

To die or not to die: early warnings of tree dieback in response to a severe drought

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Running headline: Is drought-induced dieback a critical transition?

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Abstract

1. Some disturbances can drive ecological systems to abrupt shifts between alternative stages (tipping points) when critical transitions occur. Drought-induced tree death can be considered as a nonlinear shift in tree vigour and growth. However, at what point do trees become predisposed to drought-related dieback and which factors determine this (tipping) point? We investigated these questions by characterizing the responses of three tree species, silver fir (*Abies alba*), Scots pine (*Pinus sylvestris*) and Aleppo pine (*Pinus halepensis*), to a severe drought event.
2. We compared basal-area increment (BAI) trends and responses to climate and drought in declining (very defoliated and dying) versus non-declining (slightly or not defoliated) trees by using generalized additive mixed models. Defoliation, BAI and sapwood production were related to functional proxies of tree vigour measured at the onset and end of the drought (non-structural carbohydrate concentrations, needle N content and C isotopic discrimination, presence of wood-inhabiting fungi). We evaluated whether early-warning signals (increases in synchronicity among trees or in autocorrelation and standard deviation) could be extracted from the BAI series prior to tree death.
3. Declining silver fir and Scots pine trees showed less growth than non-declining trees one to three decades, respectively, before the drought event, whereas Aleppo pines showed growth decline irrespective of tree defoliation. At the end of the drought period, all species showed increased defoliation and a related reduction in the concentration of sapwood soluble sugars. Defoliation was constrained by the BAI of the previous five years and sapwood production. No specific wood-inhabiting fungi were found in post-drought declining trees apart from blue stain fungi, which extensively affected damaged Scots pines. Declining silver firs showed increases in BAI autocorrelation and variability prior to tree death.

4. *Synthesis.* Early-warning signals of drought-triggered mortality seem to be species specific and reflect how different tree species cope with drought stress. Highly correlated declining growth patterns during drought can serve as a signal in silver fir, whereas changes in the content of sapwood soluble sugars are suitable vigour proxies for Scots and Aleppo pines. Longer growth and defoliation series, additional vigour parameters and multi-species comparisons are required to understand and predict drought-induced tree death.

Keywords: blue-stain fungi; critical transitions; drought stress; early-warning signals; forest decline; Generalized Additive Mixed Models; mortality; tree death.

1 **Introduction**

2 Forest dieback in response to warming-related drought illustrates some of the challenges that
3 global change poses for ecologists (Allen *et al.* 2010). First, dieback may happen in response
4 to previous constraints, including climate warming (Williams *et al.* 2013), that reduce tree
5 vigour or growth in the long term (predisposing factors *sensu* Manion 1991). Therefore, we
6 need new tools and approaches to quantify those long-term processes acting before drought-
7 triggered mortality occurs (Pedersen 1998). Second, we have a limited (but rapidly growing)
8 knowledge of the key physiological mechanisms responsible for dieback and tree mortality
9 which restrains our ability to identify generalizable properties of the process. Current
10 mechanistic approaches emphasize the interrelationships between carbon metabolism and
11 plant hydraulics (McDowell *et al.* 2008, 2011). It is assumed that drought induces stomatal
12 closure, leading to reduced photosynthesis and carbon starvation, whereas failure to close
13 stomata may cause catastrophic xylem embolism of the vascular system, hydraulic failure and
14 tree death through excessive water loss (Adams *et al.* 2013; Sevanto *et al.* 2014). Third, stress
15 drivers interact and tree responses may be nonlinear. For instance, temperature rise could lead
16 to a nonlinear increase in the growing-season vapour pressure deficit (atmospheric demand
17 for moisture) and, thus, amplify the negative roles of warming and drought on tree mortality
18 (Breshears *et al.* 2013).

19 Warmer temperatures and aridification trends portend increases in regional dieback
20 events and background rates of tree mortality with positive feedbacks on climate warming
21 through the reduction of land carbon sinks and canopy cover, and the loss of ecosystem
22 services (Bonan 2008; Anderegg, Kane & Anderegg 2013). Therefore there is a pressing need
23 to determine whether climate warming and increasing drought stress will increase the
24 likelihood and extent of tree mortality. However, even determining when a tree dies is not an
25 easy task (Anderegg, Berry & Field 2012). Here, we aim to gain a better understanding of

26 drought-triggered tree mortality by considering tree death as a nonlinear threshold of tree
27 vigour and growth (Fig. 1). This theoretical framework states that there are critical transitions
28 or tipping points beyond which ecological systems shift to alternative states (Scheffer *et al.*
29 2001). Such nonlinear thresholds are relatively abrupt and irreversible but it has been
30 proposed that tipping points can be preceded by detectable increases in the variability of the
31 system that can be used as early-warning signals (Dakos *et al.* 2012a, 2012b). For example,
32 climate warming and successive severe droughts should lead to growth decline, thus
33 triggering mortality; however, tree death could also be preceded by an increase in growth
34 persistence (autocorrelation) and variance. Indeed, several tree-ring studies have reported that
35 not only do dead trees usually have lower growth rates but that their growth also showed a
36 high variability and increased responsiveness to water deficit (Pedersen 1998; Ogle, Whitham
37 & Cobb, 2000; Bigler *et al.* 2006). However, even some generally assumed patterns show
38 exceptions because high growth rates in the past may predispose trees to drought-related
39 death (Jenkins & Pallardy 1995; Voltas *et al.* 2013). Hence, we need new approaches for
40 effective comparisons of past radial-growth trends of declining (very defoliated) and dead
41 trees versus non-declining (less defoliated) trees. Here, we retrospectively evaluate whether
42 those growth trends contain early-warning signals of drought-triggered mortality (critical
43 transition), namely low growth rates among declining or recently dead trees, an increase in
44 synchronicity among trees (considered a proxy of sensitivity to drought) and a rise in growth
45 autocorrelation and variability prior to tree death.

46 Using this statistical approach we aim to detect early-warning signals in growth series
47 of declining–dead versus non-declining trees. However, secondary growth is only a proxy of
48 carbon uptake and tree vigour (Fritts 2001). For that reason, we carry out a comprehensive
49 assessment of tree mortality by assessing several variables related to tree vigour (crown
50 defoliation, xylogenesis, concentrations of non-structural carbohydrates in sapwood and

51 needles, needle nitrogen concentrations and carbon isotope discrimination, and fungal
52 infection) at the onset and at the end of the severe drought in 2012, which affected several
53 conifer species across NE Spain. We characterize the responses to drought of three conifers
54 dominating areas subjected to different water deficits, ranging from relatively wet and cool
55 sites with forests of *Abies alba* Mill (silver fir), to continental *Pinus sylvestris* L. (Scots pine)
56 sites and semi-arid *Pinus halepensis* Mill. (Aleppo pine) woodlands. Specifically, we aim to:
57 (i) reconstruct and analyse the past and recent growth trends and responses to drought of co-
58 occurring declining and non-declining trees; (ii) relate those growth patterns with an
59 exhaustive description of vigour proxies that should reflect tree responses to drought-induced
60 dieback. We hypothesize that early-warning signals will be observed in the radial-growth
61 series prior to tree death such as previous growth decline, higher responsiveness to drought
62 and increased persistence and variance. We also hypothesize that drought should lead to a
63 defoliation-related reduction in non-structural carbohydrates in needles and sapwood. Finally,
64 we assess whether declining and dead trees showed symptoms of fungal infection (i.e.
65 whether fungal pathogens played a relevant role as drivers of tree death).

66

67 **Material and methods**

68 *Study site and species*

69 In early 2012, we selected three sites in NE Spain (Aragón) dominated by three different
70 conifer species: silver fir (*Abies alba*), Scots pine (*Pinus sylvestris*) and Aleppo pine (*Pinus*
71 *halepensis*). These sites were characterized by presenting abundant defoliated, dying or dead
72 trees of the dominant species (Supporting Information, Fig. S1). Such dieback coincided with
73 a severe drought that began the previous winter, as indicated by the minimum values of
74 monthly cumulative water deficit recorded during those months (Supporting Information, Fig.
75 S2). The study sites are subjected to Mediterranean and continental climatic conditions but

76 have contrasting water deficits based on climatic data obtained from nearby climatic stations
77 (see a summary of site, climate, soil and forest characteristics in Table 1). Furthermore, the
78 silver fir and Scots pine sites are located near the southernmost (“rear edge”) limit of
79 distribution of the species (Supporting Information, Fig. S3). All the stands studied had not
80 been managed since the 1950s.

81 Silver fir and Scots pine are conifers that are dominant in temperate and continental
82 sites, whereas Aleppo pine is a Circum-Mediterranean conifer better adapted to withstand
83 summer drought in dry and semi-arid areas (Camarero *et al.* 2012). The three study species
84 show isohydric regulation of their water status by avoiding hydraulic failure through rapid
85 stomatal closure, causing a decrease of photosynthetic carbon uptake and potential depletion
86 of carbon reserves (Bréda *et al.* 2006). However, this strict stomatal control or “water-saving”
87 strategy can make isohydric plants that experience prolonged droughts of severe to
88 intermediate intensity more prone to carbon starvation and death (McDowell *et al.* 2008).

89

90 *Climate data and drought index*

91 We obtained monthly climatic variables [mean maximum and minimum temperatures,
92 precipitation (P), estimated potential evapotranspiration (PET) and water balance (P-PET)]
93 from meteorological stations located near the sampling sites for the period 1950–2012 (see
94 details on the climatic stations in Table 1). The time series of the climate variables were
95 carefully controlled and homogenized. The monthly PET was estimated using monthly values
96 of mean temperature, daily temperature range and extraterrestrial solar radiation following the
97 Hargreaves–Samani method (Hargreaves and Samani 1982). To quantify the impact of
98 drought on tree growth we used the multiscalar Standardized
99 Precipitation–Evapotranspiration Index (SPEI; see Vicente-Serrano, Beguería & López–
100 Moreno 2010). The use of the SPEI is relevant to quantify the effects of droughts on radial

101 growth at different time scales (see Fig. 2), as demonstrated in a study conducted in NE Spain
102 (Pasho *et al.* 2011) and using a global dataset (Vicente-Serrano *et al.* 2013). These studies
103 show that during dry periods (negative SPEI values) tree growth declines and that the
104 strongest growth responses to SPEI usually occur at different scales at dry compared with
105 mesic sites. To determine how extreme the 2012 drought was, we determined the return
106 period of the SPEI values according to the probability of the standard normal distribution.

107

108 *Field sampling*

109 At each study site, across a sampling area ranging from 0.5 ha (silver fir, Scots pine) to 2.0 ha
110 (Aleppo pine), we randomly selected 38 trees of the dominant tree species that were at least 5
111 m apart from each other. We mapped (x and y coordinates), tagged and measured size
112 variables in all trees (dbh, diameter at breast height measured at 1.3 m; height). To
113 characterize tree vigour we estimated the percentage of crown defoliation and mistletoe
114 (*Viscum album* L.) abundance (semi-quantitative scale) using binoculars (Dobbertin 2005;
115 Sangüesa-Barreda, Linares & Camarero 2013). Mistletoe was not observed in the Scots pine
116 stand. Given that crown defoliation estimates vary among observers and places, defoliation
117 data were always taken in 5% steps by the first author by comparing every tree with a
118 reference tree with the maximum amount of foliage at each site. Tree defoliation was assessed
119 in March (prior to bud burst) and August (when both primary and radial growth were mostly
120 finished) 2012, referred to hereafter as pre- and post-drought samplings. Trees showing <50%
121 post-drought defoliation were considered to be non-declining, whereas trees with $\geq 50\%$ post-
122 drought defoliation were considered to be declining. The 50%-defoliation criterion
123 represented a robust threshold to differentiate declining from non-declining trees based on
124 recent growth data (Supporting Information, Table S2). Dead trees were regarded as those
125 completely defoliated or having only red needles at the end of the drought. Declining and

126 dead trees were grouped together in further analyses because they showed similar defoliation
127 and growth trends. Given that both pine sites formed open stands we did not try to determine
128 the current competition-intensity index. Furthermore, tree-to-tree competition is not a
129 significant driver of silver-fir decline in Pyrenean sites (Camarero *et al.* 2011).

