared to be less important to both the *Marketing* and *Overall stages*.

Furthermore, we analysed whether the patterns of efficiency obtained were persistent over time using a non-parametric test. The results provided evidence of significant persistence in the efficiency records obtained by the companies in the different

management stages. Additionally, analysing in more detail with 4x4 contingency tables, we found significant persistence of those companies included in the “Top Winners” and “Bottom Losers” clusters over several years. This finding indicated that the best-managed and worst-managed companies persist over time; this result was robust for the portfolio management and distribution of mutual funds as well as for the overall efficiency of the company as a whole. Otherwise, the study of the residuals of the main diagonals of the 4x4 contingency tables showed that the main diagonal of the *Overall Efficiency stage* provides positive and significant values in all years, indicating that the efficiency clusters remain quite similar without many changes. This finding indicated that the efficiency scores are unlikely to change to a very different efficiency cluster.

In the study of the effect of the companies’ resources on the different categories of persistence studied, we found a high persistence for the biggest companies located in the extreme cell of Top Winners–Top Winners for the *Portfolio Management stage*. Moreover, the aggregate concentration in the extreme cell of Bottom Losers–Bottom Losers shows not-significant differences in terms of some variables. An opposite pattern was found in the *Marketing and Service stage*, where size as a proxy for resources seems to play a significant and negative role in the persistence of the stage. The same pattern was also found in the *Overall Efficiency stage*, although the results are not so significantly clear for all persistence cells and variables. The above results suggest that the marketing and service of the fund companies considerably affects persistence of the overall efficiency of the companies.

In summary, the thesis contributes to the literature in several ways. First, we developed a brief but complete literature review of measures of efficiency and approaches commonly used in the literature. Based on different approaches to core competencies of the mutual fund companies, the study included an innovative model that considered multiple management stages. Within the DEA methodology, we applied a measure for the first time because basic concepts of the most sophisticated measure were developed recently. This allowed us overcome some of the limitations present in the frontier efficiency studies because of inappropriate benchmarking of the companies analysed. In addition, we found

interesting results of the persistence phenomenon in the mutual fund industry that will contribute in several ways to the studies of efficiency in the fields of banking and insurance.

Several areas can be addressed in other further studies, such as the following: the use of alternative variables to measure the inputs and outputs that lead to an efficiency and productivity complementary analysis; developing a study with a different time horizon, which could consider analyses by sectors or by investment vocations; testing efficiency scores with other models or using parametric measures; or evaluating the productivity of companies over the time, and comparing these results with other sectors or other companies in the same industry.

Resumen y Conclusiones

A continuación se presenta un resumen en español de la tesis doctoral con el fin de dar cumplimiento a la normativa de tesis doctorales escritas en otros idiomas de la Universidad de Zaragoza. El este apartado se dará una visión general de la tesis, sin entrar en detalles de la versión en inglés, y sin hacer extensivas referencias bibliográficas y/o insertar gráficos y tablas.

La tesis estuvo compuesta de dos partes principales. La primera titulada Eficiencia en las Empresas Españolas Gestoras de Fondos Mutuos: Enfoque Basado en los Slacks, estuvo compuesta de cuatro secciones o capítulos de la siguiente manera: La sección 2 ofreció una revisión de la literatura temprana del Análisis Envolvente de Datos (DEA), una breve explicación de los modelos básicos, una breve reseña de las principales contribuciones a la eficiencia en las instituciones financieras, y una discusión de los dos métodos más populares. La sección 3 describió el modelo teórico propuesto y las variables utilizadas en el análisis. Sección 4 ilustró los datos, el análisis empírico y los resultados, la influencia de las variables-retornos a escala, y los análisis de robustez. Finalmente, la sección 5 concluyó y resumió las principales conclusiones del estudio.

La segunda parte llamada Evaluación Adicional de la Eficiencia en las EGFMs, se compuso de tres secciones o capítulos, la sección 2 se examinaron los principales conceptos de las variaciones al enfoque SBM original y los resultados empíricos de estas variaciones en las EGFMs españolas. La sección 3 ilustró el fenómeno de la persistencia en los scores de eficiencia y determinó aquellos factores relevantes que podrían potencialmente conducir a los resultados de persistencia obtenidos por las empresas a lo largo de nuestro horizonte de estudio. Finalmente, la Sección 4 concluyó y resumió las principales conclusiones del estudio.

