

Drought conditions, aridity and forest structure control the responses of Iberian holm oak woodlands to extreme droughts: a large-scale remote-sensing exploration in eastern Spain

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26 **Abstract.** Understanding how Mediterranean forests respond to the increasing frequency
27 of extreme droughts and forest densification is crucial for effective land management in
28 the present context of climate change and land abandonment. We study the responses of
29 Iberian holm oak (*Quercus ilex* L.) woodlands to recent extreme droughts during 2000-
30 2019 along broad gradients of climate aridity and forest structure. To this purpose, we
31 apply large-scale remote-sensing using MODIS EVI as a primary production proxy in
32 5274 *Q. ilex* sites distributed within a 100,000 km² region in eastern Spain. These
33 woodlands were extensively affected by two extreme drought events in 2005 and 2012.
34 Resistance, assessed as the capacity of the ecosystems to maintain primary production
35 during drought, was significantly lower for semi-arid than for sub-humid and dry-
36 transition conditions. Holm oak woodlands located in semi-arid areas of the region
37 showed also poorer resilience to drought, characterized by low capacity to fully recover
38 to their pre-drought production levels. Further, drought intensity and both pre- and post-
39 drought hydric conditions controlled the variations of resistance, recovery and resilience
40 between the two analyzed extreme drought events. Drought effects were particularly
41 negative for dense *Q. ilex* stands under semi-arid climate conditions, where strong
42 competition for scarce water resources reduced drought resistance. The observed drought
43 vulnerability of semi-arid holm oak woodlands may affect the long-term stability of these
44 dry forests. Adaptive management strategies, such as selective forest thinning, may be
45 useful for improving drought responses in these more vulnerable semi-arid woodlands.
46 Conversely, natural rewilling may more appropriately guide management actions for
47 more humid areas, where densely developed *Q. ilex* woodlands show in general a high
48 ability to maintain ecosystem primary production during drought.

49 **Keywords:** aridity; drought; ecosystem stability; forest structure; holm oak; resilience.

50 **1. Introduction**

51 Forest ecosystems play a critical role for the maintenance of biodiversity worldwide and
52 the provision of many other ecosystem services, including water cycle regulation, erosion
53 control, habitat creation, carbon sequestration, the production of wood and non-wood
54 market goods and cultural/recreational services (Mori et al., 2017). Environmental data
55 records indicate that the organization and function of these landscapes is rapidly changing
56 under the influence of external drivers, such as increased water shortages and frequency
57 of droughts resulting from climate change (Simonson et al., 2014). Drought effects
58 commonly include the reduction of forest productivity, canopy defoliation, tree mortality
59 and the alteration of natural regeneration patterns (Allen et al., 2015; DeSoto et al., 2020;
60 García-Fayos et al., 2020), which together can induce large changes in the carbon cycle
61 at the forest stand level (Reichstein et al., 2013; Anderegg et al., 2020) and long-term
62 geographical displacement at the forest species level (Benito-Garzón et al., 2008; Ruiz-
63 Labourdette et al., 2012; Mauri et al., 2022).

64 The Mediterranean basin is currently perceived as a hotspot of climate change (Giorgi,
65 2006; Doblas-Reyes et al., 2021). Regional air temperatures have increased in this region
66 approx. 1.3°C since 1920, in comparison with a noticeably smaller worldwide increase of
67 ~0.85°C for the same period (Guiot and Cramer, 2005; Cramer et al., 2020). The greater
68 atmospheric evaporative demand resulting from temperature rise has also contributed to
69 an increase in the frequency, intensity and duration of droughts, especially during the last
70 5-6 decades (Vicente-Serrano et al., 2014) and will likely continue to amplify in the
71 future, in light of current climate projections of increased water shortages for the
72 Mediterranean region (Ali et al., 2022). Synchronously, land abandonment and the
73 decline of rural population in this region since the second half of the 20th century have
74 contributed to recent forest expansion and stand densification, particularly in areas of the

75 northern-shore Mediterranean basin (Lana-Renault et al., 2020). The interactions between
76 land-abandonment patterns and climate change in Mediterranean landscapes can lead to
77 complex ecosystem responses. For example, Astigarraga et al. (2020), in a large-scale
78 exploration along the Iberian Peninsula of changes in forest demography and structure
79 along 1986-2017 concluded that forest density, basal area and tree size have increased
80 since the 1980s due to land abandonment. However, this same study also indicates an
81 aggravation of the negative effects of climate change and greater stand competition due
82 to forest densification since the beginning of the 21st century, resulting in increased
83 growth reductions and tree mortality by severe droughts. How forest ecosystems react in
84 the present context of interactions between increased frequency of extreme droughts and
85 forest densification is, therefore, a question of major importance for land management
86 and adaptation in the Mediterranean region.

87 Resilience is an emergent property of ecosystems that can be described as their ability to
88 absorb and recover from disturbances , regaining the structure and functions that
89 characterize their stability domains (Gunderson, 2000). Accordingly, forest resilience to
90 extreme drought provides an integrated description of the capacity of these ecosystems to
91 maintain, recover and regain the integrity of their functions during and after the effects of
92 extreme droughts (Lloret et al., 2011; Anderegg et al., 2020). Previous studies indicate
93 that these forest responses to drought depend on complex interactions of multiple intrinsic
94 and extrinsic environmental factors. For example, while tree traits and physiology control
95 drought tolerance, drought intensity and both pre- and post-drought conditions have a
96 primary influence on drought stress and post-drought tree growth legacies, constituting
97 important determinants of resilience (McDowell et al., 2008; Greenwood et al., 2017;
98 Forner et al., 2018). Further, climate aridity, which constraints water availability and
99 forest productivity, can largely increase forest vulnerability to drought, particularly for

100 local tree populations distributed near the species' aridity tolerance limit (Camarero et al.,
101 2013; Gazol et al., 2017; DeSoto et al., 2020). Forest structure is also likely to affect forest
102 responses to drought. Specifically, intense competition for the use of water due to high
103 tree density has been associated with drought-induced tree mortality (Greenwood and
104 Weisberg, 2008; Xerra-Maluquer et al., 2018; Ogaya et al., 2020).

105 Forest response to drought is frequently assessed using tree dendrochronological analysis,
106 which offers direct metrics of the effects of climate variability on tree growth (Camarero
107 et al., 2016; Gazol et al., 2017; Sánchez-Salguero et al., 2018). The scope of these tree-
108 ring data explorations is, however, commonly limited to single trees and the forest stand
109 scale. The increasing availability of decadal series of remotely sensed data, such as
110 satellite-driven vegetation indexes that strongly correlate with ecosystem net primary
111 production and tree growth, provide excellent approaches for exploring the response of
112 forest ecosystems to drought at very broad spatial scales (Gazol et al., 2018; Vicente-
113 Serrano et al., 2019). In fact, remote sensing techniques offer unique chances for
114 overcoming the strong spatial limitations that characterize dendrochronological
115 approaches, facilitating fast exploration of forest ecosystems over vast areas and remote
116 regions.

117 *Quercus ilex* L. (holm oak) is a drought tolerant evergreen oak that constitutes a keystone
118 species for the provision of ecosystem services in the Mediterranean region (Marañón et
119 al., 2012). In fact, *Q. ilex* woodlands are one of the most conspicuous dryland forests in
120 the western Mediterranean basin, where they have been historically impacted during
121 millennia by human activities (mainly wood consumption for fuel, timber and charcoal
122 production, domestic livestock and dryland agriculture) causing reductions of their range
123 and important changes in their structure, including, among others, forest transformations
124 in intensively used coppice stands (Terradas, 1999; Serrada et al., 2017; Camarero and

125 Valerio, 2023). Since 1950, however, land abandonment has promoted holm oak
126 densification in the region (Vericat et al., 2011). Species distribution models for the
127 Iberian Peninsula and Europe have identified *Q. ilex* as one of the key forest species that
128 will progressively displace cold-temperate forest species in sub-humid landscapes at
129 medium altitudes (800-1400 m) as a consequence of climate change (Ruiz-Labourdette
130 et al., 2012; Mauri et al., 2022). However, these climate change models also warn that
131 future *Q. ilex* distribution in the Mediterranean basin could also be reduced in drier areas
132 of its present distribution range, where holm oak may be increasingly exposed to water
133 scarcity and the effects of extreme droughts (Mauri et al., 2022). These negative
134 projections are in line with results obtained by recent holm oak stand explorations in semi-
135 arid areas of the Mediterranean region, where the lack of natural regeneration and
136 increased tree decline are emerging as significant problems for their stability (Camarero
137 et al., 2016; Gentiliesca et al., 2017; García-Fayos et al., 2020).

138 Despite the very conspicuous distribution of holm oak woodlands in the Mediterranean
139 region and the important changes that these dryland forests may experience in the future
140 as a function of climate change and land abandonment, the resilience of *Q. ilex* forests to
141 extreme drought has been scarcely explored at very broad spatial scales. Important
142 knowledge gaps in the analysis of these effects are: (i) the influence of aridity along the
143 full range of climate conditions that govern present *Q. ilex* spatial distribution on the
144 capacity of holm oak woodlands to maintain, recover and regain the integrity of primary
145 production during and after drought; and (ii) the impact that forest density, largely
146 influenced by the legacy of past human use and recent land abandonment, has in these
147 drought responses. In this study, we explore the resilience of Iberian holm oak woodlands
148 to recent extreme droughts during 2000-2019, using a remote sensing approach over
149 broad-scale gradients of climate aridity and forest structure in eastern Spain. We expect

150 to find a major impact of drought characteristics and aridity on the analyzed responses to
151 extreme drought, since these factors have a direct influence on water shortages and
152 drought stress. Specifically, we hypothesize that drought vulnerability of these *Q. ilex*
153 woodlands will increase with climate aridity from sub-humid to semi-arid landscapes. We
154 also expect to find a significant effect of forest structure on drought response, particularly
155 under the most arid conditions of *Q. ilex* distribution, since forest density can largely
156 exacerbate competition for scarce water resources.