130

131 *Dendrochronological data*

132 We used dendrochronology to retrospectively characterize growth trends of trees (Fritts
133 2001). Secondary growth was measured by extracting one radial core per tree in late 2012 at
134 1.3 m using a Pressler increment borer. We considered one core sufficient to characterize
135 growth trends; so as to avoid damaging the monitored trees further (Woodall 2008). We
136 measured the length of the sapwood in the extracted cores in the field; in the case of silver fir,
137 we applied stain (bromocresol green) to distinguish the sapwood clearly. We calculated the
138 proportion of the core that was sapwood relative to the radius of the tree. Given the tight
139 relationship between the length of sapwood in the core and the proportion of the core that was
140 sapwood ($r = 0.82-0.90$, $P < 0.001$) we used the former in further analyses because it
141 produced similar results to those obtained using relative values. Wood samples were sanded
142 until tracheids were visible and then visually cross-dated. Once dated, we measured the tree-
143 ring widths to the nearest 0.01 mm using a binocular scope and a LINTAB measuring device
144 (Rinntech, Germany). The accuracy of visual cross-dating was checked with the program
145 COFECHA, which calculates moving correlations between each individual series and the
146 mean site series (Holmes 1983). We successfully dated and measured 35, 37 and 33 silver fir,
147 Scots pine and Aleppo pine trees, respectively. Tree growth measurements were derived by
148 converting the tree-ring width series into basal area increment (BAI) which accounts for the
149 geometrical constraint of adding a volume of wood to a stem of increasing radius (Biondi and
150 Qaedan 2008). We obtained the BAI by using the formula:

151
$$\text{BAI} = \pi (r_t^2 - r_{t-1}^2) \quad (1)$$

152 where r_t and r_{t-1} are the squared stem radial increments at the end and the beginning of a given
153 annual ring increment corresponding to rings formed in years t and $t-1$, respectively. In the
154 case of cores without pith, we estimated the radius by fitting a geometric pith locator to the
155 innermost rings to calculate the length of the missing part of the radius (Duncan 1989).

156

157 *Xylogenesi s*

158 Xylem phenology (xylogenesi s) can be used to assess changes in tree vigour (Gričar, Krže &
159 Čufar 2009). To compare growth potential between declining and non-declining trees we
160 extracted wood microcores from the stem at a height of 1.0 m in 20 trees in spring (late May)
161 when high rates of xylem production occur in the study species (Camarero, Olano & Perras
162 2010; Cuny *et al.* 2012). Trees were randomly selected within the declining and non-declining
163 classes of each species; completely defoliated trees or trees without living needles were
164 excluded. Microcores were preserved in ethanol, transversally cut, and then stained to
165 differentiate xylem cells (radially enlarging, wall thickening and mature tracheids) according
166 to their colour and shape (indicating different developmental phases; for further details see
167 Camarero *et al.* 2010). The three types of tracheids were counted along five radial lines of the
168 2012 ring and mean values for each vigour class were obtained for the three species.

169

170 *Carbohydrate concentrations in sapwood and needles*

171 To assess carbon storage or mobilization we quantified the concentrations of non-structural
172 carbohydrates (NSC) in sapwood and current-year needles in the pre-drought (2011 needles)
173 and post-drought (2012 needles) samplings. Stem sapwood and needles account for the largest
174 proportion of total NSC pools in conifers and could be affected by crown defoliation (Fischer
175 and Höll 1991, Oleksyn *et al.* 2000). All samples were collected between 10.00 h and 13.00 h

176 to avoid diurnal variability in NSC concentrations. Current-year needles were collected from
177 two upper and light-exposed branches from each tree with green foliage. Stem sapwood
178 samples were obtained by taking two radial core samples per tree at 1.3 m using a Pressler
179 increment borer. Needle and wood samples were transported to the laboratory in a portable
180 cooler. Needles were dried in the oven at 60°C for 72 h. Portions of sapwood (outer 5 cm)
181 were separated from each core using a razor blade. Wood samples were subsequently frozen
182 and stored at -20 °C until freeze dried. All dried samples were weighted and milled to a fine
183 powder in a ball mill (Retsch Mixer MM301, Leeds, UK) prior to chemical analyses. Soluble
184 sugars (SS) were extracted with 80% (v/v) ethanol and their concentration determined
185 colorimetrically using the phenol-sulfuric method of Dubois *et al.* (1956) as modified by
186 Buysse and Merckx (1993). Starch and complex sugars remaining after ethanol extractions
187 were enzymatically reduced. NSCs measured after ethanol extraction are referred to as
188 soluble sugars (SS) and carbohydrates measured after enzymatic digestion are referred to as
189 starch. The sum of SS and starch is referred to as NSC.

190

191 *Needle nitrogen and carbon isotope discrimination*

192 We measured needle nitrogen (N) and carbon isotope discrimination ($\delta^{13}\text{C}$) in the post-
193 drought samples. Needle N and $\delta^{13}\text{C}$ were used as surrogates of nutrient use and integrated
194 water-use efficiency (i.e. the amount of carbon gained per unit of water lost) (Farquhar,
195 Ehleringer & Hubick 1989). Samples of current-year needles from all trees with green foliage
196 in late summer were obtained and ground as already described. Needle N concentration was
197 measured using an elemental analyser (VarioMAX, Hanau, Germany); needle $\delta^{13}\text{C}$ was
198 analysed at the Stable Isotope Facility (University of California, Davis, USA) using an
199 isotope-ratio mass spectrometer (Thermo Finnigan MAT 251, Bremen, Germany). The

200 relationship between carbon stable isotopes was expressed in relation to a Pee-Dee Belemnite
201 (PDB) standard. The accuracy of the $\delta^{13}\text{C}$ measurements was 0.05%.

202

203 *Wood-inhabiting fungi*

204 To detect the presence of wood-inhabiting fungi associated with the dieback, we took two
205 core samples per tree from the stump and the upper main root of every tree in autumn 2012
206 using a Pressler increment borer. The borer was carefully sterilized after the extraction of
207 each core using ethanol 70%. Increment cores were placed in sterile plastic tubes, transported
208 to the laboratory within two days of extraction, stored at 4° C and plated out onto selective
209 media within the next three days. Based on observations in previous studies of silver fir
210 decline (Oliva & Colinas 2007), we attempted to isolate the root-rot fungi *Armillaria* and
211 *Heterobasidion* from silver fir cores by plating the cores onto malt-extract-based selective
212 media for basidiomycetes that included benomyl (10 mg l⁻¹) and chloramphenicol (200 mg
213 l⁻¹). Cores were incubated at 20–22°C in the dark and observed weekly for a period of three
214 months. Fungal colonies growing from the cores were transferred onto malt agar plates.
215 Fungal identification was performed by extracting DNA, amplifying the internal transcribed
216 spacer region, sequencing and performing BLAST searches in the GenBank and UNITE
217 databases (Benson *et al.* 2013, Kõljalg *et al.* 2005). We used 98% as a similarity threshold for
218 typing our sequences at species level. The wood samples of both pine species showed signs of
219 discoloration, particularly the Scots pine samples, which showed signs of blue stain fungi.
220 Owing to the potential involvement of blue stain ascomycetes, the selective media onto which
221 the pine cores were plated did not include benomyl. To obtain a quantitative measure of the
222 damage caused by sap stain in Scots pine, we measured the extent of blue stain in the
223 increment core from which we extrapolated the proportion of the section that was colonized
224 (Heiniger *et al.* 2011).

225

226 *Interrelationships between vigour proxies and dieback*

227 We used Structural Equation Models (hereafter SEMs) to explore multiple pathways and
228 proxies related to tree dieback (defoliation, bai decline, NSC, needle chemical variables). In
229 SEMs, researchers may enter information a priori and reformulate the models using goodness-
230 of-fit statistics, which enables the use of both deductive and inductive approaches (Bollen
231 1989). The SEM approach is based on a general linear model and enables the simultaneous
232 assessment of direct and indirect relationships among variables (Grace 2006). SEMs can test
233 hypotheses involving multiple relationships between the studied variables (standardized
234 partial regression coefficients or path coefficients) expressed as structural equations. These
235 equations can be represented in a “path” diagram where the variables are connected by arrows
236 representing the theoretical structural model for the system under consideration. In this study,
237 SEMs were used as a confirmatory statistical method because we tested how our dataset
238 resembled a knowledge-based theoretical model.

239 We developed an a priori conceptual model of dieback (Fig. 3a). We hypothesized that
240 the post-drought defoliation occurred as a result of a decline in growth in previous years (i.e.
241 the mean BAI for 2008–2012 was selected after preliminary analyses and abbreviated as
242 bai5), diminished sapwood production and reduced sapwood or needle NSC concentrations
243 (carbon depletion used to rebuild the crown), which were positively associated. A decline in
244 growth in previous years, sapwood production and wood or needle NSC concentrations could
245 indicate a lower ability of trees to tolerate further drought stress because of a loss in hydraulic
246 efficiency and impaired carbon uptake and transport (Ryan and Yoder 1997; McDowell *et al.*
247 2008). Defoliation could also lead to an increase in needle N due to declining primary growth
248 (concentration effect) and a decrease in needle $\delta^{13}\text{C}$ and water-use efficiency as a
249 consequence of stomata opening and excessive water loss through needles (McDowell *et al.*

250 2008; Galiano, Martínez-Vilalta & Lloret 2011). Scatter plots and Pearson correlation
251 analyses were used to analyse bivariate relationships between the considered variables. The *t*-
252 test was also used to assess the statistical significance of differences between declining and
253 non-declining trees in the study variables. Most of the variables compared followed normal
254 distributions (dbh, height, sapwood area, age, sapwood SS concentrations, no. of expanding
255 tracheids), and those departing from a normal distribution were log-transformed to reach
256 normality (bai 2008–2012, defoliation). A separate SEM was fitted for each species. The
257 overall fit of the model (i.e. the comparison of observed versus model-implied covariances
258 using a maximum likelihood-based minimization process) leads to the selection of parameter
259 estimates and produces a global assessment of the model, here quantified using the χ^2 statistic
260 and its associated probability level (*P*), and the Goodness of Fit Index (GFI). A high *P* value
261 and a GFI > 0.90 indicate an acceptable fit for the model (Grace 2006). We used the Amos
262 software to calculate SEMs (Arbuckle 2010).

263

264 *Growth trends and response to climate in declining and non-declining trees*

265 We used Generalized Additive Mixed Models (GAMM; Wood 2006) to study the long-term
266 BAI trends and responses to drought of declining and non-declining trees of the three study
267 species. GAMM is a semi-parametric method that allows the simultaneous modelling of linear
268 and nonlinear relationships between the response variable (here BAI) and the different
269 covariates by using regression splines. Several GAMM models were fitted and compared.
270 First, we studied the presence of non-linear growth trends and the influence of tree ontogeny
271 (represented by age), size (dbh) and the spatial structure (*x* and *y* coordinates). For each
272 species, we applied a GAMM of the form:

$$273 \quad \log(\text{BAI}_{i+1}) = s(\text{dbh}_i) + s(\text{age}_i) + s(\text{year}) + te(x,y) + Z_i B_i + \nu_i \quad (2)$$

274 where the BAI of a tree i is modelled as smooth functions (s) with different combinations of
275 the four predictors. In addition, given that BAI represents multiple measurements performed
276 in each tree, tree identity ($Z_i B_i$) is regarded as a random effect (Z_i and B_i indicate matrix
277 variables and related coefficients). For dbh, age and year, smooth terms were represented
278 using smooth splines with default package settings, whereas in the case of the spatial
279 coordinates a tensor product (te) spline was used (Augustin *et al.* 2009). Given that BAI has a
280 skewed distribution, we log-transformed this variable ($\log x + 1$). In addition, we also included
281 an error term (v_i) in the model with a first-order temporal autocorrelation [AR(1)] structure.
282 Eight potential models were calculated and ranked according to their second-order Akaike
283 information criterion (AICc). The model with the lowest AICc was selected (Burnham and
284 Anderson, 2002).

285 In a second step, we included the differences in growth between declining and non-
286 declining trees and the potential influence of drought on BAI. Specifically, we considered
287 nine GAMMs by including a factor classifying trees as declining and non-declining, the SPEI
288 June drought index, and potential interactions between these variables and the non-linear
289 trends in tree growth. For each tree species, we fitted a GAMM of the form:

$$290 \log(\text{bai}_{i+1}) = s(\text{dbh}_i) + s(\text{age}_i) + s(\text{year} * \text{declining}) + te(x,y) + \text{drought} * \text{declining} + Z_i B_i + v_i \quad (3)$$

291 which models bai considering different responses to drought in declining and non-declining
292 trees of each species. Again, models were calculated and ranked according to their AICc and
293 the one with the lowest AICc was selected. The GAMMs were performed using the function
294 *gamm* in the *mgcv* package (Wood 2011).

295

296 *Tipping points and tree dieback*

297 We used several statistical techniques to evaluate whether critical transitions can be detected
298 in a growth series before tree dieback occurs. Recent studies have suggested that the

299 proximity of a system or an organism to a tipping point (critical transition to a new state) is
300 preceded by early-warning signals that can be detected using time series analyses (Dakos *et*
301 *al.* 2012a). In particular, mathematical properties such as the AR(1) or the standard deviation
302 (SD) of a time series representing temporal trends of a natural system may vary near to a
303 tipping point and can be used as “early warnings”. In natural systems, slowing down can be
304 an indicator of changes in the system and can be represented by an increase in AR(1) or
305 variance (SD) near the tipping point (Dakos *et al.* 2012b). We also quantified tree-to-tree
306 synchronicity in growth, which should increase in response to drought stress (Fritts 2001).