Primera Parte de la Tesis

Resumen

Desde principios de 1990 hasta mediados de 2007, la industria de fondos mutuos mostró un notable crecimiento en todo el mundo. La exitosa expansión implicó tanto la proliferación de fondos mutuos y empresas de gestión, como la creación de un gran número de fondos de inversión. Más tarde, cuando los mercados experimentaron la crisis financiera de las hipotecas sub-prime, la industria de fondos de inversión sufrió una disminución significativa en el total de activos bajo gestión, sobre todo en 2008, impulsado principalmente por una caída en las inversiones financieras de los hogares. Ello se debió a una mayor aversión al riesgo provocado por la alta volatilidad y a una caída de la confianza en los instrumentos financieros como consecuencia de la crisis. Este escenario del mercado detuvo la creación de nuevas empresas de gestión y en otros casos a su fusión o cierre, y por consiguiente a la reordenación del mapa competitivo de la industria. Por consiguiente la caída de activos bajo la administración de las gestoras, al igual que el número de inversores se explica no sólo por la mala imagen de los fondos de inversión, sino también por la falta de confianza en los mercados globales. En muchos casos, esta situación obligó a las instituciones financieras a repensar sus estructuras de producción y a la creación de nuevos productos no sólo por su propia iniciativa, sino también impulsados por las nuevas exigencias legislativas y normativas impuestas por las autoridades de supervisión que intentaban restaurar la confianza en los mercados financieros.

Más recientemente, los activos de fondos de inversión de todo el mundo han estado regresando lentamente a las cifras de éxito alcanzados antes de la crisis, es decir, \$ 28.4 billones administrados en septiembre de 2012, lo que representó más de un tercio del PIB mundial. Por otro lado esta industria emplea a mano de obra calificada, tiene efectos indirectos en otros sectores, proporciona liquidez importante para el sistema financiero y da riqueza para los inversores minoristas e institucionales. El análisis de la eficiencia de la industria de fondos mutuos ha sido muy similar a la extensamente desarrollada en la banca y los seguros. Es claro afirmar que si los bancos y compañías de seguros son más

productivos, entonces pueden obtener mejor rendimiento y ofrecer así productos nuevos y más seguros a sus clientes con precios más bajos. De acuerdo con este argumento, las Empresas de Gestoras de Fondos Mutuos (EGFMs) deben trabajar para ofrecer una amplia gama de fondos de alto rendimiento con características diversas de inversión para los diferentes tipos de inversores, manteniendo las tasas y los gastos lo más bajo posible. Por lo cual la expansión de una apropiada gestión debe dar lugar a mayores niveles de eficiencia global en la industria de fondos mutuos. Es importante señalar que las actividades de los fondos de inversión reducen la exposición de los bancos, a economías de escala y por consiguiente a un aumento de rentabilidad, repercutiendo todo ello en la mejora de su rendimiento operativo (Gallo et al, 1996;. Asaftei, 2008).

El análisis de las diferencias de productividad entre las empresas en los últimos años podría hacer posible identificar el éxito o fracaso de las iniciativas de gestión y también puede poner de relieve las diferentes estrategias adoptadas por las empresas durante la crisis financiera.

En la literatura observamos que una amplia investigación se ha dedicado a la productividad en las instituciones financieras, por lo que sabemos, sólo Zhao Yue (2010) y Medeiros (2010) han estudiado la eficiencia de las empresas de fondos de inversión y compañías de fondos de pensiones, respectivamente. Por una parte, Zhao y Yue (2010) examinan la eficiencia de los fondos en china basados en sus competencias básicas, analizando tanto la inversión / investigación y los subsistemas de marketing / servicio. Mientras que Medeiros (2010) analiza los cambios en la productividad total de una muestra de empresas portuguesas de fondos de pensiones desde 1994 hasta 2007 a través las medidas DEA y el índice de Malmquist. Una posible explicación de esta escasa literatura puede ser la dificultad de identificar las variables específicas para la evaluación adecuada de estas empresas, sin llegar sólo a la réplica de los estudios anteriores centrados bancos y compañías de seguros. Desarrollar modelos de evaluación adecuados para las empresas de fondos de inversión para disponer de un abanico de posibilidades con variables específicas de la industria, complementaría los modelos analizados en otros sectores financieros. Por lo tanto, se convierte en un reto el desarrollar nuevas propuestas específicas para la industria

de fondos, que consideren adecuada selección de inputs y outputs propios para las empresas gestoras en lugar de la simple replica de los modelos previamente estudiados en banca y seguros.

