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Title: Consumer preferences for nutritional claims: An exploration of attention and choice based on an eye-tracking choice experiment

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Abstract: Nutritional claim (NC) requirements on food packages are among the most important and influential EU policy measures related to diet and have the capacity to promote healthy eating. This study combines a discrete choice experiment (DCE) method with eye-tracking (ET) technology to assess consumer preferences for multiple NCs in yogurt selection and explores the relationships between the NC preferences and the visual attention paid to these claims and the visual attention and choice decisions. The results indicate that the low-sugar NC was the least-preferred claim in all the models. Overall, the presence of NCs generally increases visual attention in terms of fixation count, which may be linked to an increased likelihood of affecting the final decision to purchase yogurts with NCs.

1 **Consumer preferences for nutritional claims: An exploration of attention and choice based**
2 **on an eye-tracking choice experiment**

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12
13 **Abstract**

14
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16 influential EU policy measures related to diet and have the capacity to promote healthy eating.
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Abbreviations: NC, nutritional claim; DCE, discrete choice experiment; ET, eye tracking; EU, European Union; NCD, non-communicable disease; FOP, front of pack; AOI, area of interest.

32 1. Introduction

33

34 Poor dietary patterns, high-energy intake, and malnutrition are some of the major triggers
35 of non-communicable diseases (NCDs), such as obesity, diabetes, cardiovascular disease, and
36 some types of cancer. According to the World Health Organization (WHO, 2018), NCDs cause
37 70 percent of deaths every year worldwide. Of the six WHO regions, Europe is the most affected
38 by NCDs, and they are increasing. The impact of NCDs in Europe has accounted for an
39 estimated 86 percent of the deaths and 77 percent of the disease burden in the last decade
40 (WHO/Europe, 2018). Given the current situation, policy makers, such as the European Union
41 (EU) and the United States Department of Agriculture (USDA), have called for transitions
42 toward healthier diets and more informed food choices (Burlingame & Dernini, 2010; Dötsch-
43 Klerk, Mela, & Kearney, 2015; UNEP, 2010). Healthiness, though, typically needs to be
44 encouraged in consumers through trustworthy information that is based on scientific evidence.

45 In this regard, the EU has introduced European Council (EC) Regulation No. 1924/2006
46 (Smith, 2015), which requires NCs¹ in food products to be based only on scientific evidence. The
47 positive impact of this regulation is that it identifies lawful claims and thereby makes it possible
48 for authorities to take action if other NCs are used in the marketplace. Partly due to this EU
49 labeling requirement, on average 85 percent of all packaged food products in Europe have NCs
50 (Prieto-Castillo, Royo-Bordonada, & Moya-Geromini, 2015). In Spain, the availability of NCs
51 reached 95 percent, making Spain one of the top countries in terms of nutritional labeling
52 (Prieto-Castillo et al., 2015). In particular, a recent study that explored the presence of nutritional
53 and health claims in five EU countries (the UK, Slovenia, the Netherlands, Germany, and Spain)
54 ranked Spain second, after the UK, regarding the presence of NCs (Hieke et al., 2016). Studies of
55 consumers' understanding and use of nutritional information have shown considerable interest in
56 NCs, but, in the case of Spain, of the 52 percent who reported a full understanding, only 21
57 percent reported using them (Prieto-Castillo et al., 2015). Hence, there is a need to investigate
58 and identify the attributes that motivate the use of NCs and their influence on the decision to
59 purchase.

¹ This regulation defines an NC as “any statement that suggests or implies that a food has specific beneficial nutritional properties.” This definition distinguishes two types of NCs. The first group refers to the content of nutrients or substances (e.g., a source of vitamin B₆), while the second group compares the product with its conventional version in terms of the content (high or low) of a nutrient or substance (e.g., high in calcium).

60 Previous literature has indicated that NCs help consumers to compare the healthfulness of
61 food products (Grunert, Wills, & Fernández-Celemín, 2010) and that generally they are willing
62 to pay premium prices for food products bearing NCs (Ballco & de-Magistris, 2018; Barreiro-
63 Hurlé, Gracia, & de-Magistris, 2010; de-Magistris, López-Galán, & Caputo, 2016; Jurado &
64 Gracia, 2017; Van Wezemaal, Caputo, Nayga, Chryssochoidis, & Verbeke, 2014). However,
65 despite these findings, there is increasing evidence that what consumers say about their
66 preferences regarding NCs is not actually reflected in what they purchase in the marketplace. To
67 illustrate, in the last few decades, the consumer demand for healthier functional food (FF)
68 products offering NCs has grown rapidly (Santeramo et al., 2018). Attracted by such market
69 growth, companies have invested in and developed new FF products (Khan, Grigor, Win, &
70 Boland, 2014). Nevertheless, 70 to 90 percent of these new FF products exited the market within
71 the first two years from their launch (Bimbo et al., 2017). This high failure rate suggests that a
72 deeper understanding of the main motives underlying consumer preferences and the
73 heterogeneity in the demand for NCs is needed. For this reason, understanding how consumers
74 make trade-offs among multiple front-of-pack (FOP) NCs is an important issue for marketing
75 and public policy purposes.

76 Recent studies have focused on exploring new approaches to investigating consumer food
77 choice behavior based on consumers' visual attention.² These approaches use eye-tracking (ET)
78 technology to analyze consumers' purchase decisions by tracking the visual attention paid to
79 areas of interest (AOIs). ET technology is considered to be one of the most powerful means to
80 determine individual choices (Balcombe, Fraser, & McSorley, 2015), especially when combined
81 with discrete choice experiments (DCEs) (Scarpa, Zanolli, Bruschi, & Naspetti, 2013).

82 This study investigates consumers' preferences for alternative NCs (fat free, low sugar,
83 high fiber, source of vitamin B₆, and source of calcium) and explores the impact of consumers'
84 visual attention on their final choice. To elicit consumers' preferences for alternative NCs, we
85 conducted a DCE, because its ability to evaluate multiple attributes simultaneously is consistent
86 with random utility theory (RUM) and very similar to the purchase decision process (Lusk,

² By definition, "attention" is the "degree to which consumers focus on a stimulus within their range of exposure" (Solomon, Bamossy, Askegaard, & Hogg, 2006).

87 2003). Visual attention was measured in terms of fixation time (milliseconds) and fixation count³
88 using ET. The fixation time was used due to its frequency of use in the extended literature
89 analyzing visual attention to food products (Antúnez et al., 2013; Ares, Mawad, Giménez, &
90 Maiche, 2014; Ares et al., 2013; Bialkova & Trijp, 2011; Bialkova et al., 2014; Fenko, Nicolaas,
91 & Galetzka, 2018; Gere et al., 2016; Grebitus & Davis, 2017; Hummel, Zerweck, Ehret, Winter,
92 & Stroebele-Benschop, 2017; Samant & HanSeok, 2016; Spinks & Mortimer, 2016; Torrico et
93 al., 2018; Uggeldahl, Jacobsen, Lundhede, & Olsen, 2016; Van Loo et al., 2015; Vu, Tu, &
94 Duerrschmid, 2016). However, the recent research by Orquin and Holmqvist (2018) suggested
95 that the total fixation duration is not recommended because it often involves inappropriate
96 aggregation data. Therefore, in our research, we also included the fixation count to compare
97 results across ET measures. This study focuses on NCs because they are a simpler way to present
98 information than nutritional tables. NCs do not list the amount of a nutrient but rather summarize
99 the information concerning a specific nutrient and communicate it to consumers in simple, easy-
100 to-process language (e.g., fat free). We chose to study yogurt claims because yogurt is
101 recommended as part of a healthy diet in many countries (Eržen, Kač, & Pravst, 2014). Most
102 notably, in a market study that we conducted on food products with NCs in Spain, yogurt was
103 found to be a product that commonly contained NCs.

