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50 Abstract

Global climate change is expected to further raise the frequency and severity of 51 extreme events, such as droughts. The effects of extreme droughts on trees are difficult to 52 disentangle given the inherent complexity of drought events (frequency, severity, duration 53 and timing during the growing season). Besides, drought effects might be modulated by trees' 54 phenotypic variability, which is, in turn, affected by long-term local selective pressures and 55 management legacies. Here, we investigated the magnitude and the temporal changes of tree-56 57 level resilience (i.e., resistance, recovery, and resilience) to extreme droughts. Moreover, we assessed the tree-, site-, and drought-related factors and their interactions driving the tree-58 level resilience to extreme droughts. We used a tree-ring network of the widely distributed 59 Scots pine (*Pinus sylvestris*) along a 2800 km latitudinal gradient from southern Spain to 60 northern Germany. We found that the resilience to extreme drought decreased in mid-61 62 elevation and low productivity sites from 1980-1999 to 2000-2011 likely due to more frequent and severe droughts in the later period. Our study showed that the impact of drought 63 64 on tree-level resilience was not dependent on its latitudinal location, but rather on the type of 65 sites trees were growing at and on their growth performances (i.e., magnitude and variability of growth) during the pre-drought period. We found significant interactive effects between 66 drought duration and tree growth prior to drought, suggesting that Scots pine trees with 67 68 higher magnitude and variability of growth in the long term are more vulnerable to long and severe droughts. Moreover, our results indicate that Scots pine trees that experienced more 69 frequent droughts over the long-term were less resistant to extreme droughts. We therefore 70 conclude that the physiological resilience to extreme droughts might be constrained by their 71 growth prior to drought, and that more frequent and longer drought periods may overstrain 72 73 their potential for acclimation.

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74 Keywords: Acclimation; latitudinal gradient; *Pinus sylvestris*; predisposition; tree rings

75 Introduction

Climate change effects are broadly characterized by elevated temperature, changed 76 precipitation regimes, and increased inter-annual variability, often resulting in more frequent 77 and intense climate extremes such as severe droughts (Dai, 2012; Spinoni et al., 2018). The 78 increased frequency and severity of droughts can significantly impact tree growth by 79 reducing their photosynthetic activity (Flexas & Medrano, 2002; Reddy et al., 2004) and 80 altering their cambial activity (Gruber et al., 2010). In addition, severe drought events have 81 82 been associated to forest decline either through direct abiotic effects leading to hydraulic failure and/or carbon starvation (Adams et al., 2017; Choat et al., 2018; McDowell et al., 83 2008) or mediated by biotic factors, such as insects (Rouault et al., 2006), fungi (Giordano et 84 al., 2009), and mistletoes (Rigling et al., 2010). These effects may ultimately induce shifts in 85 forest composition (Buras & Menzel, 2019; Walther et al., 2002) and reduction in forest 86 productivity (Ciais et al., 2005). 87

Growing recognition of the impacts of extreme droughts on forest ecosystems has 88 spurred on a number of long-term experiments and observational studies (e.g., Breshears et 89 al., 2005; Jentsch et al., 2011; Seidel et al., 2019). The results of these studies revealed a 90 large variability in pattern and magnitude of responses to extreme droughts (McDowell et al., 91 92 2008; Smith, 2011), because phenotypic acclimation to such extreme events may depend on a 93 multitude of factors and their interactions, including drought characteristics (Anderegg et al., 2015; Gazol et al., 2018), drought history of the growing environment (Vicente-Serrano et 94 al., 2013), species-specific functional traits and life-history strategies (Anderegg et al., 2016; 95 96 Greenwood et al., 2017; Lévesque et al., 2013), provenance (Sánchez-Salguero et al., 2018; Seidel et al., 2016), tree size and age (Granda et al., 2018; Magnani et al., 2000; Serra-97 Maluquer et al., 2018), tree-to-tree competition (Linares et al., 2010), nutrient imbalances 98

(Hevia et al., 2019), nutrient availability (Gessler et al., 2017), species composition and 99 stocking of the forest stand (Bottero et al., 2017; Forrester et al., 2016; Grossiord et al., 100 2014), trees neighbourhood composition (Grossiord, 2019), micro-climatic conditions related 101 to forest edge and interior (Buras et al., 2018), and growth trends prior to drought (Zang et 102 al., 2014). On the longer term, acclimation is often complemented by evolutionary genotypic 103 adaptation (Bose et al., 2020; Hamrick, 2004; Sánchez-Salguero et al., 2018) leading to 104 105 differentiation of populations and ecotypes with varying adaptive capacities to drought, often observed for marginal populations at dry species range margins (Bolte et al., 2016; Hampe & 106 107 Petit, 2005).

