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1 **SHORT-TERM FOREST RESILIENCE AFTER DROUGHT-INDUCED DIE-OFF IN**
2 **SOUTHWESTERN EUROPEAN FORESTS**

3

4 **F. Lloret^{1,2*}, L.A. Jaime¹, J. Margalef-Marrase¹, M.A. Pérez-Navarro¹, E. Batllori^{1,3}**

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6 ¹Centre de Recerca Ecològica i Aplicacions Forestals (CREAF), 08193 Cerdanyola del Vallès,
7 Barcelona, Spain

8 ²Unitat d'Ecologia, Departament de Biologia Animal, Biologia Vegetal i Ecologia, Universitat
9 Autònoma Barcelona (UAB), 08193 Cerdanyola del Vallès Barcelona, Spain

10 ³Unitat de Botànica i Micologia, Departament de Biologia Evolutiva, Ecologia i Ciències
11 Ambientals, Universitat de Barcelona, 08028 Barcelona, Spain

12

13 *corresponding author, francisco.lloret@uab.cat, +34935812700

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17 **Abstract**

18 Drought-induced die-off in forests is becoming a widespread phenomenon across biomes,

19 but the factors determining potential shifts in taxonomic and structural characteristics

20 following mortality are largely unknown. We report on short-term patterns of resilience

21 after drought-induced episodes of tree mortality across 48 monospecific forests from

22 Morocco to Slovenia. Field surveys recorded plants growing beneath a canopy of dead,

23 defoliated and healthy trees. Site-level structural characteristics and management legacy

24 were also recorded. Resilience was assessed with reference to forest composition (self-

25 replacement), structure, and changes in the climatic suitability of the replacing community

26 relative to the climatic suitability of the dominant pre-drought species. Species climatic

27 suitability was estimated from species distribution models calculated for the baseline 1970-

28 2000 period.

29 Short-term resilience decreased under higher levels of drought-induced damage to the
30 dominant species and with evidences of management legacy. Greater resilience of structural
31 features (fewer gaps, greater canopy height) was observed overall in forests with a larger
32 basal area. Less gaps were also associated with greater woody species richness after
33 drought. Overall, Fagaceae-dominated forests exhibited greater structural resilience than
34 conifer-dominated ones. On those sites that were more climatically suited to the dominant
35 pre-drought species, replacing communities tended to exhibit lower climatic suitability than
36 pre-drought dominant species. There was a greater loss of climatic suitability under a legacy
37 of management and drought intensity, but less so in the replacing communities with higher
38 woody species richness.

39 Our study reveals that short-term forest resilience is determined by pre-drought stand
40 characteristics, often reflecting previous management legacies, and by the impact of
41 drought on both the dominant pre-drought species and post-drought replacing species in
42 terms of their climatic suitability.

43

44 **Keywords**

45 climate change, drought-induced die-off, drought events, forest mortality, forest resilience,
46 species climatic suitability.

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54 **1. Introduction**

55

56 There have been an increasing number of reports on mortality in many forest types across
57 forested biomes (van Mantgem et al. 2009; Allen et al. 2015, Neumann et al. 2017). Climate
58 changes, and particularly extreme drought episodes, have often been associated with this
59 phenomenon (Allen et al. 2010; Carnicer et al. 2011; Peng et al. 2011; Williams et al. 2013;
60 Evans and Lyons 2013; Senf et al. 2020), with drought intensity being an important
61 determinant of such mortality episodes (Greenwood et al. 2017). Other drivers, such as
62 pests, management legacy and species biogeographical context, usually interact with
63 climate to increase mortality rates above the historical baseline (Clark et al. 2016;
64 Sommerfeld et al. 2018). This leads to a high degree of local idiosyncrasy in terms of the
65 extent, duration and identity of the affected tree species (Lloret and Batllori 2021). Despite
66 the relevance of these forest mortality events to the future of forests and the services that
67 they provide (Anderegg et al. 2013), their consequences in terms of forests' compositional
68 and structural characteristics have been insufficiently investigated. For instance, vegetation-
69 type conversion following drought has been reported at multi-regional scale (Batllori et al.
70 2020), but our understanding about the mechanisms of resilience in forests affected by
71 drought-induced mortality still remains largely uncertain (Martinez-Vilalta and Lloret, 2016).

72

73 From the perspective of the resilience theory, engineering resilience (*sensu* Pimm 1984; see
74 Gunderson 2000) refers to the capacity of the ecosystem to recover its pre-disturbance
75 properties, which are considered to define its ecological state of reference. In forests
76 experiencing drought-induced die-off, their state of reference can be estimated as the pre-
77 drought levels of dominance of overstory species. Alternatively, ecological resilience (*sensu*
78 Holling 1973; see Gunderson 2000) refers to the capacity of a system to remain in a state of

79 dynamical equilibrium until it reaches a tipping point that leads to a new equilibrium state.
80 In drought-affected forests, this would correspond to shifts towards a non-forested
81 community. Long-term monitoring of drought-affected areas is needed to gain conclusive
82 data about tipping points, but the assessment of the short-term replacement patterns after
83 mortality events (Batllori et al. 2020) can provide important insights into future trajectories
84 (Johnstone et al. 2016). For instance, the absence of woody species replacing dead trees
85 following drought can be considered a preliminary indicator of potential tipping points,
86 which will be confirmed if tree species are not able to establish themselves in the long term.
87 In other words, measures of engineering resilience, such as short-term replacement
88 patterns, are also informative of long-term ecological resilience.

89

90 Canopy replacement largely determines the initial stages of forest dynamics in different
91 ways after a mortality event. Drought mortality can 'push back' successional trajectories
92 towards early stages if pioneer species are favored, or it may 'push forward' successional
93 processes if replacement by late-successional species is promoted (Batllori et al. 2020).
94 Alternatively, new successional trajectories may also appear following drought if species
95 turnover leads to novel community assemblages (Hobbs et al. 2006). Gap-dynamics
96 approaches based on tree-level observations and modelling, eventually scaled-up to stand
97 level, are suitable for capturing such processes (Ibáñez et al 2019).

98 Stand characteristics - including basal area, tree density, and species richness - determine
99 forest dynamics after gap opening to some extent (e.g., Rebertus and Veblen 1993; Gray
100 and Spies 1996) and, therefore, they are likely to explain the replacement patterns of the
101 dominant species in the canopy following drought, including the self-replacement of a tree
102 species. Furthermore, stand characteristics closely reflect management legacy - including
103 planting, clearing and cutting - and forest stage, which is linked to successional trajectories.

104 While the contribution of stand-level characteristics to forest vulnerability to drought-
105 induced die-off has been widely explored (Galiano et al. 2010; Bell et al. 2014; Crouchet et
106 al. 2019), its relevance to post-drought forest dynamics remains largely unexplored. For
107 instance, large stand basal area reflects suitable growth conditions for dominant trees,
108 often associated with late successional stages, which would endow such species with a
109 greater higher capacity to recover for such species. In turn, stand tree density likely reflects
110 the availability of trees in the understory ready to replace dead trees. Finally, greater
111 richness of woody species, would likely be associated with the presence of species with
112 different light requirements, thus limiting the continuation of the open spaces which
113 resulted from tree death.

114

115 The effects of drought episodes on tree populations are also expected to be strongly
116 determined by the suitability of affected sites' climatic conditions for different species.
117 Species Distribution Models (SDMs) provide a quantitative characterization of species'
118 requirements by correlating their occurrences in a territory with their respective
119 environmental conditions (often climatic), as obtained from georeferenced sources
120 (Franklin, 2010). These models can then be projected under given sets of climatic conditions
121 to obtain climatic suitability indexes, which synthetizes the degree to which such climatic
122 conditions are adequate for the species, according to the climatic conditions that the
123 species encounter through their entire distribution range (Soberón and Peterson 2005). A
124 key feature of species distribution models is that their standardization (through suitability
125 indices) the way that different populations and species experience the climatic
126 environment. This approach has been used to explain the impact of extreme drought
127 episodes on co-occurring species (Sapes et al. 2017, Perez-Navarro et al. 2019) and on

128 populations of dominant tree species across regions (Lloret and Kitzberger 2018, Margalef
129 et al. 2020).

130

131 In this context, we could expect resilience patterns after disturbance to be determined by
132 the climatic suitability of the dominant species on a given site. For instance, on those sites
133 that have been historically suitable for a particular species, a higher availability of its recruits
134 would be expected (Sexton et al. 2009), thus enhancing self-replacement. However,
135 recruiting populations living in historically suitable and stable conditions can be less
136 successful under the harsh conditions of a drought episode (Walck et al. 2011). The balance
137 between such processes could determine the eventual relationship between historical
138 climatic suitability and resilience patterns (Lloret and Kitzberger 2018, Margalef et al. 2020).