157

158 **2. Materials and Methods**

159 **2.1. Study area**

160 The study region extends over 100,000 km² in eastern Spain, within the Iberian System
161 and surrounding areas of this mountain range, comprising the headwaters of the Douro,
162 Tagus and Guadiana rivers and extensive areas of the Ebro, Jucar, Turia and Mijares river
163 basins (Fig. 1). The climate can be classified as Mediterranean (Papadakis, 1966), with
164 two rainy periods concentrated in spring (April-June) and autumn (September-
165 November). Mean annual precipitation (MAP) and air temperature are 360-750 mm and
166 9.2-14.2°C, respectively (data derived from Ninyerola et al., 2005). Potential
167 evapotranspiration (PET) is 870-1190 mm (data derived from Trabucco and Zomer,
168 2009). Aridity (Ar), referred in this study as 1-MAP/PET, ranges in the region from 0.25
169 to 0.65 from the wettest, sub-humid conditions to the driest, semi-arid conditions,
170 respectively.

171 We selected a total of 5274 study sites (each 232 × 232 m; these dimensions are
172 determined by the pixel size of the UTM re-projected MODIS product applied later in
173 this study for the calculation of ecosystem production and drought resilience
174 components), distributed within sub-humid (175 sites, Ar 0.25-0.35), dry-transition (3165

175 sites, Ar 0.35-0.50) and semi-arid (1936 sites, Ar 0.50-0.65) climate conditions. Site
176 selection was developed by applying a GIS-based procedure, validated using high-
177 resolution aerial imagery. Specific site selection methods are detailed in Appendix A. Our
178 study sites comprise holm oak woodlands along the full climate distribution range of *Q.*
179 *ilex* in the region and a broad variety of forest structure (i.e., tree canopy density)
180 conditions, from closed forests to open woodlands. Active use of these holm oak
181 ecosystems, historically managed as coppice woodlands in the study region, ceased about
182 70 years ago. In fact, human population in the region has showed a 4-fold decrease since
183 1950, also leading to large reductions in livestock density, which may have facilitated
184 forest recovery in recent decades (Stellmes et al., 2013; Garcia-Fayos et al. 2020). All
185 study sites were selected without any signs of having been recently managed and no past
186 wildfire activity, at least since 2000. Elevation ranges from 730 to 1470 m above sea level
187 and all sites have flat topography (slope angle is equal or less than 6°). Soils are Mollic
188 Haploxeralfs (Soil Survey Staff, 2014) developed over homogeneous calcareous parent
189 lithology (massive limestones of Jurassic and Cretaceous age). The homogeneous
190 topography and soil/lithology characteristics of the study sites were imposed in the site
191 selection criteria to minimize the influence of environmental factors other than the factors
192 of interest under study (i.e., aridity, forest structure and drought characteristics).
193

194 **2.2. Drought resilience components**

195 We followed the approach described by Lloret et al. (2011) for the analysis of ecosystem
196 responses to extreme drought, using the following three components (or metrics) of
197 resilience to compare ecosystem primary production before, during and after the effects
198 of a drought event (Fig. 2):

199 (i) Resistance (R_t , dimensionless), which describes the capacity of an ecosystem to resist
200 the effects of a drought, and is calculated as the ratio of ecosystem production during a
201 drought period (EP_{Dr}) to pre-drought ecosystem production (EP_{PreDr}):

$$202 R_t = EP_{Dr}/EP_{PreDr} \quad (1)$$

203 (ii) Recovery (R_c , dimensionless), which describes the performance readjustment, in
204 terms of recovery of production, of an ecosystem after a drought, and is calculated as the
205 ratio of post-drought (EP_{PostDr}) to drought (EP_{Dr}) ecosystem production:

$$206 R_c = EP_{PostDr}/EP_{Dr} \quad (2)$$

207 (iii) Resilience (R_s , dimensionless), which describes the ability of an ecosystem to regain
208 pre-drought production performance after a drought, and is calculated as the ratio of post-
209 drought (EP_{PostDr}) to pre-drought ecosystem production (EP_{PreDr}):

$$210 R_s = EP_{PostDr}/EP_{PreDr} \quad (3)$$

211 We applied remote sensing estimation of annual net primary production, based on
212 temporal series of the enhanced vegetation index (EVI), to calculate the indices of
213 resilience for extreme drought events of the period 2000-2019. Following
214 recommendations of previous studies (Gazol et al., 2017; Sánchez-Salguero et al., 2018),
215 we used reference periods of 3 years before and after the drought events to determine pre-
216 drought and post-drought mean ecosystem production in the calculation of the R_t , R_c and
217 R_s metrics. The selected length for the reference periods agrees with reported 1-year to
218 3-year post-drought growth recovery of holm oak and other angiosperm forest species
219 (Anderegg et al., 2015; DeSoto et al., 2020; Gazol et al., 2020) and prevents drought event
220 overlap in the studied data series.

221

222 **2.3. Remote sensing proxy of ecosystem net primary production**

223 The enhanced vegetation index (EVI) is a remote-sensing vegetation index strongly
224 sensitive to leaf phenology, which minimizes the interferences caused by the effects of
225 atmospheric aerosols and the variations of soil background color (Huete et al., 2002).
226 Garbulsky et al. (2013) showed that annual EVI strongly correlates with *Q. ilex* stem
227 diametric increment in holm oak woodlands of Spain (Pearson's R=0.91, p<0.001), hence
228 providing an excellent proxy of tree growth. Furthermore, Ponce-Campos et al. (2013)
229 showed that the annual integral of EVI values provides an excellent estimator for annual
230 net primary production (ANPP) in woodlands and grasslands worldwide across arid,
231 semi-arid and sub-humid climate conditions. We, therefore, apply annual integral values
232 of the enhanced vegetation index (iEVI) as a proxy of ecosystem production for the
233 analysis of drought resilience components in our holm-oak sites.

234 We compiled two-decadal (2000-2019) series of EVI for all the study sites (16-day
235 temporal resolution) from the MODIS Terra satellite (MOD13Q1 product, collection 6)
236 using the NASA's LP DAAC Data Pool. The data were re-projected to UTM ETRS89
237 Zone 30N (232 m resolution after re-projection) and filtered for the removal of anomalous
238 EVI values (e.g., EVI values affected by snow, ice, clouds and other atmospheric
239 anomalies). We applied a weighted Savitzky-Golay adaptive filtering algorithm (Jönsson
240 and Eklundh, 2004) programmed to use the ancillary MOD13Q1 quality/reliability
241 information as criteria for identifying and filtering anomalous EVI values. The algorithm
242 was run using a window size of 5 data points and 3 iterations.

243 Previous studies have found a close synchronization of MODIS EVI trends with the
244 annual cycles of leaf phenology in Iberian woodlands (Garbulsky et al. 2013; Pasquato et
245 al. 2015). Similarly, annual EVI trends in the studied holm oak sites (Fig. C1 a-b in
246 Appendix C) showed an absolute maximum in June, synchronized with spring
247 production, followed by a summer reduction of EVI, frequently leading to a relative

248 minimum value associated with summer drought, and a late, small growth period until
249 the end of autumn. EVI trends then decrease to reach their absolute minimum values
250 during winter, in late February. Accordingly, the processed EVI data was aggregated in
251 annual phenological cycles (from the beginning of March to the end of February of the
252 next year) to obtain the final iEVI values used as ANPP proxy for the calculation of
253 drought resilience components.

254

255 **2.4. Forest structure and its relationship to the legacy of human use**

256 Tree canopy cover (TC, %) estimations for this study (Fig. 3a) were obtained applying a
257 field-calibrated and validated remote sensing data transformation equation developed
258 using 60-m resolution Normalized Difference Red-Edge 2 index (NDRE2; Barnes et al.,
259 2000) data obtained from a winter-season Sentinel-2 composite image of the region
260 (downloaded from the Copernicus Sentinel-2 Global Mosaic service), and 138 reference
261 (232 × 232 m) holm oak sites with field-validated TC information (Moreno-de-las-Heras
262 et al., 2018) that are distributed within the study area. We selected winter-season NDRE2
263 upon 4 alternative seasonal composite images and 21 alternative vegetation indexes, since
264 winter NDRE2 maximized the correlation with the reference TC data (Pearson's R=0.95;
265 p<0.001). The resulting TC-NDRE2 transformation equation (TC (%) =
266 215.9NDRE2 – 24.5) showed an excellent calibration and validation performance
267 (calibration R²=0.91, n=84 sites; validation RMSE=6%, n=54 sites). Full details for the
268 development, calibration and validation of the NDRE2-TC transformation equation
269 applied for TC estimation can be found in Appendix B.