Moreover, the effects of past drought and growing conditions (legacy effects), can 108 remain for several years and modify the tree growth and physiological responses to the 109 current drought (Anderegg et al., 2015; Kannenberg et al., 2019; Seidel et al., 2019). An 110 important question in the debate on drought and acclimation is whether individuals will be 111 able to acclimate fast enough to cope with increased frequency and severity of droughts (Dai, 112 2012; Szejner *et al.*, 2019). It is therefore important to understand how tree growth responses 113 to extreme droughts vary across sites with different productivity (Valladares et al., 2007; 114 Valladares *et al.*, 2014), since site productivity can modify trees phenotypic strategies such as 115 tree height, root to shoot ratio, and crown development for efficient conservation and 116 utilization of water (Vanninen & Mäkelä, 2005). For example, tree height which is commonly 117 118 used as an indicator of site productivity (e.g., Westoby et al., 2002) was reported to be the strongest predictor of tree mortality in southwestern USA where 1.8 million trees were 119 studied (Stovall et al., 2019). 120

Several recent studies conducted in southern and central Europe have reported
drought-induced dieback of Scots pine (Buras et al., 2018; Camarero *et al.*, 2015b; Etzold *et al.*, 2019; Galiano *et al.*, 2010; Hereş *et al.*, 2012; Sánchez-Salguero *et al.*, 2012) causing a

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shift towards the dominance of oak (Quercus spp) species (Carnicer et al., 2014; Galiano et 124 al., 2010; Rigling et al., 2013). Although the impact of various tree- and site-level factors on 125 tree growth during drought has been studied from local to global scales (e.g., Anderegg et al., 126 2015; Buras et al., 2018; Gazol et al., 2018; Zang et al., 2014), their interactive effects are 127 still not clearly understood (Maes et al., 2019). For example, some large-scale studies found a 128 low to moderate influence of drought severity on tree growth response (e.g., Gazol et al., 129 130 2017; Sánchez-Salguero et al., 2018), possibly because they did not consider interactive effects between drought characteristics and long-term tree growth performances. In addition, 131 132 large-scale studies often characterize drought according to a pre-defined meteorological season (e.g., drought in spring-summer) irrespective of local site conditions, soil moisture 133 content, and geographic location (e.g., Bottero et al., 2017; Gao et al., 2018; Gazol et al., 134 2018). As a consequence, site-specific climate-growth signals might be overlooked if a 135 particular studied season is not the most relevant period for tree radial growth (Pasho et al., 136 2011; Sánchez-Salguero et al., 2015). 137

Here we combined Scots pine tree-ring width data from 30 sites into a network to 138 determine how growth responses to extreme drought varied along a latitudinal gradient across 139 Europe stretching from southern Spain to northern Germany. Tree growth response was 140 assessed to retrospectively quantify short- and long-term drought effects on growth for 141 numerous individuals, sites, and species at annual resolution. Tree growth resilience was 142 143 defined as the capacity of a tree to reach growth rates similar to those prior to a given drought event. Thus, resilience encompasses the capacity to buffer the impact of a disturbance 144 (resistance), as well as the ability to return to pre-disturbance growth levels (recovery) (Lloret 145 et al., 2011). Specifically, we asked four research questions: (i) How does the impact of the 146 climatic water balance (i.e., precipitation minus potential evapotranspiration) of different 147 seasons on tree growth vary along a latitudinal gradient? (ii) How do radial growth rates of 148

Scots pine during drought and non-drought years vary across sites? (iii) Has tree growth resilience to extreme drought changed over the past decades due to an increased frequency and severity of droughts (Serra-Maluquer et al., 2018; Szejner et al., 2019)? (iv) How do drought characteristics, site conditions, and tree growth related variables modulate the tree growth resilience to extreme drought events? For this last research question, we considered a list of biological hypotheses based on a literature review (see Table S1: e.g., Gazol et al., 2017; Gazol et al., 2018; Sánchez-Salguero et al., 2018; Vitali et al., 2017; Zang et al., 2014).

156 Materials and Methods

157 Study sites and tree-ring data

We compiled tree-ring width data of Scots pine from 30 sites (Table S2) along an 158 approximately 2800 km long latitudinal gradient from southern Spain (Baza; 37.2° N, 4.0° 159 160 W) to north-eastern Germany (Torgelow; 53.6 °N, 14° E) (Fig 1). To avoid age-related growth effects only trees older than 30 years at the time of examined drought were selected, 161 resulting in 615 adult Scots pine trees (6 to 60 trees per site). From each tree, two to four tree 162 ring width series were included, measured from increment-cores extracted at breast height 163 (1.3 m height) and cross-dated following standard dendrochronological procedures (Grissino-164 Mayer, 2001). 165

Considering the large differences in productivity among study sites along this long 166 gradient, the sites were grouped using a hierarchical cluster analysis (Kaufman & Rousseeuw, 167 168 1990). The classification was based on site productivity index (i.e., dominant tree height at 50 years of stand age) and site elevation. Dominant tree height has been commonly used as an 169 indicator of site productivity (e.g., Bugmann, 1996; Diéguez-Aranda et al., 2005a; Westoby 170 et al., 2002) including Scots pine sites (e.g., Diéguez-Aranda et al., 2005b; Hökkä & 171 Ojansuu, 2004; Mäkinen et al., 2017; Palahi et al., 2004). The hierarchical clustering was 172 done using the *hclust* function and *ward*.*D* method in R (R Development Core Team, 2018). 173