104 This study contributes to the existing literature on consumer food choice behavior in
105 several ways. First, while most previous literature has focused on consumer preferences for
106 fewer than three NCs, this study analyzes consumer preferences and choice behavior for multiple
107 NCs. Second, this is the first study to combine ET and a DCE to investigate whether consumers
108 pay attention to alternative NCs when making food choice decisions and how their attention
109 affects their final food choices. Most researchers utilizing DCE and ET methods have explored
110 consumer preferences for different formats of nutritional labels (e.g., choice logos, monochrome
111 guidelines, daily amount nutritional labels, color coded nutritional labels, the traffic light system,
112 and information tables showing nutritional facts) displayed on the FOP (Bialkova & Trijp, 2011;
113 Bialkova et al., 2014; Graham & Jeffery, 2011; Mawad, Trías, Giménez, Maiche, & Ares, 2015)
114 and the effect of sustainability-related labels on consumers' purchase behavior (Samant &
115 HanSeok, 2016; Van Loo et al., 2015). Hence, this research contributes to the food choice

³ The fixation time is respondents' fixation duration within an AOI, and the fixation count measures participants' fixation frequency within an AOI (Duchowski, 2017).

116 literature by exploring the importance of visual attention to a selection of NCs. Finally, this study
117 offers new insights into the combination of DCEs and ET, a novel methodological approach that
118 has not yet been applied to food products in a European country such as Spain.

119 The findings from this research can be informative for producers, processors, and
120 retailers. In addition, the results can provide new insights for policy makers, assisting them in
121 designing strategies to promote healthy food choices.

122 **2. Consumer attention and food choices: Background**

123 During a purchase decision, consumers are exposed to multiple food attributes, such as
124 symbols, health-related label messages, health claims, nutritional claims, and others (Carrillo,
125 Fiszman, Lähteenmäki, & Varela, 2014; Miraballes, Fiszman, Gámbaro, & Varela, 2014). As
126 documented by Milosavljevic and Cerf (2008), consumers typically make choice decisions
127 within a few seconds; thus, they may not attend to all the information available on the food
128 package. Generally, some information is selected to be processed further while the rest is lost,
129 and, in most cases, consumers are not even aware of its presence on the label (Oliveira et al.,
130 2016). For this reason, studying consumers' attention to food labels is becoming a key aspect of
131 the design of food labels that successfully attract attention.

132 In this regard, a rapidly growing body of literature has examined the relationship
133 between visual attention and stated preference in the food sector. Table 1 contains a review of
134 previous studies using ET and discrete choice experiments and their key findings. We focus on
135 these particular studies because they combine DCEs with ET and center on consumer valuation
136 for food-labeling programs.⁴ The results of these studies are mixed regarding the extent to which
137 the degree of visual attention paid to specific attributes correlates with the actual choices.

⁴ Although we limited our literature review to food choice studies, we acknowledge that eye-tracking technology is widely used in other fields, such as psychology (Orquin & Lagerkvist, 2015; Orquin & Mueller Loose, 2013; Peschel & Orquin, 2013), marketing (Meißner, Musalem, & Huber, 2016; Pieters, 2008; Pieters & Warlop, 1999), and health economics (Ryan, Krucien, & Hermens, 2017), among others. Recently, ET has also increasingly been used to explore methodological issues related to survey design, organizational research (Meißner & Oll, 2017; Meißner, Pfeiffer, Pfeiffer, & Oppewal, 2017), visual biases, and threats (Orquin, Ashby, & Clarke, 2016; Orquin, Bagger, & Mueller Loose, 2013; Orquin, Chrobot, & Grunert, 2018; Orquin, Perkovic, & Grunert, 2018).

138 Table 1 – Summary of studies that have combined ET with stated preferences and DCEs

No.	Authors	Country	Products	Methodology	Key findings
1	Balcombe et al. (2015)	UK	A basket of goods containing a mix of foods	DCE and ET	No compelling evidence that higher- or lower-value attributes receive more or less attention.
2	Balcombe, Fraser, Williams, and McSorley (2017)	UK	A basket of goods containing a mix of foods	DCE and ET	Although respondents with higher levels of visual attendance valued specific attributes more, the results reveal weak relationships between ET and stated preference data.
3	Bialkova et al. (2014)	Netherlands	Yogurt	A combination of an experimental choice task with ET	Results suggest that attention mediates the effect of nutrition labels on choice. The longer the fixation, the higher the likelihood of being chosen.
4	Bialkova and van Trijp (2011)	Netherlands	Yogurt	Integration of the visual search paradigm (ET) with a CE	ET was found to be a promising tool for consumer research on attention to nutrition labeling information and its effect on informed healthy choices.
5	Graham and Jeffery (2011)	USA	Pizza, soup, yogurt, snacks, fruits, and vegetables	Self-reported online grocery shopping CE and ET	Participants spent longer looking at labels for foods they decided to purchase compared with foods they decided not to purchase.
6	Samant and HanSeok (2016)	USA	Chicken products	Stated preference and ET	Findings suggest that enhanced label knowledge increases consumers' visual attention to labels with a possibility of positive purchase behavior.

No.	Authors	Country	Products	Methodology	Key findings
7	Uggeldahl et al. (2016)	Denmark	Ground beef minced meat	DCE and ET	Eye movements are related to stated choice certainty.
8	Van Herpen and van Trijp (2011)	Turkey and Netherlands	Breakfast cereals	Self-reported use, recognition, ET, and CE	Although a nutrition table was evaluated most positively, it received little attention and did not stimulate healthy choices. Other types of labels enhanced healthy product choices.
9	Van der Laan, Hooge, Ridder, Viergever, and Smeets (2015)	Netherlands	Different food images	Choice screens and ET	Results show that for both the most-wanted and the least-wanted decision types, the total fixation duration was longest for the product of choice.
10	Van Loo et al. (2015)	USA	Coffee	DCE and ET	Results suggest that consumers who spend more time attending to and fixate more on sustainability attributes value them more.
11	Vu et al. (2016)	Austria	Different images food	Stated preference under time pressure, test design complexity, and ET	Highlights the importance of understanding the factors influencing gazing behavior in an ET test for better future application.

141 For example, Balcombe et al. (2015) examined visual attention in a multi-attribute DCE
142 using ET and found little evidence that visual attention in terms of fixation duration on the
143 attributes indicates the level of importance. In other words, looking longer or more often at an
144 attribute does not necessarily mean that it is of higher value to the consumer. A more recent
145 study by Balcombe et al. (2017) again examined the combination of visual attention and stated
146 preferences and found weak relationships between them. These results differ significantly from
147 those reported by Uggeldahl et al. (2016), who, through a DCE combined with ET on the
148 selection of ground beef minced meat, found that visual attention paid to the alternatives in a
149 choice task does reflect participants' stated choices. Similarly, Bialkova and Trijp (2011)
150 indicated that the combination of ET with a DCE is a promising tool for consumer research on
151 attention to nutrition labeling information and its effect on informed healthy food choices. Other
152 explanatory studies that have combined visual attention with actual choices have found a positive
153 association. More specifically, in the US, Graham and Jeffery (2011) examined visual attention
154 to nutritional labels (e.g., a nutritional fact table) for sixty-four different food products in an
155 online shopping scenario. Consumers were found to spend more time looking at the nutrients in
156 food products that they ultimately chose to purchase. Another study using an online shopping
157 purchase scenario, by Van der Laan et al. (2015), tested the effect of healthy food choices and
158 changes in visual attention on purchases. This study showed that health goals increase the
159 attention to goal-congruent items and increase the likelihood of the consumer choosing them.

160 Van Herpen and van Trijp (2011) examined consumer attention and the use of three
161 different types of nutrition labeling (a logo, a traffic-light label, and a nutritional table) in Turkey
162 and the Netherlands to investigate whether the type of label influences consumers to make
163 healthier food choices. The results in both countries suggested that, although consumers
164 evaluated the nutritional table positively, it received little visual attention and did not stimulate
165 healthy choices. However, the traffic light and especially the logo labels enhanced healthy
166 product choices. Bialkova et al. (2014) used yogurt selection in a DCE to explore whether and
167 how attention to nutritional information (a health logo, a monochrome Guideline Daily Amount
168 (GDA) label, or a color-coded GDA label) affects consumer choice. The results suggested that
169 products with long fixation times have the highest likelihood of being chosen.