307 We explored the presence of early warning signals in the growth trends of declining
308 and non-declining trees of the three species studied. First, we studied tree growth
309 synchronicity of declining and non-declining trees of each species. The mean Pearson
310 moment correlation coefficient (r) between all pairs of individual bai series was calculated
311 using moving 30-year-long time windows displaced every year starting in 1950. This was
312 done separately for declining and non-declining trees of the three species as a measure of
313 growth synchronicity within each vigour class. These correlations were plotted against year to
314 detect trends (quantified using r) or sudden changes in growth synchronicity. Second, we
315 quantified the AR (1) and the SD of residuals of the mean bai series of declining and non-
316 declining trees of each species. The mean bai series for each vigour class and species were
317 log-transformed and detrended applying a Gaussian filtering that was half the size of the time
318 period studied (30 years). The mean series of BAI residuals were then analysed using 30-
319 year-long moving windows. The nonparametric Kendall τ statistic was used to analyse the
320 trends of AR(1) and SD because it is a robust statistic against the presence of heterogeneity in
321 a time series. The Kendall τ statistic measures the strength of the trend by comparing the
322 ranks of the time data and the analysed variable reaching maximum and minimum values of
323 +1 and -1, respectively. If the variable increases as time advances then $\tau > 0$, whereas a

324 negative τ value indicates a decreasing trend. The analyses were performed using the
325 functions provided in the *earlywarnings* package (Dakos *et al.* 2012a). All statistical analyses
326 were performed using the R statistical software (R Development Core Team, 2014).

327

328 **Results**

329 *Growth responses to the extreme drought*

330 The 5- and 10-month-long SPEIs calculated in June 2012 for the silver fir, Scots and Aleppo
331 pine sites were, respectively, -2.40 , -1.90 and -1.92 , respectively, corresponding to drought
332 return periods of 122, 35 and 36 years, in that order. Therefore, these low SPEI values
333 represent extreme dry conditions for the period 1950–2012 (Supporting Information, Fig. S4).

334 We were able to successfully cross-date most tree-ring series to determine the highest
335 and lowest mean growth rates in silver fir and Scots pine, respectively (Supporting
336 Information, Table S1). On average, silver fir showed the lowest autocorrelation in radial
337 growth whereas Scots pine had the highest coherence among trees and also the maximum
338 year-to-year variability in ring width. The June SPEI was the climatic variable most tightly
339 related to BAI when considering the three species (Fig. 2, Supporting Information, Fig. S5);
340 however, this association was observed for longer time scales (10 month) at the driest site (*P.*
341 *halepensis*) and for shorter time scales (5 months) at the other two sites (*A. alba*, *P.*
342 *sylvestris*).

343

344 *Differences between declining and non-declining trees*

345 Declining–dead and non-declining trees only differed in terms of crown defoliation, sapwood
346 production, post-drought sapwood SS concentration and bai5, with significantly higher values
347 for non-declining trees except in the case of defoliation (Table 2). The rest of the analysed
348 variables did not show any significant difference between vigour classes (results not shown).

349 Overall, all trees produced a low number of expanding tracheids, indicating a low growth
350 potential except for the non-declining silver firs. In spring, a significantly lower number of
351 expanding tracheids were produced by declining trees compared with the non-declining trees
352 in all species (Table 2). The other tracheid types were not significantly different (results not
353 shown). Declining trees showed a greater response to water balance during the critical months
354 for growth (spring, previous autumn) than the non-declining trees (Supporting Information,
355 Fig. S5)

356

357 *Drivers and proxies of defoliation*

358 In all species, the strongest negative associations were observed between sapwood
359 production, post-drought sapwood SS concentration and defoliation (Figs. 3 and 4). Indeed,
360 pre-drought defoliation and sapwood SS abundance were also inversely related except in the
361 case of Scots pine trees (Fig. 4). We found higher post-drought than pre-drought sapwood SS
362 concentrations in the case of silver fir and Aleppo pine; however, these differences were not
363 significant (silver fir, $F = 3.07$, $P = 0.08$; Aleppo pine, $F = 0.74$, $P = 0.39$). The selected
364 SEMs, which accounted for more than 50% of the variance in tree defoliation, indicated that
365 the effect of sapwood on defoliation was direct in the case of silver fir (Fig. 3b), whereas in
366 Aleppo pine, sapwood had an indirect effect on defoliation through reduced sapwood SS
367 concentrations, and in Scots pine defoliation was mainly driven by sapwood SS (Fig. 3c and
368 3d). Previous BAI (bai5) was the second most important variable explaining current
369 defoliation after sapwood variables, with a significant negative effect in both pine species but
370 not in silver fir. Sapwood production was significantly and positively related to bai5 in all
371 species except Aleppo pine.

372 In Scots pine, the post-drought defoliation was positively related to needle SS
373 concentrations ($r = 0.34$, $P = 0.04$) but negatively associated with needle starch

374 concentrations ($r = -0.57$, $P < 0.001$). In Scots and Aleppo pines, post-drought defoliation and
375 needle N concentration were positively ($r = 0.34$, $P = 0.03$) and negatively ($r = -0.32$, $P =$
376 0.04) related, respectively. Defoliation was unrelated to needle $\delta^{13}\text{C}$, which only showed a
377 positive association with post-drought sapwood starch concentration in Aleppo pine ($r = 0.37$,
378 $P = 0.03$). In spite of these aforementioned relationships, no needle chemical variable (NSC,
379 N and $\delta^{13}\text{C}$) entered into the selected SEMs.

380

381 *Role of forest pathogens*

382 Scots pine was predominantly damaged by blue-stain fungi. The proportion of the blue
383 stained area at stump level was positively related to post-drought defoliation ($r = 0.54$, P
384 < 0.001) and negatively associated with sapwood SS concentrations (Fig. 5a). We could not
385 isolate the causal agent from all affected trees; however, four trees appeared to be infected by
386 the blue stain fungus *Ophiostoma minus* (Hedgcock) Sydow et P. Sydow. Other fungi isolated
387 from Scots pine wood were *Umbelopsis isabellina* (Oudem.) W. Gams and *Acremonium*
388 *strictum* W. Gams. In the case of Aleppo pine, the heart-rot fungus *Porodaedalea pini* (Brot.)
389 Murrill was the predominant pathogen, although no association with defoliation or carbon
390 reserves could be found. In the case of silver fir, the dominant pathogen was *Amylostereum*
391 *chailletii* (Pers.) Boidin, which was isolated from five dying silver fir trees. Other pathogens
392 such as *Heterobasidion annosum* (Fr.) Bref. s.l., *Ganoderma lucidum* (Curtis) P. Karst. and
393 *Trichaptum abietinum* (J. Dicks.) Ryvarden were only isolated from two trees each.
394 *Amylostereum*-infected trees had lower starch and SS sapwood concentrations than non-
395 declining trees in the pre- and post-drought periods, respectively (Fig. 5b).

396

397 *Growth trends*

398 The GAMMs revealed different BAI trends for the three species studied (Fig. 6, Table 3). The
399 non-linear trend was present in the models of all three species (Fig. 6), whereas tree size (dbh)
400 was only selected in the models of Scots pine and silver fir, and ontogeny (tree age) was only
401 significant in the case of Scots pine (Table 3). The spatial location of trees did not have an
402 influence on BAI trends (Supporting Information, Table S3).

403 The BAI trends of declining and non-declining trees started to diverge in the mid-
404 1990s in the case of silver fir and in the 1980s in the case of Scots pine (Fig. 6). In Aleppo
405 pine, both declining and non-declining have shown a growth loss since the 1980s. Indeed,
406 only in this species were defoliation and the year of cambial death (the last formed tree-ring)
407 significantly associated ($r = -0.59$, $P = 0.001$), indicating the long-term decline of these trees.
408 For example, the most defoliated trees stopped growing on average seven years before the
409 sampling year, whereas the least defoliated trees only stopped growing two years before the
410 sampling. The second selected GAMM, incorporating different BAI responses of declining
411 and non-declining trees to drought stress, demonstrated that drought is an important driver of
412 long-term BAI trends for the three study species (Fig. 6, Table 4). The models revealed that
413 the BAI response to drought in declining and non-declining trees of both pine species differed
414 but did not differ among silver fir trees (Table 4), suggesting that factors driving growth
415 decline are species specific.

416

417 *Early-warning signals of tree dieback and death*

418 The analyses of BAI performed to detect early warnings of tree dieback showed contrasting
419 results depending on the species (Fig. 7). In silver fir, there was a coherent inflection point in
420 bai data indicating a rise in synchronicity among trees, AR(1) and SD after the 1995 drought.
421 These increases in synchronicity, autocorrelation and variance were higher for declining than
422 for non-declining trees even though the trends of those variables were negative for the whole

423 study period. A similar pattern was observed in Aleppo pine with increases in synchronicity,
424 AR(1) and SD following the 2005 drought, particularly in the case of declining trees. By
425 contrast, declining Scots pines showed a loss in synchronicity and decreases in AR(1) and SD.

426

427 **Discussion**

428 *Implications on the use of early-warning signals in growth series to portend die-off*

429 We found mixed results concerning the trees' growth responses to drought. Early-warning
430 signals in the growth series of declining trees prior to their death were detected in silver fir
431 (Fig. 7). Growth trends of declining and non-declining trees diverged in all species except
432 Aleppo pine (Fig. 6). First, declining Scots pine trees have shown a lower growth rate relative
433 to non-declining trees since the 1980s, which was a warm and dry decade (Fig. 6). In this
434 species, the low growth rates and high defoliation and blue-stain levels (Table 2, Fig. 5a), the
435 high levels of synchronicity in growth among trees and responsiveness to drought, and the
436 lack of clear early-warning signals suggest that the whole Scots pine study population is
437 experiencing a chronic drought-induced decline and may be prone to disappear. This could be
438 a paradigmatic case of a warming-induced ongoing local extinction of a rear-edge population
439 triggered by drought. Second, in Aleppo pine, growth trends declined irrespective of recent
440 tree defoliation (Figs. 2 and 6). Cambial death was common in this stand during the past
441 decade which implies an active selection of drought-tolerant non-defoliated trees. Cambial
442 death (i.e. when tree-ring production ceases) precedes total crown defoliation by several years
443 or even decades (Pedersen 1998). Therefore, the extensive growth decline of Aleppo pines
444 portends increasing defoliation and mortality levels in the forthcoming years. Third, the silver
445 fir growth series offered the best opportunity to use early-warning growth indicators of
446 dieback. In this species the divergence between vigour classes coincided with a severe
447 drought in 1995 (Fig. 6) and a sharp increase in autocorrelation and variability of all trees,

448 particularly declining trees (Fig. 7). Given that declining and non-declining silver firs showed
449 similar growth responses to drought, this population could also be experiencing a long-term
450 decline characterized by the removal of drought-intolerant defoliated individuals. Several
451 Pyrenean silver fir populations subjected to sub-optimal conditions are experiencing ongoing
452 decline and high mortality rates as temperatures rise and atmospheric water demand increases
453 (Camarero *et al.* 2011). Overall, our findings concur with previous studies showing how
454 growth is nonlinearly impacted by drought (Cavin *et al.* 2013). Such threshold responses in
455 growth and vigour suggest that warmer air temperatures may exacerbate drought stress in the
456 three study species by amplifying atmospheric water demand or the duration of the water-
457 deficit period (Peguero-Pina *et al.* 2007, Linares & Camarero 2012, Alexou 2013, Poyatos *et*
458 *al.* 2013).

459 Relative growth rates and short-term growth trends prior to tree death have been
460 shown to be reliable predictors of timing of tree death in Swiss *Picea abies* (L.) Karst.
461 (Norway spruce) forests (Bigler & Bugmann 2004) and in German *Fagus sylvatica* L. (beech)
462 stands (Gillner *et al.* 2013) among others. Here we show that considering autocorrelation and
463 variability prior to tree death improves our knowledge of the already complex mortality
464 processes in silver fir and Aleppo pine. We argue that a broad comparison of a long-term
465 growth series prior to death between dead, declining and non-declining trees of several tree
466 species from contrasting biomes would give a better insight about mortality patterns and
467 processes. The evaluation of tree death as a tipping point in vigour could profit from the
468 theoretical framework offered by ecological early-warning growth signals of increased
469 synchronicity and critical slowing down (e.g., Boden *et al.* 2014).