Based on the hierarchical cluster analysis (Fig S1), three groups (i.e., site types) were 174 characterized as (1) high- elevation sites (1865-2011 m a.s.l.) with low productivity (6.0-14.0 175 m in stand dominant tree height) referred as "HELP", (2) mid-elevation sites (600-1450 m 176 a.s.l.) with low productivity (7.5-11.0 m in stand dominant tree height) referred as "MELP", 177 and (3) low-elevation sites (33-326 m a.s.l.) with high productivity (15-23.7 m in stand 178 dominant tree height) referred as "LEHP" (Fig S1). 179 180 **Analytical approaches** Addressing our four research questions, the analytical approach involved two steps: 181 182 data preparation and data analysis. The data preparation step embodied four sub-steps, (i) quantification of tree-ring width indices, (ii) quantification of drought indices, (iii) 183 identification of drought, pre-drought, and post-drought periods (i.e., years), and (iv) 184 quantification of tree growth resilience indices. The data analysis step embodied four steps, 185 i.e., one for each research question. 186 Quantification of tree-ring width indices 187 We aimed at quantifying growth responses to extreme drought events over the recent 188 50 years period roughly from year 1960 to year 2011. However, our studied trees largely 189 differed in age across sites (Table S2). Hence, ring width data were transformed into 190 dimensionless ring width indices (RWI) with both age-related growth trends and lower-191 frequency variation removed from the time series (Cook & Kairiukstis, 1990). For this, ring 192 193 width data were detrended by fitting a negative exponential curve or using a 30 year cubic

spline with a 50% frequency cut-off (Cook & Kairiukstis, 1990). In addition to these

detrending methods, we also converted the raw ring width data into basal area increment (cm^2

196 per year) (Biondi & Quedan, 2008) using the *dplR* package in R (Bunn *et al.*, 2018). We

assessed the suitability of these approaches to disentangle the drought effects on tree growth

by computing the correlation coefficient with the drought indices (*cf.* next section) and by

characterizing the trend over a 50-year period (Table S3 and Fig S2). The results showed that 199 the negative exponential detrending method performed best among the used approaches in 200 terms of the magnitude of correlation with the drought and of capturing the long-term trends 201 (Table S3 and Fig S2). We thus used the detrended negative exponential RWI (hereafter 202 referred to as RWI) for the analysis. 203 To build the site-level tree-ring chronology, we averaged the detrended individual 204 RWI series with a Tukey's biweight robust mean (Cook & Kairiukstis, 1990; Fritts, 2001). 205 The RWI and average tree-level chronology were calculated using the *detrend* and *chron* 206 207 functions, respectively, available from the *dplR* R package (Bunn et al., 2018; R Development Core Team, 2018). 208 Quantification of drought indices 209 Monthly mean temperature (°C) and total precipitation (mm) data were obtained for 210 each site from different climate data sources (Table S4). To compute the correlation 211 coefficient between drought indices and the RWI, we considered a 50-year period for all 212 sites. However, the range of years for the 50-year period varied across sites due to differences 213 in timing of data collection. 214 For drought index, we initially considered the De Martonne Index (De Martonne, 215 1926), the Standardized Precipitation Index (McKee et al., 1993), and the Standardized 216 Precipitation Evapotranspiration Index (SPEI) (Vicente-Serrano et al., 2010). The SPEI had a 217 218 stronger correlation with RWI than the other indices examined for most of the sites (see Table S5). Hence, SPEI was used for defining the drought and non-drought years. 219 The SPEI is a unitless drought index, which takes into account both precipitation and 220 potential evapotranspiration effects in the calculation of the climatic water balance (CWB), 221 and is commonly used in the literature for identifying and characterizing drought and non-222 drought years (e.g., Bottero et al., 2017; Gazol et al., 2018). The potential evapotranspiration 223

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224	was calculated using the <i>Thornthwaite</i> function of the R package SPEI (Begueria & Vicente-
225	Serrano, 2013). The SPEI was then calculated from CWB using the spei function of the R
226	package SPEI (Begueria & Vicente-Serrano, 2013). For each site, we calculated SPEI of
227	various timescales i.e., integrated over 1 to 15 months in order to represent different lengths
228	of the growing season or at least different growth sensitive periods within the current and the
229	previous growing season. We assessed the Pearson correlation between RWI (ring width
230	index) and SPEIs (i.e. the different time intervals) for identifying the most relevant SPEI (i.e.,
231	most sensitive time interval) for each site to define the drought and non-drought years (see
232	Table S6). The resulting SPEIs (i.e., those best correlated with RWI) are presented in the
233	Table S7.
234	For identifying the extreme drought year of a site, we selected the year with the
235	lowest SPEI value. For each site, we first selected the extreme drought years for the period of
236	1980-2011. We then selected the extreme drought year for the period of 1980-1999 and for
237	the period of 2000-2011.
238	Identification of drought, pre-drought, and post-drought periods
239	We characterized drought periods by single or multiple years based on SPEI \leq -1.00
240	and pre-drought or post-drought periods (i.e., without drought) based on SPEI \geq -0.99. We
241	limited the pre-drought and post-drought periods to a maximum of three years, but for
242	drought periods we considered all consecutive years with SPEI \leq -1.00 (see Table S6). We
243	identified the most extreme droughts during 1980-1999, and during 2000-2011 for all study
244	sites (see Table S6) for comparing the tree growth responses to extreme droughts during the
245	recent decade (2000-2011) with the previous two decades (1980-1999). Since many sites had
246	no drought during 1990-1999, we decided to enlarge the earlier period back until 1980.
247	Tree growth resilience indices