270 Under the homogeneous soil/lithological and geomorphological conditions of the study
271 sites (calcareous soil parental material and flat topography), maximum tree canopy cover
272 is limited mainly by climate aridity (Moreno-de-las-Heras et al., 2018; Bochet et al.,

273 2021). In fact, maximum tree canopy cover (TC_{max} , %) varies non-linearly in the study
274 area from approx. 60% in the driest explored semi-arid conditions to 90-100% in the
275 wettest sub-humid conditions of the analyzed aridity range (Fig. 3a). In order to describe
276 forest structure, we standardized their TC values as a function of TC_{max} , using the Local
277 Deforestation index (Dl, dimensionless, Moreno-de-las-Heras et al., 2018):

278
$$Dl = \frac{(TC_{max} - TC)}{TC_{max}} \quad (4)$$

279 Dl is a forest structure and human disturbance index that is independent of climate aridity
280 and ranges from 0 for sites with well preserved, maximum development of forest cover
281 for the climate conditions of the site ($TC = TC_{max}$), to 1 for treeless ($TC = 0\%$) sites that
282 are entirely deforested.

283 Our study sites comprise a broad variety of forest structure conditions (Fig. 3b),
284 encompassing from dense *Q. ilex* forests, with tree canopy cover close to their climate
285 development potential ($Dl < 0.35$; 1435 sites), to moderately deforested sites ($Dl 0.35-$
286 0.65; 2967 sites) and very open sites with a few adult, isolated *Q. ilex* trees ($Dl > 0.65$;
287 872 sites). These Dl variations are closely related to the intensity of human deforestation
288 in the past, controlled in the study region by the proximity of the sites to the nearest human
289 settlements (Fig. 3c), where traditional human activities were typically nucleated
290 (Moreno-de-las-Heras et al., 2018).

291

292 **2.5. Drought conditions**

293 The Standardized Precipitation Evapotranspiration Index (SPEI, Vicente-Serrano et al.,
294 2010) is a multiscale drought index that quantifies the hydric conditions (i.e., difference
295 between precipitation and atmospheric evaporative demand) as a standardized variable
296 (zero mean and unit variance) and allows for comparison of drought intensity through
297 time and space, independently of the climate conditions of the comparing sites or regions.

298 SPEI can be calculated for different moments of the year and using different monthly
299 time scales. Negative SPEI values indicate dry conditions, while positive values indicate
300 wet conditions. SPEI takes values below -0.83 and -1.28 (above 0.83 and 1.28) for
301 moderately and severely dry (wet) conditions with 5- and 10-year return periods,
302 respectively. These threshold values were established by detailed analysis of the
303 distribution of extreme events using long-term (up to 117 years) series of SPEI values for
304 the Iberian Peninsula (González-Hidalgo et al. 2018; Liberato et al., 2021).
305 We obtained 2000-2019 SPEI data for this study from the Historical Database of
306 Meteorological Drought Indices for Spain (<https://monitordesequia.csic.es/>). This dataset
307 offers gridded (1.1 km spatial resolution) SPEI data available in NetCDF format for the
308 full territory of peninsular Spain with weekly time resolution since 1961 and up to 48-
309 month timescale, developed using climate data of up to 2269 reference stations of the
310 Spanish Meteorological Agency (Vicente-Serrano et al., 2017).
311 In order to select an optimal SPEI time and scale for analysis, we applied multilevel
312 Pearson correlations (Hox et al., 2018) between the iEVI values of the sites and a variety
313 of SPEI metrics obtained for different months of the phenological cycle (from March to
314 February) and a variety of time aggregation scales (from 1 to 20 months) through 2000-
315 2019. Site ID was set as random-effects grouping variable in the explored multilevel
316 correlations. We selected July SPEI values with a 10-month timescale to characterize the
317 hydric conditions of the years and further analyses, as this metric maximized the
318 correlation between iEVI and SPEI for our set of 5274 *Q. ilex* sites (Multilevel Pearson's
319 R=0.47, p<0.000; Fig. C2 in Appendix C). Similarly, Peña-Gallardo et al. (2018), using
320 tree ring data in Iberian holm oak woodlands found a high sensitivity of *Q. ilex* tree
321 growth to July SPEI values with 8-12-month timescale.

322 To characterize the hydric conditions during drought events and their associated pre-
323 drought and post-drought conditions, we applied for each study site and drought year the
324 (July, 10-month scale) SPEI value of the drought event (SPEI_{Dr}) and the corresponding
325 SPEI values of the year immediately before and after the drought event ($\text{SPEI}_{\text{PreDr}}$ and
326 $\text{SPEI}_{\text{PostDr}}$, respectively).

327

328 **2.6. Data analysis**

329 **2.6.1. Identification and characterization of extreme drought events**

330 Extreme drought events were identified as those drought years within the analyzed series
331 with mean SPEI values for the full region below -1.28 that produced spatially extensive
332 (at least 40% of the sites), significant reductions in the ecosystem productivity of the
333 studied holm oak woodlands. For each site and year, a significant reduction of ecosystem
334 productivity was defined where iEVI was lower than mean minus one standard deviation
335 iEVI of the previous 3-year period. This drought year selection procedure follows the
336 philosophy of the pointer-year approach (Schweingruber et al., 1990). The applied spatial
337 threshold of 40% site drought affection is in the range of common pointer-year threshold
338 values used for analysis at both the tree and site levels (for example, Gazol et al., 2017;
339 Sánchez-Salguero et al., 2018).

340 For every site, we calculated the resistance R_t , recovery R_c and resilience R_s metrics of
341 each identified extreme drought event using iEVI as ecosystem production proxy. Cross-
342 correlation of the calculated R_t , R_c and R_s metrics was analyzed using multilevel Pearson
343 correlation, applying drought event as random-effects grouping variable. Differences
344 between extreme drought events in the measured (R_t , R_c and R_s) responses of the studied
345 sites were further characterized and tested using paired t tests.

346

347 **2.6.2. Controlling factors of holm oak woodland resilience to drought**

348 We used linear mixed-effects models (LMM) to study the controlling factors of the Rt,
349 Rc and Rs holm oak woodland resilience responses to extreme drought. Separate LMMs
350 were built for each of the three components of drought resilience, using aridity (Ar) and
351 forest structure (described using the Local Deforestation index, Di) as environmental
352 fixed-effects variables. We also included the interaction between aridity and forest
353 structure (Ar:Di), since the effects of forest structure may change as competition for the
354 use of scarce water resources intensifies with aridity. Pre-drought SPEI ($SPEI_{PreDr}$) and
355 drought event SPEI ($SPEI_{Dr}$) were included in the LMM structure for Rt analysis, while
356 both Rc and Rs models included the pre-drought, drought and post-drought SPEI metrics
357 ($SPEI_{PreDr}$, $SPEI_{Dr}$ and $SPEI_{PostDr}$, respectively) as additional fixed-effects environmental
358 variables. Last, site ID was included in all models as random-effects factor to control for
359 the effects of repeated measures between the extreme drought events included in the
360 analysis. We optimized the model structure comparing model configurations of increased
361 complexity (i.e., number of fixed effects), using the Akaike's Information Criterion (AIC;
362 Akaike, 1974), which provides a trade-off between model complexity and goodness of fit
363 (details in Tables C1, C2 and C3 in Appendix C).

364 Analysis of residual values of the selected optimal models revealed the presence of
365 significant spatial autocorrelation (SAC) in the data. We applied the Residuals
366 Autocovariate approach (Crase et al., 2012; Bardos et al., 2015) to control for the effects
367 of SAC in the optimal Rt, Rc and Rs models. This approach requires the calculation of a
368 spatial autocovariate from the model residuals (RAC, for residual autocovariate), which
369 is later included in the model structure with the environmental explanatory variables and
370 interaction effects, as an additional fixed effect variable. For the selected configurations
371 of the Rt, Rc and Rs models, we applied focal calculation of RAC values (Crase et al.,

372 2012) within a 500-m neighborhood. We selected a 500-m distance for RAC calculation
373 since this neighborhood length resulted in the best model performance in terms of SAC
374 reduction. RAC inclusion in the optimal Rt, Rc and Rs models successfully removed
375 SAC, as judged by the analysis of spatial cross-correlograms (Bjornstad and Falck, 2001)
376 in the final model residuals. For the final, SAC corrected models, we determined the
377 standardized coefficients (Std. β) and significance of all (environmental and RAC) fixed
378 effects. We also determined for the final models the marginal (fixed-effects only) and
379 conditional (fixed plus random effects) R^2 (Nakagawa and Schielzeth, 2013).

380 Data processing and analysis was carried out using R 4.1.1 (R Core Team, 2021).

381

382 **3. Results**

383 **3.1. Extreme drought events: the 2005 and 2012 severely dry episodes**

384 Different dry and wet episodes were identified along the study period, as assessed by the
385 (July, 10-month scale) SPEI values of the study sites (Fig. 4a). Years 2005, 2012 and
386 2017 can be considered dry (mean SPEI of the sites is lower than -0.83), while years 2004,
387 2010 and 2013 can be considered wet (mean SPEI of the sites is higher than 0.83). Two
388 severely dry episodes (mean SPEI lower than -1.28) were identified for years 2005 (SPEI
389 -1.99 ± 0.21 SD) and 2012 (SPEI -1.57 ± 0.23 SD). The 2005 drought event was preceded
390 by a year with moderately wet conditions (SPEI 0.88 ± 0.25 SD) and succeeded by a
391 normal year (SPEI -0.53 ± 0.38 SD). Differently, the 2012 drought was preceded and
392 succeeded by a normal year (SPEI 0.20 ± 0.36 SD) and an exceptionally wet year (SPEI
393 1.27 ± 0.26 SD), respectively.