170

171 Regarding sustainability-related label claims, Samant and HanSeok (2016) determined
172 the effect of label education on consumers' purchase behavior by combining visual attention and
173 sustainability label claims on chicken products. The findings provided empirical evidence that
174 enhanced label knowledge increases consumers' visual attention to labels, with the possibility of
175 positive purchase behavior. Lastly, Van Loo et al. (2015) analyzed the importance of
176 sustainability labels on coffee (e.g., Fairtrade, Rainforest Alliance, USDA Organic, and carbon
177 footprint) by combining the visual attention paid to these labels with a DCE. Their results
178 indicated that greater importance associated with sustainability labels results in increased visual
179 attention and willingness to pay (WTP) for coffee with these labels.

180 Based on the findings of earlier studies, we hypothesize the following:
181 (H1). Providing NCs on yogurt packages may provide a signal detection assumption that an
182 increase in participants' visual attention may result in an increased probability of the product
183 being purchased.

184 Because consumers have raised concerns about their health and are shifting toward food
185 products that are low in calories (Carrillo, Varela, & Fiszman, 2012; de-Magistris & Gracia,
186 2016; Jurado & Gracia, 2017), we also hypothesize that:

187 (H2). Low-calorie⁵ yogurts (e.g., fat free and low sugar) will generate greater utility in
188 participants than other nutritional claims.

189 **3. Materials and methods**

190 191 *3.1 Choice experiment: Product and attribute selection*

192
193 The product for the experiment was selected based on market research on food products
194 bearing NCs sold in local supermarkets between July and September 2015. The foods were
195 included in the database according to their importance in the shopping basket of Spanish
196 families.⁶ An examination of the products showed that yogurt carried the most NCs. In total, 251
197 yogurts that contained 1 NC on the FOP that corresponded to the official EU definitions

⁵ According to the previous literature, low-calorie yogurts are mostly low fat, fat free (i.e., skimmed or semi-skimmed), and low in sugar (Peres, Esmerino, da Silva, Racowski, & Bolini, 2018; Pinheiro, Oliveira, Penna, & Tamime, 2005).

⁶ According to the Ministry of Agriculture and Fisheries, Food and Environment's (MAPAMA, 2014) consumer survey in Spain, 89 percent of the per capita consumption of packaged food was liquid milk, processed meat, yogurt, cheese, industrial bread, and biscuits.

198 (Regulation (EC) No. 1924/2006) were considered for further analysis as well as a full-fat
 199 unlabeled yogurt. We used the 500 g package (4 containers, each with 125 g), because it is the
 200 size with the greatest presence in the market. All the products used were natural yogurts (no
 201 added flavor), with no fruits, except the one with fiber, which contained several types of cereal
 202 (oats, barley, wheat, and wheat bran). We included the high-in-fiber yogurt because of the high
 203 demand and the large variety of cereal-fiber-source yogurt in the local market (Cuevas, 2012;
 204 Fontecha, Recio, & Pilosof, 2009; Sah, Vasiljevic, McKechnie, & Donkor, 2016). The NCs
 205 included in the study are shown in Table 2.

206 Table 2 – Nutritional claims used in the study

N°	Natural yogurts with NCs	Frequency of NC
1°	Fat free	42.78%
2°	Source of calcium	21.25%
3°	Full-fat unlabeled (reference) ^a	12.26%
4°	Low sugar	11.99%
5°	Source of vitamin B ₆	10.63%
6°	High fiber	1.09%

207 Note: ^a The unlabeled product is a full-fat natural yogurt with no added flavor and no NC on the FOP.

208 Following Bialkova and vanTrijp (2011), Bialkova et al. (2014), and Carlsson, Kataria,
 209 and Lampi (2010), we excluded the price attribute by asking consumers to assume that the price
 210 was the same as the yogurt that they regularly consume, since yogurt is regularly consumed in
 211 Spanish households (Ministry of Agriculture and Fisheries, Food and Environment (MAPAMA),
 212 2014) and individuals are aware of the price variations (which are not large except for the
 213 reference full-fat, no-NC yogurt) among different types of yogurt. Following the experimental
 214 design of Bialkova and van Trijp (2011) and Bialkova et al. (2014), a full factorial design (i.e.,
 215 nutritional claims in our case) resulted in a combination of 15 choice questions (or choice tasks),
 216 each with 2 alternatives. To each choice task, we also added a non-buy option. The product
 217 location (either left or right in the two-alternative choice set) of the two products was systematically
 218 varied. A computer program (Tobii X2-30 ET) randomized the sequence of appearance of the 15
 219 choice tasks. The participants had 15 seconds⁷ to observe the 2 products in each task and then

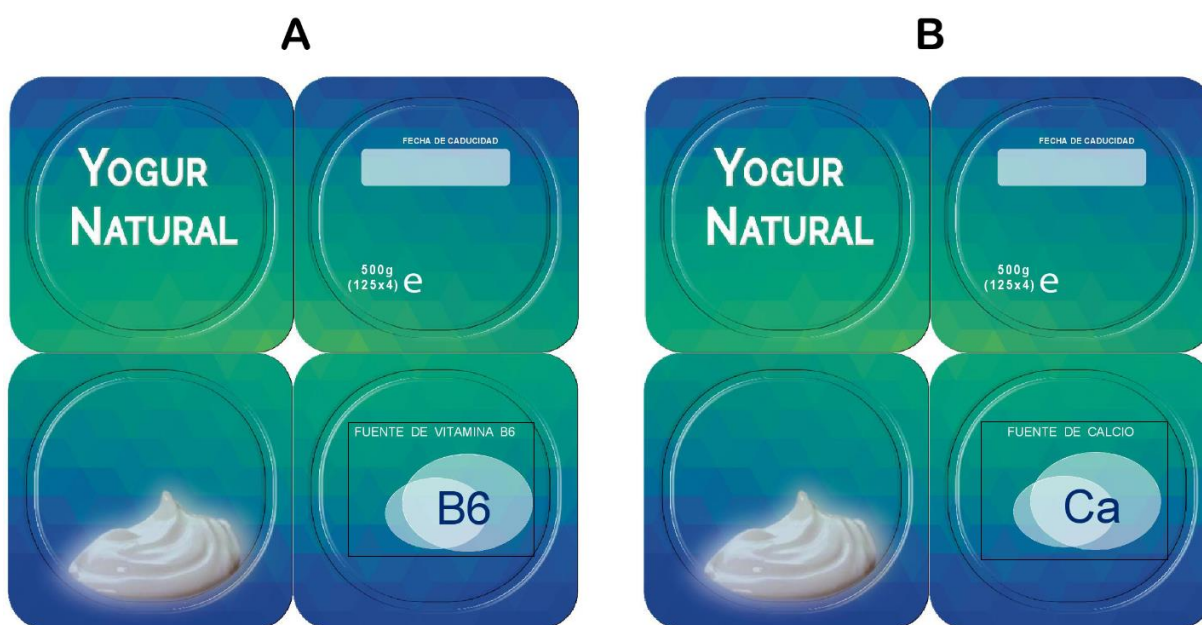
⁷ We used a fixed exposure time to measure the fatigue effect from the 15 choice tasks and to examine the fixation process through the 15-second exposure time. However, due to the main focus of this paper, the results from this analysis are not included here. As for the set-up time, we considered studies in which the times varied from short periods of 2.5 seconds (Piqueras-Fiszman, Velasco, Salgado-Montejo, & Spence, 2013) to 10 seconds (Orquin & Scholderer, 2011) and up to 30 seconds (Strasser, Tang, Romer, Jepson, & Cappella, 2012). In addition, from a pretest of 20 participants, we observed that participants needed an average exposure time of 13 seconds to choose

220 were asked to choose their preferred yogurt. Oral answers were recorded through an evaluation
221 form that appeared on the screen after 15 seconds. Then, the moderator, using a parallel screen,
222 selected the preferred alternative defined by the participant (A, B, or no buy). See the evaluation
223 form in Appendix A (Figure A1).

224 3.2 Eye-tracking procedure and measures

225 To capture the visual attention during the DCE, we replicated the work of Van Loo et al.
226 (2015) using a totally different product, yogurt, and measured preferences without considering
227 the price attribute. For the analysis of the eye movement data, we defined a set of AOIs to
228 capture the eye fixations, in terms of fixation time and fixation count, on the NCs (see Figure 1).