For tree growth resilience, we computed three resilience indices as suggested by 248 Lloret et al. (2011): resistance, recovery, and resilience. The resistance quantifies the ratio 249 between growth during a drought period and growth during the preceding non-drought 250 period, representing thus the capacity of the trees to buffer the stress and maintain growth 251 during drought. The recovery quantifies the growth reaction following the drought period and 252 is defined by the ratio between growth during the post-drought period and growth during the 253 254 drought period. The resilience quantifies the ratio between growth during the post-drought period and growth during the pre-drought period, which represents the capacity of trees to 255 256 recover and regain the growth of the pre-drought period. We quantified resistance, recovery, and resilience for all trees of all sites during the most extreme droughts in 1980-1999 and in 257 2000-2011 (see Table S6). 258

259 Research question 1: Impact of seasonal drought (SPEI) on tree growth

Based on the results of preliminary analysis (i.e., correlation between RWI and 260 different SPEIs), we identified the eight best correlated SPEIs for understanding the 261 magnitude (i.e., degree of correlation) and pattern (i.e., type of correlation) of influences of 262 drought on RWI, and how that magnitude and pattern of correlation varied across the 263 latitudinal gradient examined in this study. The selected SPEI timescales were August-15 264 (i.e., from previous June to current August), May-12 (i.e., from previous June to current 265 May), May-9 (i.e., from previous September to current May), May-6 (i.e., from previous 266 267 December to current May), May-3 (i.e., spring, from current March to current May), August-6 (i.e., summer, from current March to current August), August-3 (i.e., from current June to 268 current August), and November-6 (i.e., from current June to current November). 269

270 Research question 2: Tree growth rate in drought and non-drought years

For understanding the absolute tree radial growth performances during drought and non-drought years, we modelled absolute tree radial growth (non-detrended tree ring width)

as a function of site types (three levels: LEHP, MELP, and HELP), drought status (two
levels: drought years and non-drought years), and the interaction between site types and
drought status. For understanding the potential role of tree age on absolute tree radial growth,
we considered tree age as a covariate in this analysis.

277 Research question 3: Temporal change in tree growth resilience to extreme droughts

We modelled resistance, recovery, and resilience as a function of time period (two levels: 1980-1999 and 2000-2011), site types (three levels: LEHP, MELP, and HELP), and the interaction between time period and site types.

281 Research question 4: Factors affecting tree growth resilience to extreme drought

For this research question, we selected the most extreme drought during the entire 282 study period 1980-2011 and used the corresponding resistance, recovery and resilience 283 indices as response variables in a mixed-effects model (cf. next section). We considered 284 several tree-, site-, and drought-level explanatory variables and various two-way interaction 285 terms (see Table S1). The variables included tree size (i.e., tree diameter at breast height 286 (DBH) inside bark at the drought year), tree growth and tree growth variability prior to 287 drought representing the average and standard deviation of RWI, respectively of 10 288 consecutive years prior to the extreme drought excluding the years considered as pre-drought 289 period for quantifying the three resilience indices, site types, elevation, latitude, drought 290 291 severity (measured by the average SPEI during the drought period), drought duration 292 (measured by the length of the drought period in years), and drought frequency (measured by the number of drought years (SPEI \leq -1.00) within 10 years preceding the maximum drought 293 period). 294

295 Statistical analyses

We used a linear mixed-effect modelling approach for research question 2, 3, and 4 in which our variables of interest were considered as fixed effects and trees nested within sites

were considered as random effects. The modelling was performed using the function lme of 298 the R package *nlme* (Pinheiro *et al.*, 2014; Pinheiro & Bates, 2000). The response variables 299 were log-transformed to normalize residuals and homogenize variances and we checked the 300 assumptions of normality of the residuals and homogeneity of the variances. Preliminary 301 analysis indicated that an additional error structure to account for plot spatial autocorrelation 302 did not improve model performance and thus was not incorporated into the final model. We 303 304 also assessed potential multicollinearity among explanatory variables using the Variance Inflation Factor (VIF) and discarded variables when VIF > 2.0. The VIF was calculated using 305 306 the function vif of the R package car (Fox & Weisberg, 2011). The post hoc Tukey multiple comparison test was performed to detect the statistical differences (Hothorn et al., 2008). 307 For research question 4, we used the information-theoretic approach (Burnham & 308 Anderson, 2002; Johnson & Omland, 2004), which provides a measure of strength for each 309 310 candidate model that represents a plausible hypothesis relative to the entire set of candidate models considered (Mazerolle, 2006). In the context of our research question (i.e., what are 311 the factors driving the tree growth resilience to extreme drought?), we considered 16 312 hypotheses (i.e., candidate models) (Table S1), which were developed based on the current 313 understanding resulted from different studies that examined tree growth resilience to extreme 314 droughts. Model selection was performed using the AICcmodavg package of R (Mazerolle, 315 2011). Candidate models were compared using Akaike's Information Criterion corrected for 316 317 small sample sizes (AICc). Akaike weights were computed to assess the support for each model. We used multi-model inference to compute the model-averaged estimates of the 318 explanatory variables and their 95% confidence intervals (Burnham & Anderson, 2002). A 319 confidence interval excluding 0 indicated that the corresponding explanatory variable had an 320 effect on the response variable (Burnham & Anderson, 2002; Mazerolle, 2006). In addition to 321 our candidate models we also considered a null model and a full model. The coefficient of 322