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141 Climatic suitability indexes also make it possible to scale up from coexisting populations of
142 different species to community-level estimates. Community-weighted means can be
143 calculated from the abundance of the species present on a site in order to estimate the
144 overall suitability of the community's climatic conditions. Higher values of community
145 climatic suitability would indicate that most species growing on that site find optimal
146 climatic conditions there. In this case, the climatic disequilibrium, defined as the difference
147 between the climate existing on that site and the climate inferred from the community
148 composition (Davis 1986; Svenning and Sandel 2013; Blonder et al. 2015), would be
149 minimal. After disturbance, differences between the climatic suitability of the previously
150 dominant tree species and the climatic suitability of the replacing species would indicate
151 trends in the emerging community's suitability with respect to the recent climate.

152 Accordingly, negative trends (i.e., lower climatic suitability of the replacing community)

153 would entail an increase in climatic disequilibrium, in relation to the historical conditions. In
154 contrast, positive suitability trends after disturbance (i.e., higher climatic suitability of the
155 replacing community) would indicate that the formerly dominant species exhibited greater
156 climatic disequilibrium than the potential replacers. Both positive and negative changes in
157 community-level suitability and, therefore, climatic disequilibrium trends ultimately reflect
158 the filtering effects of the disturbance event on community composition (Keddy 1992). For
159 instance, when a drought event filters those species living closer to their climatic tolerance
160 limits, climatic disequilibrium would decrease (Perez-Navarro et al. 2021). By contrast, if the
161 filtering is produced in species living under high climatic suitability (Cavin and Jump 2017;
162 Lloret and Kitzberger 2018), the climatic disequilibrium increases – i.e., a loss of community
163 climatic suitability occurs. Finally, equal values of climatic suitability in the replacing
164 community and the former dominant community (i.e., no change in climatic disequilibrium)
165 would imply high community resilience in terms of community climatic suitability, even if
166 both the initial and the replacing communities are not compositionally equivalent.

167

168 Here we study short-term patterns of resilience (engineering resilience) to drought-induced
169 episodes of tree mortality across 48 monospecific forests from Morocco to Slovenia,
170 dominated by 18 different tree species. Resilience estimates were based on plant-by-plant
171 replacement field data. These estimates referred to 1- taxonomic composition (mainly self-
172 replacement), 2- stand structure (gaps, canopy height) and 3- changes in the climatic
173 suitability (estimated from MaxEnt algorithms; Phillips and Dudick 2008) of the replacing
174 community. Stand structural characteristics (basal area, density, woody richness) and
175 management legacy were also recorded in the field.

176

177 We analyze short-term resilience of forest taxonomic composition and structure in relation
178 to characteristics of the drought event, the stand characteristics and the climatic suitability
179 of the dominant pre-drought species. As regards the characteristics of the drought event, we
180 hypothesize that resilience overall tends to be lower under high drought intensities (both
181 during and after the drought episode), which will have an impact on both overstory and
182 understory plants from the dominant tree species. With respect to stand characteristics, we
183 hypothesize that self-replacement tends to increase with greater basal area, as a proxy for a
184 population's good performance in a locality, and with higher density, indicating more
185 recruitment of saplings. Gaps will decrease with greater woody richness, because
186 complementarity between species will promote canopy cover, while canopy height will tend
187 to increase with basal area. Milder management (i.e., preserving regeneration) and absence
188 of pests/pathogens will favor overall resilience. As for the climatic suitability of the
189 dominant pre-drought species, we hypothesize that resilience in terms of self-replacement
190 and structural features tends to be greater on sites with a greater historical suitability of the
191 dominant pre-drought species, due to better performance and therefore the existence of
192 surviving recruits and neighboring canopy trees from this species; alternatively, recruits can
193 be equally as affected as adults, or even more affected, thus jeopardizing dominant species'
194 persistence after the event. Furthermore, we address short-term resilience in terms of
195 changes in the climatic suitability of the replacing community relative to the suitability of
196 dominant pre-drought species. We hypothesize a decrease in community climatic suitability
197 and, therefore, a greater climatic disequilibrium following drought, as it is expected that
198 dominant pre-drought species are more suited to historical site conditions than the post-
199 drought replacements.

200

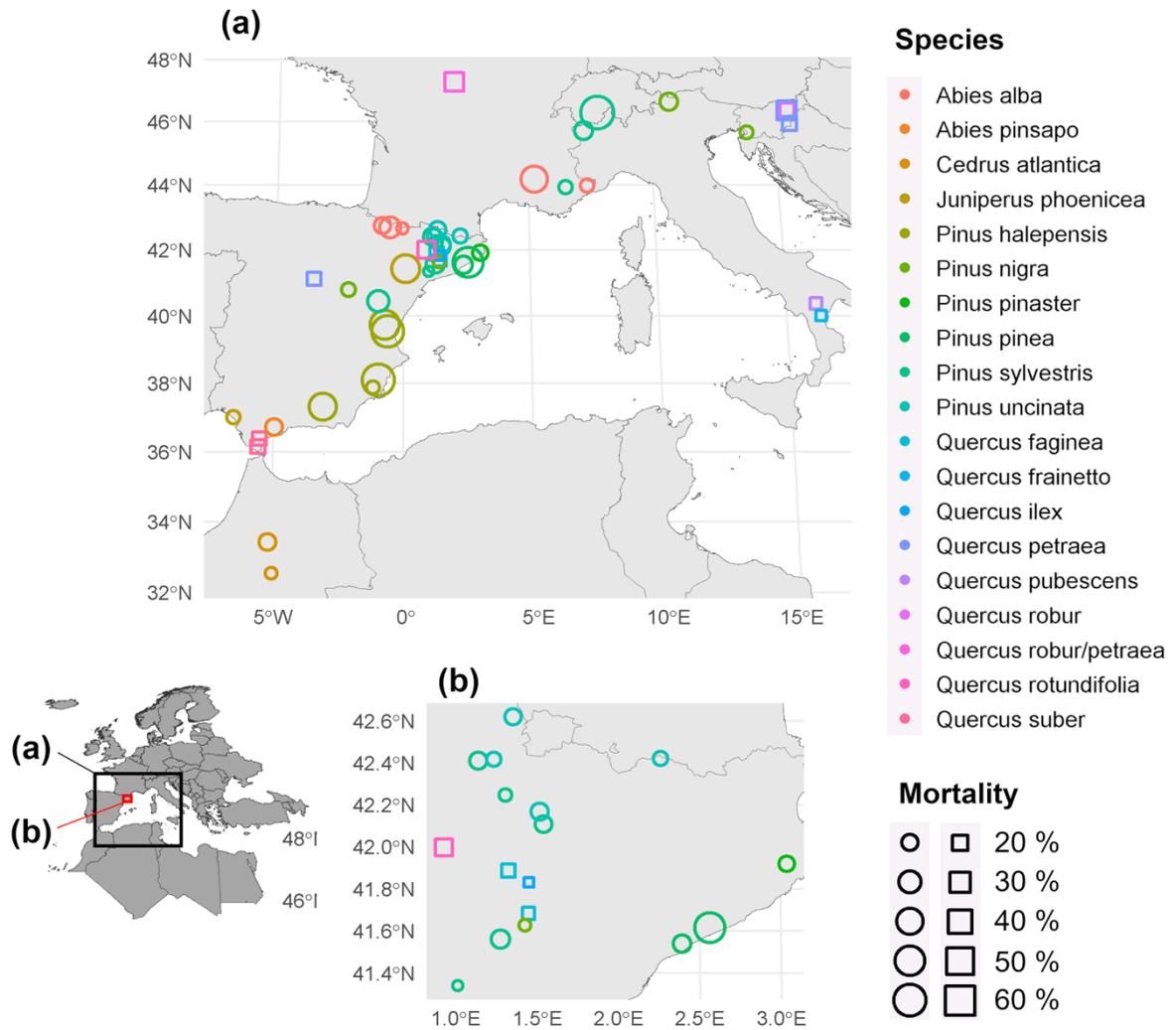
201 **2. Methods**

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2.1. Location and description of the study sites

From 2014 to 2019, we surveyed 48 sites distributed across Morocco, Spain, France, Switzerland, Italy and Slovenia (Fig. 1), corresponding to semi-natural forest dominated by single tree species. A total of 2,400 trees were sampled. All sites showed conspicuous standing mortality of the dominant tree species, ranging from 14.5% to 60.6%. Overall, the study sites encompassed 18 different tree species: ten conifers (*Abies*, *Cedrus*, *Juniperus* and *Pinus* genera), and eight belonging to the Fagaceae family (*Quercus* genus) (Supplementary Material Table S1). The sites did not show any evidence of recent disturbances (e.g., wildfires, windstorms causing tree fall), planting, or other intensive management practices (pruning, grazing, resin or cork exploitation).