394 The 2005 and 2012 severely dry episodes extensively affected holm oak woodlands in
395 the territory (Fig. 4 b-c), producing significant iEVI declines (as compared to the
396 corresponding mean minus one SD iEVI of their precedent, reference 3-year periods) for

397 90% and 51% of the sites, respectively. The 2005 extreme drought affected
398 homogeneously the holm oak woodlands of the territory (Fig. C3a in Appendix C), while
399 the effects of the 2012 extreme drought were distributed more heterogeneously in space.
400 In fact, inland holm oak woodlands located in western areas of the territory, particularly
401 those of sub-humid and dry-transition climate conditions (Ar 0.25-0.50) lying on the
402 mountain range, were less affected by the 2012 drought than *Q. ilex* woodlands distributed
403 in the central and eastern regions of the territory (Fig. C3b in Appendix C).
404 The 2005 drought impacted woodland productivity, in terms of iEVI drop, more
405 intensively than the 2012 drought, as evidenced by the drought resistance metric (Rt, Fig.
406 5a). Conversely, post-drought iEVI increase was significantly stronger for the 2005
407 drought (Rc, Fig. 5b). However, the studied holm oak woodlands recovered pre-drought
408 iEVI levels following the 2012 drought significantly better than following the 2005
409 extreme event, as indicated by the drought resilience metric (Rs, Fig. 5c). In fact, the 2012
410 event showed a very good post-drought response, with 15% of the sites exceeding
411 significantly their pre-drought (mean plus one standard deviation) iEVI levels after
412 drought. Overall, Rt showed a strong negative correlation with Rc (Multilevel Pearson's
413 R=-0.73; p<0.001). Less strong, positive correlations were found between Rs and both Rt
414 and Rc (0.42 and 0.31, respectively; p<0.001).

415

416 **3.2. Holm oak woodland responses to extreme drought: controlling factors**

417 **3.2.1. Rt: controlling factors of woodland resistance to drought**

418 The best model structure for the analysis of the controlling factors of woodland resistance
419 to drought included all the (fixed-term) explanatory variables (Ar, Di, SPEI_{PreDr}, SPEI_{Dr})
420 and interactions (Ar:Di) originally included in the full model (Table C1 in Appendix C).
421 All these effects remained statistically significant at p<0.001 after correcting the optimal

422 model to control for the effects of spatial autocorrelation (Table 1). Aridity (Ar) affected
423 negatively drought resistance. In other words, woodland resistance to the effects of
424 extreme droughts (i.e., the capacity to preserve antecedent, pre-drought primary
425 production levels during drought) decreased along the studied aridity gradient (Table 1;
426 Fig. 6a). In addition, the interaction between Ar and Dl positively affected Rt, so that the
427 effect of forest structure (assessed by the local Deforestation index, Dl) was dependent
428 on the aridity level (Table 1). In fact, although drought resistance was higher for holm
429 oak woodlands with well-preserved forest cover (i.e., closed forest structure with canopy
430 cover near the maximum climate potential of the sites) under sub-humid and dry-
431 transition conditions, this tendency was reversed for woodland sites with semi-arid
432 climate conditions, where Rt decreased with forest density. A boundary was found
433 between these two drought resistance trends at the limit between dry-transition and semi-
434 arid conditions (Ar approx. 0.5, Fig. 6a)

435 Pre-drought SPEI and drought SPEI also affected Rt (Table 1; Fig. 6b). In fact, woodland
436 resistance was negatively affected by the hydric conditions of the antecedent, pre-drought
437 year (i.e., the more positive the $\text{SPEI}_{\text{PreDr}}$ values, the lower the Rt of the sites) and
438 decreased with the intensity of the drought events (i.e., the more negative the SPEI_{Dr}
439 values, the lower the Rt of the sites).

440

441 **3.2.2. Rc: controlling factors of woodland recovery to drought**

442 The best model structure for the analysis of the controlling factors of woodland recovery
443 to the effects of extreme droughts included aridity, forest structure and their interactions
444 (Ar, Dl, and Ar:Dl, respectively) as well as the pre-drought and drought SPEI metrics
445 ($\text{SPEI}_{\text{PreDr}}$, SPEI_{Dr} , respectively) as fixed term explanatory variables (Table C2 in
446 Appendix C). All these environmental effects on Rc, which remained significant at

447 p<0.001 after correcting the model for the effects of spatial autocorrelation (Table 1),
448 showed opposite trends to those described for Rt. In fact, woodland recovery to drought
449 (i.e., the increase of primary production after drought) increased along the aridity gradient
450 from sub-humid to semi-arid conditions (Table 1; Fig. 7a). In addition, drought recovery
451 of the studied holm oak sites under sub-humid and dry-transition conditions was stronger
452 for heavily deforested, open woodlands than for closed woodlands with well-preserved
453 forest structure (Table 1; Fig. 7a). These effects of forest structure on Rc were less intense
454 under dry-transition conditions and became neutral for semi-arid holm oak sites.

455 Pre-drought SPEI affected positively drought recovery, while drought year SPEI affected
456 negatively Rc (Table 1; Fig. 7b). In other words, the better the hydric conditions of the
457 pre-drought year (i.e., the more positive the SPEI_{PreDr} values) and the more intense the
458 drought event (i.e., the more negative the SPEI_{Dr} values), the higher is the post-drought
459 increase of woodland primary production (i.e., the higher the Rc of the sites).

460

461 **3.2.3. Rs: controlling factors of woodland resilience to drought**

462 The optimal model structure for the analysis of the controlling factors of woodland
463 resilience to drought included aridity (Ar), drought year SPEI (SPEI_{Dr}) and post-drought
464 SPEI (SPEI_{PostDr}) as explanatory variables (Table C3 in Appendix C), which remained
465 significant at p<0.001 after correcting the model for the effects of spatial autocorrelation
466 (Table 1). Ar negatively affected Rs, reducing the ability of the studied holm oak
467 woodlands to fully recover their pre-drought primary production levels from sub-humid
468 to dry-transition and semi-arid conditions (Table 1; Fig. 7a). Both, drought year SPEI and
469 pre-drought SPEI affected positively Rs (Table 1; Fig. 8b). In fact, full recovery of the
470 pre-drought levels of productivity in the studied woodlands was hindered by drought
471 intensity (i.e., the more negative the SPEI_{Dr} values, the lower the Rs of the sites), but was

472 boosted by the hydric conditions of the subsequent post-drought year (i.e., the more
473 positive the SPEI_{PostDr} values, the higher the Rs of the sites).

474

475 **4. Discussion**

476 Two extreme droughts with spatially wide impacts on the productivity of the studied holm
477 oak woodlands were identified along 2000-2019: the 2005 and 2012 events. Vicente-
478 Serrano et al. (2014) in an extensive exploration of drought severity in southern Europe
479 identified the 2005 dry event as the most intense and spatially extensive drought that has
480 affected the Iberian Peninsula between the decades of the 1960s and the 2010s.
481 Accordingly, our results indicate that the extreme drought of 2005 affected almost entirely
482 all *Q. ilex* woodlands analyzed within the 100,000 km² study region. Numerous studies
483 have reported large impacts in tree growth and forest productivity caused by this extreme
484 drought event, including canopy decay and tree mortality, particularly for pines and cold-
485 temperate broadleaf species (Sánchez-Salguero et al., 2018; Serra-Maluquer et al., 2018;
486 Gazol et al., 2018). Although the 2012 extreme drought was less intense than the 2005
487 drought, it also affected a high percentage of holm oak woodlands in the study region.
488 Similarly, previous studies have reported significant effects of the 2012 drought for
489 different *Pinus* and *Quercus* species growing in forests and woodlands within the Iberian
490 Range (Camarero et al., 2016; Forner et al., 2018; García-Barreda et al., 2023).

491

492 **4.1. Controlling effects of *Q. ilex* woodland responses to extreme drought**

493 Determining how Mediterranean forest ecosystems resist, recover and regain their
494 functionality during and after drought is becoming a major ecological issue for the
495 management of these ecosystems in the present context of climate change and land
496 abandonment. Overall, our results indicate that climate aridity and forest density largely

497 controlled the resistance and recovery of the analyzed *Q. ilex* woodlands to the effects of
498 the 2005 and 2012 extreme drought events, while the ability to regain their primary
499 production levels after drought was further controlled by aridity.

500 Drought resistance, which describes the capacity of holm oak woodlands to maintain
501 primary production during drought, decreased with climate aridity, which is in
502 accordance with results from other forest species in Spain (Camarero et al., 2013; Gazol
503 et al., 2017; Sánchez-Salguero et al., 2018). During drought, the maintenance of tree
504 growth and ecosystem primary production is highly dependent on the presence of
505 transient reserves of water resources in deep soil horizons, particularly for Mediterranean
506 woodlands dominated by deep-rooted oak species such as *Q. ilex* (Baldocchi and Xu,
507 2007; Cubera and Moreno, 2007; Moreno-de-las-Heras et al., 2018). While the presence
508 of these deep reserves of soil water resources is feasible in sub-humid landscapes and, to
509 a major extend, in dry-transition areas, their existence is unlikely under drier, semi-arid
510 conditions, which can explain the impact of aridity on drought resistance in our study
511 region. Further, the denser the holm oak woodlands, the better their production levels
512 were maintained during drought under sub-humid and dry-transition climate, while the
513 opposite was true under semi-arid conditions. These contrasting effects of forest structure
514 can be explained because deep-rooted holm oak trees in densely developed sub-humid
515 and dry-transition stands can facilitate maintaining the pre-disturbance ecosystem
516 primary production levels during drought by accessing deep soil water resources
517 (Moreno-de-las-Heras et al., 2018). Conversely, in semi-arid sites where these soil
518 reserves of water are unlikely, a high stand density can only contribute to strongly
519 increase tree competition for scarce water resources, thus reducing resistance to drought
520 (Astigarraga et al., 2020).