- variation (R^2) for fixed and random effects were calculated using the function
- *r.squaredGLMM* of the MuMIn package in R (Bartoń, 2013).
- 325 **Results**

326 Impact of seasonal drought (SPEI) on tree growth

Our results showed significant differences in the response of tree growth to the 327 different time periods of SPEI. The current year summer to autumn (June-November) SPEI 328 significantly controlled tree growth at LEHP sites of northern Germany (Fig. 2), while tree 329 growth at MELP sites was driven by SPEI of spring (March-May), summer (June-August), 330 331 and spring and summer combined. Tree growth in HELP sites was either non-related or negatively correlated with different time periods of the SPEI (Fig. 2; Table S3). Overall, the 332 magnitude of correlation between RWI and different SPEIs was higher for MELP than the 333 two other site types (Table S3). Three sites of HELP site type had a negative correlation with 334 SPEIs while one site of HELP site type was not significantly correlated with any SPEI 335 considered in our analysis (Table S3). 336

337 Tree growth rate in drought and non-drought years

In drought and non-drought years, tree radial growth was higher at LEHP than at HELP and at MELP sites (Fig. 3). The MELP sites had significantly lower tree radial growth in drought years than in non-drought years (Table S8 and Fig. 3). Contrary to MELP, tree growth was not significantly different between drought and non-drought years at LEHP and at HELP sites (Table S8 and Fig. 3). Tree age was negatively associated with the radial growth (p<0.0001) irrespective of site types (Table S8).

344 Temporal change in tree growth resilience to extreme droughts

Tree growth resistance to extreme drought for all site types (i.e., HELP, LEHP, and MELP) did not change over the two periods (i.e., 1980-1999 and 2000-2011) (Fig. 4A).

Nevertheless, resistance was higher at HELP than at LEHP, and higher in the latter compared
to MELP, irrespective of the period (Table S9 and Fig. 4A).

Tree growth recovery changed significantly over the two periods for all site types, where recovery decreased from 1980-1999 to 2000-2011 at MELP and HELP sites, while increased from 1980-1999 to 2000-2011 at LEHP sites (Table S9 and Fig. 4B). In 1980-1999, recovery was significantly higher at MELP compared to the two other site types irrespective of period (Table S9 and Fig. 4B).

Tree growth resilience changed significantly over the two periods for LEHP and MELP sites, but not for HELP sites. Resilience decreased from 1980-1999 to 2000-2011 at MELP sites, while it increased from 1980-1999 to 2000-2011 at LEHP sites (Table S9 and Fig. 4C).

358 Factors affecting tree growth resilience to extreme drought

The model that included additive and interaction effects of all variables considered in the analysis had full support of Akaike weight for resistance (Table 1). A lower resistance was associated with higher pre-drought growth rate (Table 2). In addition, a lower resistance was associated with greater drought frequency, and with longer drought but depending upon pre-drought growth rate (Table 2). Resistance was higher at HELP and LEHP sites than at MELP sites (Table 2 and Fig. 5A).

The model that included drought severity and site types, and the interaction between the two variables had the highest support of Akaike weight for recovery (0.74; Table 1). Recovery was lower at HELP and LEHP sites than at MELP sites (Table 2 and Fig. 5B). In addition, the recovery was higher where trees experienced a higher frequency of droughts (Table 2).

The model that included pre-drought growth rate and drought duration, and theinteraction between the two variables had the highest support of Akaike weight for resilience

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372 (Table 1). Resilience was negatively associated with pre-drought growth rate and pre-drought
373 growth variability and there was no difference across the three site types (Table 2 and Fig.
374 5C).

375 **Discussion**

Using tree ring width data from 30 sites along a 2800 km latitudinal gradient across 376 Europe we analysed whether tree growth resilience to extreme drought depended on the 377 geographical location of the tree (Isaac-Renton et al., 2018) and if resilience to extreme 378 drought decreased over time due to more frequent drought events in recent years (Serra-379 380 Maluquer et al., 2018; Spinoni et al., 2014). We examined these questions on Scots pine, one of the most widely distributed tree species in the world which is also considered vulnerable to 381 extreme drought conditions (Camarero et al., 2015b; Galiano et al., 2010; Matías et al., 2017; 382 Rigling et al., 2013). Our study shows that tree-level resilience to drought was not dependent 383 on the latitudinal location, but rather on the type of site they were growing at and their growth 384 performance (i.e., magnitude and variability of growth) during the pre-drought period. Our 385 results indicate that trees with higher magnitude and variability in growth are more 386 vulnerable to long and severe droughts. In addition, we found that tree growth resilience to 387 extreme drought was lower during 2000-2011 than during 1980-1999 at mid-elevation and 388 lower productivity sites. These results may indicate that more frequent drought events that 389 occurred in 2000-2011 than in earlier period make Scots pine trees more vulnerable to 390 391 extreme droughts. However, we found high variability in tree-level responses (Fig. 5) as detected by previous studies (e.g., Gazol et al., 2018; Maes et al., 2019). This high tree-level 392 variability (Fig 5) and low marginal R² values (Table 2) may indicate that combining tree 393 information from different sites without direct measurements of local site-related factors 394 (e.g., soil water content, nutrient availability, stand stocking, and rooting depth) compromises 395

the predictive power of the models (DeSoto *et al.*, 2020; Gazol et al., 2018; Grossiord et al.,

397 2014).