470 It is possible that our hypothesis referred to a defoliation-related reduction in needle
471 and sapwood NSC concentrations (Figs. 3 and 4). At the end of the drought period, all species
472 showed a defoliation-related reduction in the concentration of sapwood SS (Fig. 4). We also

473 observed increasing post-drought sapwood SS concentrations in the case of Aleppo pine, and
474 particularly silver fir, irrespective of tree defoliation (Fig. 4). Apparently, this could contradict
475 the isohydric behaviour of these species, which rapidly close stomata and reduce carbon
476 uptake to avoid hydraulic failure. However, we interpret such accumulation of sapwood SS as
477 a response to drought-induced declining sink activity (Sala *et al.* 2010), which would be most
478 relevant in silver fir. In addition, an increased concentration of sugars, derived from starch
479 stored in parenchyma cells, has been linked to the refilling of embolized vessels (Secchi &
480 Zwieniecki 2012). Interestingly, previous growth and sapwood production predetermined the
481 final levels of defoliation, which can be regarded as a multiannual process (Fig. 3). Given that
482 SS are among the most drought-responsive metabolites, their concentration changes could
483 reflect either an impairment of photosynthetic carbon uptake or a reduction in starch
484 hydrolysis. In Scots pine, the concentration of SS in needles increased and starch decreased as
485 response to defoliation, suggesting an active release of these molecules, which could represent
486 an early-warning stress signal to trees given that SS correlate negatively with leaf
487 photosynthesis (Franck *et al.* 2006) and act as osmoprotectants during water deficit stress
488 (Rodriguez-Calcerrada *et al.* 2011). We speculate that the general reduction of sugars in the
489 stem sapwood could be linked to an increase in the abundance of these metabolites in the
490 needles; however, but this association was only detected in Aleppo pine ($r = -0.39$, $P = 0.02$).
491 Thus, the negative effect of drought on sugar concentrations can have multiple and complex
492 effects and does not necessarily imply a link between tree death and carbon starvation (Sala,
493 Piper & Hoch 2010; Dietze *et al.* 2014). For instance, rapid death due to hydraulic failure may
494 be associated with a lack of adequate tissue carbohydrate content required for osmoregulation
495 owing to shifts in carbon allocation priorities or phloem dysfunction (Sevanto *et al.* 2014). In
496 addition, drought-triggered defoliation can lead to carbon depletion of roots in conifers
497 (Oleksyn *et al.* 2000; Hartmann, Ziegler & Trumbore 2013) while above-ground tissues use

498 sugars for osmoregulation or to rebuild the crown (Saffell *et al.* 2014). Overall, our findings
499 confirm that the sapwood concentrations of SS reflect crown defoliation in the two pine
500 species whereas needle chemical variables (NSC, N, $\delta^{13}\text{C}$) do not.

501 In Scots pine, the decline in sapwood sugar content was not only linked to crown
502 defoliation but also to the abundance of blue-stain fungi in the wood (Fig. 5a). These
503 symptoms confirm that this population of trees was in decline and highlight the value of
504 combining defoliation and incidence of blue-stain fungi as early-warning indicators of forest
505 decline. Blue-stain fungi may be important contributing factors (*sensu* Manion 1991) involved
506 in the long-term decline of Scots pine populations, as demonstrated by Heiniger *et al.* (2011)
507 and theoretically proposed in Oliva, Stenlid & Martínez-Vilalta (2014). In silver fir, we could
508 not associate the susceptibility of trees to drought stress with previous fungal infection
509 because few trees were infected with the same pathogen species (Fig. 5b). Given that
510 pathogens take a long time to kill a tree and the diversity of pathogens that attack trees, future
511 studies should use broader sampling designs than the one used in this study. For
512 instance, the presence of *H. annosum* and *Armillaria* spp. has not only been linked to dieback
513 but also to thinning in Pyrenean silver fir stands (Oliva & Colinas 2007; Oliva, Suz & Colinas
514 2009). Unlike previous findings in *Pinus mugo* Turra infected by *H. annosum*, which died
515 rapidly, possibly because of fungal attack (Cherubini *et al.* 2002), the declining silver firs in
516 this study, with or without fungal infection, presented diverse growth patterns.

517

518 *Species-specific responses of tree species to drought-induced die-off*

519 Statistical rarity was one aspect of the severe 2012 drought, which was characterized by low
520 levels of precipitation, a long dry period and also high temperatures. However, such an
521 unusual climatic episode was also an ecological extreme event (*sensu* Smith 2011) given that
522 the trees' responses were outside the typical or normal variability of the system. Our study

523 can be described as “opportunistic” (*sensu* Smith 2011) since we evaluated the trees’
524 responses to a climate extreme (2012 drought) during the course of an ongoing observational
525 study. Using a retrospective approach as we did can still capture large spatial and temporal
526 scale information, even though we cannot control for characteristics of climatic extremes
527 (type, timing, magnitude) or for other interacting factors (e.g., previous droughts). The length
528 of the 2012 drought exceeded the ability of the trees to tolerate drought stress leading to long-
529 term impacts such as ongoing defoliation, growth decline, mortality and changes in the
530 abundance of the dominant tree species. Such drought-triggered community shifts will have
531 cascading effects on forest composition and ecosystem services. Early-warning signals such
532 as an increase in growth persistence and variability were useful as death forecasts for silver fir
533 (Fig. 7), whereas the two pine species presented other dieback prognoses and responses to
534 defoliation (decrease in sapwood production and SS concentrations, blue-stain fungi). Future
535 dieback assessments could focus on those idiosyncratic proxies to determine the vulnerability
536 and resilience of forests in response to warming and drought stress. We suggest that more
537 attention should be focussed on the interactions between growth, defoliation and sapwood
538 function (including NSC use and transport) to forecast dieback and mortality as a function of
539 drought stress using process-based models. Finally, these simulation exercises could be scaled
540 up to the ecosystem level by using, for instance, ecohydrological models to evaluate the
541 extent to which drought-induced dieback and the resulting loss of canopy cover lead to:
542 decreased evapotranspiration and carbon sink potential, increased streamflow, and altered
543 water yield as a function of site aridity and post-mortality successional dynamics.

544

545 **Acknowledgements**

546 This study was supported by projects CGL2011-26654 (Spanish Ministry of Economy and
547 Competitiveness), 1032S/2013 and 387/2011 (Organismo Autónomo Parques Nacionales,

548 Spanish Ministry of Agriculture and Environment, Spain). We thank ARAID for supporting
549 J.J.C. and the AEET for providing climatic data. We also thank M. Maestro and E. Lahoz for
550 performing the chemical analyses and R. Hernández, A.Q. Alla and E. González de Andrés
551 for their help in the field.

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746

748 **Table 1. Main features of the three study sites and tree species. Values are means \pm SE.**

Site	Paco Ezpela	El Carrascal	Vedado de Peñaflo
Species	<i>Abies alba</i> Mill.	<i>Pinus sylvestris</i> L.	<i>Pinus halepensis</i> Mill.
Latitude (N)	42° 45'	40° 26'	41° 47'
Longitude (W)	0° 52'	0° 58'	0° 44'
Elevation (m a.s.l.)	1230	1303	375
Aspect	NE	NW	NE
Slope (°)	35	25	5
Soil type, soil texture	Cambisol, loamy	Cambisol, loamy	Regosol, loamy
Soil pH	7.1	7.7	8.1
Annual water balance (mm) ¹	531	-210	-468
No. trees	38	38	38
No. dead trees pre- / post-drought ²	8 / 16	7 / 25	4 / 7
No. non-declining / declining trees ²	23 / 12	13 / 24	19 / 14
No. trees with tree-ring data	35	37	33
Distance to nearest sampled tree (m)	10.6 \pm 1.5	8.5 \pm 1.3	29.2 \pm 4.8
Dbh (cm) ³	37.3 \pm 1.2	27.3 \pm 1.0	32.3 \pm 1.4
Height (m)	24.0 \pm 0.5	8.5 \pm 0.3	7.8 \pm 0.3
Age (years) ⁴	85 \pm 3	138 \pm 6	94 \pm 4
Crown cover (%)	59.9 \pm 5.9	58.0 \pm 4.9	55.4 \pm 4.7
Sapwood area (%)	15.2 \pm 2.2	8.6 \pm 1.2	17.0 \pm 2.8
Frequency of trees severely infested by mistletoe (%) ⁵	29	0	29
Basal area (m ² ha ⁻¹)	18	15	8
Tree-ring width for period 1950–2012 (mm)	2.00 \pm 0.02	0.40 \pm 0.01	0.85 \pm 0.02
Main co-occurring tree species	<i>Fagus sylvatica</i> L.	<i>Pinus nigra</i> subsp. <i>salzmannii</i> (Dunal) Franco	<i>Juniperus thurifera</i> L.
Climatic station (distance to sampling site in km)	Jaca, 42° 34' N, 0° 33' W, 818 m (30)	Teruel, 40° 21' N, 1° 06' W, 915 m (15)	Zaragoza, 41° 39' N, 0° 53' W, 208 m (24)

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750 ¹The water balance is defined as the difference between precipitation and potential evapotranspiration.751 ² Dead trees were regarded as those completely defoliated or having only red needles at the end of the drought.752 Non-declining trees were considered those presenting post-drought defoliation of < 50%; declining trees were
753 considered those presenting post-drought defoliation of \geq 50%. These trees were counted after the 2012 drought.754 ³Diameter at breast height (dbh) measured at 1.3 m.755 ⁴Age estimated at 1.3 m.756 ⁵Trees with abundant mistletoe individuals growing in at least one-third of the crown.

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760 **Table 2. Comparison of relevant variables calculated for declining and non-declining**
761 **trees in the three study species and based on *t* tests (significant *t* values with *P* < 0.05 are**
762 **in bold; negative *t* values indicate higher value of variables for declining than for non-**
763 **declining trees and vice versa). Values are means ± SE.**
764

Species	Variable	Declining trees	Non-declining trees	<i>t</i>	<i>P</i>
<i>Abies alba</i>	Dbh (cm)	26.97 ± 1.45	27.70 ± 1.33	0.35	0.730
	Height (m)	8.62 ± 0.48	8.58 ± 0.42	-0.05	0.962
	Defoliation (%)	78.23 ± 4.41	21.46 ± 2.26	-12.75	<0.0001
	Sapwood length (cm)	3.17 ± 0.33	5.91 ± 0.35	2.09	0.045
	Age (years)	146 ± 11	133 ± 7	-0.97	0.336
	Post-drought sapwood SS (%)	0.29 ± 0.02	0.39 ± 0.01	2.08	0.046
	BAI 2008–2012 (cm ²)	1.65 ± 0.37	3.51 ± 0.41	2.97	0.005
	No. expanding tracheids	3 ± 1	10 ± 2	6.88	<0.0001
<i>Pinus sylvestris</i>	Dbh (cm)	36.93 ± 1.89	36.36 ± 1.64	-0.21	0.833
	Height (m)	24.13 ± 0.97	23.41 ± 0.66	-0.62	0.540
	Defoliation (%)	87.92 ± 5.82	18.83 ± 2.38	-13.00	<0.001
	Sapwood length (cm)	2.50 ± 0.60	7.70 ± 0.68	5.02	0.001
	Age at 1.3 m (years)	91 ± 4	88 ± 3	-0.51	0.615
	Post-drought sapwood SS (%)	0.21 ± 0.02	0.30 ± 0.01	4.22	0.001
	BAI 2008–2012 (cm ²)	3.41 ± 0.78	12.02 ± 1.33	4.44	0.001
	No. expanding tracheids	3.5 ± 0.8	4.1 ± 1.9	1.51	0.136
<i>Pinus halepensis</i>	Dbh (cm)	29.03 ± 2.70	34.12 ± 1.57	1.73	0.094
	Height (m)	7.56 ± 0.55	7.95 ± 0.35	0.63	0.535
	Defoliation (%)	70.14 ± 4.39	26.21 ± 3.74	-7.63	<0.001
	Sapwood length (cm)	4.57 ± 0.73	6.68 ± 0.68	2.09	0.045
	Age at 1.3 m (years)	78 ± 5	93 ± 6	1.83	0.072
	Post-drought sapwood SS (%)	0.63 ± 0.06	0.79 ± 0.04	2.24	0.033
	BAI 2008–2012 (cm ²)	0.12 ± 0.08	1.20 ± 0.41	2.20	0.036
	No. expanding tracheids	2.6 ± 0.9	3.5 ± 1.7	1.56	0.123

765 BAI, basal area increment; dbh, diameter at breast height; SS, soluble sugars.

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768 **Table 3. Models selected to explain yearly log-transformed basal area increment [log**
769 **(BAI+1)] of declining and non-declining trees as a function of smoothing terms (s) and**
770 **interactions (*). Two models are shown for each species. The first model was fitted by**
771 **selecting the best structural model, including only trend, tree diameter (dbh) and age**
772 **and spatial coordinates. Eight potential models were compared and the best one was**
773 **selected according to the Δ AIC (difference in AIC between the best and the second-best**
774 **model). The second model was fitted including the best first model, the influence of**
775 **summer drought (June SPEI calculated at 5- and 10-month-long scales) and the**
776 **potential difference in growth, trend and response to drought of non-declining vs.**
777 **declining trees. Nine potential models were compared and the best one was selected**
778 **according to the Δ AIC (difference in AIC between the best and the second-best model).**
779 **The adjusted R^2 of each model is also shown. The non-linear and linear coefficients of**
780 **the second-best model are shown in Tables S2 and 4 respectively. The adjusted bai of the**
781 **second-best model is shown in Figure 6.**

782

Species	No. rings	Models	AIC	Δ AIC	Adj R^2
<i>Abies alba</i>	2142	s(year) + s(dbh)	373	3.8	0.39
		s(year * declining) + s(dbh) + declining + June SPEI ₅	320	1.3	0.48
<i>Pinus sylvestris</i>	2327	s(year) + s(dbh) + s(age)	2397	3.9	0.52
		s(year * declining) + s(age) s(dbh)+ declining * June SPEI ₅	2083	1.6	0.58
<i>Pinus halepensis</i>	1974	s(year)	1286	1.1	0.32
		s(year) + declining * June SPEI ₁₀	698	2.0	0.39

783 AIC, Akaike information criterion; dbh, diameter at breast height; GAMM, generalized
784 additive mixed models; SPEI, standardized precipitation–evapotranspiration index.

