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1 **Scots pine (*Pinus sylvestris* L.) mortality is**
2 **explained by the climatic suitability of both host tree**
3 **and bark beetle populations**

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17

18 **Abstract**

19 Higher temperatures and extreme drought events are promoting insect-driven tree mortality.
20 However, there is great uncertainty about the impact of extreme climatic variations on the
21 susceptibility to infestation, as this depends on the suitability of climatic conditions to both
22 host trees and insects. So, the consequences of infestation would likely be more intense in
23 tree populations living closer to the limits of species' climatic tolerance, where resource
24 allocation to defenses could be compromised by other functions essential to tree survival.
25 In this article, we explored 22 Scots pine populations (*Pinus sylvestris* L.) in the northeast of
26 the Iberian Peninsula that experienced a tree mortality episode involving bark beetle
27 infestation and drought. We hypothesized that the infestation and eventual tree mortality
28 varied according to the climatic suitability of a particular location for the host trees and bark
29 beetles. Climatic suitability values were estimated by using species distribution models
30 (MaxEnt). Then, we developed generalized linear mixed-effects models and partial least
31 squares-generalized linear models to study tree mortality and stand bark beetle infestation
32 in relation to host and insect suitability, intensity of bark beetle attack, tree characteristics,
33 and stand structure. Intensity of bark beetle attack and tree size were the main factors
34 determining the probability of tree mortality. Smaller trees in particular exhibited a higher
35 probability of dying at lower values of attack intensity. At stand level, bark beetle infestation
36 was negatively associated with the density of Scots pine and the basal area, and positively
37 with tree species richness. Taken together, the response of Scots pine populations to tree
38 characteristics and stand structure suggested an attack by bark beetles with a low
39 population density that preferred weakened smaller trees. Remarkably, our results showed
40 that Scots pine populations established in higher climatic suitability areas were more
41 vulnerable to infestation, particularly when locations were also suitable for these insects.
42 Therefore, previous drought events and differential resource allocation for growth and
43 defenses among central and peripheral tree populations seem to regulate infestation

44 patterns, as mediated by the suitability for both hosts and insects. Our study demonstrates
45 the importance of assessing species' climatic suitability to better understand and anticipate
46 tree mortality episodes driven by bark beetles, particularly under climate scenarios with
47 increasingly frequent extreme climatic events.

48

49 Keywords: bark beetles; Scots pine; drought; climatic suitability; tree mortality; stand
50 infestation; disturbance interactions.

51

53 **1 Introduction**

54 Climate change has been linked to increased mortality of coniferous tree species
55 around the globe due to extreme drought, heat, and associated insect outbreaks (Allen et
56 al., 2010; Das et al., 2013; García de la Serrana et al., 2015; Logan et al., 2003; Raffa et
57 al., 2008). Specifically, the contribution of bark beetles (Curculionidae: Scolytinae) to forest
58 mortality has emerged as a major disturbance under the context of global climate change
59 (Anderegg et al., 2015). Unprecedented outbreaks, in which some bark beetle species have
60 killed billions of coniferous trees across vast areas (e.g., Bentz et al., 2010; Meddens et al.,
61 2012; Rouault et al., 2006; Seidl et al., 2016) and the expansion of the geographic range of
62 native beetle species into new areas (Logan and Powell, 2001; Raffa et al., 2013; Weed et
63 al., 2013) indicate that disturbance regimes associated with insect-driven tree mortality are
64 changing. Furthermore, extreme climatic events such as drought episodes are likely to
65 affect local tree populations differently across their geographic distribution range (Hampe
66 and Petit, 2005). However, there are still many uncertainties about how drought interacts
67 with other co-drivers, particularly bark beetle infestation, in modulating the vulnerability and
68 mortality of host tree populations across climatic gradients (Creeden et al., 2014; Gaylord
69 et al., 2013).

70 Several studies of insect-driven tree mortality have determined that the emergence
71 of outbreaks of infestation depend on the habitat conditions, such as tree characteristics and
72 forest structure and composition (Bentz et al., 2010; Jenkins et al., 2014; Raffa et al., 2008).
73 More specifically monospecific stands with high basal area and high density of the host tree
74 are the most susceptible to bark beetle infestation (Chapman et al., 2012; Fettig et al., 2007;
75 Perkins and Roberts, 2003; Shore et al., 2000), as beetles are more likely to find a
76 susceptible tree and achieve a successful attack (Raffa et al., 2016). Furthermore, trees
77 exhibit differences in their defensive ability to regulate bark beetle populations (Raffa and
78 Berryman, 1983) according to their age, size, and the level of stress they support (Krokene,

79 2015). The insect-host system is also highly sensitive to climatic variability (Seidl et al.,
80 2009). Warming increases beetles' overwinter survival and accelerates their growth rates
81 (Bentz et al., 2010; Creedon et al., 2014; Temperli et al., 2015), whilst severe drought stress
82 disrupts host tree defenses, thereby increasing trees' overall susceptibility to infestation
83 (Gaylord et al., 2013; Netherer et al., 2015). This is especially relevant in the ongoing
84 warming scenario and the predicted increase in extreme climatic conditions that will affect
85 the habitat suitability of both insects and hosts (Kirschbaum, 2000; Walther et al., 2002).
86 Therefore, the insect-host system could be modified at the edges of the geographical range
87 of tree populations when these correspond to the limits of species' climatic tolerance
88 (Thuiller et al., 2008). For instance, the emergence of climates favoring bark beetle
89 populations would enhance their attack in places where infestations have not previously
90 been recorded (Buotte et al., 2017; Lantschner et al., 2017).

91 According to the classic niche theory (MacArthur, 1972), perturbations in the
92 phenology, abundance, and ecological interactions of species are more evident in localities
93 with less suitable environmental conditions and limitations on population performance due
94 to harsher climatic conditions (Lynch et al., 2014; Morin et al., 2007; Raffa et al., 2015),
95 which in turn often correspond to the edges of the range of a species' distribution (Parmesan,
96 2006). However, some recent studies have suggested that locally adapted tree populations
97 living close to the edges of a species' climatic range can be more resistant to extreme
98 drought episodes than core populations (Cavin and Jump, 2017; Lloret and Kitzberger,
99 2018; Rehm et al., 2015), probably due to genetic adaptations and/or higher plastic
100 responses (Aitken et al., 2008; Hampe and Petit, 2005; Jump and Peñuelas, 2005;
101 Savolainen et al., 2007). In contrast, populations living in localities with more suitable
102 environmental conditions could be less tolerant to extreme climatic fluctuations since they
103 have to invest a higher proportion of resources to maintain high growth rates, basal area or
104 stand density (Anderegg et al., 2015; Gutschick and BassiriRad, 2003; Jump et al., 2017;

105 McDowell et al., 2008). Based on these premises, tree populations probably respond
106 differently to biotic disturbances if such disturbances take place under environmental
107 conditions within the historical range of variability or during extreme drought episodes. In
108 this work, we propose that the climatic suitability of both the host trees and the insects are
109 involved in the response of tree populations to bark beetle attacks. This approach will
110 considerably further understanding of the consequences of climate change on host-insect
111 dynamics across biogeographical gradients and therefore improve anticipation of specific
112 management and conservation actions on tree populations.