On 28 of the sites (~60%), tree mortality had been explicitly related to drought in previous studies (Supplementary Material Table S1). Moreover, we checked for climatic drought concomitant with tree mortality, according to the Standardized Precipitation-Evapotranspiration Index SPEI (Vicente-Serrano et al. 2010; Begueria et al. 2010). Minimum SPEI values over the ten years before the sampling date were below -1.5 on almost all the studied sites (except for two sites where this value was -1.4) (see below for details on this index) (Supplementary Material Table S1). Overall, these SPEI values correlated positively with the degree of forest die-off across the whole set of sites (Supplementary Material Table S2), endorsing the contribution of drought to forest decay, although other factors, such as pests and pathogens, could also be involved.



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229 Figure 1. Map of sampled sites in the Western Mediterranean region (a), with detail of
230 locations in Northeastern Spain (b). Sign color identifies the different species: circles
231 correspond to Conifers and squares to Fagaceae. Sign size describes the magnitude of the
232 standing mortality (%) recorded at each site. Additional information for each site is provided
233 in Supplementary Material Table S1.

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235

236 **2.2. Survey description and structural variables**

237

238 At each site a set of close linear transects (located less than 250 m apart from each other)
239 were randomly selected to include at least 50 adult trees with non-overlapping canopies of
240 the dominant pre-drought species within any of the following die-off categories: (1) dead
241 trees - no green leaves, or in the case of *Quercus*, trees with few leaves on some sprouts,
242 but without the capacity to restore a tree's mid-term viability; (2) defoliated trees - less than
243 70% of green canopy, and (3) healthy trees - more than 75% of green canopy
244 (Supplementary Material Figures S1, S2). In all cases, sampled trees were part of the forest
245 overstory. An estimate of the standing mortality of the sites' dominant species was obtained
246 from the whole set of recorded trees (Supplementary Material Table S1). For each selected
247 tree, we measured its height (tip of the upper branches) and we identified the tallest woody
248 plants established beneath the canopy. Young seedlings were not considered, while close
249 tree neighbors with canopies occupying the space potentially released by focal trees were
250 included. In order to maintain the same criteria for identifying potential replacers across the
251 three die-off tree categories, we visually estimated the perimeter of the former tree canopy,
252 in case of focal dead trees. When no woody species was present beneath the canopy of the
253 affected trees, no potential replacer was recognized and a "gap" was recorded. The relative
254 height of the canopy of the replacing plants in relation to the former canopy (*Hrel*) was
255 calculated as:

256

257
$$Hrel = 100 - \left(\left(\frac{H_{dom} - H_{rep}}{H_{dom}} \right) \times 100 \right)$$

258

259 where H_{dom} is the height of the selected trees of the dominant species and H_{rep} is the
260 height of the tallest woody plants established beneath the canopy of each selected tree.

261

262 Moreover, on each site we estimated the die-off percentage of the dominant species, as
263 well as the total forest density and basal area. To do this, we randomly established 10 to 14
264 points within 5 m of the transect lines, stratified according to the length of the different
265 transects. At each of these points, two intersecting perpendicular lines were drawn over the
266 ground to demarcate four 90° quadrants. For each quadrant, the distance was measured
267 from the central point to the nearest tree, regardless of its species identity, and the
268 percentage of green canopy and the diameter at breast height (DBH) of all these trees'
269 stems (larger than 2 cm diameter) were measured (Supplementary Material Figure S2).

270 These measures were used to calculate:

271

272 - Dominant die-off (% die-off of the dominant tree species in the locality) was calculated as:

273

274

$$275 \quad \text{Dominant die - off} = \left(\frac{\sum_{i=1}^k (ABD_i - (RGC_i \times ABD_i))}{\sum_{i=1}^k ABD_i} \right) \times 100$$

276

277 where ABD_i is the basal area of all trees (from $i=1$ to k) of the dominant species sampled in
278 the 90° quadrants, and RGC_i corresponds to the proportion of each tree's remaining green
279 canopy. This measure can be considered an estimate of drought damage, which is
280 equivalent to the severity of the drought disturbance.

281

282 - Density was calculated by considering the inverse of the circular area determined by a
283 radius equal to the distance of the nearest tree, irrespective of species, in each 90° quadrant
284 to the central point (d_i) (Pollard, 1971) as:

$$285 \quad \text{Density} = \left(\frac{4 \times (N - 1)}{\pi \sum_{i=1}^k d_i^2} \right)$$

286
287 where N is the total number of trees (irrespective of species) measured in the 90° quadrants
288 around all central sampling points established on a given site.

289
290 - Basal Area (stand basal area) was calculated as the sum of the DBH-derived basal area of
291 all trees, weighted by Density, as:

$$292 \quad \text{Basal Area} = \left(\sum_{i=1}^k AB_i \right) \times \text{Density}$$

293
294
295 where AB_i is the basal area of the nearest tree sampled in each 90° quadrant (from $i = 1$ to k
296 = 4), irrespective of species.

297
298 - Richness, computed as the number of different woody species found as nearest neighbors
299 within the four 90° quadrants around each sampling point.

300
301 Furthermore, on each site we recorded information about the management legacy. This
302 information was based on on-site visual evidence of previous planting (e.g., regular spatial
303 and tree size structures), logging (e.g., stumps), pruning, strong grazing (e.g. trampling),
304 resin or cork exploitation, as well as evidence of major pests (Scolytidae, *Mastococcus* sp.,

305 processionary moth) or fungal pathogens (*Heterobasidium* sp., *Armillaria* sp., *Phytophthora*
306 sp.). For instance, a legacy of logging or high thinning would correspond to a forest site with
307 abundant stumps. Given the diversity of the studied forests, a one-size-fits-all threshold to
308 define the occurrence of management legacy (e.g., stumps density) is hard to establish and
309 that is why such estimates rely on the authors' observations in the study areas and they
310 result in a qualitative estimation (presence or absence of clear evidence) of such legacy.

311

312

313 **2.3. Drought characterization**

314

315 The drought conditions experienced on each site were characterized using the SPEI index
316 (Vicente-Serrano et al. 2010; Begueria et al. 2010), an index of climatic drought based on
317 the difference between precipitation and potential evapotranspiration. This index has been
318 demonstrated to correlate with other ecological proxies of drought (Vicente-Serrano et al
319 2013). This multi-scalar index calculates water deficit deviations for a given period in
320 relation to the reference timespan (in our case, 1950-2019), considering different temporal
321 windows of values (encompassing 1 to 48 months, SPEI1 and SPEI48, respectively) (accessed
322 March 2020; <https://monitordesequia.csic.es>). In our case, drought conditions on each site
323 were estimated from 10-year averages of SPEI24 prior to the sampling date. This wide
324 temporal window is appropriate for exploring the relationships between climate and forest
325 dynamics, since tree growth and mortality responses to climatic variability are often
326 buffered over time and lagged and cumulative responses are common (Tei and Sugimoto
327 2018). A short temporal window would prevent to correlation of our variables of resilience
328 and climatic suitability changes with the climate conditions that have actually determined
329 the observed patterns of occupancy of the space left by trees damaged by the extreme

330 drought episode. The drought characteristics were then estimated for each site by
331 calculating the following SPEI24-derived variables: (1) minimum SPEI value (drought
332 intensity), (2) total number of months with SPEI<-1.5 (drought frequency), (3) number of
333 consecutive months with SPEI<-1.5 around the minimum SPEI (drought length) and (4)
334 number of months from the sampling date to the minimum SPEI value (time since drought).
335 We opted to use SPEI values under -1.5 as being indicative of drought conditions, as this
336 value appropriately characterizes intense drought in secular climatic series for the
337 considered SPEI window (Vicente-Serrano et al. 2010).