398 Impact of seasonal drought on tree growth

Our results reveal important seasonal differences across lower and higher productivity 399 sites in terms of SPEI-RWI correlations. Scots pine trees growing at mid-elevation lower 400 productivity (MELP) sites showed a greater sensitivity to spring-summer SPEI while trees 401 402 growing at low-elevation higher productivity (LEHP) sites of northern Germany were more sensitive to summer-autumn SPEI (Fig. 2). This is consistent with results from Pasho et al. 403 404 (2011), who studied eight tree species including Scots pine for the period of 1950-2005 using the Standardized Precipitation Index at timescales from 1 to 48 months. Our results showed 405 that Scots pine trees growing in MELP are more sensitive to drought (measured by SPEI) 406 407 than trees growing in LEHP and HELP sites (Fig. 3 and Table S3), which is consistent with 408 the findings of other studies on Scots pine (Lévesque et al., 2014; Pasho et al., 2011; Sánchez-Salguero et al., 2015). Trees at MELP sites are likely to grow under water limitation 409 during extended periods of the year, and changes in precipitation or evaporative demand 410 during these periods will directly affect the water availability of the trees and thus their 411 physiological responses, resulting in lower growth (Cabon *et al.*, 2020). A tree at a less dry 412 site in contrast is growing mostly without water limitations. During such periods with plenty 413 of water supply there will be no strong direct effect of precipitation (and at least within a 414 415 certain range of evaporative demand) on tree physiology and on growth (Martínez-Vilalta et al., 2009; Sterck et al., 2008). Thus, it is reasonable to consider that the climatic control of 416 tree growth is stronger at MELP which are relatively drier than LEHP sites (Table S2). 417 However, we need to improve our understanding on local drought characteristics because the 418 climatic water balance and SPEI do not consider the soil water holding capacity and the depth 419

420 of water table which are important parameters for estimating the soil water available for trees421 (Zang *et al.*, 2019).

422 Tree growth rate in drought and non-drought years

We found higher annual growth among trees at lower-elevation sites than trees at 423 higher-elevation sites (Fig. 3). Trees at low-elevation sites were relatively younger than trees 424 at high-elevation sites and thus more likely in their full growth ontogenetic stage. However, 425 426 Scots pine trees at high-elevation sites had no growth difference between our selected drought and non-drought years (Fig. 3) irrespective of the fact that they experienced similar 427 428 drought severity and frequency as the two other site types in terms of SPEI (Fig. 6). Therefore, tree growth at high-elevation Scots pine forests is probably controlled by other 429 factors than changes in water availability during drought (Carrer et al., 2007; Cudlín et al., 430 2017). In our dataset, three out of six high-elevation sites had negative correlation with SPEI 431 (Table S3) indicating that tree radial growth at high-elevation sites examined in our study was 432 more related to other factors such as ontogeny and temperature (Camarero et al., 2015a; 433 Körner, 2003) than to water availability (Hagedorn *et al.*, 2014). Our high-elevation site type 434 has three sites with trees that were older than 150 years during our selected drought period 435 and older than the trees at any other site examined in our analysis. Besides, the mean annual 436 temperature at four out of six high-elevation sites was <7.0 °C which is almost 2 °C cooler 437 than at any other site examined in our analysis. Trees in our high-elevation sites are perhaps 438 well adapted to such low mean annual temperature and thus short growing season, confirming 439 the findings of other studies conducted in southern Europe (Peñuelas et al., 2008) and in 440 southwestern United States (Adams & Kolb, 2004). It is also important to mention that our 441 drought intensity index may not fully capture the absolute water balance differences between 442 drought and non-drought years because of the elevational differences between the sites and 443

climate stations (Table S4), which may partly explain the lack of growth differences betweendrought and non-drought years for our high-elevation sites.

446 Temporal change in tree growth resilience to extreme droughts

Although Scots pine is a relatively drought-tolerant species, our results showed a 447 decreased growth recovery and resilience to extreme drought at mid-elevation and lower 448 productivity sites. This is largely consistent with the patterns reported by Serra-Maluquer et 449 450 al. (2018) in three pine species (including Scots pine) growing in NE Spain. In our case, this result likely reflects the higher frequency and severity of droughts events prior to the extreme 451 452 drought occurring in 2000-2011 than prior to the extreme drought occurring in 1980-1999 (Fig. 6). Moreover, trees became older and larger over time. A larger tree carries higher non-453 photosynthetic biomass, which requires a greater investment for defence and maintenance 454 (Ryan & Yoder, 1997; Scholz et al., 2011) and increasing tree height may increase hydraulic 455 constraints and xylem cavitation risks under drought (Olson et al., 2018). This reduced tree 456 resilience over time may challenge the physiological potential for acclimation to more 457 intensive and frequent drought events that we expect in the future (Dai, 2012). The decreased 458 tree growth resilience at MELP sites may indicate that drought hardening (i.e., physiological 459 processes by which a tree becomes more acclimate to drought conditions) (Villar-Salvador et 460 al., 2013) is not very important and does not allow trees to acclimate to frequent and 461 prolonged drought events irrespective of their drought experiences. 462

In contrast to MELP sites, tree growth recovery and resilience to extreme droughts increased at LEHP sites during the recent period (i.e., 2000-2011) compared to the previous period (i.e., 1980-1999). Although drought severity and drought frequency did not change significantly at LEHP sites over these two periods (Fig 6), the duration of the extreme drought was longer in 1980-1999 than in 2000-2011. For example, seven sites of the LEHP

site type experienced a two-year long drought during 1980-1999, while none of the droughtsoccurring during 2000-2011 was longer than one year at that site type (Table S6).