113 Scots pine (*Pinus sylvestris* L.) is one of the most widespread conifer species in the
114 world and one of the most abundant trees in Europe, representing a keystone species for
115 many ecosystems. This species reaches its southern distribution limit in the Mediterranean
116 Basin, where it is primarily restricted by a combination of high temperatures and summer
117 drought and thus comes closer to the limits of its climatic tolerance (Carlisle and Brown,
118 1968). Scots pine populations situated in this region may therefore be particularly sensitive
119 to extreme climatic events when these are combined with insect attack (Hampe and Petit,
120 2005; Matías and Jump, 2012). In fact, drought-induced Scots pine dieback has recently
121 been observed in the Iberian Peninsula (Hódar et al., 2003; Martínez-Vilalta and Piñol, 2002;
122 Galiano et al. 2010), along with a variable intensity of insect attacks, particularly of bark
123 beetle species belonging to the *Ips* and *Tomicus* genera (Lieutier et al., 2004). Nevertheless,
124 the contribution of bark beetle attacks to the mortality of Scots pine populations exhibiting
125 drought-induced decline has not been fully addressed (Dobbertin et al., 2007; Krams et al.,
126 2012; Wermelinger et al., 2008).

127 In this study, we considered the climatic suitability of the host and insect species
128 inferred from of species distribution models (SDMs), in order to evaluate post-drought tree
129 mortality and bark beetle infestation in Scots pine populations from the northeast of the
130 Iberian Peninsula. The assumption behind SDMs is that species occurrences represent the

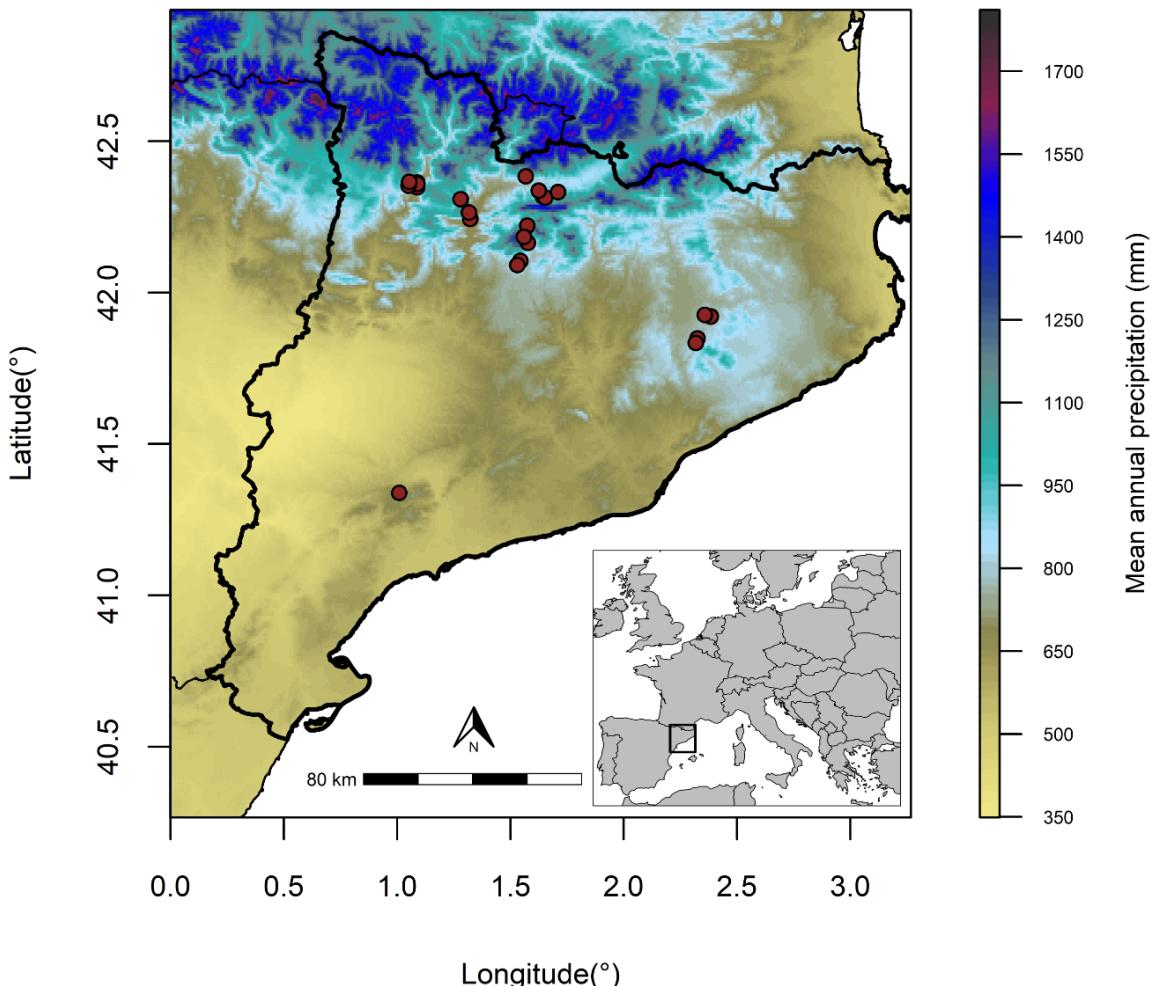
131 environmental conditions that are suitable for a given species to reproduce and survive
132 (Soberón and Peterson, 2005). Significantly, model output can be interpreted as a species-
133 specific index of the climatically suitable conditions across the landscape (Franklin, 2010)
134 and can be used to evaluate intra- and inter-specific responses (e.g., mortality, decline, and
135 insect infestation) to short-term climatic events (Lloret and Kitzberger, 2018; Pérez Navarro
136 et al., 2018; Sapes et al., 2017). We explore the response of tree mortality and bark beetle
137 infestation at tree and stand level to the climatic suitability of both the host tree and insect,
138 tree characteristics, and stand structure. Specifically, we examined whether 1) larger and
139 closer trees with evidence of bark beetle attack have a greater probability of mortality; 2)
140 denser and monospecific stands of the host tree are more prone to be infested; 3) trees
141 living in localities with low values of climatic suitability are more vulnerable to die, and 4)
142 high insect climatic suitability results in a higher probability of stand infestation by bark
143 beetles, and how this is modulated by the interaction of host-insect suitabilities.