338

339

340 **2.4. Estimation of species climatic suitability**

341

342 We used MaxEnt v3.4.1 (Phillips and Dudik 2008) to compute the climatic suitability of all
343 the dominant pre-drought species and post-drought replacing species (N = 111) on the
344 study sites, based on species occurrence data and six climatic variables: annual
345 isothermality, annual maximum and minimum temperatures, total annual precipitation,
346 precipitation seasonality and aridity. Occurrence data were obtained from GBIF (accessed
347 September 2019; <https://doi.org/10.15468/dl.sm1oc0>). For each species, we selected
348 occurrence records after 1950 from the original GBIF occurrence records. We retained only
349 one randomly selected occurrence per 1×1 km² pixel over the grid defined by the climatic
350 data (detailed below) to obtain a spatially-homogeneous data set, accounting for potential
351 geographical biases in the sampling effort. Subsequently, we used a 95% density distribution
352 kernel to systematically remove outliers in relation to the bulk of each species' occurrence
353 records and we thus removed occurrence points far from the major clusters of the species'
354 distributions. The final number of occurrences ranged from 62 (*Ononis reuteri*) to 67,768

355 (*Quercus robur*). Finally, the selected occurrences for each species were used to extract the
356 climatic information of the six climatic variables mentioned above. As for aridity, we used an
357 aridity index to quantify precipitation availability over atmospheric water demand,
358 computed as Aridity Index (AI) = MAP / MAE, where: MAP = Mean Annual Precipitation and
359 MAE = Mean Annual Potential Evapo-Transpiration. According to this formulation, the AI
360 index values increase for more humid conditions, and decrease with more arid conditions.
361 Such variables provide a good characterization of the overall precipitation and temperature
362 regime of the study areas. Climatology-type temperature and precipitation data for the
363 period 1970-2000 were obtained at a 1-km² spatial resolution from WorldClim 2.0 (Fick and
364 Hijmans 2017) and aridity data were obtained from the CGIAR-CSI Global-Aridity and Global-
365 PET Database (accessed March 2020; <http://www.cgiar-csi.org>) (Zomer et al. 2007, 2008).

366

367 Climatic suitability models for each species were computed on the basis of species
368 occurrence data and pseudo-absences or background points. Background points were
369 randomly located over the geographical area in which each species is present (Acevedo et
370 al. 2012). First, 100 model replicates were fitted with species occurrence data and 1,000
371 different background points for each species. The models' predictive accuracy was assessed
372 by means of the area under the receiver-operator curve (AUC; mean value for all assessed
373 species 0.989; range 0.938 - 1), which is interpreted as the probability that a randomly
374 chosen species occurrence location has greater suitability than a randomly chosen
375 background point (Merow et al. 2013). Second, species climatic suitability within each study
376 site was computed as the average suitability of the 100 MaxEnt model replicates fitted for
377 each species on the site.

378

379

380 **2.5. Analyses**

381

382 *2.5.1. Resilience of forest taxonomic composition and structure*

383

384 The resilience of taxonomic composition was assessed by checking for each site for
385 differences between the sets of replacing woody species growing under dead, defoliated
386 and healthy trees. For this purpose we performed pair-comparison of species abundance
387 (number of individuals) between such habitats using a Pearson correlation coefficient.

388 Relationships between these R Pearson values and drought characteristics - see above- were
389 explored using Spearman's rank correlation test. We excluded from the analysis seven
390 localities with less than four replacing species. We also checked for differences in species
391 richness (S) and Shannon diversity index (H') between each site's sets of replacing woody
392 species under dead, defoliated and healthy trees by repeated-measures ANOVA (no
393 transformation was needed).

394

395 We built different generalized linear mixed models (GLMMs) to evaluate short-term
396 resilience with respect to the persistence of the dominant species (self-replacement) and
397 structural stand characteristics. For each site, these models' response variables, which
398 exhibited normal error distributions, were (1) self-replacement, defined as percentage of
399 cases where dead trees plus trees experiencing die-off presented individuals of the same
400 species beneath its canopy; (2) gaps (%), corresponding to those cases in which the space
401 beneath the tree canopy was exclusively occupied by bare soil, herbs, or scrubs, (3) relative
402 height of the canopy of the replacing plant in relation to the former canopy (Hrel); and (4) a
403 combined estimation of the three previous variables, which corresponded to the

404 coordinates of sites on the first axis of a Principal Component Analysis (PCA) considering
405 self-replacement, gaps and Hrel.

406

407 These models' explanatory variables were: climatic suitability of the dominant species,
408 proportion of die-off in dominant species, density, basal area, richness, management legacy
409 (a binomial variable, with presence or absence of evidence), pest/pathogens (a binomial
410 variable, yes/no), taxonomic category of the dominant species (Fagaceae or Pinaceae,
411 Conifers hereafter), and drought characteristics (including the four SPEI-derived variables:
412 minimum SPEI, drought frequency, drought length, time since drought). Dominant pre-
413 drought species were included as a random factor.

414

415

416 *2.5.2. Changes in species climatic suitability*

417

418 Resilience in terms of changes in the forest climatic suitability after drought-induced die-off
419 was estimated by comparing the replacing species' climatic suitability to that of the
420 dominant pre-drought species in each site. We considered different sets of replacing species
421 according to their role in forest structure (1) tree species likely to attain the forest
422 overstory, including the dominant pre-drought species, (2) shrubs that will remain in the
423 understory and (3) all woody species, including the aforementioned trees and shrubs. For
424 the purposes of this comparison, we calculated Relative Climatic Suitability Change (RCSC,
425 hereafter) as:

426

$$427 \quad RCSC = \left(\frac{(DomSuit - WSCS)}{DomSuit} \right)$$

428

429 where *DomSuit* for a given site was the climatic suitability of the dominant species
430 (Dominant suitability hereafter) and *WSCS* (Weighted Species Climatic Suitability) for a given
431 site was the average climatic suitability of the species belonging to the different forest levels
432 (overstory, understory) weighted by its relative abundance in each set, resulting in several
433 values of RCSC: $RCSC_{over}$ for the overstory, $RCSC_{under}$ for the understory and $RCSC_{woody}$ for the
434 whole set of woody species. Higher positive values of RCSC indicate that the dominant pre-
435 drought species show higher climatic suitability than the replacing community, in relation to
436 the 1970-2000 reference period, evidencing a trend towards a loss of community climatic
437 suitability and an increase in climatic disequilibrium. In contrast, negative values of RCSC
438 indicate that the replacing community has higher climatic suitability than the dominant pre-
439 drought species, pointing to a gain in climatic suitability and a decrease in climatic
440 disequilibrium. The rationale behind using a measure of relative change was to minimize the
441 bias produced by the suitability values of the dominant species, which generally exhibited
442 higher values of climatic suitability than most of the co-existing species, thus largely
443 determining the value of the difference between dominant suitability and WSCS.

444

445 We also aimed to incorporate changes in suitability due to the absence of woody species
446 under the canopy of affected trees, i.e. gaps. We thus built a PCA that considered the values
447 of three variables on each site: $RCSC_{over}$, $RCSC_{under}$ and gaps (percentage). Then, we used the
448 first and second PCA axis (PCA1 RCSC and PCA2 RCSC, respectively) as an integrative
449 measure of forest change on each site associated with species climatic suitability.

450

451 Finally, we built GLMMs with the variables describing changes in climatic suitability as
452 dependent variables (i.e., $RCSC_{over}$, $RCSC_{under}$, $RCSC_{woody}$; first and second axis of PCA with

453 RCSC_{over}, RCSC_{under} and gaps) and the same explanatory and random factor variables as in
454 the aforementioned GLMMs exploring self-replacement and structural stand features.
455
456 Transformations of data were applied when required to obtain a normal error distribution:
457 arc-sinus to self-replacement, Hrel, richness, drought length, and time since drought, and
458 logarithmic to gaps (ln+1), dominant die-off, basal area and density. In all models, variables
459 were centered and scaled by SD, and then the minimum value of each variable was added to
460 all measures to avoid negative values. We checked in all models for the inflation factor,
461 which always had values under 1.5, and for the percentage of the random factor's
462 contribution to the total variance. We included all explanatory variables and bivariate
463 interactions in full models, and we selected the best fitted model by following a backward
464 procedure and applying the AIC criterion. Calculations were made with JMP10.0.0. (@2012
465 SAS Institute Inc.), following REML procedure.

466

467 A table with the information on all the variables used in these models is provided in
468 Supplementary Material Table S3.

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473 **3. Results**

474

475 ***3.1. Resilience of taxonomic composition and stand structure***

476

477 ***3.1.1. Changes in species composition***

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479 The composition of replacing species under the canopy of dead and healthy trees was
480 significantly different on only nine of the 41 sites, thereby endorsing the overall resilience of
481 understory taxonomic composition. Similarly, when comparing dead and defoliated trees,
482 we found differences in only ten cases, and when comparing defoliated and healthy trees
483 we found differences in only seven cases (Supplementary Material Table S1). Three localities
484 (Arenys and Cabrera, both with *Pinus pinea*, and Jimena, with *Quercus suber* as pre-drought
485 dominants) exhibited different compositions of replacing species when comparing the three
486 states of the effect of drought on the dominant species. There was a negative correlation
487 between the R Pearson values of the dead vs defoliated tree comparison of replacing
488 composition and minimum SPEI value (i.e. drought intensity) (Spearman $r_o = -0.359$, $P =$
489 0.023), this correlation being marginally significant when comparing dead vs healthy trees
490 (Spearman $r_o = -0.280$, $P = 0.076$). These results support our hypothesis that that more
491 intense droughts correlate with less resilience in terms of taxonomic composition, i.e., with
492 taxonomic similarities between the replacing species under dead trees and those under less
493 affected trees. Finally, we did not find any significant differences in the species richness (F
494 ratio = 0.80, $P = 0.453$) and the H' diversity index (F ratio = 0.25, $P = 0.771$) of replacing
495 plants under dead, defoliated and healthy trees.