En el presente estudio se llena este vacío de la literatura financiera, pretendiendo arrojar una luz adicional mediante el análisis de la eficiencia de las empresas de fondos de inversión en España, la cual es una de las industrias de fondos más relevantes en el mercado Europeo. La coexistencia de pocas, muy grandes y bien diversificadas EGFMs, junto con un gran número de pequeños gestores especializados hace que sea difícil obtener evaluaciones apropiadas de la industria. Por lo tanto, se plantó una metodología propia para el estudio y una selección de variables específicas para analizar adecuadamente de modo heterogéneo el conjunto de empresas españolas del sector de fondos de inversión.

Para llevar a cabo el estudio, fue aplicado Análisis Envolvente de Datos (DEA), uno de los métodos más populares de las últimas décadas para la evaluación de la eficiencia en el sector financiero (ver por ejemplo estudios como los de: Berg et al, 1991; Berg et al, 1993 ; Schaffnit et al, 1997; Mlima y Hjalmarsson, 2002; Cummins et al, 2004; Casu et al, 2004; Cummins y Xie, 2008; Cummins et al, 2010; Holod y Lewis, 2011), y por otro lado el análisis de rendimiento de las carteras institucionales con enfoque alternativo a las tradicionales medidas de desempeño, es decir el desempeño de las carteras se trabajó con las relaciones funcionales entre rentabilidad y riesgo asociadas con la hipótesis de comportamiento (ejemplos de estudios: Murthi et al, 1997; Basso y Funari, 2001; Gregoriou et al, 2005; Eling, 2006; Lozano y Gutiérrez, 2008a, 2008b).

En esta primera parte del estudio se aplicó el modelo original (Tone, 2001) y un conjunto único de variables específicas propias para la industria de fondos que complementan los modelos tradicionales de la banca y los seguros, lo cual permitió una evaluación precisa y completa de la eficiencia global de las EGFMs.

Se utilizó este enfoque innovador para abordar una serie de cuestiones en relación con la eficiencia de las EGFMs en el mercado español y para discutir más a fondo las

implicaciones de los resultados obtenidos en este análisis se plantearon las siguientes preguntas: ¿Cuáles son las etapas claves de gestión que impulsan la eficiencia de una empresa gestora de fondos mutuos?, Es robusta la eficiencia a través de las diferentes etapas de gestión dentro de una empresa gestora de fondos mutuos?, ¿Cómo afecta los resultados las escalas de eficiencia?

Conclusiones

A nuestro entender, este estudio es el primer análisis de eficiencia en las EGFMs en una industria de fondos europeos relevante. Con base en el enfoque de sub-DMU de Holod y Lewis (2011), la tesis desarrolla un modelo que incluye tres subsistemas de gestión que interactúan dentro de una EGFMs originalmente propuestas por Berkowitz y Qiu (2003): Gestión de la Cartera, Marketing y Servicio, y Eficiencia Global. La interacción entre los inputs y outputs de las diferentes etapas trata de superar el debate tradicional entre el enfoque basado en la producción y el enfoque de la intermediación en el sector financiero, sobre todo en las empresas bancarias y de seguros.

Discutimos un conjunto específico de variables que se incluyeron en el modelo de subsistemas-múltiples. Las medidas de estas variables no fueron una sencilla repetición de los inputs / outputs tradicionalmente considerados en los estudios de banca y seguros, ya que fue necesario tener en cuenta las características particulares de las empresas de fondos para desarrollar un modelo adecuado. La consistencia de los resultados de eficiencia obtenidos utilizando diferentes medidas para las variables demostró robustez en nuestros resultados.

La aplicación de un enfoque de frontera eficiente no-orientado basado en los Slacks (SBM) propuesto por Tone (2001) se discutió en detalle para justificar su como una herramienta adecuada para obtener los scores de eficiencia de las EGFMs.

Los scores de eficiencia demostraron el bajo impacto de la capacidad de gestión de carteras en la eficiencia del proceso de comercialización y venta de los fondos de inversión

gestionados por la empresa. Por otra parte, los índices de eficiencia más bajos se encontraron en la etapa de comercialización y venta y ello pareció afectar a los beneficios totales reportados a los accionistas de la compañía. Es decir, los resultados apoyan la evidencia de que los fondos mejor administrados no son vendidos de manera más eficiente en la etapa comercialización de las empresas, lo que redujo considerablemente la eficiencia global de las mismas. Este importante hallazgo de esta primera parte es bastante robusto considerando los efectos de escala en las empresas y a lo largo de todos los años considerados en nuestro horizonte de análisis. Sin embargo, la influencia de los retornos a escala sugirió que el tamaño parece ser una variable importante para el éxito la venta de fondos. Este problema es especialmente relevante en un mercado concentrado, con participación amplia de la banca, como es la industria de fondos española.