144 **2 Material and Methods**

145 **2.1 Description of the study area**

146 The study was carried out on the northeast Iberian Peninsula (Catalonia, Spain), in
147 locations with evidence of recent Scots pine drought-induced mortality and bark beetle
148 infestation. Scots pine reaches the southernmost and driest limit of its wide natural
149 distribution range in the Iberian Peninsula, where the species occupies large areas in the
150 Pyrenees, along with some marginal populations in central and southern mountain regions
151 (Matías and Jump, 2012). In 2005, the Iberian Peninsula experienced an extreme drought
152 episode that was followed by several dry years (EEA, 2017). This drought event affected
153 Scots pine populations in Catalonia, among other conifer species; various symptoms were
154 detected including tree defoliation, mortality (Galiano et al., 2010), and bark beetle
155 infestation (Blanco et al., 2009). Our study focused on the post-drought period that took
156 place during 2005-2013 (Serra-Maluquer et al., 2018; Tejedor et al., 2017).

157 We selected 22 different Scots pine populations throughout the species distribution
158 range in Catalonia, to compare the tree mortality and infestation processes on sites with
159 contrasted climate conditions (Fig. 1). The populations' elevation ranged from 600 to 1,600
160 m a.s.l, with mean annual temperatures from 6.5 °C to 13 °C and annual precipitation
161 between 700 and 1,100 mm/year (WorldClim, version 2.0., Fick and Hijmans, 2017). Sites
162 included pure stands of Scots pine (*Pinus sylvestris* L.) and mixed stands of Scots pine and
163 holm oak (*Quercus ilex* L.) or mountain pine (*Pinus uncinata* Ramond ex DC. in Lam. and
164 DC) with an understory dominated by strawberry tree (*Arbutus unedo* L.), kermes oak
165 (*Quercus coccifera* L.), boxwood (*Buxus sempervirens* L.) and common juniper (*Juniperus*
166 *communis* L.).



167

168 **Figure 1.** Map of the study area in the northeast of Spain and location of the 22 sampled Scots pine populations
169 (red dots) across the precipitation gradient (mean annual precipitation (mm)).

170 **2.2 Field data collection and stand attributes**

171 During 2013-2014, a circular stand (of 314 m²) was established in each of the 22
172 selected populations. On each site, we recorded the size of all adult trees (diameter at breast
173 height, dbh > 2.5 cm for Scots pine, and dbh > 5 cm for other tree species), identified all
174 trees to species, and registered their status (living or dead). We also recorded the distance
175 from all Scots pine trees to their nearest conspecific neighbor. A total of 962 Scots pines
176 were sampled.

177 Scots pine trees were naturally attacked by four bark beetles species within the study
178 region: *Ips acuminatus* (Gyll.), *Ips sexdentatus* (Börner), *Tomicus minor* (Hart.), and
179 *Tomicus piniperda* (L.). The identification of each bark beetle species was principally based
180 on galleries and/or bark beetle imagoes. For all Scots pine individuals, we measured the
181 intensity of attack in a bark sample area of 20 × 30 cm at 130 cm height by counting the
182 number of bore holes. Entrance holes are a common estimate of attack intensity (e.g.,
183 Borkowski and Skrzecz, 2016; Lieutier et al., 1995). However, exit holes are often difficult to
184 differentiate from entrance holes, so we considered both exit and entrance holes as a
185 measure of the intensity of attack per m². We thus opted to potentially overestimate insect
186 attack, associated with infestation success, rather than underestimating it. Moreover, we
187 recorded the presence/absence of bark beetle species in other parts of the tree, including
188 signs of galleries in the branches and the upper part of the stem.

189 For each stand, we calculated the following forest attributes: Scots pine density
190 (trees ha⁻¹), Scots pine basal area (m² ha⁻¹), tree species richness, and a nearest neighbor
191 (R) based on the distances to the nearest neighbor recorded for all Scots pine trees
192 (following Clark and Evans, 1954). The calculation of the nearest neighbor was used as a
193 measure of the clustering or ordering of tree spatial distribution within each stand. High

194 values of this index ($R > 1$) suggest a regular distribution of trees, low values ($R < 1$) suggest
195 a clustered pattern, and values equal or close to 1 indicate a random distribution. Finally,
196 bark beetle infestation in each stand was considered as the proportion of dead trees
197 presenting signs of a successful attack (i.e., bore holes and signals of maternal galleries).

198 **2.3 Climatic suitability modeling**

199 We assessed the contribution of the climatic suitability of the host tree and beetle
200 species in the Scots pine mortality and infestation processes. We used species distribution
201 models (SDMs) built at the European scale to obtain estimates of the probability of species'
202 occurrences, both for Scots pine and bark beetles, under the climatic conditions that
203 characterize our study sites.

204 *Occurrence data*

205 Scots pine records were obtained from European Forest data (Mauri et al., 2017),
206 amounting to a total of 9,368 occurrences. For bark beetle species, we collated species
207 occurrences from different sources, due to the scarcity of records; these sources included
208 the Global Information Facility (GBIF 2018, <http://www.gbif.org/>), the ICP Forests Level I
209 (Eichhorn et al., 2016), and previously published studies carried out in Europe and the
210 Mediterranean Basin (Borkowski and Skrzecz, 2016; Gallego and Galián, 2001; Horn et al.,
211 2009; Jankowiak, 2012; Kerdelhué et al., 2002; Kohlmayr et al., 2002; Moraza et al., 2013;
212 Olsovsky et al., 2013; Pineau et al., 2017; Ritzerow et al., 2004; Rossi et al., 2009;
213 Vasconcelos et al., 2006). However, sufficient records could only be obtained for two of the
214 four beetle species present in our study stands: 73 occurrences for *T. piniperda* (including
215 the 1998-2014 period) and 100 occurrences for *I. sexdentatus* (including the 1970-2017
216 period). Despite the absence of records of the other two bark beetle species (*I. acuminatus*
217 and *T. minor*), our analyses were not seriously compromised since the largest part of the
218 Scots pine infestation was attributed to *I. sexdentatus* and *T. piniperda*. Furthermore, these
219 two species are considered two of the most damaging pests in Europe and cause significant

220 forest loss every year (Sauvard, 2004). We therefore considered them good representatives
221 for analyzing the infestation process that occurred in our study.