521 Differently to resistance, drought recovery increased in our holm oak woodlands with
522 climate aridity. This may reflect the strong dependence of productivity in semi-arid
523 ecosystems on the annual variations of water availability, where the lack of between-year
524 reserves of soil water resources typically causes a strong coupling and reactive response
525 of primary production to precipitation variations (Huxman et al., 2004; Poulter et al.,
526 2014; Moreno-de-las-Heras et al., 2018; Vicente-Serrano et al., 2019). Alternatively,
527 phenotypic plasticity and possible local adaptation of semi-arid *Q. ilex* populations (e.g.,
528 smaller baseline crowns, greater ability of physiological recovery) may also explain, at
529 least in part, the positive impact of climate aridity on drought recovery. Interestingly, low
530 tree density increased drought recovery in the studied holm oak woodlands under sub-
531 humid and dry-transition climate conditions, while these effects of forest structure tended
532 to disappear under semi-arid conditions. This may be due to rapid, intense growth after
533 drought of dense grass vegetation in open *Q. ilex* woodlands under sub-humid and dry-
534 transition conditions. In fact, perennial grass vegetation, which typically exhibits quick
535 and strong growth responses to precipitation pulses (Garcia et al., 2010; Moreno-de-las-
536 Heras et al., 2015), is an essential component of the open holm oak woodlands under the
537 less dry conditions of the explored climate aridity gradient (Bochet et al., 2021).
538 Contrastingly, tree interspaces in open *Q. ilex* woodlands under semi-arid conditions of
539 the study region are scarcely covered by small woody shrub species (e.g., *Thymus*
540 *vulgaris*, *Helianthemum violaceum*, *Helianthemum marifolium*, *Genista scorpius*) that
541 have a slow growth response compared to grass vegetation and a rather low contribution
542 to landscape-level net primary production. In this sense, it is important to acknowledge
543 that the net primary production (NPP) proxy applied in this work (MODIS iEVI) not only
544 captures the contribution of the tree layer of vegetation but also the NPP contributions of
545 the underlying shrub and grass components of the studied woodlands.

546 In spite of the detected strong post-drought recovery of production of semi-arid *Q. ilex*
547 woodlands, these dry ecosystems showed a worse capacity to regain their pre-drought
548 production levels compared to sub-humid and dry-transition *Q. ilex* woodlands. DeSoto
549 et al. (2020) suggested that the higher resilience of drought-affected forests and trees in
550 wetter sites is caused by their higher capacity to resist the initial drought impact and keep
551 their pre-disturbance tree status. In this way, the availability of soil water reserves that
552 largely facilitate the maintenance of ecosystem production during drought in wet sites
553 (Serra et al., 2017; Moreno-de-las-Heras et al., 2018; Preisler et al., 2019) can also play
554 an important role for improving drought resilience under sub-humid and dry-transition
555 climate conditions. In turn, the lack of these soil water reserves in dry sites may explain
556 both the lower resistance and resilience of the analyzed semi-arid woodlands.

557 We acknowledge that, differently to the observed responses of drought resistance and
558 recovery, resilience was not affected by forest structure in the analyzed *Q. ilex*
559 ecosystems. Consistently, Castagneri et al. (2022) in a recent metanalysis over 166
560 published studies on the influence of stand-level competition on tree growth responses to
561 drought concluded that tree basal area significantly reduces resistance and increases
562 recovery, but does not affect resilience. For the analyzed semi-arid woodlands, the
563 observed lack of influence of forest density on drought resilience may be due to
564 competition-release compensation effects. In fact, Serra-Maluquer et al. (2018) argued
565 that tree decay and mortality during drought can result in competition release in dense
566 forest stands, so that improved water availability after drought can support relatively high
567 growth rates, potentially compensating for the initial negative effects of forest density on
568 drought resistance. These competition-release compensation effects may also be
569 supported by field observations in densely developed holm oak stands of dry areas
570 distributed in NE Spain (Hoya de Huesca region) reporting severe canopy defoliation and

571 significant stem mortality during the 2005 and 2012 drought events followed by higher
572 than expected post-drought growth (García-Barreda et al., 2023). Differently, for wetter
573 areas distributed in our study region, where higher availability of soil water resources
574 reduce competition and increase resistance to drought, strong post-drought growth in
575 open *Q. ilex* woodlands of both tree and between-tree perennial grass vegetation may
576 compensate the benefits that larger densities of deep-rooted holm oak trees accessing deep
577 soil water resources may have for maintaining landscape-level primary production during
578 drought.

579 Besides the discussed effects of climate aridity and forest structure, drought intensity and
580 both pre- and post-drought hydric conditions also conditioned the analyzed drought
581 responses, explaining to a large extent the observed differences in resistance, recovery
582 and resilience of the sites for the 2005 and 2012 drought events. For example, the 2005
583 event was characterized by higher drought intensity than the 2012 drought, thus resulting
584 in larger woodland productivity reductions. These negative effects of drought intensity
585 on resistance, frequently observed in other forest ecosystems (Gazol et al., 2017;
586 Greenwood et al., 2017; Anderegg et al., 2020), may have also be reinforced for the
587 analyzed *Q. ilex* woodlands by the wet pre-drought hydric conditions that took place
588 immediately before the 2005 drought. Zhang et al. (2021) explain this reinforced effect
589 as a function of structural overshoot, where surplus production caused by wet conditions
590 before drought can largely exceed the biomass that can be later maintained during
591 drought.

592 Contrary to drought resistance, the strength of the post-drought recovery of production in
593 the explored holm oak woodlands was increased by both drought intensity and pre-
594 drought hydric conditions, which overall reflects the common interdependence between
595 drought resistance and post-drought recovery (Lloret et al., 2011; Gazol et al., 2018;

596 Manrique-Alba et al., 2022). More importantly, the ability of the studied woodlands to
597 fully regain their pre-drought production levels was negatively affected by drought
598 intensity and further favored by wet post-drought hydric conditions. These effects can
599 explain the very positive resilience observed for the 2012 drought, where 15% of the sites
600 not only recovered, but significantly improved their pre-drought primary production
601 levels after drought. In fact, while the lower drought intensity of the 2012 event may have
602 limited tree damage, providing better conditions for tree post-drought recovery than for
603 the 2005 drought, the exceptionally wet post-drought hydric conditions that took place in
604 2013 very likely promoted strong vegetation growth after the 2012 drought. Accordingly,
605 Forner et al. (2018), in a dendrometer-based localized study developed within our study
606 region, also highlighted the advantages of the exceptionally wet conditions that took place
607 immediately after the 2012 dry event for improving *Q. ilex* drought resilience.

608

609 **4.2. Implications for the stability and management of *Q. ilex* woodlands**

610 Overall, our results highlight the feasibility and potential of our broad-scale remote
611 sensing approach for monitoring the health and drought vulnerability of forest
612 ecosystems, which can largely facilitate their management in large areas with low
613 availability of field data. We observed a strong control of local climate on drought
614 vulnerability, characterized by contrasted responses to extreme drought of *Q. ilex*
615 woodlands along the explored climate aridity gradient. On one hand, the semi-arid *Q. ilex*
616 woodlands of the region show a low resistance to drought and a poor capacity to further
617 regain their pre-drought ecosystem production levels. On the other hand, *Q. ilex*
618 woodlands in sub-humid and dry-transition areas of the region, where wetter climate
619 conditions may alleviate water stress and competition, show higher resistance and
620 resilience to the effects of drought. Interestingly, several studies point to active holm oak

621 expansion over sub-humid areas of the present distribution range, where *Q. ilex* positive
622 performance during more frequent and intense dry periods is favoring large compositional
623 shifts displacing less drought tolerant species (Peñuelas and Boada, 2003; Navarro-
624 Cerrillo et al., 2019).

625 Holm oaks in semi-arid areas operate at the limits of hydraulic safety, which largely
626 explains the emergence of more recurrent *Q. ilex* canopy dieback and tree mortality
627 episodes in the present context of increased aridity and drought frequency (Corcuera et
628 al., 2004; Camarero and Valeriano, 2023). In fact, the observed drought sensitivity that
629 characterizes semi-arid *Q. ilex* woodlands in this study region concurs with previous
630 results obtained from localized field observations, dendrochronological analyses and both
631 experimental and modelling studies. For example, Camarero et al. (2016), in a semi-arid
632 *Q. ilex* forest stand located in the Mijares river basin (east of our study region; MAP ~400
633 mm), observed very intense reductions of basal area increments accompanied by severe
634 defoliation and stem mortality caused by the 2005 and 2012 drought events. Furthermore,
635 Ogaya and Peñuelas (2021) in a 21-year long rainfall exclusion experiment performed in
636 a less dry holm oak forest of the Prades Mountains (NE Spain; MAP ~600 mm) observed
637 that exclusion of 30% annual precipitation caused large stem growth reductions and
638 increases of tree mortality during dry and hot years of the series, also promoting a
639 progressive replacement of *Q. ilex* by more drought resistant shrub species (e.g., *Phillyrea*
640 *latifolia* L). Overall, these results are in line with predictions obtained by forest species
641 distribution models that foresee future mismatches, as the climate becomes more arid,
642 between the *Q. ilex* ecological niche and the local climate conditions of semi-arid areas
643 within its present distribution range (Benito-Garzón et al., 2008; Mauri et al., 2022).