Segunda Parte de la Tesis

Resumen

La aplicación de un enfoque de frontera eficiente no-orientado utilizando el SBM dentro de la metodología DEA, propuesto por Tone (2001), se discutió en detalle en la Parte I para justificar esta técnica como una herramienta inicial y apropiada para obtener los índices de eficiencia de las EGFMs. La consistencia de los resultados de eficiencia obtenidos mediante el uso de diferentes medidas de las variables más polémicas incluidas en nuestro modelo demostró la solidez de nuestros resultados. Sin embargo, el DEA y sus diversos modelos no pueden identificar a los competidores «más similares» que pueden servir como punto de referencia para las empresas analizadas a la luz de las diferencias notables que potencialmente se encuentren entre las empresas evaluadas. Este aspecto podría limitar la exactitud de algunos resultados en la literatura, además es especialmente importante en aquellos sectores donde los competidores muestran características variadas, tales como la industria de fondos española. Por lo cual, el conjunto empresas que conforman la frontera eficiente del DEA puede estar formado por grandes empresas de fondos de propiedad de

bancos y ello puede no ser una referencia adecuada para los pequeños gestores independientes, repercutiendo entonces en engañosos rankings de eficiencia.

En segunda parte de la tesis, logramos superar esta limitación utilizando las variaciones recientes e inexploradas del SBM original, propuestas por Tone (2010). Estas variaciones permitieron la comparación de las empresas de fondos con los competidores de su verdadera referencia «más similares» de acuerdo con los recursos y los objetivos de gestión, logrando con ello trabajar con las diversas características de la industria de los fondos de inversión españoles. Como un primer paso para cuantificar el sesgo debido a la utilización de los conjuntos de referencia inadecuados de los competidores, se analizó el efecto de las mencionadas variaciones en los scores de eficiencia obtenidos por el SBM original. La aplicación de estas nuevas técnicas del SBM mejoró la precisión de los resultados y complementó la simple consideración de los retornos a escala variable (VRS) para evaluar la eficiencia de las DMUs con diferentes características de escala. Estas variaciones propuestas por Tone (2010) basan en los hiperplanos en lugar de los vértices de la frontera, lo le permite el método descubrir los competidores (facets) más adecuados con respecto cada DMU analizada.

Por último, la aplicación del proceso de agrupamiento (clustering) propuesto por Tone (2010), nos permitió una evaluación refinada de empresas eficientes, centrando el análisis únicamente en los competidores de similares características. Nuestra propuesta de agrupamiento se basó en los activos administrados y esfuerzo en personal de las EGFMs, ya que, como se encontró en la Parte I (Cuadro I-3), un gran número de empresas pequeñas de fondos gestionan una cuota de mercado residual y por otro lado un número reducido de grandes empresas de fondos dominan la industria. Bajo estos criterios de agrupamiento, asumimos la hipótesis de que las empresas de fondos con tamaño homogéneos deberían tener las mismas oportunidades de alcanzar la eficiencia en todas las fases de gestión, resolviendo así los posibles efectos de escala en la etapa de distribución y comercialización que se hayan contemplado en la Parte I. Adicionalmente el uso de esta variación basada en agrupamientos de Tone (2010) permitió la identificación de las empresas localmente eficientes en relación con los competidores con características similares de agrupamiento.

Adicionalmente se planteó la necesidad de comprobar en esta segunda parte del estudio si los patrones de eficiencia obtenidos por las técnicas anteriores son persistentes en el tiempo, es decir, si los resultados de eficiencia obtenidos por las diferentes etapas corresponden a patrones estables del proceso de gestión. De lo contrario, si los resultados de eficiencia estuvieron sujetos a una variabilidad considerable a lo largo del tiempo, que pondría en duda las conclusiones obtenidas para cada año y en determinadas EGFMs; es decir, no sería posible diferenciar las estrategias de gestión eficaces de otros factores temporales. El resultado buscado fue la separación de aquellas empresas que siguen claramente los patrones de gestión eficientes de otras empresas con resultados de gestión mucho más erráticos en relación con sus competidores. Este fenómeno de persistencia se ha debatido ampliamente en la literatura sobre los fondos de inversión, pero, hasta donde sabemos, nunca aplicado a las EGFMs. Para probar esta hipótesis, se utilizó un enfoque no paramétrico basado en grupos de eficiencia en lugar de corrientes cuartiles o quintiles de clasificación empleados en la literatura, ello nos permitió identificar grupos homogéneos de eficiencia para proporcionar una mayor fiabilidad a nuestras conclusiones.