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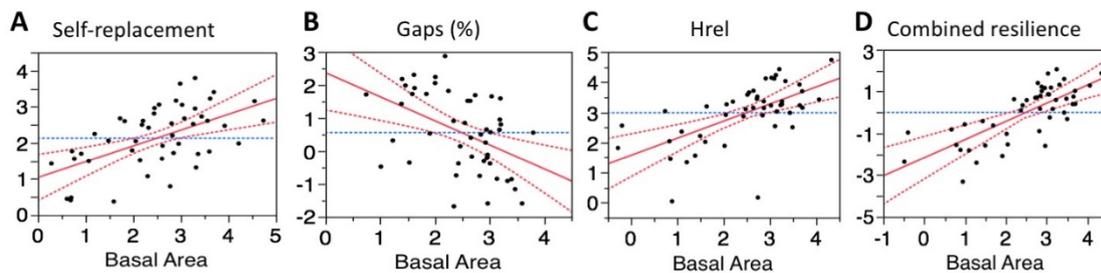
497 3.1.2. Self-replacement

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499 Self-replacement significantly decreased with a higher proportion of die-off of the dominant
500 species, and it increased with a higher basal area (Figure 2A), supporting our hypotheses.
501 However, contrary to from our expectations, self-replacement was not significantly
502 determined by the climatic suitability of the dominant species or by drought characteristics

503 (Table 1). The final model also included a positive effect of the interaction between
 504 dominant die-off and density, due to the elimination of the negative relationship of
 505 dominant die-off with self-replacement when high density occurs (Figure 3). Management
 506 legacy also influenced self-replacement, decreasing its value from 38.21% \pm 6.72SE (no
 507 evidence of management legacy) to 17.1% \pm 3.45SE (management legacy) (Supplementary
 508 Material Figure S3A). This finding also supported our hypothesis.

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 517 Figure 2. Leverage plot of the effect of stand basal area on several variables describing
 518 resilience: self-replacement (A), gaps (B), Hrel (C) and combined resilience (coordinates in
 519 the first axis of a PCA considering self-replacement, gaps and Hrel (D). Residuals of the
 520 response variable are plotted against the residuals of the respective explanatory variable.
 521 Solid line shows least square fit, while dashed lines show 5% confidence bands for the
 522 mean. The dashed horizontal blue line represents the mean of the response variable
 523 leverage residuals. The slope and P-values correspond to the respective estimate in the
 524 model (Table 1). Hrel: relative distance of the canopy of the replacing plant from the former
 525 canopy.

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Table 1. GLMM results of self-replacement, gaps, Hrel and combined resilience as response variables and dominant suitability and die-off, stand characteristics, drought characteristics, management legacy and taxonomic category of the dominant species as explanatory variables. Combined resilience was calculated as the first axis of a PCA considering, self-replacement, gaps and Hrel. R²: conditional correlation coefficient. Var ratio: ratio between variance attributed to the random effect (species) and variance attributed to the residual. See Methods for a description of the full model and variable transformations. Hrel: relative distance of the canopy of the replacing plant from the former canopy.

Self-replacement

(R² = 0.412, AIC = 135.09, Var ratio = -0.092)

	Estimate (SD)	F ratio	P
Intercept	1.040 (0.357)		
Dominant die-off	-0.344 (0.114)	9.05	0.005
Basal area	0.437 (0.121)	13.10	0.001
Density	-0.403 (0.145)	7.71	0.035
Dominant die-off : Density	0.355 (0.127)	7.87	0.008
Management legacy [Yes]	-0.518 (0.131)	15.76	<0.001

Gaps

(R² = 0.496, AIC = 156.51, Var ratio = 0.138)

	Estimate (SD)	F ratio	P
Intercept	1.750 (0.331)		
Basal area	-0.728 (0.205)	12.59	0.003
Richness	-0.526 (0.178)	8.77	0.005

Hrel

(R² = 0.370, AIC = 129.92, Var ratio = -0.021)

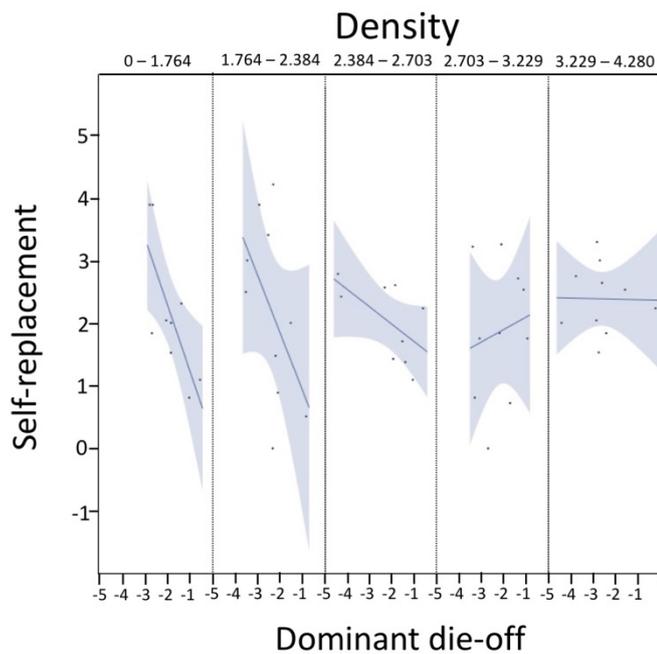
	Estimate (SD)	F ratio	P
Intercept	1.750 (0.331)		
Basal area	0.567 (0.127)	19.83	<0.001
Taxonomic category [Fagaceae]	0.385 (0.134)	8.27	0.011

Combined resilience

(R² = 0.559, AIC = 143.37, Var ratio = -0.035)

	Estimate (SD)	F ratio	P
Intercept	-2.111 (0.421)		
Basal area	0.870 (0.156)	31.09	0.001
Taxonomic category [Fagaceae]	0.577 (0.165)	12.25	0.010
Management legacy [Yes]	-0.363 (0.195)	3.46	0.087

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542 Figure 3. Self-replacement in relation to the affectation of the pre-drought dominant species
543 (Dominant die-off) at different intervals of stand Density. Intervals include similar number of
544 cases, illustrating the significant interaction between Dominant die-off and stand Density,
545 but they did not intervene in the model's calculations. See Methods for scaling and
546 transformation of variables and for a complete description of the model (GLMM). Shadow
547 indicates 95% confidence of fit.

548

549

550 3.1.3. Stand structure

551

552 The mean percentage of gaps (no woody species beneath the canopy of the affected trees)

553 appearing after the drought-induced episodes of tree mortality was 10.4% (± 2.5 SE). As

554 expected, gaps became significantly less frequent with increasing richness (Supplementary

555 Material Figure S4) and basal area (Figure 2B) (Table 1), and sites with a greater basal area

556 also had replacing plants with higher Hrel (Figure 2C). In Fagaceae-dominated forests, the

557 replacing plants also grew closer to the remaining canopy (height of replacing plants

558 corresponding to 83.7% ± 4.61 SE of the height of the former canopy) than in Conifer-

559 dominated forests ($66.2\% \pm 4.76SE$), as supported by the significant effect of taxonomic
560 category on Hrel (Table 1) (Supplementary Material Figure S3B). Gaps and Hrel were not
561 significantly determined by the climatic suitability of the dominant pre-drought species or
562 drought characteristics.

563

564 *3.1.4. Combined self-replacement and stand structure*

565

566 The first axis of the PCA considering self-replacement and resilience of structural
567 characteristics (gaps, Hrel) accounted for 59.6% of the variance (eigenvalue = 1.79) and was
568 mostly related to greater resilience in terms of replacing canopy (Hrel) and, to a lesser
569 extent, to greater self-replacement and less gaps. The second axis of the PCA only
570 accounted for 25.5% of the variance; it roughly correlated with self-replacement and gaps,
571 but the eigenvalue was only 0.76 and we excluded it from further analyses (Supplementary
572 Material Figure S5).

573

574 The model explaining the ordination of sites on this first PCA axis - indicating greater short-
575 term resilience in terms of self-replacement and replacing canopy height - showed a
576 significant positive correlation of stand basal area with the first PCA axis (Figure 2D) in
577 keeping with the assumption that greater basal area - as a proxy for a population's good
578 performance in a locality - would be positively related to short-term resilience. Fagaceae-
579 dominated forests also showed significantly higher values than Conifers-dominated ones on
580 this first axis, while management legacy had a marginal effect, resulting in lower values on
581 the first PCA axis (Table 1) (Supplementary Material Figure S3C,D). PCA axes were not
582 significantly determined by the climatic suitability of the dominant pre-drought species or

583 drought characteristics in agreement with previous results on self-replacement and stand
584 structure.