222 Both host tree and beetle occurrence data were filtered for collection year (since
223 1970) and location to remove taxonomic and geographic inconsistencies. We also filtered
224 occurrences by a minimum distance of 15 km, which corresponds to the coarsest resolution
225 of species' records in our database, in order to deal with aggregated records that may
226 introduce model biases or overfitting in the SDMs (Veloz, 2009). Moreover, we used a land
227 cover layer (Corine Land Cover 2000, version 18.5) to compile appropriate areas of possible
228 hosts for bark beetles (Lantschner et al., 2017) and we thus created a mask, which was then
229 used to select the distribution of each bark beetle generated from SDMs. The mask included
230 the following land cover classes: agro-forestry areas, different types of forest, and semi-
231 natural areas of transitional woodland-shrub with herbaceous vegetation.

232 *Environmental data*

233 SDM climatic predictor variables were selected from the 19 bioclimatic variables
234 available in WorldClim database version 2.0 (Fick and Hijmans, 2017) at a resolution of 30
235 seconds (~1 km²). We selected them on the basis of the host tree and insect ecological
236 requirements (Lieutier et al., 2004; Matías and Jump, 2012). To prevent multicollinearity
237 among them, variables were also filtered by means of a Variance Inflation Factor (VIF) < 10
238 as a critical threshold (Quinn and Keough, 2002) and a cluster dissimilarity < 0.8. Finally, we
239 retained six climatic variables as predictors of the distribution of the two bark beetles
240 species: isothermality (mean diurnal temperature range/temperature annual range),
241 temperature seasonality, mean temperature of the driest quarter, mean temperature of the
242 coldest quarter, precipitation seasonality, and precipitation of the driest quarter. For Scots
243 pine, we used isothermality, temperature annual range, mean temperature of the warmest
244 quarter, annual precipitation, precipitation seasonality, and precipitation of the warmest
245 quarter.

246 *Model algorithm and settings*

247 Given the nature of the data, especially the low number of bark beetle occurrences
248 and the lack of absence data, the algorithm selected to fit the SDMs was MaxEnt software
249 (version 3.4.1, Phillips et al., 2006). This algorithm implements the maximum entropy for
250 modeling species distributions with presence-only data and with a good performance for a
251 small number of occurrences (Pearson et al., 2007). We standardized the modeling criteria
252 using the same algorithm for the Scots pine and the two bark beetle species.

253 We experimentally calibrated the three MaxEnt models by tuning the feature classes
254 and regularization multiplier (R package ENMeval, Muscarella et al., 2014). The final
255 settings selected to run each model are described in Table 1. To test the models, we used
256 a 5-fold cross-validation for the two bark beetle species models and a 10-fold cross-
257 validation for the Scots pine model (Fielding and Bell, 1997). Then, climatic suitability was
258 obtained by projecting the models over the selected climatic layers. Outputs of the models
259 were averaged across 500 model replicates and a log raw transformation was applied to
260 obtain a habitat suitability value between 0 and 1, as the relative likelihood of species'
261 occurrence in each cell. The models' predictive accuracy was evaluated using the area
262 under the receiver-operating characteristic curve (AUC; Fielding and Bell, 1997), and we
263 also applied jackknife test to measure percent contribution of each variable to the final
264 model.

265 **2.4 Tree mortality and infestation modeling**

266 Statistical analyses were conducted at the tree- and stand-level. To improve the
267 stability and the accuracy of parameter estimates, we standardized the values of all
268 explanatory variables into dimensionless variables by subtracting the mean value of each
269 variable and dividing by two standard deviations (Gelman, 2008).

270 At the tree-level, we used generalized linear mixed-effects models (GLMM, R
271 package lme4, Bates et al., 2015) to analyze the probability of tree mortality as a function of

272 tree size, distance to the nearest neighbor, intensity of bark beetle attack, and climatic
273 suitability of Scots pine. The binary response of tree mortality (living or dead tree) was
274 modeled as a binomial error distribution with a logit-link function (Bolker et al., 2008). We
275 built a null model, from which we created a set of models by adding different combinations
276 and interactions between predictors. Note that presence/absence of bark beetle species
277 was finally not included in the models because it introduced a large amount of over-
278 dispersion. All the GLMMs were fitted by Laplace approximation and included 'stand' as a
279 random effect. After ranking our models, we selected the best one according to Akaike's
280 information criterion (AIC; Akaike, 1973), Bayesian information criterion (BIC; Schwarz,
281 1978), and the marginal and conditional R^2 (MuMIn package, Nakagawa and Schielzeth,
282 2013). The R^2 marginal can be considered as the proportion of variance explained only by
283 fixed effects, whereas R^2 conditional represents the variance explained by both fixed and
284 random effects. To avoid inflated measures of explained variance, we inspected model
285 assumptions and possible model overdispersion in the selected best models (R package
286 blmeco, Harrison, 2014).

287 At the stand-level, we modeled the proportion of bark beetle infestation as a function
288 of tree density, basal area, richness of tree species, nearest neighbor index, Scots pine
289 suitability, and the suitability of the two bark beetle species (i.e. *Ips sexdentatus* and *Tomicus*
290 *piniperda*). Since we found a high negative correlation between Scots pine and *I.*
291 *sexdentatus* suitabilities, we used the PLS (Partial Least Squares) approach for the analysis.
292 PLS generalizes and combines principal component analysis and multiple regression
293 features to model relationships between the response and explanatory variables. The
294 components are obtained through the combination of the original predictors that maximize
295 the explained variance (R^2) in the response variable (Abdi, 2003). Furthermore, this
296 technique is particularly useful for analyzing correlated predictor variables with a small
297 sample compared with the number of predictors (Carrascal et al., 2009). Thus, the response

298 variable infestation was assessed as a binomial variable formulated as a two-column matrix
299 holding the number of successes and failures, with an infested dead tree a success and a
300 non-infested dead tree a failure. Therefore, since our response variable followed a binomial
301 error distribution rather than a normal distribution, we used a PLS-GLM with a logit-link
302 function (R package *plsRglm*, Bastien et al., 2005) to model the infestation process. To build
303 the model, we focused on the independent effect of each predictor and the possible 2- and
304 3-way interactions between Scots pine and bark beetle species suitability. The meaning of
305 PLS components was derived from the weights of the original predictors on each
306 component. Predictors' weights explain the signal of their effect and its square value the
307 magnitude and importance of the effect on the component. The significance level for
308 predictors in the analysis was established at $P < 0.05$ and the coefficient of non-significant
309 predictors was set at 0. Once the PLS components were obtained, we checked their
310 significance by building a binomial response GLM with them. We selected the best candidate
311 model, considering the smallest AIC and checked model fit and performance by inspecting
312 model assumptions and possible model overdispersion.