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789 **Table 4. Linear terms of the second-best model (see abbreviations and symbols for**
790 **variables in Tables 2 and 3). For each species the associated coefficient of each linear**
791 **predictor (SE), the *t*-statistic and its probability (*P*) value are shown.**

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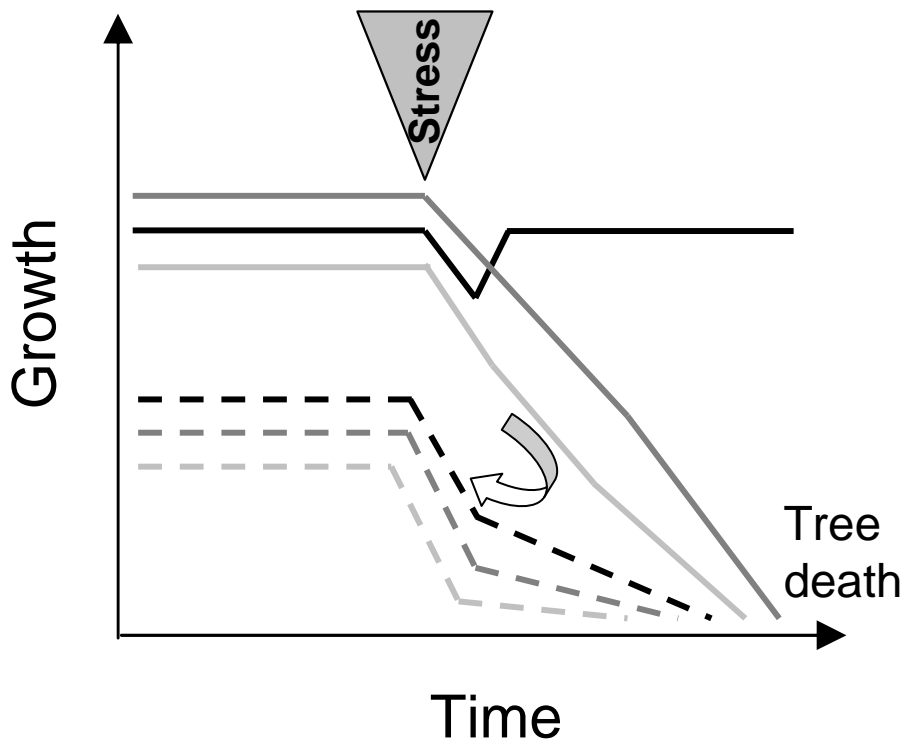
Species	Model	Coefficient	SE	<i>t</i>	<i>P</i>
<i>Abies alba</i>	Non-declining vs. declining	0.17	0.09	1.96	0.048
	June SPEI ₅	0.03	0.01	5.12	<0.001
	Non-declining * June SPEI ₅	–	–	–	–
<i>Pinus sylvestris</i>	Non-declining vs. declining	0.26	0.14	1.84	0.053
	June SPEI ₅	0.13	0.01	9.58	<0.001
	Non-declining * June SPEI ₅	0.03	0.02	1.89	0.059
<i>Pinus halepensis</i>	Non-declining vs. declining	0.11	0.07	1.53	0.127
	June SPEI ₁₀	0.12	0.01	15.55	<0.001
	Non-declining * June SPEI ₁₀	0.02	0.01	2.23	0.026

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795 **Figures**

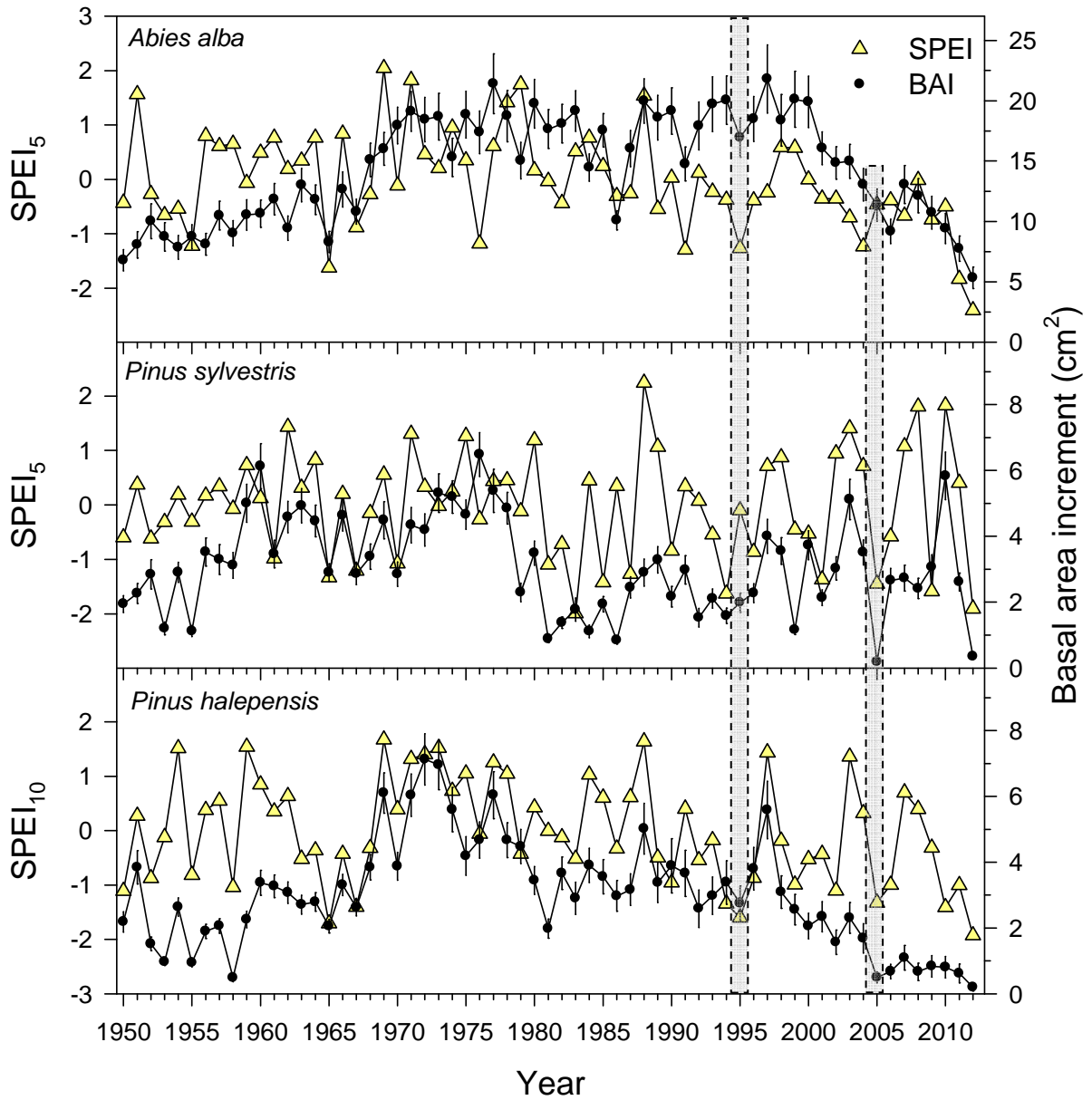
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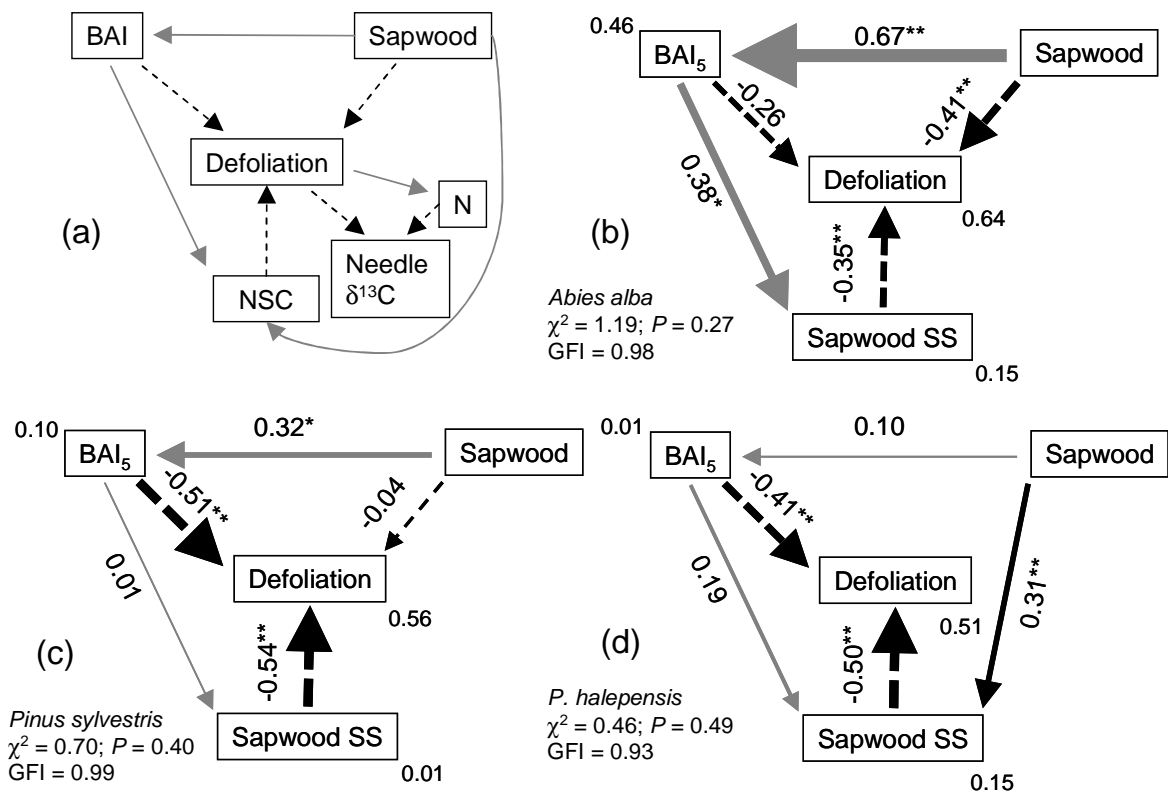
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799 **Figure 1.** A conceptual model of tree mortality based on the tipping-point concept where the
800 arrow indicates a post-drought critical transition (tipping point) from non-declining to
801 declining growth trajectories in response to a stress factor such as a drought (modified from
802 Pedersen 1998 and Manion 1991). Solid lines represent non-declining trees; broken lines
803 represent declining trees. Both types of tree can die (no growth) showing diverse rates and
804 paces of growth reduction after a stress event (inciting factor *sensu* Manion 1991) such as a
805 severe drought. The arrow indicates a post-drought critical transition from non-declining to
806 declining growth trajectories.



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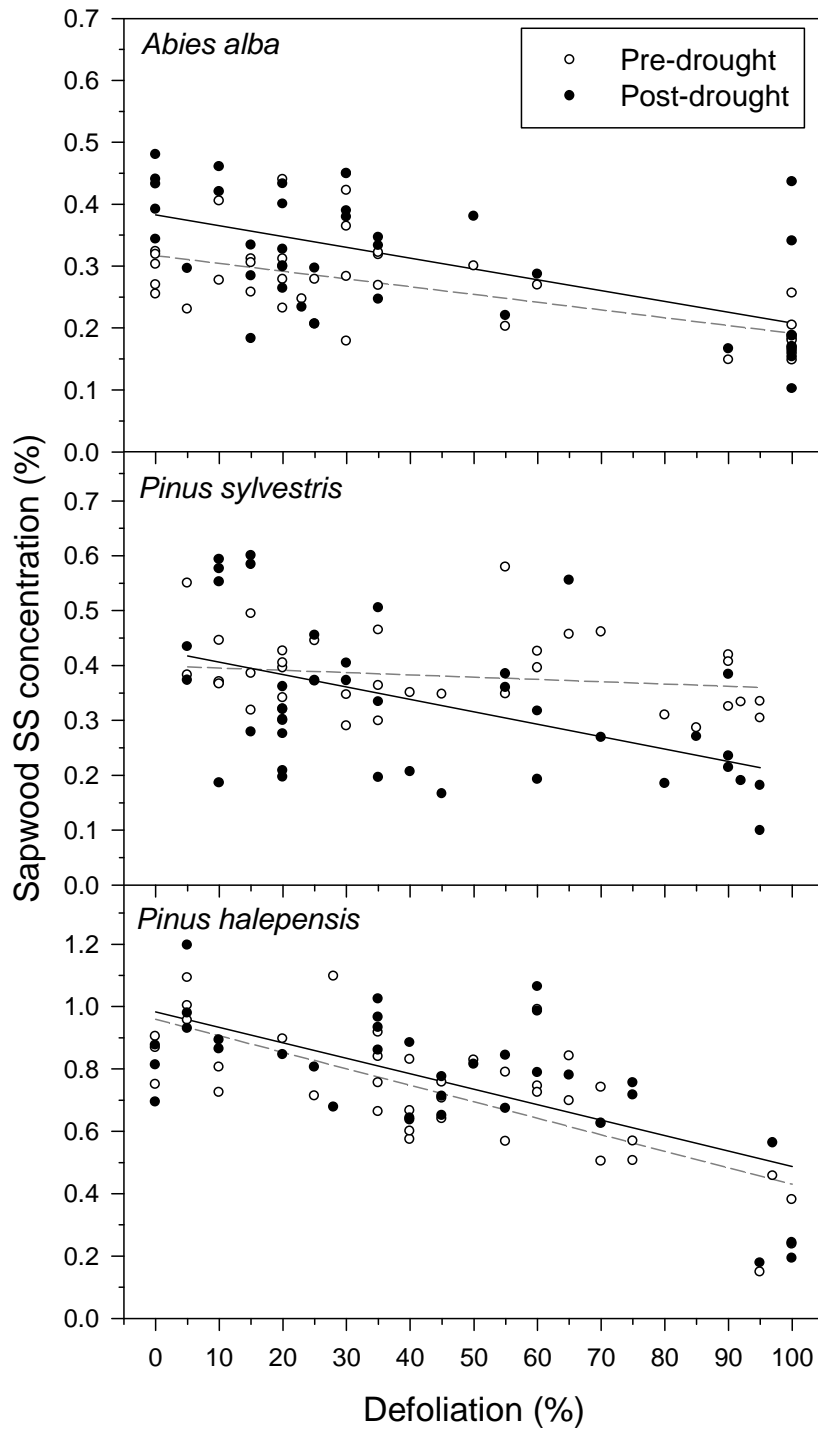
808 **Figure 2.** Growth patterns (mean basal area increment \pm SE) of the three study species as
 809 related to drought severity: June Standardized Precipitation–Evapotranspiration Index
 810 calculated at scales of 5 (SPEI₅) for *A. alba* and *P. sylvestris* or 10 months (SPEI₁₀) for *P.*
 811 *halepensis*. The vertical grey boxes indicate the severe 1995 and 2005 droughts.