644 Our results indicate that the effects of extreme droughts on primary production can be
645 particularly negative for dense forest stands of semi-arid holm oak woodlands, where

646 competition for the use of scarce soil water resources may be particularly intense. These
647 more vulnerable dry forests are commonly structured in the study region as dense coppice
648 *Q. ilex* stands formed by multiple vegetative stems sprouting from the root network of
649 aged holm oak individuals that were affected in the past by repeated logging. Their aged
650 root architecture and imbalanced root/shoot ratios, which magnify water competition
651 within multiple stems of the same individuals (Camarero et al., 2016), predispose these
652 coppice semi-arid woodlands to drought-induced tree decay and mortality (Gentilesca et
653 al., 2017; Serrada et al., 2017). Tree density reduction through selective thinning has been
654 proposed as a key adaptive strategy for improving drought responses of Mediterranean
655 forests (Molina et al., 2021). The benefits of thinning have been recently tested in
656 overstocked Mediterranean pine plantations and densely developed holm oak stands,
657 obtaining significant tree growth increases and mortality reductions (Ogaya et al., 2020;
658 Molina et al., 2021; Manrique et al., 2022). Selective forest thinning may play a
659 significant role as adaptive management strategy for improving drought responses of the
660 analyzed semi-arid holm oak forests. Conversely, the favorable responses to drought
661 observed in sub-humid and dry transition areas of the study region, where holm oak
662 woodlands with closed forest structure showed in general a high ability to maintain
663 ecosystem primary production during drought, suggest that management actions in wet
664 areas of the *Q. ilex* climate distribution range may be more appropriately directed to
665 facilitate natural rewilding, including holm oak forest expansion and stand densification.

666

667 **5. Conclusions**

668 This study applied remote sensing analysis of the responses of landscape-level ecosystem
669 production of Iberian *Q. ilex* woodlands to recent extreme droughts during 2000-2019,
670 including the spatially extensive 2005 and 2012 severely dry events, along broad-scale

671 gradients of climate aridity and forest structure in eastern Spain. Drought resistance and
672 post-drought recovery were controlled by climate aridity and forest structure, which
673 largely regulated resource availability and competition for the use of water. The ability to
674 regain the pre-drought production levels after drought was further negatively affected by
675 climate aridity in these holm oak woodlands. All these responses were also affected by
676 drought intensity and both pre- and post-drought hydric conditions, which controlled the
677 variations of the analyzed *Q. ilex* woodland responses between the 2005 and 2012
678 droughts.

679 Semi-arid *Q. ilex* woodlands, located in dry areas of the explored climate aridity gradient,
680 showed a high sensitivity to extreme drought. Their resistance, assessed as the capacity
681 of the woodlands to maintain primary production during drought, was low. Although
682 these dry ecosystems showed a strong post-drought recovery of production, they also
683 showed a poor resilience, characterized by a low capacity to fully regain their pre-drought
684 production levels. Contrarily, holm oak woodlands in sub-humid and dry transition areas
685 of the study region, where wetter climate conditions may alleviate water stress and
686 competition during dry periods, showed a high resistance and resilience to the effects of
687 droughts. Drought vulnerability was particularly high for dense holm oak stands
688 developed under semi-arid climate conditions, where strong competition for scarce water
689 resources largely reduced the ability of the woodlands to maintain landscape-level
690 ecosystem production during drought.

691 The observed drought vulnerability of *Q. ilex* woodlands under semi-arid climate
692 conditions may affect the long-term stability of these dry ecosystems, particularly in the
693 present context of rapid increases of climate aridity and severity of droughts that is taking
694 place in the Mediterranean region. Selective forest thinning may help, as an adaptive
695 management tool, to reduce drought vulnerability of semi-arid *Q. ilex* woodlands in dry

696 areas of the territory, particularly for those more densely developed forest stands that
697 show a low resistance to the effects of extreme drought. Conversely, natural rewinding
698 may more appropriately guide management actions for more humid areas of the *Q. ilex*
699 distribution range, where densely developed holm oak woodlands show in general a high
700 ability to maintain ecosystem primary production during drought.

701

702 **Acknowledgements**

703 We would like to thank the NASA's LP DAAC service and ESA's Copernicus program
704 for providing the MODIS and Sentinel-2 data used in this study. We also thank Luis
705 Cayuela Delgado for statistical advice and two anonymous referees for their thoughtful
706 comments. This work was supported by the caRRascal project (grant RTI2018-095037-
707 B-I00), funded by MCIN/AEI/10.13039/501100011033 and "ERDF A way of making
708 Europe". Mariano Moreno-de-las-Heras is beneficiary of a Serra Hunter fellowship on
709 Physical and Environmental Geography, funded by Generalitat de Catalunya (UB-LE-
710 9055), and grant 2021SGR00859, awarded by the Agència de Gestió d'Ajuts Universitaris
711 i de Recerca de la Generalitat de Catalunya (SGR2021-2024).

712

713 **Author contributions**

714 Mariano Moreno-de-las-Heras performed conceptualization, methodology, formal
715 analysis and writing – original draft. Esther Bochet and Patricio García-Fayos performed
716 funding acquisition, project coordination, conceptualization and writing – reviewing &
717 edition. Sergio M. Vicente-Serrano, Tiscar Espigares, Maria J. Molina, Vicente Monleón,
718 José M. Nicolau, and Jaume Tormo performed conceptualization and writing – reviewing
719 & edition.

720

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Table 1. Linear mixed model results (model fit and fixed effects) for the drought resistance (R_t), recovery (R_c) and resilience (R_s) metrics. All three final models are corrected to control for the effects of spatial autocorrelation using the Residual Autocovariate (Crase et al., 2012; Bardos et al., 2015) approach.

Resistance (R_t)					
	St. β	S.E.	d.f.	t	P
Marginal R^2	0.83				
Conditional R^2	0.85				
<i>Environmental variables</i>					
Ar	-0.410	0.012	5270	-35.6	<0.001
Dl	-0.501	0.032	5270	-15.8	<0.001
Ar:Dl	0.465	0.034	5270	13.6	<0.001
SPEI _{PreDr}	-0.282	0.005	5271	-64.2	<0.001
SPEI _{Dr}	0.397	0.004	5271	85.7	<0.001
<i>Spatial autocovariate</i>					
RAC	0.681	0.004	5271	172.1	<0.001
Recovery (R_c)					
Marginal R^2	0.80				
Conditional R^2	0.81				
	St. β	S.E.	d.f.	t	P
<i>Environmental variables</i>					
Ar	0.311	0.012	5270	25.7	<0.001
Dl	0.441	0.033	5270	13.2	<0.001
Ar:Dl	-0.371	0.036	5270	-10.4	<0.001
SPEI _{PreDr}	0.278	0.005	5271	57.7	<0.001
SPEI _{Dr}	-0.261	0.005	5271	-51.9	<0.001
<i>Spatial autocovariate</i>					
RAC	0.748	0.004	5271	173.0	<0.001
Resilience (R_s)					
Marginal R^2	0.80				
Conditional R^2	0.81				
	St. β	S.E.	d.f.	t	P
<i>Environmental variables</i>					
Ar	-0.128	0.005	5272	-23.4	<0.001
SPEI _{Dr}	0.136	0.008	5271	16.3	<0.001
SPEI _{PostDr}	0.185	0.008	5271	23.7	<0.001
<i>Spatial autocovariate</i>					
RAC	0.817	0.005	5271	173.2	<0.001

Abbreviations: Ar, aridity; Dl, local deforestation index (forest structure); Ar:Dl, Ar-Dl interaction; SPEI_{PreDr}, pre-drought SPEI; SPEI_{Dr}, drought event SPEI; SPEI_{PostDr}, post-drought SPEI; RAC, residual autocovariate; St. β , standardized coefficients; S.E., standard error; d.f., degrees of freedom.

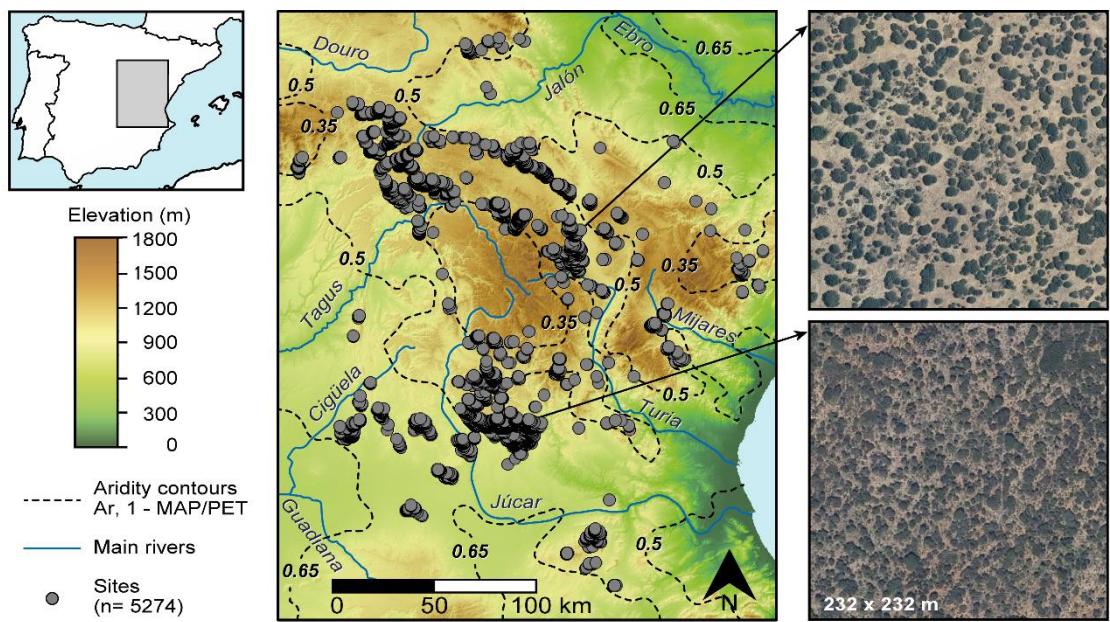


Figure 1. Location map (total 5274 sites) and aerial view (0.5 m resolution images of the Spanish National Orthophoto Program) of two *Quercus ilex* woodland sites (each 232 × 232 m size). Ar represents climate aridity (calculated as 1 - MAP/PET, where MAP and PET are mean annual precipitation and potential evapotranspiration, respectively).