313 **3 Results**

314 **3.1 Climatic suitability modeling**

315 The SMDs for the two bark beetle species showed good performance accuracy, with
316 mean AUC values higher than 0.75 and a standard deviation lower than ± 0.04 (Table 1).
317 According to the relative contribution and jackknife test of variables importance, the mean
318 temperature of the coldest quarter had the highest contribution in both the *I. sexdentatus*
319 and *T. piniperda* models, followed by mean temperature of the driest quarter, temperature
320 seasonality, and isothermality. The Scots pine model showed a poorer performance than
321 the bark beetle models, with a mean AUC value of 0.67 and a standard deviation of ± 0.005
322 (Table 1). According to the relative contribution and jackknife test, the mean temperature of
323 the warmest quarter had the highest contribution, followed by precipitation of the warmest

324 quarter, isothermality, and annual temperature range. Despite the suboptimal statistical
 325 accuracy of the Scots pine model, the results obtained yielded a distribution range of the
 326 host tree consistent with the ecological requirements of the studied species (Matías and
 327 Jump, 2012).

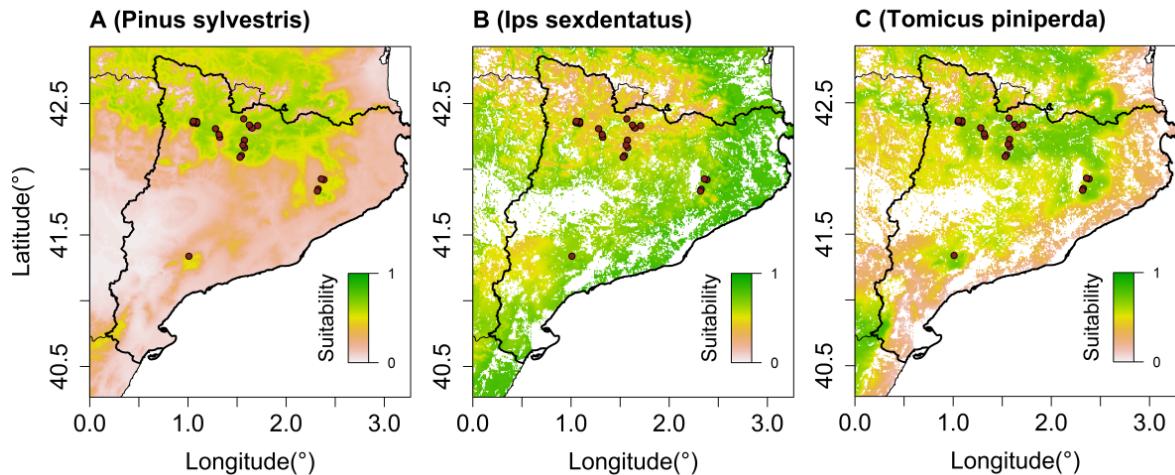
328 Table 1. Description and evaluation of the climatic suitability models built for Scots pine and the two bark beetle
 329 species, *T. piniperda* and *I. sexdentatus*. In each case the model description includes: number of occurrences
 330 (N occur.), number of background points (N backg.), feature classes (Feat.) for model fitting (L: Linear, Q:
 331 Quadratic, H: Hinge), regularization multiplier (RG), number of K-fold cross-validation, number of climatic
 332 variables used (N var.), and the top variables with the highest contribution for the model.

Species	Model description						Model evaluation	
	N occur.	N backg.	Feat.	RG	Cross-validation	N var.	Top variables	Mean AUC
<i>Tomicus piniperda</i>	73	10,000	LQH	3.0	5-fold	6	Mean temperature of coldest quarter; Mean temperature of driest quarter; Temperature seasonality; Isothermality	0.81 (±0.03)
<i>Ips sexdentatus</i>	100	10,000	LQH	2.5	5-fold	6	Mean temperature of coldest quarter; Mean temperature of driest quarter; Temperature seasonality; Isothermality	0.79 (±0.04)
<i>Pinus sylvestris</i>	9,368	20,000	LQH	0.5	10-fold	6	Mean temperature of warmest quarter; Precipitation of warmest quarter; Temperature annual range; Isothermality	0.67 (±0.01)

Standard errors are in brackets

333

334 Within the surveyed stands, the SDM-derived climatic suitability values ranged from
 335 0.37 to 0.73 for *I. sexdentatus*, from 0.33 to 0.68 for *T. piniperda*, and from 0.34 to 0.64 for
 336 Scots pine, thus indicating great variability between the sites (Fig. 2). *T. piniperda*
 337 occurrence in particular was expected to be more likely at higher altitudes with colder
 338 temperatures than *I. sexdentatus* occurrence. Scots pine did not present high suitability
 339 values on our study sites, probably because this tree species reaches its southern limit of
 340 distribution in the Iberian Peninsula.



341
 342 Figure 2. Habitat-suitability maps obtained in MaxEnt models for *P. sylvestris* (A), *I. sexdentatus* (B), and *T.*
 343 *piniperda* (C). Red dots correspond to the 22 study stands.

344 **3.2 Determinants of tree mortality and infestation**

345 A total of 56.2% of the sampled Scots pine trees corresponded to dead trees, of
 346 which 72% showed signs of bark beetle infestation (Table 2). In contrast, only 9% of the
 347 living trees showed evidence of bark beetle attack. Stands showed different severities of
 348 mortality and infestation, ranging from 14% to 82% when considering dead trees in relation
 349 to total trees, and from 0% to 100% when considering dead infested trees in relation to total
 350 dead trees.

351 At tree level, the best GLMM model fitted to describe Scots pine mortality included
 352 tree size, distance to the nearest neighbor, intensity of attack, Scots pine suitability, and the
 353 2-way interactions between the predictors (Table 3). Tree mortality was mostly explained by
 354 the intensity of bark beetle attack and tree size. Higher levels of tree mortality were
 355 associated with a greater intensity of attack and lower tree size (Fig. 3; A and B). The
 356 interaction between the intensity of attack and tree size revealed that small trees were
 357 particularly prone to die with increasing intensity of attack, even with a low number of attacks
 358 (Fig. 3; C). In contrast, at low levels of attack intensity, Scots pine mortality tended to

359 increase more markedly at higher rather than lower values of tree suitability (Fig. 3; D). The
 360 interactions between distance to the nearest neighbor and intensity of attack, and between
 361 distance to the nearest neighbor and tree size, were not significant (Table 3).