229 Figure 1 – An example of the areas of interest



NINGUNO

230 Note: Option A refers to the Spanish version of a yogurt with a *source of vitamin B₆*, and option B refers to the
231 yogurt with a *source of calcium*. AOIs were not marked in black in the original evaluation choice task. “Ninguno” is
232 the “non-buy” option.
233

234 The FOPs were consistent in terms of AOI size (width and height). For each of these
235 AOIs, we calculated the mean of the fixation time spent and the fixation count. The combination
236

between alternatives. Therefore, based on the previous research and the results from the pretest, we decided to use an exposure time of 15 seconds.

237 of images was presented in full color on a 24” computer screen with 1920×1080 pixel resolution.
238 Eye positions were sampled at 50 Hz with a remote ET device (Tobii X2-30 ET) positioned
239 under the computer screen on which the stimuli were displayed. Before recording the eye
240 movements, we ran a 9-point calibration procedure and familiarized the participants with the
241 process using an example of a 2-alternative choice task in which they were asked to choose “out
242 loud”⁸ A, B, or no buy. Then, we ran another calibration procedure before recording their eye
243 movement for the experiment. The distance between the ET device and the participants’ eyes
244 was 58–60 cm.

245
246 *3.3 The experiment*
247

248 The experiment consisted of three stages: (i) recruiting and sampling, (ii) ET in
249 combination with the DCE, and (iii) a follow-up questionnaire aimed at capturing yogurt
250 purchase behavior, consumption habits, attribute importance, general attitudes toward yogurts
251 with NCs and HCs, general health interest, and socio-demographic consumer characteristics. The
252 experiment was carried out in different periods of time (morning and afternoon) and on different
253 days (from Monday to Saturday). The sessions consisted of 1 participant at a time. Upon their
254 arrival at the lab, the respondents received information about the main purpose of the experiment
255 (stage 1). A 9-point calibration procedure was used to calibrate participants’ eye vision with the
256 eye-tracking device before the example warm-up task and after starting the data collection. The
257 respondents faced 15 choice tasks (stage 2). For each task, they were asked to choose their most-
258 preferred option (A, B, or neither). They were reminded each time to imagine that they were in a
259 supermarket to buy yogurt and that the price reference was the price of the yogurt that they
260 habitually purchase. Finally, the participants completed a follow-up questionnaire capturing their
261 yogurt purchase behavior, consumption habits, attribute importance, general attitudes toward
262 yogurts with NCs and HCs, general health interest, and socio-demographic consumer
263 characteristics (stage 3).

264
265

⁸ The choice of the product was indicated orally based on the applied methodology from two previous studies (Bialkova & van Trijp, 2011; Bialkova et al., 2014). In addition, since we followed a stratified sample approach, we used the oral choice to avoid any possible choice mistake due to a lack of computer skills (almost 10 percent of the sample was older than 70 years).

266 3.3.1 Recruitment and sample characteristics

267

268 The experiment was conducted from September to November 2016 in a medium-sized
 269 town in Spain that is widely used by food marketers and consulting companies because the
 270 socio-demographic characteristics are representative of the Spanish Census of Population (see
 271 Appendix B (Table B1)). The participants were recruited via email by a recruiting agency and
 272 were selected by random stratification with proportional allocation for age, gender, and
 273 education to avoid under/overrepresentation of consumer profiles. To discover distinctive groups
 274 with similar preferences, we performed a cluster analysis (Section 4.1). Table 3 shows the
 275 characteristics of the final sample of respondents and the segments from the cluster analysis.

276 Table 3 – Descriptive analysis of the sample and socio-demographic characteristics (percentages)

	Reference population, Spain ^a	Sample	Segment1	Segment 2
Sample size	-	n = 100	n = 39	N = 61
<i>Gender</i>				
Female	51.00	52.00	46.15	55.74
Male	49.00	48.00	53.85	44.26
<i>Age groups</i>				
18–34**	22.24	18.00	15.38	26.23
35–44**	19.55	23.00	10.26	21.13
45–54	18.28	19.00	17.95	16.39
More than 54	39.93	40.00	56.41	36.07
<i>Educational level^b</i>				
Primary	24.88	27.00	33.33	22.95
Secondary*	47.64	42.00	51.28	39.34
University**	27.48	31.00	15.38	37.70
<i>Household income</i>				
Less than €900–€1500*	-	9.00	51.28	26.23
€1501–€3500**	-	55.00	43.59	62.30
€3501–more than €4500	-	36.00	5.13	11.48

277 Note: ^a Data obtained from the Register (INE, 2017) on January 1, 2017 (www.ine.es). ^b OECD (2014). * The
 278 correlation is significant at the 0.05 level based on the χ^2 test between segments. ** The correlation is significant at
 279 the 0.01 level based on the χ^2 test between segments.

280

281 The final sample consisted of 100⁹ adults out of 113¹⁰ in total, who were older than 18
 282 years and without eye problems. Compared with previous ET studies, this sample is rather large.

⁹ For an eye-tracking study, this is a rather large sample, taking into account that past ET studies employed far fewer subjects (e.g., 53 in Ares et al., 2013; 71 in Ares et al., 2014; 40 in Balcombe et al., 2015; 99 in Balcombe et al., 2017; 10 in Bialkova & van Trijp, 2011; 24 in Bialkova et al., 2014; 48 in Fenko, et al., 2018; 59 in Gere et al.,

283 Most respondents were female (51 percent). With respect to age and education, our sample is
284 similar to the population in Spain, with approximately one-quarter of the respondents being
285 between 35 and 44 years old and 40 percent being more than 55 years old. Around half of the
286 sample had completed secondary studies.

287 3.3.2 *Measurement of the importance of yogurt attributes and nutritional claims to the* 288 *participants*

289 After completing the DCE and ET study, the respondents answered a set of questions
290 aimed at capturing the importance that they attach to the following eight yogurt attributes: price,
291 taste, brand, healthiness, convenience, health claims, nutritional claims, and natural ingredients.
292 Food choice motives and the related importance that consumers attach to product attributes are
293 valuable bases for segmentation (Haley, 1968; Jadczačková, 2013), because they determine to a
294 large extent the food choices that consumers make and the arguments and information to which
295 they are sensitive (Bellows & Hallman, 2010). Therefore, the insights gained by segmenting
296 consumers based on these importance ratings can help to identify effective marketing strategies
297 aimed at promoting healthy food consumption (Verain, Sijtsema, & Antonides, 2016).

298 The eight yogurt attributes were included based on previous studies on different food
299 categories (Grunert, Hieke, & Wills, 2014; Van Loo et al., 2015). The importance of yogurt
300 attributes was scored on a 5-point scale ranging from “not at all important” (1) to “extremely
301 important” (5), and the attributes were merged into one construct (Cronbach’s $\alpha = 0.70$). In
302 addition to measuring the importance of yogurt attributes, we asked the participants to rate how
303 important it is to them that the yogurt that they usually purchase contains one of the following
304 NCs: low sugar, fat free, source of calcium, source of vitamin B₆, and high in fiber. The
305 importance of each NC was scored on a 5-point scale ranging from “not at all important” (1) to
306 “extremely important” (5), and the NCs were merged into 1 construct (Cronbach’s $\alpha = 0.69$).

307
308

2016; 29 in Samant & HanSeok, 2016; 32 in Spinks & Mortimer, 2016; 22 in Van der Laan et al., 2015; 81 in Van Loo et al., 2015; 81 in Van Loo, Nayga, Campbell, Seo, & Verbeke, 2017; 50 in Varela, Antúnez, Cadena, Giménez, & Ares, 2014; and 39 in Zhang & Seo, 2015).

¹⁰ It should be noted that 13 participants were not able to complete the entire experiment due to problems with their vision.