585

586

587 The percentage of variance components corresponding to the random effect (species) was
588 negligible in the models evaluating self-replacement, Hrel and ordination on the first PCA
589 axis combining self-replacement and structural characteristics. In contrast, this percentage
590 had a value of 12.16 in the model with gaps as the dependent variable.

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597 ***3.2. Changes in species climatic suitability***

598

599 The mean values of $RCSC_{over}$, $RCSC_{under}$, $RCSC_{woody}$ were $0.004 \pm 0.019SE$, $0.005 \pm 0.021SE$ and
600 $0.0017 \pm 0.0152SE$, respectively, representing a range that included gain and loss of climatic
601 suitability of the replacing community in relation to the suitability of the dominant pre-
602 drought species. $RCSC_{over}$ was positively related to dominant suitability, indicating that
603 replacing communities on sites where dominant species live in more suitable conditions
604 tended to reduce their climatic suitability more than the dominant pre-drought species
605 (Table 2), overall supporting our overall hypothesis. This trend tends to disappear under
606 very intense drought conditions, as indicated by the significant interaction between
607 dominant suitability and minimum SPEI (maximal drought intensity) (Figure 4A). Indeed,

608 under stronger drought conditions, $RCSC_{over}$ tended to be higher, as revealed by the
609 marginally negative relationship of $RCSC_{over}$ with minimum SPEI (maximal drought intensity)
610 (Table 2). Loss of climatic suitability in the understory-replacing community (higher values of
611 $RCSC_{under}$) was also positively related to the suitability of the dominant pre-drought species
612 and negatively related to tree density (Supplementary Material Figure S6). Management
613 legacy tended to diminish the climatic suitability of the replacing understory ($RCSC_{under}$
614 values of $0.006 \pm 0.038SE$ and 0.004 ± 0.025 for managed and unmanaged forest, respectively)
615 (Supplementary Material Figure S3E). Evidence of pests or pathogens reduced $RCSC_{under}$,
616 pointing to an increase in the climatic suitability of the replacing understory relative to that
617 of the pre-drought species ($RCSC_{under}$ values of $-0.017 \pm 0.035SE$ and $0.019 \pm 0.026SE$ for
618 forests with and without signs of biotic disturbance, respectively) (Supplementary Material
619 Figure S3F). Finally, the time since drought significantly affected the loss of the replacing
620 understory's climatic suitability (Supplementary Material Figure S6) (Table 2). The $RCSC_{woody}$
621 model reproduced the results obtained in the $RCSC_{over}$ model (Table 2). The explanation of
622 these variables was greatly affected by the different performances of dominant pre-drought
623 species, since much of the variance (35.5% in the $RCSC_{over}$ model, 77.1% in the $RCSC_{under}$
624 model, and 45.5% in the $RCSC_{woody}$ model) was attributed to the random effect (species).
625
626 The first axis of the PCA considering $RCSC_{over}$, $RCSC_{under}$ and gaps (PCA1 RCSC) accounted for
627 60.4% of the variance (eigenvalue = 1.82) and was positively related to all three variables,
628 particularly $RCSC_{under}$. The second axis of the PCA (PCA2 RCSC) accounted for 29.8% of the
629 variance (eigenvalue = 0.89) and correlated positively with gaps and negatively with $RCSC_{over}$
630 (Supplementary Material Figure S7). The model explaining the ordination of sites in PCA1
631 RCSC - corresponding to higher $RCSC_{over}$ and $RCSC_{under}$ and more gaps - showed a positive
632 relationship with dominant suitability, but this trend was reversed at low stand densities, as

633 indicated by a significant interaction between dominant suitability and stand density (Figure
634 4B). Also, management legacy correlated positively with this first PCA axis (Supplementary
635 Material Figure S8A) (Table 2). Again, the percentage of the variance component
636 corresponding to the random effect (species) was remarkably high in this model (69.1%).
637 The model explaining the ordination of sites in PCA2 RCSC - with higher values for lower
638 RCSC_{over} and more gaps - showed a significant negative relationship with stand basal area
639 (Supplementary Material Figure S8B) (Table 2); in this model, only 11.8% of the variance was
640 attributed to the random effect (species).
641

Table 2. GLMM results of relative changes in the climatic suitability (RCSC) of the recruiting community relative to the dominant species' climatic suitability. RCSC_{over}: overstory community, RCSC_{under}: understory community, RCSC_{woody}: all woody species, PCA1 RCSC and PCA2 RCSC: first and second axis of PCA built with RCSC_{over}, RCSC_{under} and percentage of gaps (PCA1 was positively related with all three variables, while PCA2 positively correlated with gaps and negatively with RCSC_{over}). Explanatory variables included species dominant suitability and die-off, stand features, drought characteristics, signs of pest or pathogens and management legacy. Lower values of Minimum SPEI correspond to higher drought intensity. R²: conditional correlation coefficient. Var ratio: ratio between variance attributed to the random effect (species) and variance attributed to the residual. See Methods for a description of the full model and variable transformations. SPEI: Standardized Precipitation Evapotranspiration Index.

RCSC_{over}

(R² = 0.647, AIC = 131.10, Var ratio = 0.550)

	Estimate (SD)	F ratio	P
Intercept	2.949 (0.603)		
Dominant suitability	0.406 (0.125)	10.45	0.002
Minimum SPEI	-0.238 (0.129)	3.38	0.074
Dominant suitability : Minimum SPEI	0.528 (0.161)	10.75	0.002

RCSC_{under}

(R² = 0.949, AIC = 81.89, Var ratio = 3.375)

	Estimate (SD)	F ratio	P
Intercept	1.589 (0.340)		
Dominant suitability	0.784 (0.083)	87.52	<0.001
Density	-0.426 (0.097)	19.21	<0.001
Management legacy [Yes]	0.361 (0.106)	11.73	0.002
Pest/Pathogens [Yes]	-0.187 (0.068)	7.63	0.011
Time since drought	0.317 (0.076)	17.15	<0.001

RCSC_{woody}

(R² = 0.707, AIC = 130.67, Var ratio = 0.835)

	Estimate (SD)	F ratio	P
Intercept	2.766 (0.580)		
Dominant suitability	0.439 (0.122)	13.00	<0.001
Minimum SPEI	-0.120 (0.118)	2.87	0.098
Dominant suitability : Minimum SPEI	0.467 (0.158)	8.73	0.005

PCA1 RCSC

(R² = 0.909, AIC = 85.03, Var ratio = 2.233)

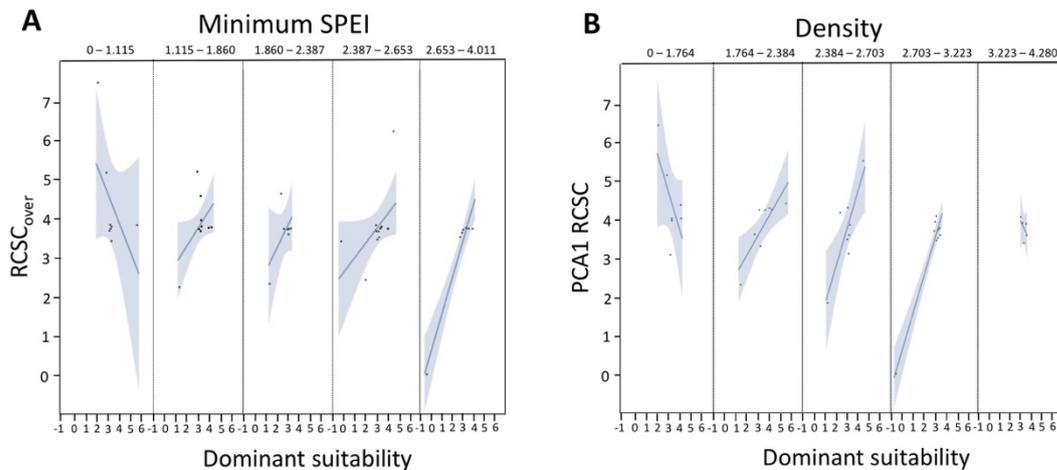
	Estimate (SD)	F ratio	P
Intercept	1.686 (0.414)		
Dominant suitability	0.702 (0.105)	44.59	<0.001
Density	-0.155 (0.122)	1.65	0.213
Management legacy [Yes]	0.302 (0.137)	4.83	0.035
Dominant suitability : Density	0.971 (0.220)	19.43	<0.001

PCA2 RCSC

(R² = 0.357, AIC = 107.95, Var ratio = 0.133)

	Estimate (SD)	F ratio	P
Intercept	2.691 (0.435)		

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Figure 4. Relative change in the climatic suitability of the overstory recruit community compared to the dominant species' climatic suitability ($RCSC_{over}$), in relation to the climatic suitability of the dominant species (Dominant suitability) at different intervals of maximal drought intensity (minimum SPEI) (A), and first axis of a PCA built with $RCSC_{over}$, $RCSC_{under}$ and percentage of gaps - positively related with all these three variables - (PCA1 $RCSC$) in relation to the climatic suitability of the dominant species at different intervals of stand Density (B). Intervals include similar number of cases, illustrating significant interaction, but they did not intervene in the model's calculations. See Methods for scaling and transformation of variables and for a complete description of the models (GLMMs). Shadow indicates 95% confidence of fit. SPEI: Standardized Precipitation Evapotranspiration Index.