470 Factors affecting tree growth resilience to extreme drought

Several studies conducted at local to global scales provided understanding on tree,
site, and drought related variables influencing tree-level resilience to extreme drought (e.g.,
Gao et al., 2018; Gazol et al., 2017; Gazol et al., 2018; Sánchez-Salguero et al., 2018; Taeger *et al.*, 2013; Vitali *et al.*, 2017; Zang et al., 2014). To our knowledge, however, none of these
studies compared explicitly the model performances with and without interaction effects
among all these variables.

Our results showed that the top-ranked model (according to Akaike weights) for all 477 the three resilience indices (resistance, recovery, and resilience) included interaction effects 478 between tree and drought-related variables (Table 1). This means that the impact of drought 479 on tree-level resilience is not independent, but rather dependent on how the trees were 480 growing (i.e., magnitude and variability of growth) during the pre-drought period and on the 481 type of site they were growing at. In addition to drought severity, which was usually found to 482 significantly affect tree-level resilience at various spatial scales variables (Gazol et al., 2017; 483 Gazol et al., 2018; Zang et al., 2014), we also considered drought duration and drought 484 frequency as explanatory variables as suggested by Gao et al., (2018). We indeed found a 485 significant negative effect of drought frequency on tree resistance (Table 2), suggesting that 486 487 trees that experienced more frequent droughts were less resistant to extreme droughts. Therefore, trees that display higher variability in growth are not only sensitive to extreme 488 drought but also to frequent drought occurrence as e.g. shown by Seidel et al. (2016) for 489 490 different Scots pine provenances. Higher above-ground biomass growth can be related to lower or at least non-proportional biomass allocation to roots (Gessler et al., 2017) and might 491 consequently increase tree-level sensitivity to upcoming drought periods (Martínez-Vilalta et 492

al., 2012). Long or repeated droughts may reduce the number of living branches and, because 493 of needle multi-year life-span, the leaf area per branch (Galiano et al., 2011; Vennetier et al., 494 2013). This holds back short-term tree leaf area recovery and may drive tree growth to very 495 low levels during and after drought. Although a reduced leaf area limits water stress during 496 and after droughts and may favour resilience, it also hampers carbohydrate reserves build-up 497 and may lead to carbon starvation. Accordingly, higher growth variability may be related to a 498 499 higher vulnerability to the upcoming stresses (Cailleret et al., 2019; DeSoto et al., 2020; McDowell et al., 2008; Ogle et al., 2000). For instance, Ogle et al., (2000) noticed a 1.5 times 500 501 higher growth variability in dead pinyon pine trees relative to surviving ones in drought years preceding mortality across forests of southwestern USA. 502 Although the pre-drought growth of Scots pine had a strong relationship with 503 resistance, it had no-relationship with recovery. This lack of relationship between pre-drought 504 growth and recovery resulted in a relatively weak relationship (although statistically 505 significant) between pre-drought growth and resilience (Fig 5), because the resilience is 506 507 mathematically related to both resistance and recovery (Lloret et al., 2011), and in the context of our study, recovery was more closely related to resilience than the relationship between 508 resistance and resilience (Fig S3). Scots pine trees growing at MELP sites displayed lower 509 resistance but higher recovery than trees growing at LEHP and HELP sites (Table 2 and Fig. 510 5), suggesting different growth strategies exercised by trees from different sites to cope with 511 512 drought (Sánchez-Salguero et al., 2018). However, resistance and recovery are relative indices and do not allow comparison across trees in terms of their absolute growth 513 performances. Although resistance and recovery hold a negative mathematical relationship 514 (Lloret et al., 2011), they provide useful insights on disentangling trees that tried to remain 515 firm during drought years (i.e., higher resistance and lower recovery) from trees that tried to 516 conform drought impact (i.e., lower resistance and higher recovery) (Gazol et al., 2017a). 517

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These two different strategies to extreme droughts by trees from high-productivity and low-518 productivity sites are also reported by other local (Vitali et al., 2017; Zang et al., 2014) and 519 global scale studies (Gazol et al., 2017). For instance, a multi-species comparison across 520 Spain found higher resistance for species dominating mesic sites than for species from more 521 xeric sites, which presented a higher recovery (Gazol et al., 2018). At our LEHP, Scots pine 522 trees were on average taller and larger than MELP sites (Table S2). These larger trees in 523 524 mesic sites most likely developed larger crowns, which might be able to still provide sufficient photosynthates to sustain growth during the drought period. Alternatively, these 525 526 larger trees had sufficient reserves to compensate for the drought-induced reduction in photosynthesis. Both mechanisms would result in a higher resistance compared to trees in 527 lower productivity sites (Martínez-Vilalta et al., 2009). However, taller trees associated with 528 529 larger crowns and concomitantly greater water demand can be more vulnerable to drought induced hydraulic failure (McDowell et al., 2008; Olson et al., 2018) and the required post-530 drought investment of assimilates for restoring their hydraulic system could slow down their 531 recovery process (Brodribb et al., 2010). 532