309 3.4 Data analysis

310

311 3.4.1 Statistical analysis of yogurt attributes and eye-tracking variables

312

313 The yogurt attributes and ET variables were analyzed using STATA 12 (StataCorp.,
314 Texas, TX). The scale construct reliability was tested with Cronbach's α , while the correlations
315 between the attributes and the ET variables were tested with Spearman's correlation coefficients.
316 The yogurt attributes were used as segmentation variables in cluster analysis. Cluster analysis
317 allows the grouping of observations into segments in which the preferences within the same
318 segment are similar while the preferences between segments are dissimilar (Wedel & Kamakura,
319 2000). As suggested by Van Loo et al. (2015) and Verain et al. (2016), we applied a two-step
320 procedure. First, a hierarchical agglomerative clustering procedure defined the number of
321 clusters and the cluster centroid (Ketchen & Shook, 1996). Second, a non-hierarchical (k-means)
322 approach was used to group the respondents into the optimal number of clusters using the
323 centroids of the sub-clusters found in the first step as initial starting points (Ketchen & Shook,
324 1996). Two distinct segments with relatively homogeneous importance ratings were identified as
325 the optimal solution. Cross-tabulations with student t-test statistics were used to determine the
326 associations between the categorical variables, while an Anova F-test and Bonferroni post hoc
327 test were used for the comparison of mean scores.

328 3.4.2 Econometric analysis of the choice experiment and eye tracking

329

330 The DCE method is consistent with the random utility theory and the theory of consumer
331 demand (Lancaster, 1966). A random utility function may be defined as follows:

332
$$U_{njt} = V_{njt} + \varepsilon_{njt} \tag{1}$$

333

334 where U_{nj} is the n^{th} utility from the consumer's choice of alternative j ; V_{nj} is the systematic or
335 representative portion of the utility function, which depends on the product attributes and their
336 values for alternative j ; and ε_{nj} is the stochastic Gumbel distributed error term (unobserved and
337 treated as random). To estimate the consumer preferences for the multiple NCs, we used a
338 random parameter logit (RPL) model (Train, 2003). More specifically, we estimated an RPL
339 model, named RPL1, which accounts for both random taste variation and correlation patterns

340 across random parameters. Given our choice experiment, the utility function that individual n
 341 derives from alternative j in choice situation t is defined as follows:

$$\begin{aligned}
 342 \quad U_{njt} = & \text{OptOut} + \beta_1 Ffat_{njt} + \beta_2 Lsugar_{njt} + \beta_3 Hfiber_{njt} + \beta_4 SvitB6_{njt} + \\
 343 \quad & \beta_5 Scalcium_{njt} + \varepsilon_{njt}
 \end{aligned} \tag{2}$$

344 where n is the number of respondents, j represents the available choices in the choice tasks (two
 345 experimentally designed yogurt profiles and the opt-out option), and t is the number of choice
 346 situations. *OptOut* is the alternative-specific constant representing the opt-out option. The
 347 variables related to the five NCs (fat free, *Ffat*; low sugar, *Lsugar*; high fiber, *Hfiber*; source of
 348 vitamin B₆, *SvitB₆*; and source of calcium, *Scalcium*) enter the model as dummy variables, and
 349 “full fat – unlabeled” yogurt represents the product of reference.

350 To investigate the effects of visual attention on consumer choice behavior and preferences,
 351 we estimated two additional RPL models that incorporate the visual attention data into the utility
 352 function. In particular, RPL2 adds to RPL1 by including visual attention in terms of fixation time
 353 expressed in milliseconds, and RPL3 adds to RPL1 by including visual attention in terms of
 354 fixation count. In line with Grebitus, Roosen, and Seitz Carolin (2015) and Van Loo et al.
 355 (2015), we rescaled the fixation time spent and fixation count to have a zero mean. For RPL2
 356 and RPL3, the utility function specified for individual n , alternative j , in choice situation t , is
 357 defined as follows:

$$\begin{aligned}
 358 \quad U_{njt} = & \text{OptOut} + \beta_1 Ffat_{njt} + \beta_2 Lsugar_{njt} + \beta_3 Hfiber_{njt} + \beta_4 SvitB6_{njt} + \\
 359 \quad & \beta_5 Scalcium_{njt} + \gamma_{Ffat}(FtFfat * Ffat_{njt}) + \gamma_{Lsugar}(FtLsugar * Lsugar_{njt}) + \\
 360 \quad & \gamma_{Hfiber}(FtHfiber * Hfiber_{njt}) + \gamma_{SvitB6}(FtSvitB6 * SvitB6_{njt}) + \\
 361 \quad & \gamma_{Scalcium}(FtScalcium * Scalcium_{njt}) + \varepsilon_{njt}
 \end{aligned} \tag{3}$$

362 where γ_{Ffat} is the coefficient of the interaction term between the fat-free attribute and the
 363 fixation time *FtFfat* for the fat-free attribute and so on for the other attributes. Thus, in RPL2,
 364 the *FtFfat* variable is the mean-centered fixation time spent on the fat-free nutritional claim,
 365 whereas, in RPL3, *FcFfat* is the mean-centered fixation count. Similarly, the other γ s are the
 366 coefficients of the interaction terms between the attribute and the visual attention mean-centered
 367 variables. The remaining variables are as specified in (2).
 368

369 In all the models, it is assumed that the coefficients of the five NCs (*Ffat*, *Lsugar*, *Hfiber*, *SvitB6*,
 370 and *Scalcium*) are random and follow a normal distribution. In the RPL2 and RPL3 models, the
 371 interaction terms are also assumed to be random and to follow a normal distribution.

372 4. Results

373 4.1 Consumer segmentation and stated importance of yogurt attributes

374
 375 The results from the questionnaire reveal that, when evaluating yogurt attributes,
 376 participants attach the highest level of importance to the health aspect of the product, followed
 377 by taste and nutritional and health claim labels (Table 4).

378 Table 4 – Importance of yogurt attributes

No.		Mean	Standard deviation
1	Health ^a	4.16	0.81
2	Taste	4.12	0.91
3	NC labels	4.11	0.91
4	HC labels	3.95	1.11
5	Natural ingredients	3.85	0.99
6	Price	3.66	1.01
7	Brand	3.09	1.04
8	Convenience ^b	2.72	1.16

379 Note: Measured on a 5-point scale from 1 (not at all important) to 5 (extremely important). ^a Health means that
 380 consumers might choose the product because of the health properties that it holds. ^b Convenience means that it can
 381 be found easily, there is a large variety, and it can be combined easily with other food.

382
 383 This result suggests that NCs are perceived as being less important than health and taste
 384 and more important than health claims, natural ingredients, price, brand, and convenience. From
 385 the cluster analysis using the importance of yogurt attributes, we obtained two distinct consumer
 386 segments. The segment sizes and scores are reported in Table 5.

387 Segment 1 (39 percent of the sample) attaches the greatest importance to the *fat-free* type
 388 of nutritional claim followed by the *source of calcium* and *source of vitamin B6* types of
 389 nutritional claims when purchasing yogurt. Segment 2 (61 percent of the sample), on the other
 390 hand, attaches the greatest importance to the *source of calcium* NC followed by the *fat-free* and
 391 *source of vitamin B6* types of claims. The *high in fiber* type of claim is the least valued claim by
 392 both segments. With respect to the importance attached to yogurt attributes, both segments do
 393 not attach importance to any of the yogurt attributes mentioned in Table 5. The χ^2 test revealed
 394 no significant differences across the segments in terms of the socio-demographic variables

395 gender, age group (45–54 and older than 54), education (primary), and income (from €3501 and
 396 above €4500) (Table 3). To describe the segments further, the importance of NCs on the yogurt
 397 packaging (Table 5) was compared with the visual attention data (Sections 4.2, 4.3, and 4.4).