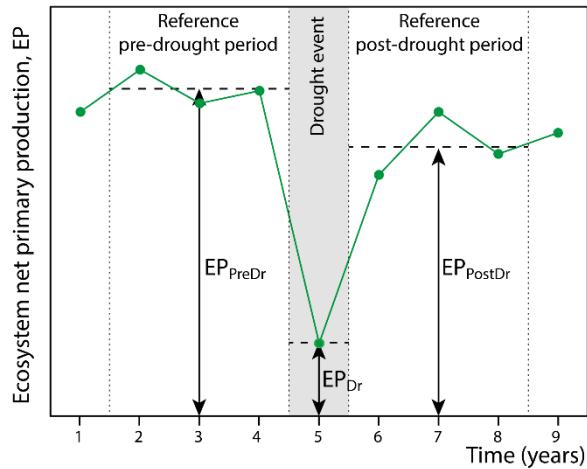


Figure 2. Schematic representation of the annual dynamics and response of ecosystem net primary production (EP) to the effects of a drought event. EP_{PreDr} , EP_{Dr} and EP_{PostDr} represents pre-drought, drought and post-drought net primary production, respectively.

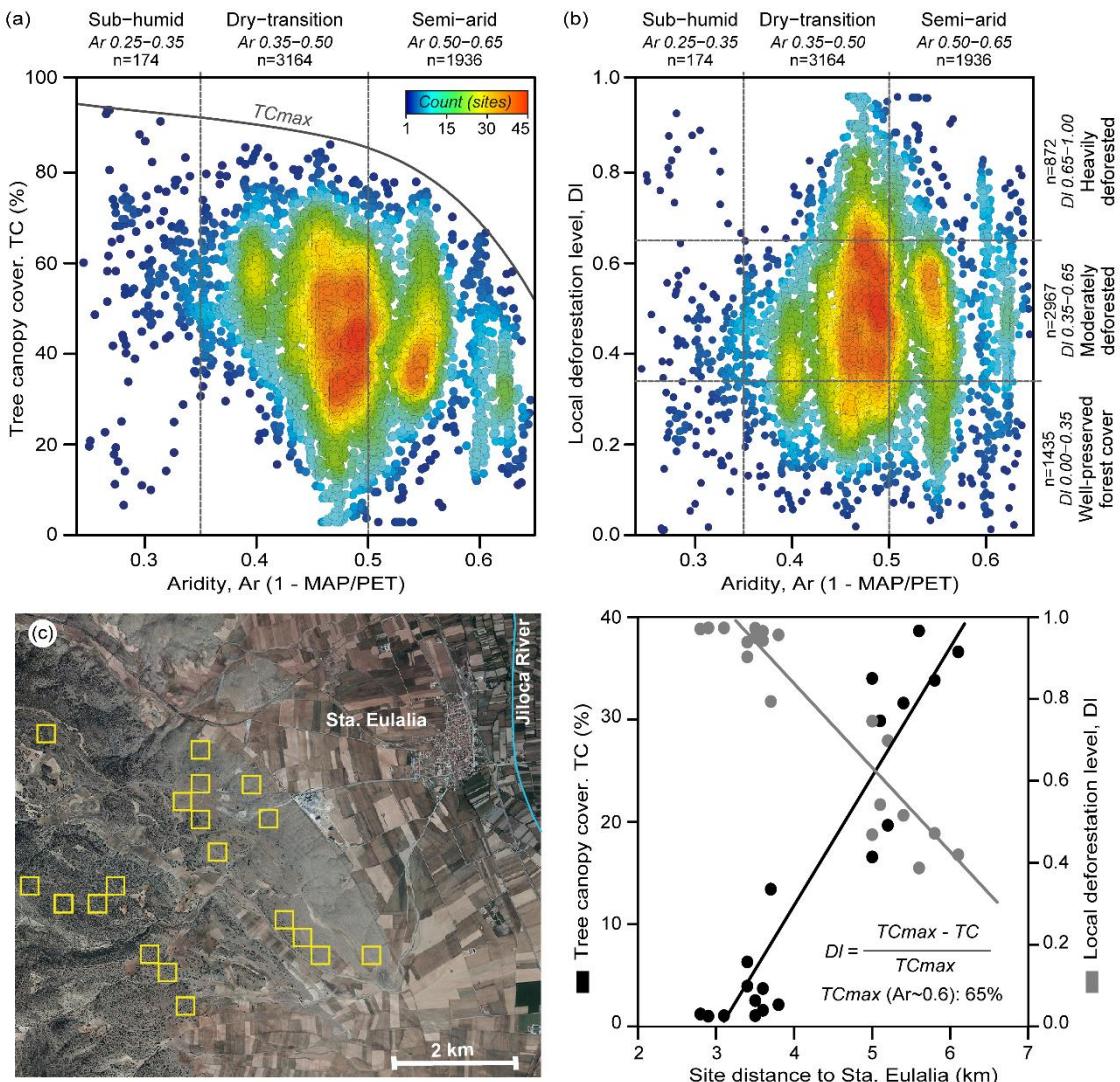


Figure 3. Forest structure and its relationship to climate aridity and the legacy of human use in the analyzed *Quercus ilex* woodlands: distribution of (a) tree canopy cover (TC, %) and (b) local deforestation level (DI) of the study sites along the explored climate aridity gradient; (c) tree canopy cover and local deforestation level variations in a set of 19 *Q. ilex* sites distributed at increasing distances from a rural settlement in the study region (Santa Eulalia, Teruel). The envelope TC_{max} (%) line in (a) represents maximum tree cover along the climate aridity range of the study holm oak sites. The aerial view (Spanish National Orthophoto Program) and TC-DI spatial trends in (c) are adapted from Moreno-de-las-Heras et al. (2018).

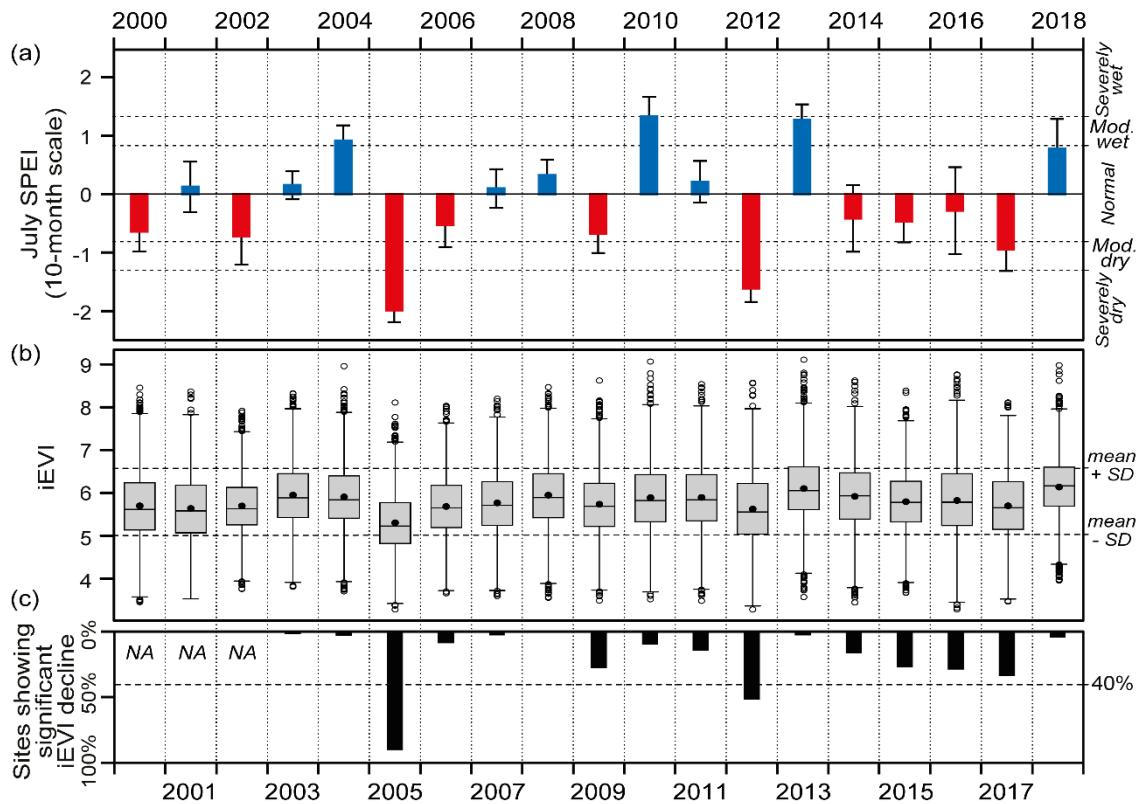


Figure 4. Drought conditions and ecosystem production along 2000-2019 for the studied *Quercus ilex* woodlands: (a) mean \pm SD hydric conditions (10-month scale July SPEI; blue and red colors indicate wet and dry conditions, respectively); (b) ecosystem production (iEVI); (c) proportion of sites showing significant drought impacts on ecosystem production (i.e., iEVI of present annual cycle is lower than mean minus one SD iEVI of the reference, three previous years). Classification of (severely, moderately) wet/dry and normal hydric conditions in (a) follows Liberato et al. (2021). NA in (c) indicates that iEVI decline was not evaluated for 2000-2002 (reference period is shorter than three years).