4. Discussion

4.1. General patterns

Our study shows that, over a broad geographical range and different types of forest, short-term resilience after drought-induced tree mortality is influenced by both the degree of severity of the disturbance experienced in the forest and stand characteristics. These stand features include species diversity, tree density and forest basal area, which reflects

667 management legacy. The impact of drought on the replacing community is shown to be a
668 major mechanism for explaining the relevance of these factors. Interestingly, the relevance
669 of these factors is not affected by the historical climatic suitability of the dominant species
670 on each site. In other words, dominant tree populations living close to their optimal climatic
671 conditions may be vulnerable to these drought episodes. In turn, populations living near the
672 limits of species climatic suitability may exhibit mechanisms that allow them to persist.
673 Nevertheless, our models showed a limited statistical capacity to explain short-term
674 resilience in the studied forests. This may be due to the existence of other unexplored
675 factors, to our low ability to quantify the studied factors (e.g., management legacy), and,
676 especially, to the inherent variability within the types and range of the studied forests,
677 which cover a broad range of biogeographical and land use contexts.

678

679 ***4.2. Resilience of taxonomic composition***

680

681 Self-replacement constitutes a major mechanism for forest resilience after disturbances
682 (Johnstone et al. 2016). Here, we found that self-replacement was jeopardized by drought
683 damage (% dominant die-off), equivalent to the severity of the drought disturbance. This
684 indicates that recruits from the dominant species, in addition to adults, were likely affected.
685 Previous studies have demonstrated that forest understory can also be impacted in
686 drought-induced mortality episodes (Lloret et al. 2004; Suarez and Kitzberger 2008). In fact
687 higher mortality rates have been reported in juvenile and small individuals than in adults in
688 these episodes (Lloret et al. 2004), as well as in drought experiments (Hanson et al 2001).
689 This suggests that the physiological mechanisms leading to drought-induced mortality
690 (hydraulic failure, carbon starvation, Adams et al. 2017) could operate similarly in adults and
691 juvenile plants (e.g., Sapes et al. 2019). Overall, this sensitivity of recruiting populations is

692 consistent with the common demographic trends of high mortality rates in younger plants
693 and of regeneration climatic niche as subsets of adult niche (Dobrowski et al. 2015).
694
695 However, climate variables characterizing drought, together with species climatic suitability
696 derived from SDMs failed to explain self-replacement rates. The above-mentioned
697 limitations inherent to our study may explain this failure. In addition, several mechanisms
698 may mask the relationship between estimates of climatic suitability and demographic
699 processes associated with drought episodes. Moreover, acclimation of populations to
700 specific conditions can be important to understand short-term resilience, and this process is
701 not well considered in SDMs, which assume that species respond homogeneously to climate
702 across their range (Benito Garzon et al. 2011). Populations living close to a species' climatic
703 optimum may have produced a large stock of recruits ready to replace dying trees ("core
704 hypothesis" Lloret and Kitzberger 2018), but these populations may also suffer from
705 extreme climatic conditions due to their acclimation history. They may be less efficient
706 under these extreme conditions (Anderegg et al. 2016), or they may be vulnerable if they
707 have built structures such as stems, branches or leaves that require a significant amount of
708 resources for their maintenance (Jump et al. 2017). In contrast, plants - including recruits -
709 living close to the climatic limits of species suitability (i.e., marginal populations) may have
710 experienced acclimation or selection to extreme conditions (Valladares et al. 2014; Solarik
711 et al. 2018). The balance between sensitivity vs. acclimation to extreme conditions may
712 explain similar levels of self-replacement across the broad range of conditions and species
713 suitability assessed here. This is consistent with the reported occurrence of tree die-back
714 and mortality in both core and edge populations (Lloret and Kitzberger 2018; Margalef et al.
715 2020) and with patterns of forest dynamics after drought-induced mortality (Martinez-
716 Vilalta and Lloret 2016, Batllori et al. 2020).

717

718 Self-replacement patterns were also explained by stand characteristics, across a wide range
719 of forests types. A large basal area likely reflects good habitat quality - as determined by soil
720 depth and nutrient availability, for instance -, which should promote tree growth and the
721 existence of saplings and young adults of the dominant species (Vayreda et al. 2013).
722 Moreover, a large basal area also characterizes forests that have not recently experienced
723 extensive disturbances; in these undisturbed forests, recruits are more likely to be found.
724 This is consistent with the negative effect of management legacy on self-replacement, also
725 reflected in the first axis of the PCA-combined estimation of resilience, likely due to its
726 impact on potential recruits. Management is a major contributor to forest dynamics in semi-
727 natural forests, as is the case on most of the studied sites, and it has been reported as
728 determining ecological short-term trajectories after drought-mortality episodes, often
729 leading to increasing prevalence of shrubs (Batllori et al. 2020). The importance of recruits
730 would also be reflected by higher levels of self-replacement in stands with high density,
731 which would likely shelter more recruits. Accordingly, in high density stands the
732 correspondence between high dominant die-off and low self-replacement tends to
733 disappear.

734

735 Although the composition of replacing species that grew beneath dead, defoliated and
736 healthy trees was different in some forests, a consistent pattern was not observed. This
737 could be attributable to a level of replication insufficient to capture site variability.
738 However, our sampling was able to distinguish changes in the resemblance of the replacing
739 community under trees with different degrees of die-off in relation to drought intensity.
740 Thus, assuming that the understory was relatively uniform across each site before the
741 episode, and that it was affected by drought, our results suggest that drought impacted the

742 understory to some extent as well. This impact would correspond to a sorting effect caused
743 by environmental stress, as proposed by community assembly theory (White and Jentsch
744 2004). In this case, drought would be the driver of such community sorting, as has been
745 observed in Mediterranean shrublands close to some of the study sites (Perez-Navarro et al.
746 2019).

747

748 Our appreciation of resilience, particularly of taxonomic composition including self-
749 replacement, corresponds to a short temporal window of forest dynamics whose capacity to
750 predict the long-term fate of forests subjected to extreme drought episodes is clearly
751 limited. Alternatively, pre-established long-term programs of forest inventories have also
752 been able to document tendencies in taxonomical and functional composition likely related
753 to increasing levels of climate dryness (Zhang et al. 2018; Ruiz-Benito et al. 2017), but the
754 monitoring timespan of most of these programs to date has been too short to produce
755 conclusive results, and they find it difficult to capture a sufficient number of plots in areas
756 strongly affected by drought. Our approach has the advantage of focusing on forests that
757 have experienced such affectation. However, oncoming disturbance regimes, climate
758 change - including further drought episodes -, biotic interactions - pests, exotic species - and
759 management will largely determine the trajectories of the studied forests. Nevertheless,
760 many recent studies highlight the importance of the initial stages after disturbances in
761 forest dynamics (Turner et al. 1998; Seidl et al. 2014; Johnstone et al. 2016), particularly
762 when successional trajectories are not disrupted by any further disturbances or
763 management. Species sorting in these initial stages may be determined by different, non-
764 exclusive mechanisms (White and Jentsch 2004). Species badly adapted to arid conditions
765 can get filtered (Perez-Navarro et al. 2019). Alternatively, gap opening may initiate
766 successional trajectories in accordance with the existing sapling bank and the differential

767 tolerance of species to light and competition (Prach and Walker 2020), causing succession to
768 move back toward earlier stages (Pickett and White 1985). Another possibility is that the
769 death of trees that are dominant in the canopy may accelerate the transition to later stages
770 through species replacement (Rigling et al. 2013). This could be further assessed by an
771 accurate analysis of a species' successional status and plant functional traits in the replacing
772 community (Batllori et al 2020). Although we cannot compare the differences between
773 these stages and those occurring in paired stands of unaffected forests, our individual-based
774 comparison between trees with different degrees of drought disturbance reveals that initial
775 trajectories following drought exhibit noticeable differences from one forest to another.