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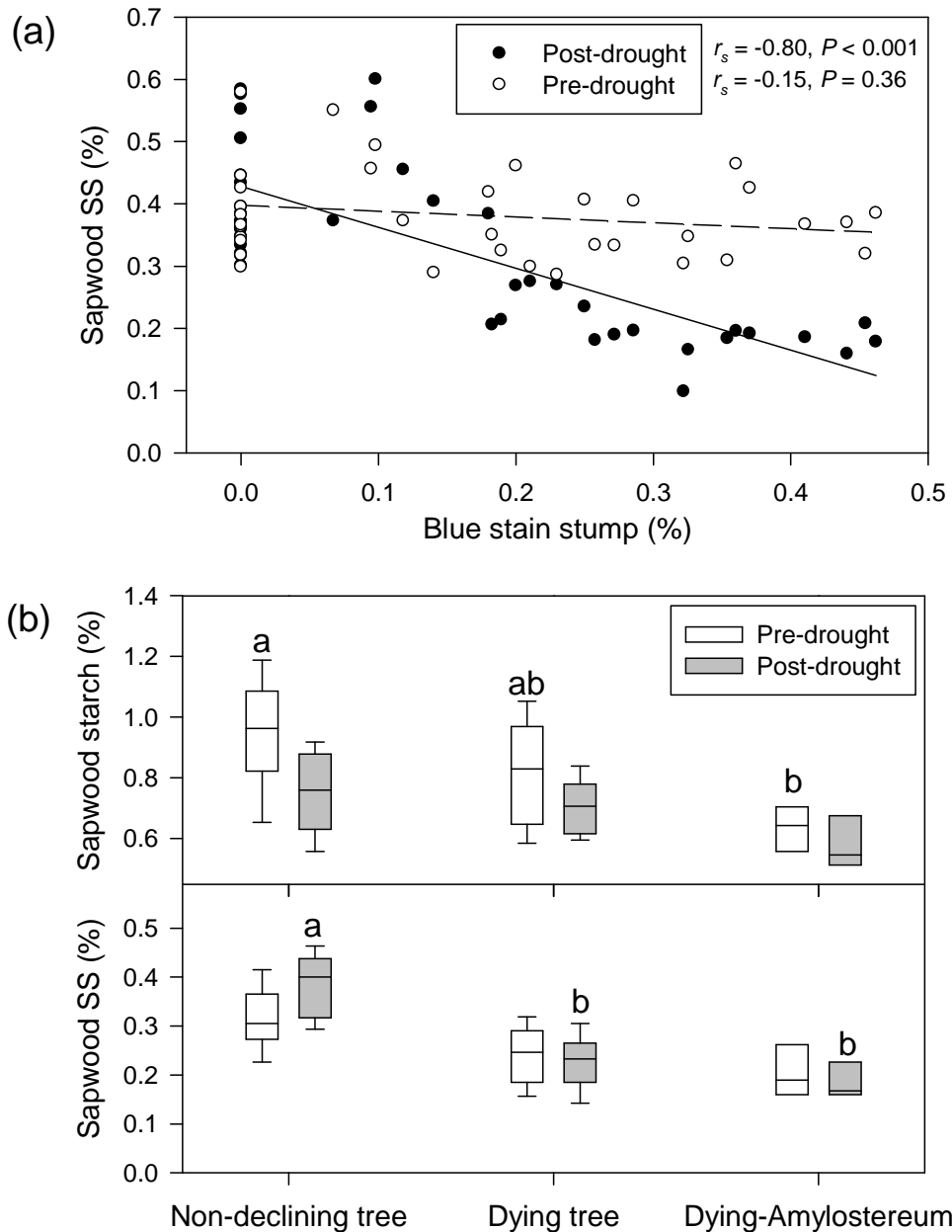
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815 **Figure 3.** Conceptual functional model of tree decline (a) relating different variables (BAI₅,
 816 basal area increment of the previous five years; sapwood length; N, needle N concentration;
 817 $\delta^{13}\text{C}$, needle carbon isotopic discrimination; NSC, total non-structural carbohydrate
 818 concentrations in sapwood or needles) and structural equation models (SEM) fitted to silver
 819 fir (*Abies alba*) (b), Scots pine (*Pinus sylvestris*) (c) and Aleppo pine (*Pinus halepensis*) (d)
 820 trees. The broken and solid arrows indicate negative and positive relationships (significance
 821 levels are * $P < 0.05$ and ** $P < 0.01$). Arrow sizes are proportional to the magnitude of
 822 standardized path coefficients (numbers located above arrows). Explained variance of each
 823 observed variable is indicated in corners near boxes. In each SEM, three fitting statistics are
 824 shown in the lower left inset (χ^2 statistic and its corresponding P value; GFI, Goodness of Fit
 825 Index). Note that some variables shown in the conceptual model were not selected in the fitted
 826 models or were only available for some of the study species (e.g. mistletoe abundance or
 827 presence of fungi).



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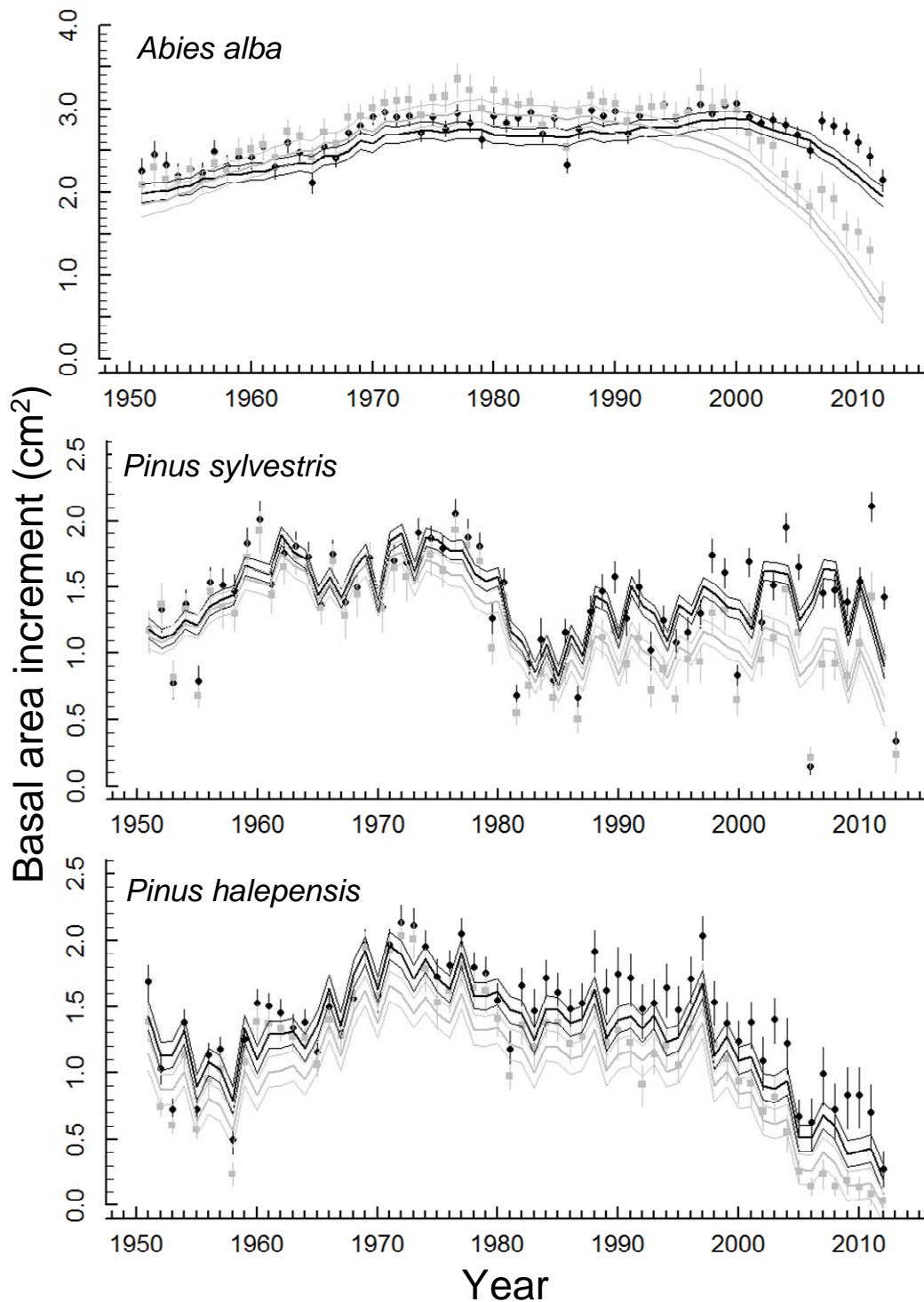
Figure 4. Negative associations observed for the three study species between tree defoliation and the concentration of soluble sugars (SS) in the sapwood before and after a severe drought event (see also Fig. 2). All relationships are highly significant ($P < 0.01$) apart from the one corresponding to the pre-drought period in Scots pine ($r = -0.18$, $P = 0.28$).



836

837 **Figure 5.** Main findings relating to tree vigour, sapwood non-structural carbohydrate
 838 concentrations (starch and soluble sugars, SS) before and after drought and presence of fungi
 839 in *Pinus sylvestris* (a) and *Abies alba* (b). In (a), the relationships between variables are
 840 described using the Spearman correlation coefficient (r_s). In (b), different letters indicate
 841 significant differences between tree types within each season (pre- or post-drought
 842 comparisons) according to Tukey HSD tests.

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844

845 **Figure 6.** Observed (symbols with error bars) and modelled (log-transformed) basal-area
 846 increment trends (mean \pm SEs are shown) for declining (grey symbols and lines) and non-
 847 declining (black symbols and lines) silver fir (*A. alba*), Scots pine (*P. sylvestris*) and Aleppo
 848 pine (*P. halepensis*) trees. Trends were based on the best-fitted generalized additive mixed
 849 models for the period 1950–2012 (see statistical parameters in Table 3).