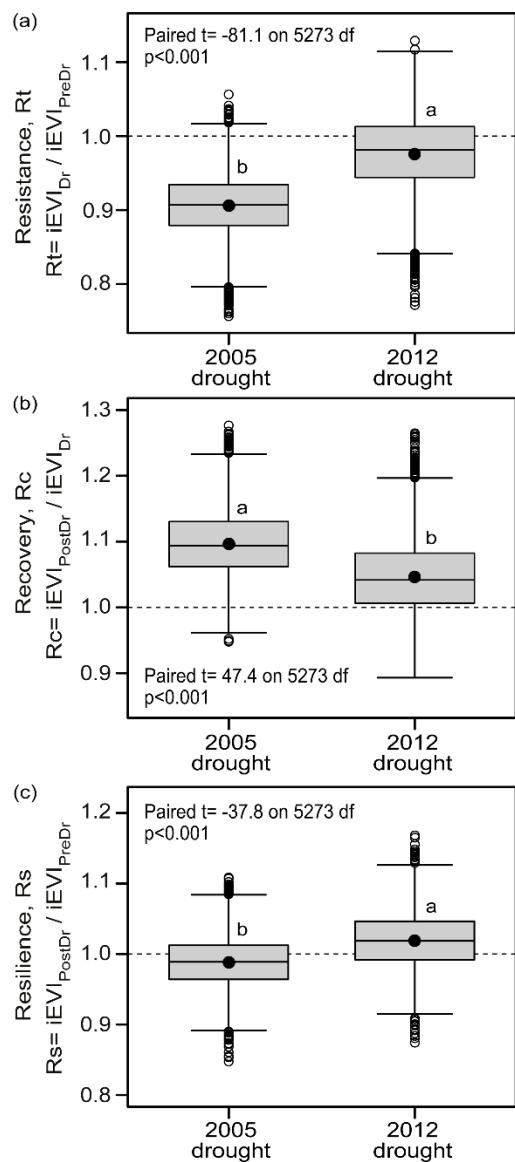


Figure 5. Responses of the study *Quercus ilex* sites to the 2005 and 2012 extreme drought episodes: between drought comparisons of (a) drought resistance (R_t), (b) recovery (R_c), and (c) resilience (R_s). Different letters indicate statistical differences between drought events (tested using paired t tests). $iEVI_{PreDr}$, $iEVI_{Dr}$ and $iEVI_{PostDr}$ in the (R_t , R_c and R_s) equations indicate pre-drought, drought and post-drought ecosystem production, respectively.

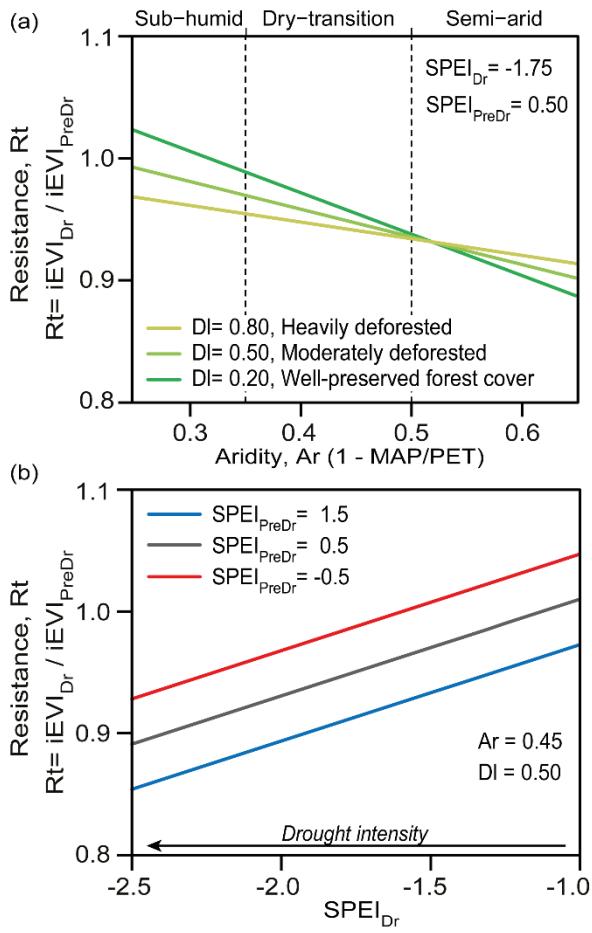


Figure 6. Controlling factors of *Quercus ilex* woodland resistance to drought (Rt): (a) linear mixed model (LMM) effects of climate aridity (Ar) and forest structure (DI, local deforestation index); (b) LMM effects of drought event SPEI ($SPEI_{Dr}$) and pre-drought SPEI ($SPEI_{PreDr}$). $iEVI_{Dr}$ and $iEVI_{PreDr}$ in the Rt equation indicate drought and pre-drought ecosystem production, respectively.

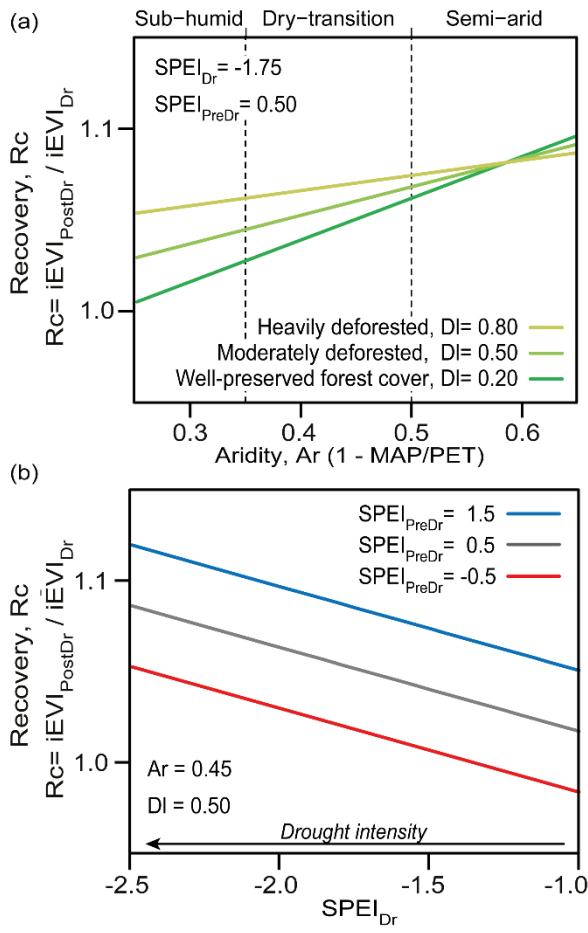


Figure 7. Controlling factors of *Quercus ilex* woodland recovery to drought (Rc): (a) linear mixed model (LMM) effects of climate aridity (Ar) and forest structure (DI, local deforestation index); (b) LMM effects of drought event SPEI ($SPEI_{Dr}$) and pre-drought SPEI ($SPEI_{PreDr}$). $iEVI_{PostDr}$ and $iEVI_{Dr}$ in the Rc equation indicate post-drought and drought ecosystem production, respectively.

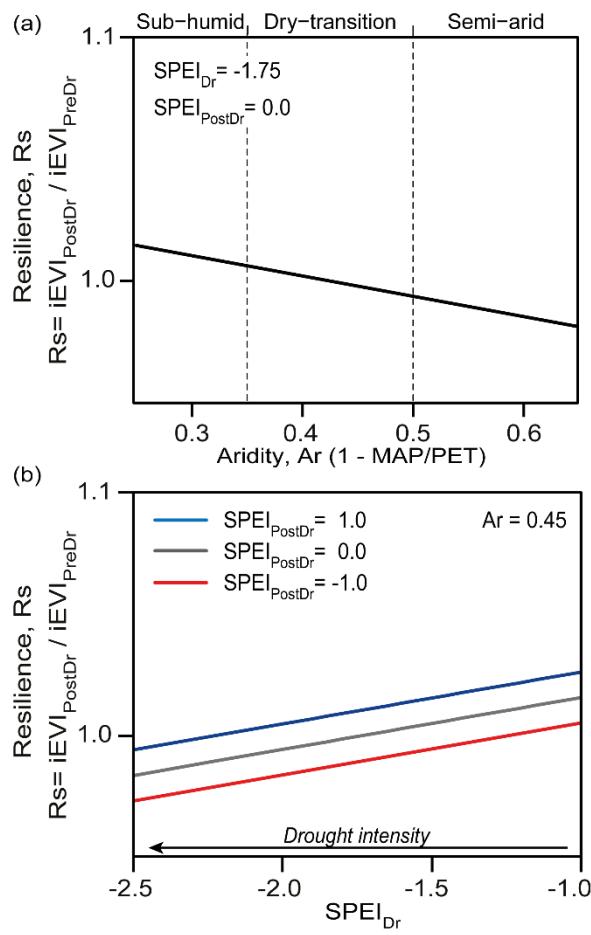


Figure 8. Controlling factors of *Quercus ilex* woodland resilience to drought (Rs): (a) linear mixed model (LMM) effects of climate aridity (Ar); (b) LMM effects of drought event SPEI ($SPEI_{Dr}$) and post-drought SPEI ($SPEI_{PostDr}$). $iEVI_{PostDr}$ and $iEVI_{PreDr}$ in the Rs equation indicate post-and pre-drought ecosystem production, respectively.