Por último, en esta segunda parte de la tesis se identificaron los principales factores que parecen para conducir a los resultados de eficiencia de las EGFMs. Este estudio toma especial importancia en mercados financieros con los competidores y características variadas, tales como la industria de fondos española.

Conclusiones

Con base en el enfoque de sub-DMUs de Holod y Lewis (2011) y teniendo en cuenta la interacción entre los diferentes subsistemas de gestión dentro de una EGFMs propuestas por Berkowitz y Qiu (2003), la primera parte de la tesis se evaluó la eficiencia de las EGFMs españolas mediante la aplicación de la medida basada en el enfoque de holgura (SBM), propuesto por Tone (2001). En esta segunda parte, se llegó más allá mediante el desarrollo de un estudio que supera la limitación de tomar un conjunto de referencia como identificación adecuada para el universo de empresas evaluadas. Vimos como este

problema es especialmente importante en la industria de fondos española debido tanto a la alta concentración como a las características variadas las EGFMs. Se utilizaron para ello las variaciones inexploradas y recientemente propuestas por Tone (2010) en el modelo SBM, ello nos permitió comparar las EGFMs con sus verdaderos competidores «más similares» que contaban con recursos de gestión equivalentes.

Se analizaron los efectos de estas variaciones (sobre la base de los hiperplanos en lugar de los vértices de la frontera eficiente) basados en los scores de eficiencia previamente obtenidos por el SBM originales. En primer se encontró que la Variación I mejora los scores medios de eficiencia (scores) para cada etapa (subsistemas-múltiples planteados) como consecuencia del nuevo enfoque, aunque el rango de correlación de Spearman para ambos SBM y Variación I fue cercano a 1 para cada etapa y año analizado. Por otro lado, las comparativas para la etapa de comercialización y etapa de eficiencia global proporcionaron evidencia del alto grado de similitud entre los scores del SBM original y la Variación I; es decir, se encontró evidencia del efecto residual de Variación I en los scores de eficiencia de SBM originales obtenidos previamente. El hallazgo mencionado plantea preguntas acerca de la relevancia de esta variante en los resultados empíricos de nuestro estudio.

De manera similar a los hallazgos de SBM original, los scores de eficiencia de la Variación I también muestran un patrón de eficiencia decreciente en el sector de los fondos españoles. En la misma línea, esta variación destaca el hecho de que las empresas parecen ser peores vendiendo de fondos mutuos que gestionando sus carteras, lo que reduce considerablemente las ganancias generales de accionistas de las compañías.

Continuando con las variaciones del SBM propuestas por Tone (2010), se aplicó SBM Variación II, la cual minimiza el score del SBM obtenido a partir de todas las “facets” que encontradas. SBM-Variación II se calculó utilizando los “maximal friends facets” obtenidas aleatoriamente (pasando a ser Variación IV) y por otra parte por la agrupación de la “facets” (pasando a ser Variación III) como se plantea en el modelo propuesto por Tone (2010). Por una parte la Variación IV mejora de nuevo los índices de eficiencia promedio para todas las

etapas de gestión, mostrando de nuevo un alto y significativo rango de correlación de Spearman entre los rankings de eficiencia (SBM originales y Variación IV). Sin embargo, la correlación entre la variación de los rankings de la Variación I y IV, es algo menor que la existente entre SBM original y Variación I.

Por otro lado la Variación IV también mostró una relación positiva y significativa entre la capacidad de comercialización y distribución las empresas y la eficacia global para todo el horizonte de tiempo. Por lo tanto, esta nueva medida propuesta por Tone (2010) confirma los principales resultados obtenidos en la primera parte con el SBM original. Este resultado verifica la robustez de las conclusiones del primer estudio, como son la aparente dicotomía entre la eficiencia de la gestión de las carteras y de la capacidad de comercialización de la industria de fondos de inversión española.