362 Table 2. Description of the tree- and stand level attributes (mean \pm standard deviation) within the study sites.

	Living trees	Living infested trees	Dead trees	Dead infested trees	Total stands
Number of trees	421	37	541	387	962
Mean dbh (cm)	14.6 (\pm 9.9)	20.9 (\pm 12.7)	15.2 (\pm 8.3)	15.4 (\pm 8.1)	14.9 (\pm 9.1)
Mean dnn (cm)	1.4 (\pm 1.0)	1.6 (\pm 1.3)	1.2 (\pm 1.0)	1.2 (\pm 1.0)	1.3 (\pm 1.0)
Mean iat (n holes/m ²)	75 (\pm 72)	143 (\pm 180)	207 (\pm 211)	284 (\pm 198)	123 (\pm 191)
Basal area (m ² /ha)	--	--	--	--	33.4 (\pm 14.4)
Density (n/ha)	--	--	--	--	1429.5 (\pm 909.9)
Richness (n tree species)	--	--	--	--	2.9 (\pm 1.5)
Nearest neighbor index	--	--	--	--	0.8 (\pm 0.2)

363

364 Table 3. Results of the GLMM describing the process of mortality at tree level. The model was carried out with
 365 mortality as a binary response variable (dead and living trees), eight fixed predictor variables (including two 2-
 366 way interactions), and stand as a random factor. In bold, significant fixed effects ($p < 0.01$).

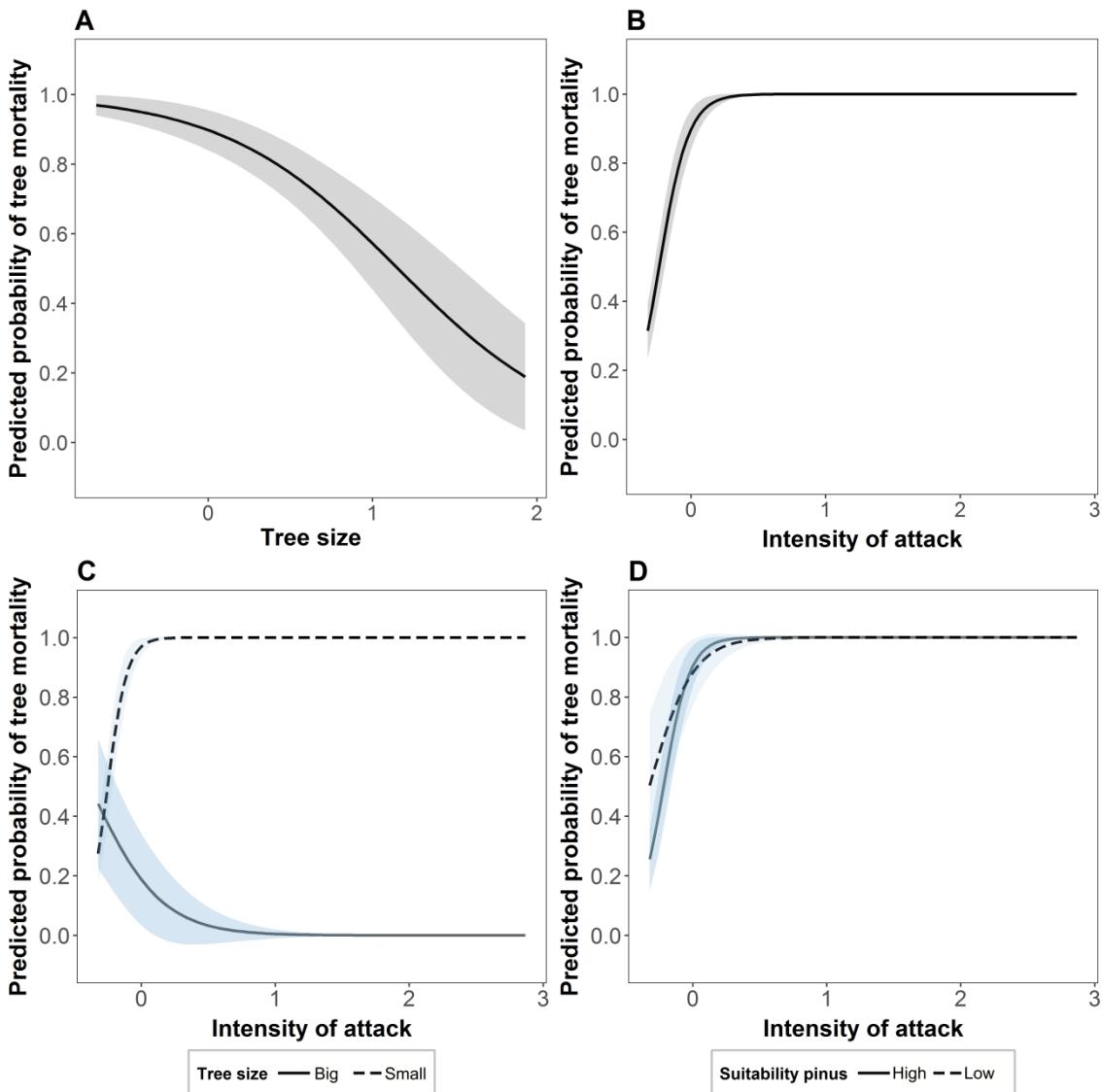
Fixed effects	Coefficient	z value	Pr(> Z)
Intercept	2.17 (\pm 0.32)	6.81	<0.001
Tree size (dbh)	-1.88 (\pm 0.34)	-5.51	<0.001
Distance nearest neighbor (dnn)	-0.01 (\pm 0.32)	-0.32	0.97
Intensity of attack (iat)	9.18 (\pm 0.93)	9.85	<0.001
Scots pine suitability (sps)	0.11 (\pm 0.35)	0.32	0.75
iat x dbh	-6.74 (\pm 0.96)	-6.99	<0.001
iat x dnn	0.97 (\pm 0.99)	0.98	0.33
dbh x dnn	-0.56 (\pm 0.35)	-1.59	0.11
iat x sps	2.03 (\pm 0.76)	2.67	<0.01

Only the final model is shown. AIC = 835; BIC = 884; R² marginal = 0.83; R² conditional = 0.85; Dispersion statistic = 0.90; N = 962.
 Standard errors are in brackets

367

368 At stand level, the PLS-GLM model built to describe Scots pine infestation (i.e., the
 369 proportion of dead infested trees) provided two significant components that together
 370 explained 77% of the original variance (Table 4), revealing that insect and host climatic
 371 suitability are significantly related to the infestation success. The first component (explaining
 372 42% of the variance) positively related infestation to Scots pine suitability, *T. piniperda*
 373 suitability, and the positive interaction between Scots pine and *I. sexdentatus* suitabilities.