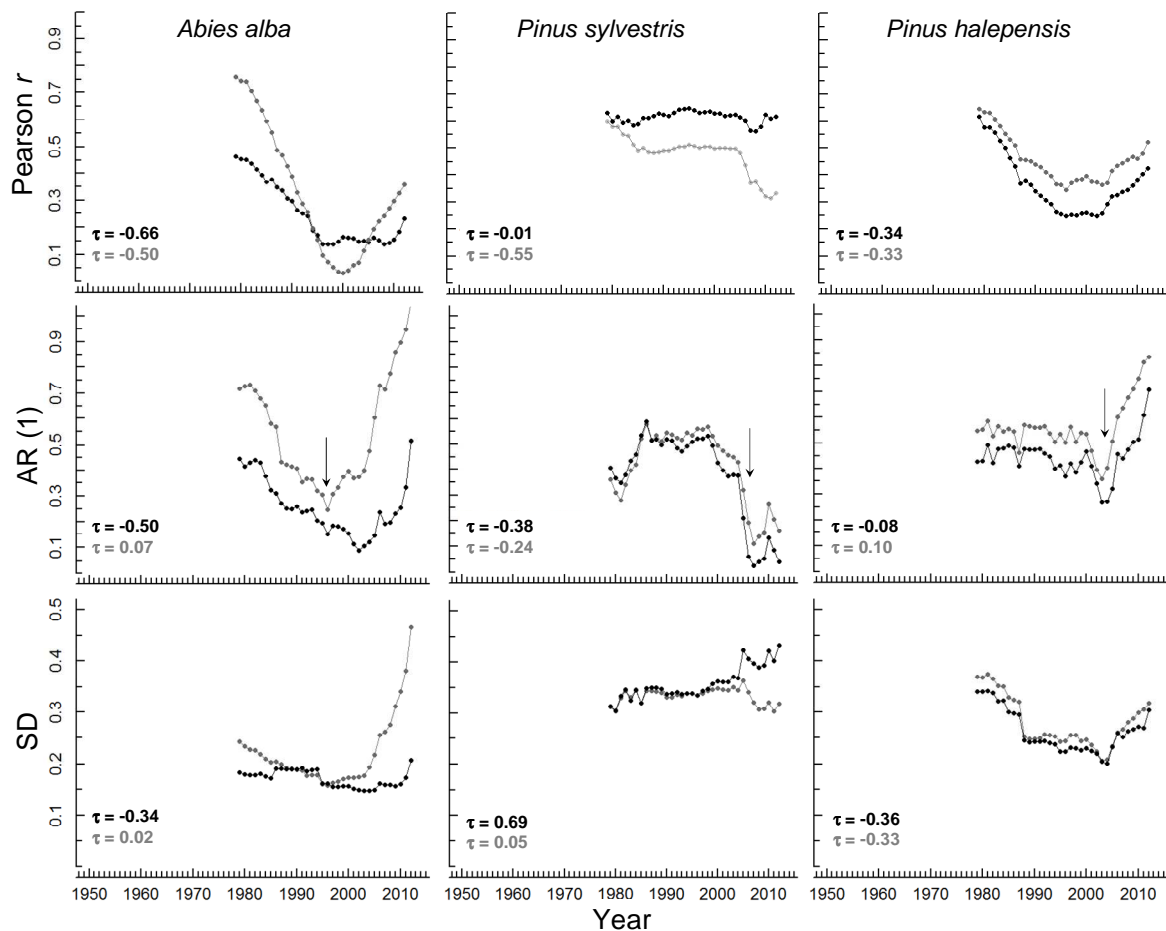


Figure 7. Early-warning signals (moving Pearson correlations; AR(1), first-order autocorrelation; SD, standard deviation) obtained using 30-year moving windows for basal-area increment (bai) series of declining (grey lines) and non-declining trees (black lines) of the three study species (*A. alba*, *P. sylvestris* and *P. halepensis*). The statistics were calculated for bai individual series (upper graphs) or bai residuals of mean series for declining and non-declining trees (middle and lower graphs) after extracting mid-term trends using 30-year-long windows. The symbols correspond to the last year of each 30-year interval (e. g, 1979 corresponds to the interval 1950–1979). The Kendall τ statistics indicate the trends for each variable. The arrows indicate severe droughts in 1995 and 2005 (see Fig. 1).

Supporting Information

Table S1. Dendrochronological statistics (means \pm SE) of the tree-ring width series.

Species	No. trees	First-last years	Tree-ring width (mm)	Correlation with mean series	First-order autocorrelation	Mean sensitivity ¹
<i>Abies alba</i>	35	1886-2012	1.99 \pm 0.11	0.59 \pm 0.02	0.23 \pm 0.01	0.22 \pm 0.01
<i>Pinus sylvestris</i>	37	1817-2012	0.70 \pm 0.04	0.78 \pm 0.02	0.62 \pm 0.02	0.49 \pm 0.01
<i>Pinus halepensis</i>	33	1869-2012	1.05 \pm 0.08	0.73 \pm 0.02	0.68 \pm 0.02	0.44 \pm 0.01

¹A measure of the year-to-year variability in width of consecutive tree rings (Fritts 2001).

Table S2. Sensitivity analysis calculated for selected variables (BAI, basal area increment) showing significant differences when comparing declining vs. non-declining trees using two thresholds to classify trees according to their defoliation (40% and 60%). Overall, we observed similar results when using the aforementioned defoliation thresholds and comparing the variable as when considering the 50% defoliation threshold (see Table 2), except two variables in the case of *P. halepensis* (bold values shown in the table).

Species	Variable	40% defoliation threshold		60% defoliation threshold	
		<i>t</i>	<i>P</i>	<i>t</i>	<i>P</i>
<i>Abies alba</i>	Defoliation (%)	-11.31	<0.0001	-11.78	<0.0001
	Sapwood length (cm)	2.35	0.024	2.28	0.031
	BAI 2008–2012 (cm ²)	3.34	0.002	3.56	0.001
<i>Pinus sylvestris</i>	Defoliation (%)	-13.00	<0.001	-14.07	<0.001
	Sapwood length (cm)	5.01	0.001	4.23	0.001
	BAI 2008–2012 (cm ²)	4.44	0.001	3.84	0.001
<i>Pinus halepensis</i>	Defoliation (%)	-7.04	<0.001	-7.08	<0.001
	Sapwood length (cm)	2.05	0.048	1.54	0.133
	BAI 2008–2012 (cm ²)	2.64	0.013	1.49	0.147

Negative and positive *t* values indicate higher values for declining and non-declining trees, respectively.

Table S3. Statistical parameters of the non-linear terms of the second best GAMM fitted to basal area increment of declining and non-declining trees (see Table 2). For each species (*P. sylvestris*, *A. alba* and *P. halepensis*) the degrees of freedom of the smooth term (edf), the associated *F* statistic and the *P* value of each selected term are shown.

Species	Terms	edf	<i>F</i>	<i>P</i>
<i>Pinus sylvestris</i>	Trend (year) declining	7.88	30.77	<0.001
	Trend (year) non-declining	8.63	28.64	<0.001
	age	3.25	12.20	<0.001
	dbh	1.00	51.32	<0.001
	Spatial location	–	–	–
<i>Abies alba</i>	Trend (year) declining trees	6.02	27.73	<0.001
	Trend (year) non-declining trees	6.65	11.12	<0.001
	age	–	–	–
	dbh	1.00	67.32	<0.001
	Spatial location	–	–	–
<i>Pinus halepensis</i>	Trend (year)	8.29	46.31	<0.001
	age	–	–	–
	dbh	–	–	–
	Spatial location	–	–	–

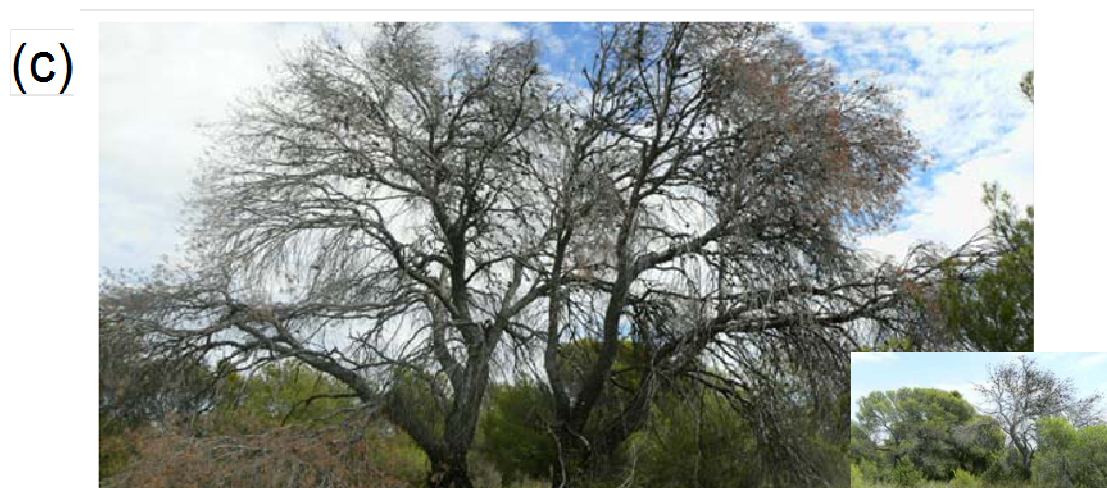
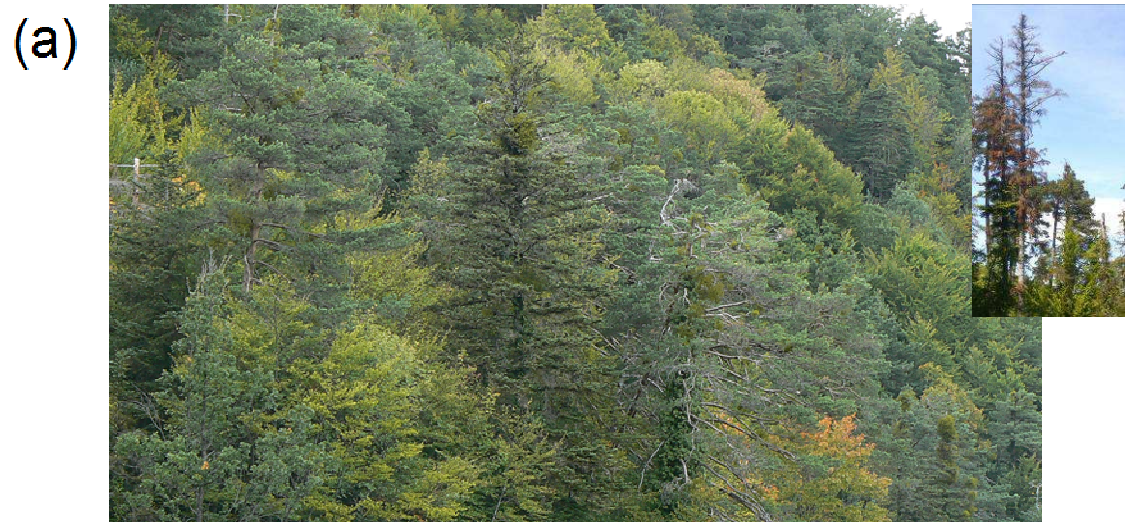


Figure S1. Illustrative views of drought-induced dead or dying silver fir (*Abies alba*, a), Scots pine (*Pinus sylvestris*, b) and Aleppo pine (*Pinus halepensis*, c) stands and trees.

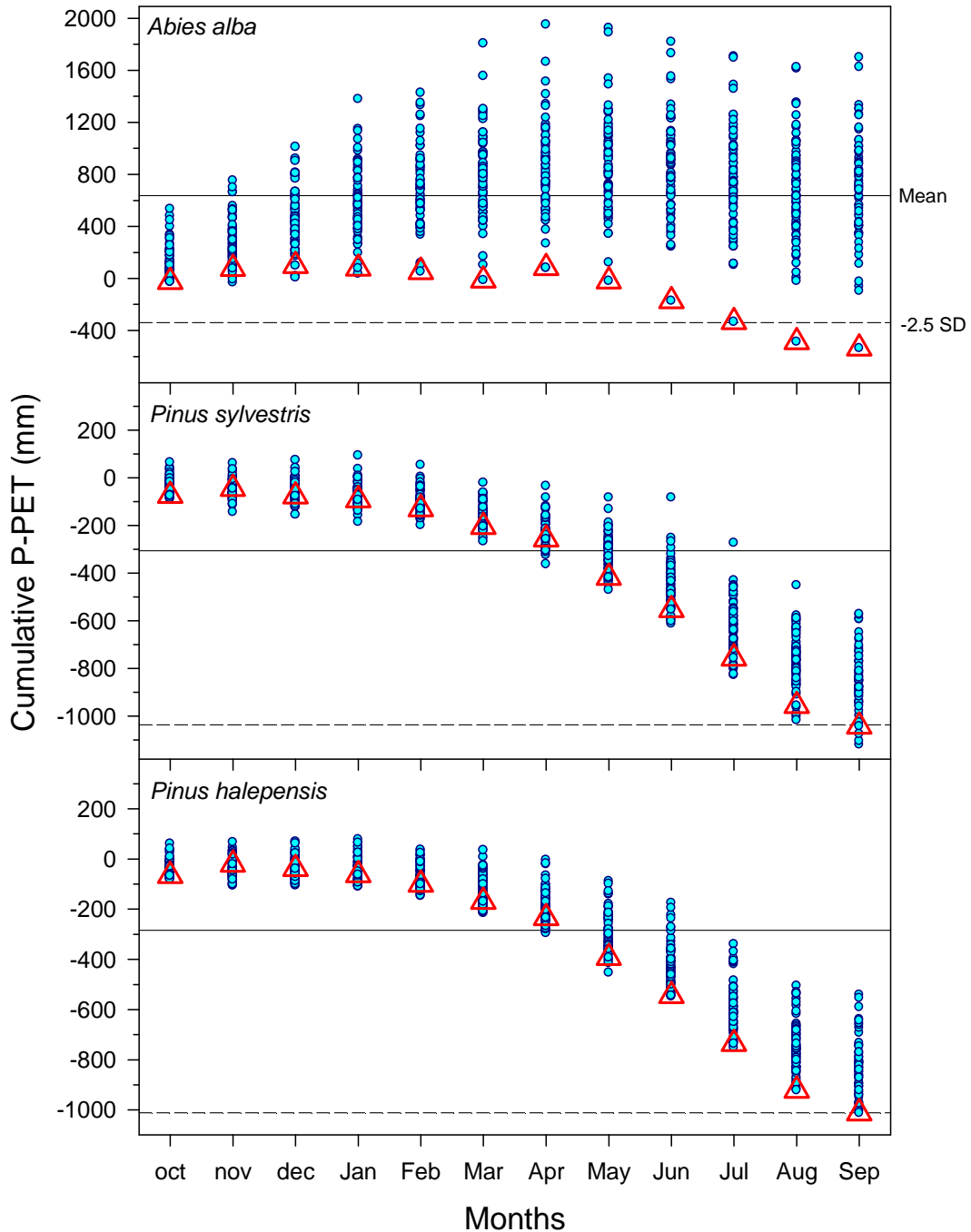


Figure S2. Monthly cumulative water balance (P-PET, difference between precipitation and potential evapotranspiration) estimated for the three study species and calculated from the year prior to growth (months abbreviated by lowercase letters) up to the end of the growing season in 2012 (months abbreviated by uppercase letters). Data correspond to the period 1951–2012. The triangles show the 2012 drought while the solid and broken horizontal lines show the mean and the mean minus two times the standard deviation (-2.5 SD), respectively.