Adicionalmente se realizó una búsqueda refinada de las empresas localmente eficientes dentro de los grupos encontrados (Variación III), ello realizado con base en la técnica jerárquica divisiva y bajo diferentes dimensiones de medición (activos gestionados y número de empleados de cada EGFMs). Este enfoque nos permitió encontrar algunas evidencias vinculadas a que la adecuada evaluación en la etapa de gestión de carteras, debe considerar conjuntos homogéneos de empresas con recursos de gestión similares. Por lo tanto, cuando las empresas de referencia incluidas en la frontera eficiente no pertenecen al mismo grupo que la empresa objetivo, los scores del SBM original podrían estar sesgados.

El análisis de la persistencia en el tiempo de los scores de eficacia obtenidos, por medio de un enfoque innovador y completo debido a la inclusión de grupos de eficiencia consistentes en lugar de la sencilla consideración de la mediana o cuartil con igual número de empresas, nos permitió encontrar evidencia de persistencia. En primer lugar, con el uso de tablas de contingencia 2x2, se halló un fenómeno importante de la persistencia en los grupos de eficiencia, tanto en la etapa de gestión de cartera como en la etapa eficiencia global, aunque se observó menos importancia entre 2008 y 2009. La prueba de Cochran también confirmó esta evidencia para todo el período 2005-2009. Estos hallazgos apoyan la hipótesis subyacente de que las empresas que gestionan sus fondos de inversión mejor que

sus competidores suelen ser los mismos durante todo el horizonte de tiempo analizado. Estas conclusiones también podrían extenderse a aquellas empresas con scores de eficiencia peores que sus competidores.

En cuanto a la persistencia en la etapa de comercialización y distribución, los resultados mostraron la persistencia significativa para los dos primeros períodos comparativos (2005-2006, 2006-2007). Sin embargo, se observó falta de persistencia en 2007-2008 y 2008-2009, lo que pudo haber sido impulsado por las consecuencias de la crisis financiera.

Por otra parte, las tablas de contingencia 4x4 proporcionaron pruebas de una persistencia significativa en los scores de eficiencia obtenidos por las empresas en las diferentes etapas de la gestión. Encontramos un significativo valor de la prueba chi-cuadrado para todos los períodos comparativos y para cada EGFMs. Además, encontramos persistencia significativa tanto de las mejores y peores empresas en dos años consecutivos. Este resultado es robusto para la etapa de gestión de cartera y para la etapa de distribución y comercialización, así como para la eficiencia global de las empresas en su conjunto. Un análisis más detallado de los residuos de estas tablas de contingencia 4x4 indica que los índices de eficiencia tienen poca probabilidad de cambio a un clúster de eficiencia muy diferente.

Adicionalmente se ha encontrado que el tamaño (recursos gestionados) parece jugar un papel positivo y significativo en la persistencia de la etapa de gestión de cartera en las mejores EGFMs. Este resultado fue consistente en cuanto al número de empleados, los activos de cada empresa, el patrimonio y la cantidad de fondos ofrecidos. Por otra parte, encontramos que las empresas con menos recursos de gestión son consistentemente las peores EGFMs, especialmente en términos de número de fondos y activos administrados.

Un patrón opuesto fue encontrado en la etapa de comercialización y servicio, donde el tamaño, como “proxy” de los recursos, parece jugar un papel negativo y significativo en la persistencia de esta etapa. Por lo tanto, cuanto más grande es una empresa, más problemas tiene vendiendo y distribuyendo eficientemente sus fondos de inversión. Este resultado es robusto en términos de número de empleados, de los activos, del patrimonio y de la

cantidad de fondos. Es de destacar que el mismo patrón se encuentra también en la etapa de eficiencia global, aunque los resultados no son tan significativamente claros para cada celda y variable. Lo anterior podría adicionalmente sugerir que la comercialización de las EGFMs también está afectando considerablemente la persistencia de la eficiencia global de las empresas.

Por último, observando con más detalle la diferencia en los recursos de gestión de los diferentes grupos de persistencia extrema (Top de ganadores y los más bajos perdedores), encontramos resultados robustos con los proporcionados por las comparaciones anteriores, lo que sugiere que las empresas con más recursos son mejores en la gestión de su cartera que en la venta de las mismas, y por lo tanto generan un impacto importante en los niveles de eficiencia general. Las diferencias en estas categorías de persistencia extrema son de significativa importancia la etapa de comercialización y servicio.

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