374 In contrast, *I. sexdentatus* suitability was negatively related to infestation. Scots pine
375 suitability and its interaction with *I. sexdentatus* suitability had the highest contribution in the
376 first component (35% and 29%, respectively), followed by *I. sexdentatus* suitability (14%),
377 and *T. piniperda* suitability (11%). These four predictors retained 89% of the information
378 within the Component 1. In the second PLS-GLM component (explaining 35% of the
379 variance), the infestation was negatively associated with stand density and basal area, and
380 positively with richness of tree species, *Ips* suitability, and the three-way interaction between
381 Scots pine, *Ips*, and *Tomicus* suitabilities. This component was best explained by the three-
382 way interaction, which alone retained nearly half of the information content (43%), whereas
383 *Ips* suitability, tree species richness, basal area, and stand density retained 16%, 15%, 11%,
384 and 9%, respectively. All these variables together explained 85% of Component 2.



385

386 Figure 3. Plots depicting the significant relationships obtained within the tree-level mortality model. A: Correlation
 387 between tree size (dbh scaled) and predicted probability of tree mortality; B: Correlation between intensity of
 388 attack (number of holes scaled) and predicted probability of tree mortality. Black lines represent prediction of the
 389 model and gray shaded areas represent limit of the confidence intervals. C: Effect of the two-way interaction
 390 between intensity of attack (number of holes scaled) and tree size (maximum and minimum of dbh scaled) on
 391 predicted tree mortality; D: Effect of the two-way interaction between intensity of attack (number of holes scaled)
 392 and Scots pine suitability (maximum and minimum of suitability scaled) on predicted tree mortality. Black and
 393 dashed lines represent predictions of the model and blue shaded areas represent limit of the confidence intervals.
 394 See methods for details on variable standardization.

395 Table 4. PLS-GLM weights estimates of stand-level variables to explain the proportion of dead infested trees. In
 396 bold, significant predictors ($p < 0.05$). R^2 : proportion of the variance in the response variable accounted in the
 397 model by each component; p value: signification of each component obtained in GLM analysis. The coefficient
 398 of non-significant predictors was set to 0.

Variables	Component 1	Component 2
Density (den)	-0.00 (0.00)	-0.30 (0.09)
Basal área (ba)	-0.28 (0.08)	-0.34 (0.11)
Richness of tree species (r)	0.00 (0.00)	+0.38 (0.15)
Spatial pattern (sp)	0.00 (0.00)	0.00 (0.00)
Scots pine suitability (sps)	+0.59 (0.35)	0.00 (0.00)
Ips suitability (ips)	-0.37 (0.14)	+0.40 (0.16)
Tomicus suitability (tos)	+0.33 (0.11)	0.00 (0.00)
sps x ips	+0.54 (0.29)	0.00 (0.00)
sps x tos	-0.19 (0.03)	0.00 (0.00)
Ips x tos	0.00 (0.00)	-0.24 (0.06)
sps x ips x tos	0.00 (0.00)	+0.66 (0.43)
R^2	0.42	0.35
p _value	<0.001 ***	<0.001 ***

Only the final model is shown. AIC = 98.96; Dispersion statistic = 1.98; N = 22.
 Square value of predictor's weights in brackets

399

400 4 Discussion

401 The climatic suitability of both the host tree and bark beetles species, in combination
 402 with tree characteristics and stand structure, explained a large portion of the Scots pine
 403 mortality and bark beetle infestation patterns across our study area. However, we
 404 unexpectedly identified patterns of tree mortality driven by bark beetles inconsistent with
 405 host tree populations close to reaching a beetle eruptive phase (i.e., outbreak). Previously
 406 reported predisposing factors for insect outbreak are related to the quantity (e.g., high stand
 407 basal area and high abundance of host individuals) and quality (e.g., high level of nutrients)
 408 of host trees (Negron and Wilson, 2003; Perkins and Roberts, 2003; Shore et al., 2000).
 409 Nevertheless, it is well known that not all populations of susceptible host trees are infested
 410 with the same severity, suggesting that causes other than those related to the defensive
 411 ability of individual trees (Raffa et al., 2016; Raffa and Berryman, 1983) modulate the
 412 emergence of bark beetle outbreaks. Interestingly, our study shows that the climatic

413 suitability of the host tree may also have a relevant role in tree mortality and insect infestation
414 patterns at regional scales.

415 As regards tree characteristics, Scots pine mortality was related to tree size and
416 intensity of beetle attack in our study but, contrary to our expectations, smaller trees
417 experienced a higher probability of dying at lower values of intensity of attack. This was
418 evidenced by the significant and negative interaction between tree size and intensity of
419 attack. In contrast, the contribution of beetle attack in larger trees was scarcely relevant to
420 any explanation of their death. This conforms to the reported effects under low beetle
421 density, where the attack pattern on small trees is related to incipient bark beetle population
422 levels (Boone et al., 2011). When beetle population density is too low for a cooperative
423 attack, insects exploit weakened trees (Ryan et al., 2015), such as small and suppressed
424 individuals with competitive disadvantages for light interception and belowground water
425 uptake, due to their lower rooting depth (Bravo-Oviedo et al., 2006; Dawson, 1996; Lloret et
426 al., 2004). Furthermore, the defenses in small trees are less substantial than those of larger
427 and more vigorous trees, so the former would require lower levels of attack intensity to be
428 successfully colonized by insects (Raffa et al., 2016). Furthermore, the absence of any
429 significant effect exerted by the distance to the nearest conspecific neighbor supports this
430 interpretation, as well as indicating a scattered pattern of pine mortality.

431 At stand level, infestation patterns and their relationship to stand structure also
432 suggest the lack of an intense bark beetle outbreak in the studied populations. The
433 probability of dead Scots pines being infested was negatively associated with Scots pine
434 density and basal area, and positively related to the richness of tree species. Several studies
435 have reported that dense and monospecific stands are more susceptible to bark beetle
436 infestation (Chapman et al., 2012; Fettig et al., 2007; Perkins & Roberts, 2003), although
437 this occurs when beetle populations reach eruptive dynamics (Raffa et al., 2008). In other
438 words, once the beetle population surpasses a threshold of critical population level they can

439 increase their intensity of attack and select larger and more vigorous trees, despite having
440 better defenses (Krokene et al., 2015). However, if the beetle populations levels are
441 incipient, they preferentially attack and kill trees in stands where the hosts are more
442 susceptible (Bone et al., 2013), as we found at tree level, where small trees exhibited a
443 higher vulnerability. On the other hand, the mechanisms underlying the positive relationship
444 between richness and bark beetle infestation appear to be complex. Baier et al. (2002) found
445 that Norway spruce from mixed stands presented lower resin defenses to deal against beetle
446 attack. In contrast, some studies have suggested that mixed stands reduced the probability
447 of host trees being colonized by bark beetles because other tree species produce an
448 olfactory mask (Zhang, 2001). In our study populations, the absence of a significant
449 relationship between tree spatial pattern and infestation likely reflects the lack of the typical
450 clustered pattern of an eruptive phase (i.e. outbreak) of bark beetle attack. Furthermore, the
451 size of the sampled stands may not have been large enough to include complete clumps of
452 infestation, embedded within the forest matrix (Carrer et al., 2018).