398 Table 5 – Two-cluster solution and profiling of consumer segments (n = 100)

	Segment 1		Segment 2
<i>Segment size (n)</i>	39 (39.00%)		61 (61.00%)
<i>Importance of yogurt attributes^b</i>			
Taste	4.23 (0.78) ^a	Health	4.23 (0.76)
Health claims	4.10 (0.99)	Nutritional claims	4.11 (0.95)
Nutritional claims	4.10 (0.85)	Taste	4.05 (0.99)
Health	4.05 (0.89)	Health claims	3.85 (1.18)
Natural ingredients	3.85 (1.01)	Natural ingredients	3.85 (0.98)
Price	3.72 (0.94)	Price	3.62 (1.05)
Brand	3.00 (1.10)	Brand	3.15 (1.00)
Convenience	2.64 (1.20)	Convenience	2.77 (1.13)
<i>Importance of NCs' attributes^b</i>			
Fat free*	3.69 (1.30)	Source of calcium*	3.64 (1.20)
Low sugar	3.54 (1.39)	Low sugar	3.57 (1.16)
Source of calcium*	3.31 (1.16)	Fat free*	3.33 (1.22)
Source of vitamin B ₆ **	3.15 (1.16)	Source of vitamin B ₆ **	2.72 (1.29)
High fiber	2.92 (1.35)	High fiber	2.64 (1.08)

399 Note: * The correlation is significant at the 0.05 level based on the student t-test between segments. ** The
 400 correlation is significant at the 0.01 level based on the student t-test between segments. ^a Mean (standard deviation).
 401 ^b Measured on a 5-point scale from 1 (not at all important) to 5 (extremely important).
 402

403 4.2 Visual attention to NCs based on eye-tracking measures

404 The participants had the highest fixation count on the *low-sugar* NC with an average of 9
 405 fixations and 2146 milliseconds of fixation time, suggesting that *low sugar* is the most important
 406 attribute when customers make their choices. On average, *source of calcium* and *high fiber*
 407 received fewer fixations than the other NCs. The fixation time and fixation count are reported in
 408 Table 6.

409 Table 6 – Average eye-tracking measures for the total of 5 stimuli (n = 100)

AOIs	Fixation time (ms) ¹				Fixation count			
	Mean	Std Dev.	Min.	Max.	Mean	Std Dev.	Min.	Max.
Fat free	2057.15	1630.92	118	8544	8.30	5.20	1	26
High fiber	1314.83	1046.70	113	4665	5.37	3.63	1	18
Low sugar	2145.85	1555.14	101	7826	8.96	5.29	1	25
Source of calcium	1787.37	1245.8	129	4978	7.85	4.68	1	18
Source of vitamin B ₆	1957.87	1257.26	116	5405	8.75	4.58	1	21

410 ¹ Milliseconds.

411 *4.3 Relationship between visual attention and nutritional claims' importance*

412 The results show several relationships between the total fixation count and fixation time within
 413 an AOI and the stated importance of the NCs (Table 7).

414

415 Table 7 – Pearson correlation coefficients between stated importance and visual attention to
 416 yogurts with NCs

Stated importance ²	Fixation time (ms) ¹					Fixation count				
	Fat free	High fiber	Low sugar	Source of calcium	Source of vitamin B ₆	Fat free	High fiber	Low sugar	Source of calcium	Source of vitamin B ₆
Fat free	0.141	0.178	0.176	0.239	0.182	0.153	0.145	0.165	0.218	0.171
(p-values)	(0.161)	(0.076)	(0.079)	(0.017)	(0.070)	(0.130)	(0.151)	(0.101)	(0.029)	(0.089)
High fiber	0.086	0.138	0.195	0.201	0.186	0.061	0.139	0.170	0.218	0.140
(p-values)	(0.393)	(0.172)	(0.053)	(0.045)	(0.064)	(0.546)	(0.167)	(0.091)	(0.030)	(0.165)
Low sugar	-0.002	0.075	0.057	0.090	0.074	0.021	0.101	0.066	0.010	0.060
(p-values)	(0.984)	(0.461)	(0.573)	(0.373)	(0.467)	(0.839)	(0.317)	(0.514)	(0.339)	(0.554)
Source of calcium	0.172	0.159	0.240	0.202	0.215	0.164	0.157	0.269	0.211	0.209
(p-values)	(0.087)	(0.114)	(0.016)	(0.044)	(0.032)	(0.103)	(0.120)	(0.007)	(0.035)	(0.037)
Source of vitamin B ₆	0.138	0.162	0.279	0.231	0.199	0.168	0.195	0.310	0.292	0.211
(p-values)	(0.171)	(0.107)	(0.005)	(0.021)	(0.048)	(0.094)	(0.052)	(0.002)	(0.003)	(0.035)

417 Note: ¹ Milliseconds. ² The stated importance attributes are measured on a 5-point scale from 1 (not at all important)
 418 to 5 (extremely important).

419 There is a positive significant relationship between the stated importance and the fixation
 420 count or fixation time for two NCs: *source of calcium* and *source of vitamin B₆*. This finding
 421 suggests that those stating that they attach a high degree of importance to these two NCs when
 422 purchasing yogurt truly do pay more attention to these attributes when making choices. With
 423 respect to the rest of the visual attention and NC attributes, we observe a small positive
 424 correlation (e.g., low sugar fixation time and high fiber (0.053), high fiber fixation count and
 425 source of vitamin B₆ (0.052)); however, this correlation is weak and is not significant at the 5
 426 percent level. This suggests that the relationship suggested by the correlation between these
 427 variables could have happened by chance. Therefore, we accept the null hypothesis and conclude
 428 that there is no correlation between these and the rest of the variables above the 5 percent
 429 significance level.

430 *4.4 Differences in visual attention across segments*

431 The differences in visual attention across segments that attach different degrees of
 432 importance to NC attributes for yogurt are reported in Table 8.

433 Table 8 – Visual attention degree of importance to NC attributes for yogurt

<i>Segment size (n)</i>	Segment 1 39 (39.00%)		Segment 2 61 (61.00%)
<i>Fixation count</i>			
Low sugar***	13.97 (4.16)	Source of vitamin B ₆ ***	6.15 (2.87)
Fat free***	12.90 (4.72)	Low sugar***	5.75 (2.90)
Source of vitamin B ₆ ***	12.82 (3.72)	Fat free***	5.36 (2.83)
Source of calcium***	12.28 (3.55)	High fiber***	3.46 (1.75)
High fiber***	8.36 (3.81)	Source of calcium***	4.97 (2.66)
<i>Fixation time (ms)¹</i>			
Low sugar***	3671.33 (1305.22)	Source of vitamin B ₆ ***	1204.89 (649.66)
Fat free***	3500.28 (1620.93)	Low sugar***	1170.54 (657.13)
Source of vitamin B ₆ ***	3135.62 (1057.34)	Fat free***	1134.49 (711.06)
Source of calcium***	3004.97 (974.11)	Source of calcium***	995.95 (608.50)
High fiber***	2255.28 (1031.71)	High fiber***	713.55 (437.16)

434 Note: * The correlation is significant at the 0.05 level based on the student t-test. ** The correlation is significant at
 435 the 0.01 level based on the student t-test. ¹ Milliseconds.

436
 437 The fixation time and count for the various attributes are indicators of their relevance to
 438 participants’ purchase decisions. Therefore, we expect the segments that attach greater
 439 importance to various attributes also to have stronger visual attention in terms of fixation time
 440 and count. We find significant differences in the fixation time and count for the various NCs
 441 between S1 and S2 (Table 8). Although there are differences in the visual attention between the
 442 two segments, S1, albeit smaller, has greater visual attention in terms of fixation time and count
 443 for all the NCs than S2. The participants in this segment showed the strongest visual attention in
 444 terms of fixation time to the *fat-free* and *low-sugar* NCs followed by the *source of vitamin B₆*
 445 claim. On the other hand, in terms of the fixation count, the participants paid the most attention
 446 to the *low-sugar* and *fat-free* NCs, followed by the *source of vitamin B₆* claim. The visual
 447 preferences in S2 seem to be slightly different from those in S1; however, they are consistent in
 448 terms of fixation time and count visual attention. More specifically, regarding both fixation time
 449 and fixation count, the participants paid the most attention to the *source of vitamin B₆* and *low-*
 450 *sugar* NCs followed by the *fat-free* claim. Overall, the *high-fiber* NC is the least-valued NC for
 451 both eye-tracking measures.

452
 453

454 4.5 Effect of visual attention to nutritional claims on choice behavior for yogurt

455

456 RPL1, the baseline model, assumes random taste heterogeneity and correlation patterns
457 across random parameters, while RPL2 and RPL3 add the interaction terms between the NCs and
458 the visual attention measures fixation time and count¹¹ to RPL1. Hence, RPL2 and RPL3 allowed
459 us to determine whether consumers who pay more attention to an attribute value it more. As
460 expected, the results show that the coefficient of the opt-out option is negative and statistically
461 significant in all the models, indicating that consumers gain more utility from choosing one of
462 the experimentally designed yogurt profiles rather than the opt-out choice. The coefficients of the
463 five NCs (i.e., *fat free*, *low sugar*, *high fiber*, *source of vitamin B₆*, and *source of calcium*) are
464 also all positive and statistically significant at the 1 percent and 5 percent significance levels in
465 all the models, indicating that consumer utility increases when these claims are reported on
466 yogurt packages.