776

777

778 ***4.3. Resilience of stand structure***

779

780 In stands with a larger basal area, recruits and pre-established trees were closer to the
781 canopy level of the damaged trees, and gap occurrence after drought mortality was lower,
782 reflecting better habitat quality or a late successional stage of stands. Interestingly, gap
783 occurrence also diminished in stands with higher richness of woody species, likely indicating
784 the existence of complementarity effects of diversity (Loreau and de Mazancourt 2013).
785 Thus, the overall forest canopy would be more likely to be maintained if the total set of
786 species is able to live in a wide range of climatic conditions - including those occurring
787 during the experienced drought episodes. This hypothesis corresponds with the increasing
788 evidence that tree species richness promotes forest resilience, particularly under strong
789 climatic fluctuations (Lloret et al. 2007; Hutchinson et al. 2018; Sousa-Silva et al. 2018).

790

791 Also, Fagaceae-dominated forests tended to be more resilient, especially in terms of canopy
792 height, than Conifer-dominated ones. This higher resilience of Fagaceae-dominated forests
793 is consistent with recent patterns of forest growth in the Iberian Peninsula in response to
794 drought (Vidal-Macua et al. 2017). This pattern can be related with distinct functional
795 characteristics of Fagaceae and Conifers (Carnicer et al. 2013), involving a greater legacy of
796 drought impact on radial growth in Pinaceae than in Fagaceae (Cailleret et al. 2017), higher
797 photosynthetic rates in evergreen angiosperms than in Conifers (Lusk et al. 2003), or a
798 greater ability to maintain NSC pools in deciduous species, which are common among
799 Fagaceae, than in evergreen ones (Piper 2020). Also, Fagaceae species produce larger seeds
800 than Conifers, which eventually confer them with a greater recruiting capacity (Carnicer et
801 al. 2014).

802

803 Furthermore, Fagaceae have a greater resprouting capacity - from stumps or epicormic buds
804 - than Conifers (Bond and Midgley 2001; Zeppel et al. 2015). Resprouting is recognized as a
805 key trait for plant recovery after disturbances (Bond and Midgley 2001; Pausas et al. 2016),
806 including drought-induced mortality episodes (Zeppel et al. 2015; Batllori et al. 2020). It has
807 also been proposed that resprouters show a more conservative strategy of water use than
808 non-resprouters (Zeppel et al. 2015), which would give them a particular would confer them
809 advantage during intense drought (Pausas et al. 2016). Moreover, post-disturbance
810 regrowth tends to be faster in species resprouting from surviving plants than in non-
811 resprouting ones, which must recruit new individuals. Overall, resprouter species, such as
812 *Quercus* spp., *Populus* spp., *Eucalyptus* spp., tend to become more prevalent than non-
813 resprouters ones after episodes of drought-induced mortality in temperate and boreal
814 forests (Batllori et al. 2020).

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818 **4.4. Changes in species climatic suitability**

819

820 Changes in the climatic suitability of the replacing community also reflect the resilience of
821 forests affected by drought. Substantial changes in the climatic suitability of the replacing
822 community would indicate that such forests are in the initial stage of a sorting process, in
823 which species turnover results in an altered climatic suitability in relation to the historical
824 climate. The fact that the time since drought tends to increase $RCSC_{under}$ suggests that, in
825 the studied forests, the initial pulses of reduction of climatic suitability are consolidated
826 over the long term, and that the demographic impact of drought on the understory persists
827 for some time (Lloret et al. 2004). The changes observed in replacing communities' climatic
828 suitability were strongly determined by species idiosyncrasy, as shown by the large
829 contribution of the random factor (species) in the models. Considering that the average
830 suitability values of the replacing species on each site were not significantly different from
831 the suitability value of the dominant pre-drought species (t ratio = 0.90, $P = 0.372$), these
832 changes should mostly be attributed to shifts in the species' relative abundance.

833

834 The variability explained by the rest of the models' fixed variables ($RCSC_{over}$, $RCSC_{under}$,
835 $RCSC_{woody}$, PCA1 RCSC and PCA2 RCSC) shows that the greater the suitability of the dominant
836 pre-drought species, the greater the loss of suitability in the replacing community, and the
837 greater the increase in community climatic disequilibrium. This may be explained by the
838 assumption that dominant species are well suited to a region's climatic conditions (but see
839 Sexton et al. 2009). In fact, when drought intensity was higher, this relationship between
840 both $RCSC_{over}$ and $RSCS_{woody}$ and the historical suitability of the dominant species tended to

841 disappear, probably because of the concomitant impact of drought on other species.
842 Alternatively, in a context of climate change, a sudden short-term reduction of community
843 climatic disequilibrium may also occur as a result of a decrease in those species less
844 adjusted to the harsh climatic conditions concurrent with the disturbance (Perez-Navarro et
845 al. 2021), or a greater abundance of those species better suited to the new conditions. This
846 reflects the widely acknowledged notion that species assemblages match the climatic
847 conditions of the sites on which they occur, and that climatic changes are followed by
848 corresponding changes in species distribution and community composition (Blonder 2015,
849 Svenning and Sandel 2013, Gauzere et al. 2018). Thus, climatic disequilibrium would occur
850 when community composition does not track the climate promptly or accurately (Blonder et
851 al 2015) - in our case, this corresponds to higher RCSC -. Both management and the effects
852 of pests and pathogens on RCSC_{under} may also reflect human-driven modification in
853 understory composition, thus altering any such climate tracking. Nevertheless, the
854 relevance of all these processes should be considered with caution, given the low
855 explanatory power of the model when discounting the contribution of the species (random)
856 effect.

857

858 Moreover, the overall failure of climatic suitability in explaining resilience to drought of
859 stand structure and taxonomic composition could also be due to a low capacity in our
860 procedure for quantitative discrimination of climatic suitability across species. The MaxEnt
861 algorithm is recognized as being robust for species comparisons, performing particularly
862 well when occurrences are low across the species distribution area (Pearson et al. 2007). It
863 has also been proved to be useful for explaining differences between species' demographic
864 responses to drought episodes (Sapes et al. 2017, Lloret and Kitzberger 2018), even more
865 than other algorithms (Pérez-Navarro et al. 2019), even though their outputs may be

866 sensitive to the selected climatic variables. In our case, the 10% and 90% percentile values
867 of climatic suitability for pre-drought dominant trees obtained from MaxEnt models were
868 0.407 and 0.700, respectively. This reduced capacity of the fitted models to discriminate
869 between species likely affected our calculations of changes in climatic suitability before and
870 after drought, and this may have contributed to the low explanatory capacity of our
871 statistical models. Predictions obtained by assembling different algorithms are supposed to
872 be advantageous for predicting populations' suitability (Araujo and New 2007) but these
873 ensembles can be sensitive to the predictive capacity of the selected algorithms. Therefore,
874 the inclusion of models with low predictive accuracy could reduce the performance of
875 consensus predictions in comparison with individual models (Elith and Graham 2009; Hao et
876 al. 2020). A more accurate assessment of community climatic disequilibrium could be
877 achieved in the future via other procedures, such as building the climatic environment's
878 multivariate space, including yearly variability, and then using the distance between the
879 species assemblies and observed climate (Blonder et al. 2015; Perez-Navarro et al. 2020)
880

881 **5. Conclusions**

882

883 Our study reveals that the self-replacement of dominant species after drought-induced
884 mortality episodes is not guaranteed. In fact, the percentage of gaps uncovered by woody
885 vegetation may noticeably increase in the short term after a single die-off episode.

886 Regardless of the historical climatic suitability of the affected species, the short-term
887 community dynamics and the recovery of major structural forest features depend on the
888 pre-existing bank of potential replacers - both in the understory and the overstory. This
889 result contrasts with our expectations of higher resilience in terms of self-replacement in

890 those localities with high climatic suitability for the dominant pre-drought species. As
891 hypothesized, other essential drivers of resilience are related to the structural conditions of
892 the forest, such as basal area - likely reflecting habitat quality or successional/maturity stage
893 - stand density and, to a lesser degree, species richness, all of which are closely linked to
894 management legacy. Also, the replacing community is affected by the impact on pre-
895 drought understory species, supporting our hypothesis about the role of drought
896 characteristics in the short-term patterns of resilience. Therefore, drought could exert a
897 filtering effect on community composition in the early stages of the subsequent
898 successional trajectories. To what extent this climatic filtering can alter expected
899 successional pathways remains to be elucidated. A decrease in the climatic suitability of the
900 replacing community, was observed, particularly on sites climatically suitable for dominant
901 species, thus supporting our overall hypothesis of greater climatic disequilibrium following
902 drought. However, our results were largely determined by a highly idiosyncratic
903 performance in which both species composition and the structural state of forests would be
904 determining factors. Therefore, forest management should reinforce resilient trajectories by
905 controlling such parameters.

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