453 It should be noted that a relevant number of dead trees (28%) did not show any signs
454 of bark beetle attack. Although there may be an underestimate of attacks due to the
455 limitations of our visual survey, this value suggests that causes other than insects
456 contributed to tree mortality on the study sites. In effect, recent drought periods have been
457 demonstrated to be a major driver of Scots pine die-off and mortality in the region (Barba et
458 al., 2018; Galiano et al., 2010; Martínez-Vilalta et al., 2012; Vilà-Cabrera et al., 2011). Tree
459 mortality is recognized to be a multifactorial process with positive feedbacks between insect
460 attack and tree weakness due to environmental stress, such as drought (Anderegg et al.,
461 2015; Hartmann et al., 2018). In this context, species climatic suitability in sites where hosts
462 and insects interact acquires special relevance.

463 In opposition to the notion that tree populations are more vulnerable to disturbances
464 when living closer to the species' climatic tolerance limits, our results indicate that Scots pine

465 populations living in more climatically suitable areas were more vulnerable to bark beetle
466 infestation. These remarkable results could be explained by trade-offs between tree growth
467 and resource allocation to resin defenses (Ferrenberg et al., 2015; Lorio, 1986; Stamp,
468 2003). Some studies have reported that Scots pine populations living in wetter, more
469 suitable sites in the Iberian Peninsula exhibit fast-growing trends and higher vulnerability to
470 intense drought (Martínez-Vilalta and Piñol, 2002), whereas resistance to drought appears
471 to be associated with low growth rates in populations of Scots pine living in the south of this
472 region, due to genetic adaptations (Alía et al., 2001). Therefore, a higher pre-drought
473 investment of resources in growth in areas more suitable to the hosts might reduce a
474 subsequent investment in resin defenses against bark beetles, while slow-growing patterns
475 in less suitable localities could allow trees to allocate more resources to defense, with
476 respect to growth rates. This hypothesis concurs with the results of Rigling et al. (2003), who
477 reported that the radial growth of Scots pine is positively associated with wetter and cooler
478 climatic conditions, while the production of resin ducts is stimulated by hot summer
479 temperatures. Moreover, other studies have found that greater tree growth results in fewer
480 resin ducts for defense and more susceptibility to bark beetles (Ferrenberg et al., 2014).

481 Interestingly, infestation in our Scots pine populations was particularly intense on
482 sites where the climatic suitability for *I. sexdentatus* and *T. piniperda* was greater. Therefore,
483 maximum infestation levels were only attained when high suitability values coincided in both
484 the beetle and the host tree. Even though the two studied bark beetle species have different
485 climatic requirements, these results suggest a synergistic interaction between both beetle
486 species when the climatic suitability of the host is high. This line of reasoning concurs with
487 the findings of Amezaga and Rodríguez (1998), who found niche overlaps between *T.*
488 *piniperda* and *I. sexdentatus* on sites where resources were less limited. Moreover, it is
489 known that interspecific competition between beetle species is in many cases minimized by
490 using different parts of a common resource (Paine et al., 1981), allowing them to coexist in

491 the same host population. *Ips sexdentatus* typically infests the branches and upper parts of
492 large trees, and it can potentially kill small trees. In contrast, *T. piniperda* concentrates its
493 attack and brooding near the bole base, where the bark is thicker (Sauvard, 2004). In an
494 eruptive phase, these two species would probably compete for food resources and habitat,
495 reducing coexistence within the same stand (Light et al., 1983; Rankin and Borden, 1991).
496 However, given the likely low densities of the beetle population in our stands, the two
497 species' niche differentiation within the tree may enable them to cohabit in the same host or
498 in trees of a different size, helping to overcome tree defenses and promoting higher
499 infestation levels.

500 **5 Conclusions**

501 This study reveals that the climatic suitability of both host trees and bark-beetle
502 species explain regional patterns of tree mortality and the success of beetle infestation.
503 Species distribution models have proved useful for this purpose by providing indexes of
504 climatic suitability that can be combined to explain population-level processes such as bark
505 beetle infestation. Interestingly, our results evidence a higher vulnerability to bark beetle-
506 driven mortality after drought episodes in tree populations established in climatically suitable
507 sites, particularly when these sites are also suitable for such insects. Furthermore, the
508 relationships recorded between tree characteristics and stand structure with tree mortality
509 and stand infestation suggested attacks by low-density populations of bark beetles in which
510 smaller trees were more vulnerable to the attack. In fact, the stage of the infestation process
511 also appears relevant to interpreting the relationship between the success of an attack and
512 the climatic suitability of both the host tree and the bark beetle.

513 Models of insect-driven mortality that incorporate relationships between the radial
514 growth of Scots pine, the production of resin ducts (Ferrenberg et al., 2014; Rigling et al.,
515 2003), and the density of beetle populations (Colombari et al., 2012; Wermelinger et al.,
516 2008) might be particularly useful for predicting more precisely whether trade-offs of

517 resource allocation are modulating the vulnerability of tree populations to infestation along
518 their range of climatic suitability. Furthermore, the local insect-driven mortality patterns
519 observed in this work suggest, in line with recent studies (Cavin and Jump, 2017; Lloret and
520 Kitzberger, 2018), that tree populations situated closer to the species' central climatic ranges
521 could also be at risk under future climate scenarios. Additional studies should confirm
522 whether mortality patterns in the climatically central versus peripheral tree populations are
523 maintained in areas where a massive attack has occurred, such as recent outbreaks in North
524 America (Meddens et al., 2012; Weed et al., 2013). Overall, our study demonstrates the
525 uncertainty around tree mortality and bark beetle infestation processes in a context of
526 climate change in which drought and warming temperatures are expected to increase in
527 many regions, potentially altering insect-host dynamics. Further studies examining the
528 climatic niche of both the insect and host species are therefore needed to better anticipate
529 the potential effects of these interactions on future forest dynamics.

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