467 The corresponding standard deviations are also statistically significant, suggesting that
468 consumers' preferences for these five attributes are heterogeneous. According to the results from
469 RPL1, consumer utility is greater when a yogurt bears the *fat-free* NC, followed by the *high-fiber*
470 and *source of calcium* claims, in comparison with the unlabeled yogurt. On the other hand,
471 yogurt that bears the *source of vitamin B₆* or the *low-sugar* claim is the least preferred.
472 Participants' utility changes when we look at the visual attention results. In both models (RPL2
473 and RPL3), four of the five interaction terms are statistically significant: those related to *calcium*,
474 *fat*, *fiber*, and *vitamin B₆* contents. This result indicates that a longer fixation time or higher
475 fixation count is related to greater utility for these attributes. In other words, people who visually
476 attend more to these types of NCs are more likely to choose yogurt that carries them. Table 9
477 reports the coefficient estimates from the three RPL models.¹²

478 A model fit comparison of the information criteria shows that RPL1 and RPL3 improve the
479 model performance. This result suggests that the incorporation of visual attention in terms of
480 fixation count information as covariates improves the model fit (see the model fit comparison in
481 Appendix C (Table C1)).

¹¹ The fixation time and fixation count are in the utility model as dummy variables. They take the value of 1 when the individuals' fixation time (milliseconds) or fixation count is equal to or higher than the centered mean of each attribute and 0 otherwise (e.g. the fat-free yogurt takes the value of 1 if the time fixation is equal to or higher than 2057 ms or 0 otherwise).

¹² The results from the Cholesky matrix are available on request.

482 Table 9 – Results of three random-parameter logit model specifications

<i>Parameters</i>	RPL 1	RPL 2	RPL 3
	-	Fixation time	Fixation count
	β (z)	β (z)	β (z)
Opt-out	-1.34 (-8.06)***	-1.38 (-7.98)***	-1.37 (-7.93)***
Fat free	3.13 (8.57)***	3.30 (8.46)***	3.44 (7.93)***
Standard deviation	4.01 (9.56)***	4.20 (8.17)***	4.26 (8.08)***
Low sugar	0.76 (2.08)**	1.07 (2.49)**	1.15 (2.24)**
Standard deviation	2.71 (8.37)***	4.14 (5.54)***	3.84 (4.65)***
High fiber	2.39 (7.08)***	2.42 (6.84)***	2.76 (6.77)***
Standard deviation	2.99 (8.38)***	3.68 (7.42)***	3.57 (7.85)***
Source of vitamin B₆	1.22 (3.94)***	1.12 (3.50)***	0.77 (2.14)**
Standard deviation	3.04 (8.8)***	3.46 (5.08)***	1.96 (4.79)***
Source of calcium	2.09 (4.82)***	0.93 (2.75)***	1.00 (2.77)***
Standard deviation	2.12 (6.15)***	1.56 (4.36)***	2.02 (4.53)***
Int. 1 – Fat	-	2.55 (2.81)***	2.66 (4.23)***
Standard deviation		1.56 (4.36)***	2.02 (4.53)***
Int. 2 – Sugar	-	-0.41 (-0.77)	-0.25 (-0.42)
Standard deviation		1.22 (2.41)**	0.17 (0.39)
Int. 3 – Fiber	-	2.35 (3.76)***	1.43 (2.46)**
Standard deviation		1.15 (2.11)**	0.91 (1.89)*
Int. 4 – Vitamin B₆	-	0.64 (1.70)*	1.33 (2.96)***
Standard deviation		1.23 (2.43)**	1.12 (3.09)***
Int. 5 – Calcium	-	2.61 (5.22)***	3.36 (6.83)***
Standard deviation		1.53 (3.40)***	1.23 (3.09)***
N	4500	4500	4500
Log likelihood	-934.08	-895.10	-868.14
AIC	1.274	1.282	1.246

483 Note: Significance levels at *** 1%, ** 5%, and * 10%.

484 **5. Discussion and final remarks**

485 This study combined a DCE and ET regarding yogurt selection to assess consumers’
 486 valuation of multiple NCs and to investigate whether attention is related to food choice decisions
 487 in one European country (Spain). Consumer heterogeneity was taken into account through
 488 consumer segmentation, which entailed the classification of the participants into two segments
 489 by consumer characteristics. Those in segment 1, compared with those in segment 2, are more
 490 likely to be male, to be between 18 and 34 years old, to have completed secondary studies, and to
 491 have a low income. This segment attached a high level of importance to the *fat-free* NC followed
 492 by a *source of calcium and a source of vitamin B₆*. Segment 2 is characterized by females aged

493 between 18 and 34 years with a higher income than segment 1 who had completed secondary
494 education. For this segment, the most important NCs considered when purchasing yogurts were
495 the *source of calcium* type of claim followed by the *fat-free* and *source of vitamin B₆* claims. The
496 preferences of segment 2 are consistent with the interaction terms (i.e., fixation count visual
497 attention and choice) of the RPL 3 model, which also had the best model fit.

498 In terms of the importance attached to yogurt attributes, we did not find any statistically
499 significant differences between segments. This result suggests that there is homogeneity in the
500 importance given to these attributes between our two segments. The first four most important
501 attributes to the participants of both segments when purchasing yogurt were taste, nutritional
502 claims, health claims, and health. These findings are consistent with the results of previous
503 studies that defined taste as one of the most important attributes in the decision to purchase food
504 products (Carrillo et al., 2012; Insch & Jackson, 2014; Markovina et al., 2015; Sautron et al.,
505 2015). Moreover, the results are consistent with a previous study by Rebollar, Lidón, Guzmán,
506 Gil, and Martín (2017), who found healthfulness to be one of the most important attributes in
507 yogurt for Spanish consumers.

508 Taking the aforementioned into consideration, food companies should be willing to
509 differentiate their products according to these preferences. These results can be informative and
510 challenging to producers and processors: informative in terms of promoting the *source of*
511 *calcium*, *fat-free*, and *source of vitamin B₆* types of NCs as a differentiation strategy and
512 challenging in terms of combining taste and health (i.e., two intrinsic attributes) to reduce the
513 “halo” effect of the common belief that “healthy” in most cases equals less tasty food products.
514 Since taste has been found to be one of the most important determinants of repeated purchases
515 (Elbel, Gyamfi, & Kersh, 2011; Holmquist, McCluskey, & Ross, 2012), a strategy that would
516 allow consumers to taste the food product before purchasing it may generate repurchases in the
517 case of satisfaction and may be seen as a form of differentiation. This strategy is common in
518 some stores in the US (e.g., Costco) and has proven to be effective in increasing sales (Pinsker,
519 2014).

520 In terms of the extent to which providing NCs on yogurt packages may provide a signal
521 detection assumption that increasing participants’ visual attention may result in increasing the
522 probability of the product being purchased (H1), we showed that visual attention in terms of

523 fixation count may increase the likelihood of a product being purchased. This finding is in line
524 with the overall results of previous studies that suggest that visual attention plays a role in
525 explaining choice behavior (Bialkova & van Trijp, 2011; Bialkova et al., 2014; Graham &
526 Jeffery, 2011; Samant & HanSeok, 2016; Uggeldahl et al., 2016; Van der Laan et al., 2015; Van
527 Loo et al., 2015, 2017; Vu et al., 2016). This finding is consistent with Orquin and Holmqvist
528 (2018), who suggested that the total dwell time may threaten the external validity of the study.
529 Our results partially confirm that greater utility is generated when the *fat-free* and *low-sugar*
530 claims (H2) are present on the yogurt package compared with the other claims. Overall, the
531 results from the interactions of the DCE and ET suggest that the *fat-free* claim received the
532 second-strongest visual attention, after *source of calcium*, and was the most chosen among the
533 claims. This result is consistent with the attribute preferences from the cluster analysis (segment
534 2) and is in line with the previous studies by Krystallis and Chrysochou (2012) and Van
535 Wezemael et al. (2014), who found that consumers have positive perceptions of and attach
536 higher values to NCs related to fat content and saturated fat.