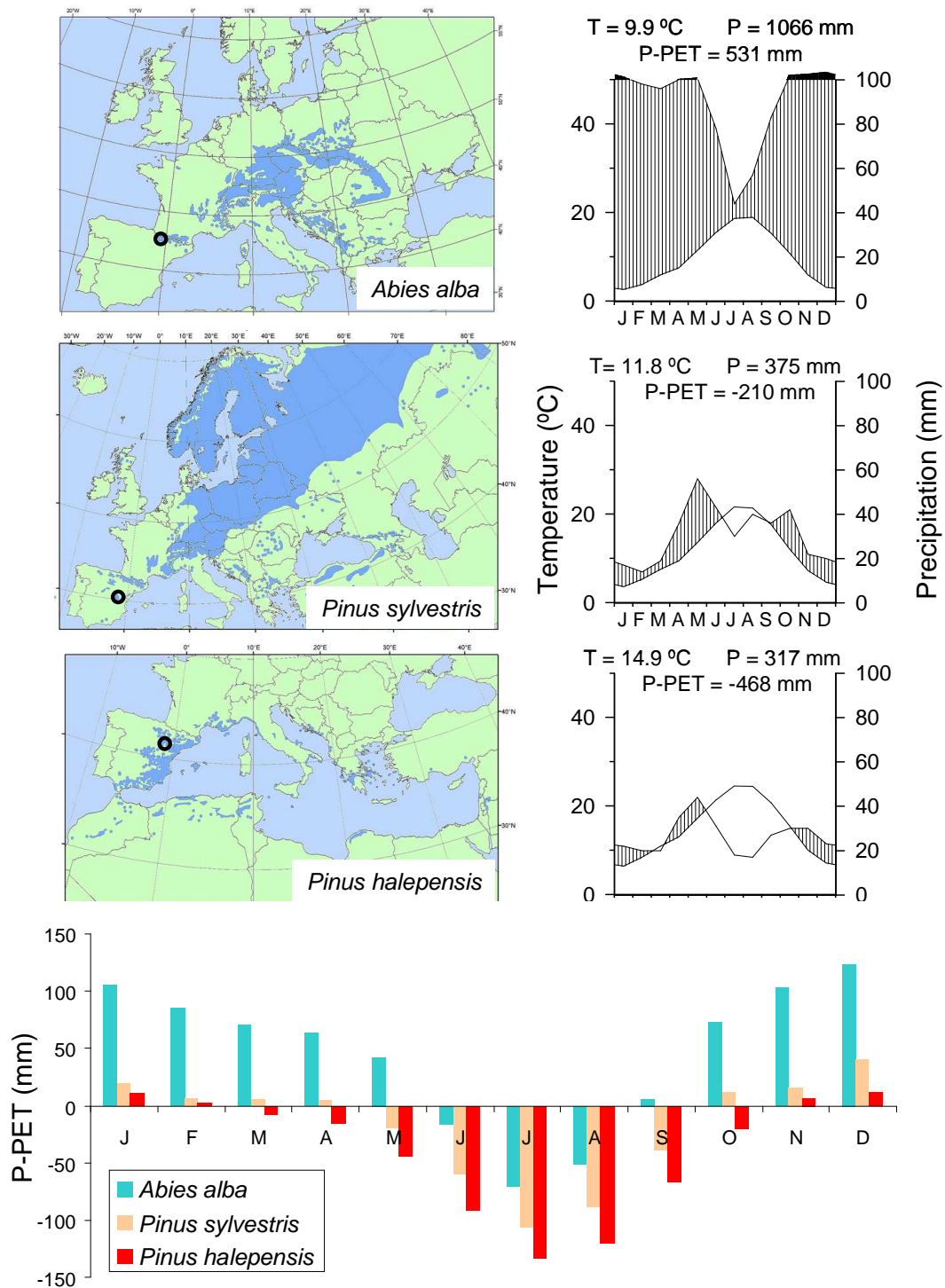


Figure S3. Geographical location (maps modified from those available at http://www.euforgen.org/distribution_maps.html) and climatic characteristics (climatic diagrams, water balance or difference between precipitation –P– and potential evapotranspiration –PET) of the three study sites. Mean (T, temperature) or total (P, PET lower line) annual values are indicated for each climatic diagram.

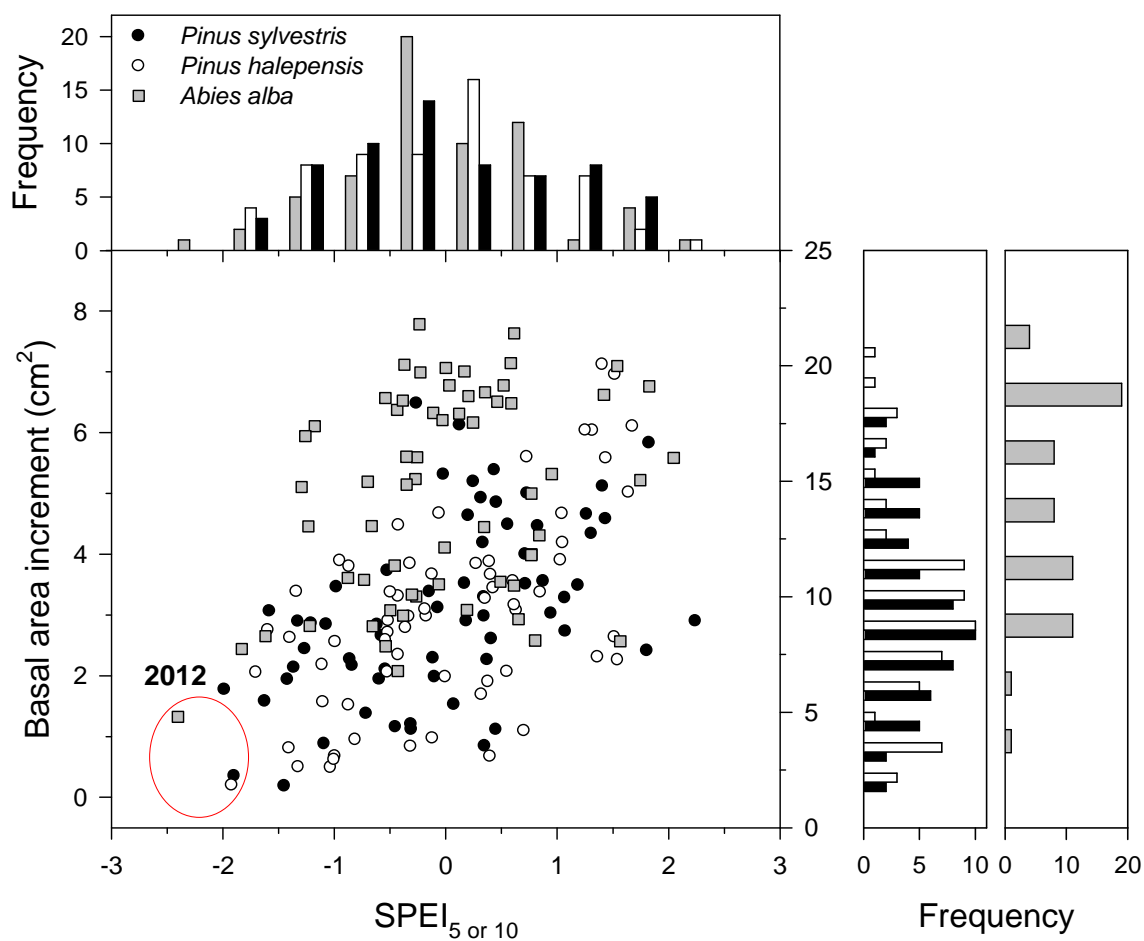


Figure S4. Relationships between mean basal area increment (BAI) of the three study species (different symbols and fill colours) and the June 5-months (*A. alba*, *P. sylvestris*) or 10-months (*P. halepensis*) long Standardised Precipitation–Evapotranspiration Index (SPEI). The histograms show the distributions of the BAI and SPEI values. The ellipse emphasizes the extreme BAI and SPEI values corresponding to the 2012 drought event.

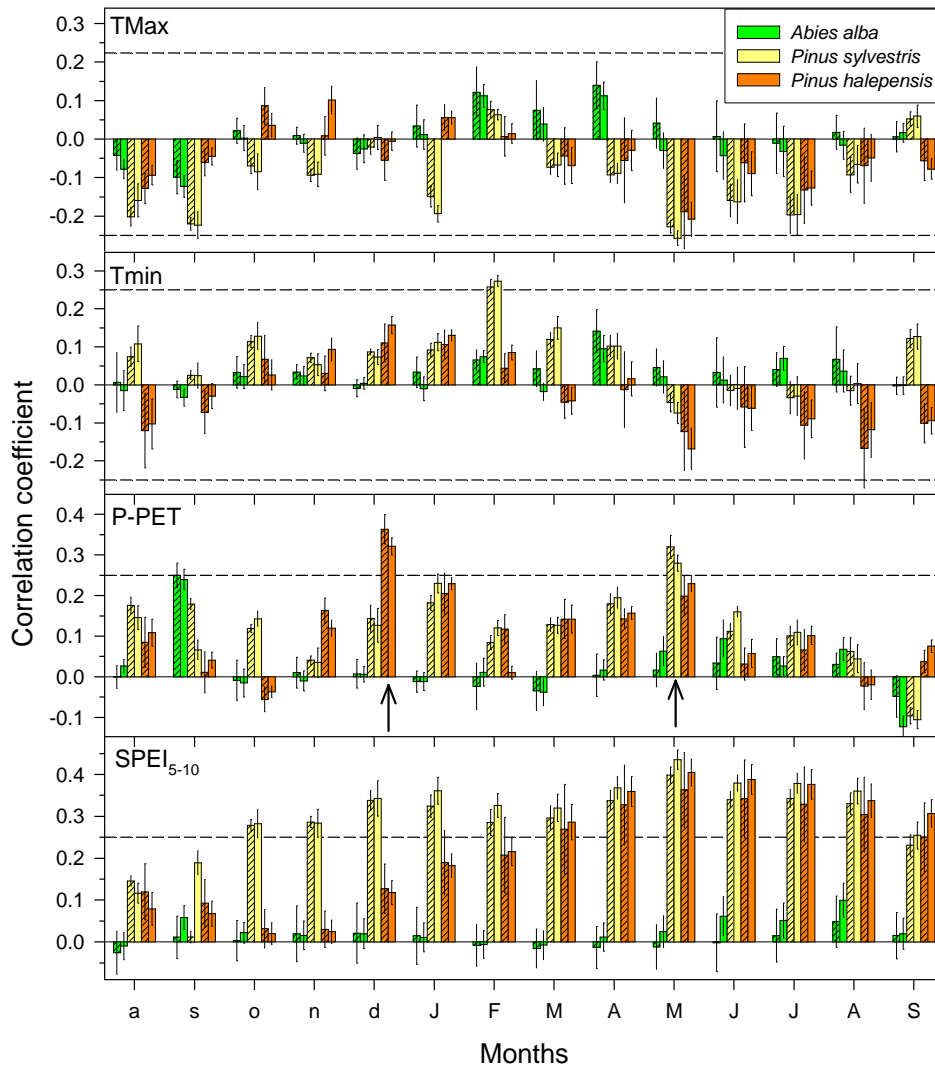


Figure S5. Associations (Pearson correlation coefficients, means \pm SE) calculated between individual growth series (basal-area increment residuals) and monthly climatic variables (TMax, mean maximum temperature; Tmin, mean minimum temperature; P-PET, water balance or difference between precipitation and potential evapotranspiration; SPEI₅₋₁₀, Standardised Precipitation–Evapotranspiration Index calculated at scales of 5 –*A. alba*, *P. sylvestris*– or 10 months –*P. halepensis*) for declining (hatched bars) and non-declining (empty bars) trees of the three study species. Correlations were obtained for months prior to the year of growth (abbreviated by lowercase letters) and for months corresponding to the year of tree-ring formation (abbreviated by uppercase letters). The arrows indicate the climatic variables showing significantly ($P < 0.05$) different impacts of climatic variables on growth of declining and non-declining trees for each tree species. Dashed horizontal lines show the 0.05 significance thresholds.