537 The *low-sugar* NC, on the other hand, was the least-preferred claim in all the models.
538 This result also confirms the increasing evidence that what consumers say about their
539 preferences regarding NCs is not actually reflected in what they finally purchase in the
540 marketplace. One reason for rejecting the *low-sugar* NC may be that consumers reject sugar-
541 reduced products that do not meet their sensory preferences, even if they are more healthful than
542 regular products (Civille & Oftedal, 2012). Therefore, emphasizing sugar reduction may create
543 negative sensory effects and decrease the value of a product (e.g., yogurt) (Brunner, Horst, &
544 Siegrist, 2010; Lähteenmäki et al., 2010; Raghunathan, Naylor, & Hoyer, 2006). Although the
545 *fat-free* NC was the most valued by both clusters and produced the greatest utility in terms of
546 visual attention and final choice in yogurt, producers, processors, and retailers should carefully
547 consider the type of food product and modify the sensory characteristics related to the NCs
548 accordingly (e.g., fat reduction in meat products, in general, reduces the sensory quality, the
549 texture, and the acceptance of the final product; Méndez-Zamora et al., 2015).

550 This study has some limitations that constitute areas for further research. The first
551 limitation is that, even though we found that the presence of NCs on yogurts' FOP increases
552 attention, we cannot prove this with certainty but can only assume that attention might be linked

553 to an increased likelihood of affecting the final decision to purchase yogurts with NCs. As
554 defined by Orquin and Holmqvist (2018), it is difficult to support an eye–mind assumption,
555 because researchers cannot know whether the presence of fixation implies that the object has
556 been processed or not and vice versa. Therefore, whilst we maintain that eye tracking is useful,
557 we argue that more research is needed to understand the extent to which ET data can be used to
558 improve stated preference research. The second limitation is that this research was carried out in
559 only one European country due to the limitation in funding; hence, it should be replicated in
560 other countries to provide more evidence. Future research using eye tracking should be
561 developed not only in lab conditions but also in a real supermarket context using eye-tracking
562 glasses to test the consumers’ attention in terms of preferences and decision making in different
563 contexts.

564 Finally, since each NC has its own effect on people’s health, it would also be interesting
565 to explore groups of consumers with similar shopping goals (e.g., fat-free products for
566 consumers who are concerned about reducing their cholesterol level) and discover whether their
567 taste preference is more important than their health goals.

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576 **Conflict of interest**

577 The authors declare no conflict of interest.

578

579 **Highlights:**

580

- 581 • Two clusters profile consumer segments for Spanish yogurts with nutritional claims.
- 582 • The presence of NCs on yogurts’ front of pack increases the attention of consumers.
- 583 • The *low-sugar* claim was the least valued of the claims.
- 584 • Visual attention (fixation count) increases the likelihood of purchase decisions.

585

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587

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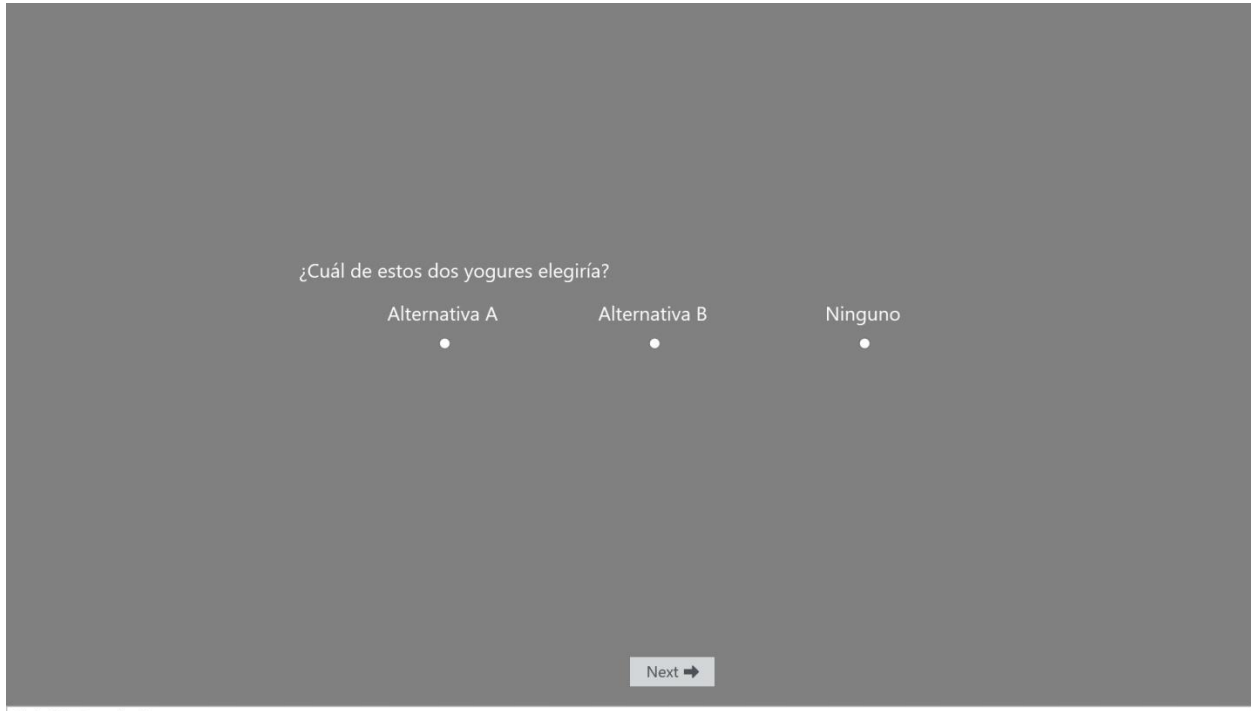
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Appendixes

Appendix A

Figure A1 – An evaluation form of the most-preferred yogurt



Note: The question is translated from Spanish as follows: “Which of these two yogurts would you choose?” “Alternativa A” refers to option A, “Alternativa B” refers to option B, and “Ninguno” is the “no-buy” option.

Appendix B

Table B1 – Population in Spain and Zaragoza (%)

Total	Sex ^a		Age						
	Female	Male	0–14	15–34	35–54	55–64	65–84	85 and above	
Spain	46,624,382	51	49	15.06	22.59	32.20	11.76	15.60	2.79
Zaragoza	1,317,847	50	50	14.06	21.13	31.53	12.24	17.24	3.80

Source: Spanish Census of Population, 2017, www.ine.es. ^a In percentages.

953 **Appendix C**

954
955 The model fit information criteria, such as the Akaike Information Criterion (AIC) and the
956 Bayesian Information Criterion (BIC), as well as the log-likelihood values, can be used to
957 discuss the relative fit of the various models (Table C1). The lower the information criteria, the
958 better the model fit. It is known that using the BIC (AIC) tends to under-fit (over-fit) models,
959 while evidence presented in previous studies (Caputo, Nayga, & Scarpa, 2013; Dias, 2006)
960 shows that AIC3 (with three instead of two weights for parameter penalization) outperforms the
961 other two, correcting for the over-fitting.

962
963 Table C1 – Comparison of the information criteria

Model	Choices	Log-Lik.	Parameters	BIC/N	AIC/N	AIC3/N
MNL	1499	-1227.45	6	1.650	1.646	1.650
RPL1	1499	-934.08	21	1.261	1.274	1.288
RPL2	1499	-895.10	66	1.334	1.282	1.326
RPL3	1499	-868.14	66	1.298	1.246	1.290

964
965 Nevertheless, the BIC assumes that one of the models is the true one, which is unlikely to be the
966 case here, while the AIC aims at finding the model that approximates the unknown data-
967 generating process (by minimizing the expected estimated Kullback–Leibler divergence). All
968 three, BIC, AIC, and AIC3, favor RPL1 and RPL3 over the competing models. The combined
969 evidence from ruling out RPL2 and preferring RPL1 and RPL3 suggests that these two are
970 indeed the best models. In addition, the log-likelihood is closer to zero and the information
971 criteria are lower in RPL1 and RPL3 than in RPL2, implying that the incorporation of visual
972 attention in terms of fixation count information as covariates improves the model fit.