In this analysis, we considered tree, site, and drought level variables and their 533 interactions, but available data did not allow to quantify the effects of forest stand-level 534 variables including tree-to-tree competition (Serra-Maluquer et al., 2018), site conditions 535 including soil and topography (Vennetier et al., 2018; Zalloni et al., 2019), species mixtures 536 537 (Pretzsch et al., 2013), trees neighbourhood composition (Grossiord, 2019), and stand stocking (Bottero et al., 2017; D'Amato et al., 2013), which will be the subject of subsequent 538 analyses. Moreover, we did not include possible sub-species variation due to local 539 540 evolutionary adaptation induced by drought-related selection (Hampe and Petit 2005). We detected large tree-level variability within and across sites, and a larger proportion of the 541 variance of our models was explained by the random effects (i.e., trees nested within sites) 542

than by the fixed effect variables (Table 1). Therefore, these points need to be consideredwhen interpreting our results.

To conclude, we show that tree radial growth responses of Scots pine to extreme 545 drought depend on site condition, tree growth prior to drought, and the number of droughts 546 that a tree experienced within the 10 years before the selected drought. Our study identified a 547 reduced tree growth resilience to extreme drought in Scots pine trees growing at mid-548 549 elevation and low productivity sites likely driven by the more frequent and severe drought events that occurred at those sites in recent years. We show that the assessment of tree growth 550 551 responses in terms of resistance, recovery, and resilience to extreme drought using radial growth data is challenging along large ecological and biogeographical gradients, since a 552 multitude of location-specific tree-, site-, and drought-related factors and their interactions 553 drive tree growth performances. 554

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578 **Conflict of Interest**

579 None is declared.

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Models	Hypotheses	References	AICc	Δ AICc	AICc	R ²	R ²
					weight	(Fixed)	(Fixed an
							random)
Resistance (RT)	Tree resistance to drought is affected by						
RT ~ all variables	Full model		117.9	0.0	1.00	0.33	0.49
RT~D_INT*PGR+	the intensity of the drought, but depending upon the growth prior to drought	Adapted from Gazol et al.,	130.4	12.5	0.00	0.22	0.46
D_INT*ST	and site types	(2018)					
RT~D_FRE*PGR	the frequency of the drought, but depending upon the growth prior to drought	Adapted from Gao et al.,	130.5	12.6	0.00	0.14	0.47
		(2018)					
Recovery (RC)	Tree recovery after drought						
RC~D_INT*ST	is affected by the intensity of the drought, but depending upon the site types	Adapted from Gazol et al.,	301.9	0.0	0.74	0.16	0.36
		(2018)					
RC~ST	decreased with site types	Sánchez-Salguero et al., 2018	307.0	5.1	0.06	0.07	0.37
RC~D_INT+ D_FRE+	is affected combinedly by intensity of drought, duration of drought, and	Gao et al., 2018	308.4	6.5	0.03	0.08	0.37
D_DUR	frequency of drought						
Resilience (RS)	Tree resilience to drought is						
RS~ D_DUR*PGR	affected by the duration of the drought, but depending upon the growth prior	Adapted from Taeger et al.,	342.4	0.0	0.45	0.03	0.35
	to drought	(2013)					
RS~PGR	negatively associated with the growth prior to drought	Zang et al., 2014; Ruijven and	344.3	1.9	0.17	0.02	0.35
		Berendse 2010					
RS~D_FRE*PGR	affected by the frequency of the drought, but depending upon the growth prior	Adapted from Gao et al.,	344.6	2.2	0.15	0.02	0.36
	to drought	(2018)					

949 Table 1. Results of the best models explaining tree growth resistance, recovery and resilience of Scots pine trees along the studied gradient. From the 16 tested models, only the three with the highest Akaike's

977 productivity)), D_FRE= number of drought years within the past 10 years from the studied drought, D_INT= intensity of drought, and D_DUR= duration of drought, *indicates an interaction term and +indicates an

978 additive term, PGR was quantified from tree growth performances during the 10 consecutive years prior to drought excluding the years considered as pre-drought period quantifying the three indices (i.e., resistance,

979 recovery, and resilience).

981 Table 2. Log-transformed estimates of predictor variables and 95% confidence intervals (CI) based on model averaging for the three response variables resistance, recovery, and resilience. Only predictor variables that

982 had a strong effect (i.e., a 95 % confidence interval excluding 0) are presented.

Parameters	Types of effect	Estimate (β)	Lower 95% CI	Upper 95% CI
Resistance				
Pre-drought growth	Additive	-0.13	-0.22	-0.04
High-elevation and lower productivity vs mid-elevation lower productivity	Additive	0.31	0.09	0.52
Low-elevation higher productivity vs mid-elevation lower productivity	Additive	0.19	0.02	0.36
Drought intensity*Pre-drought growth	Interaction	0.005	0.002	0.008
Drought duration*Pre-drought growth	Interaction	-0.22	-0.35	-0.09
Drought frequency	Additive	-0.13	-0.20	-0.06
Recovery				
High-elevation and lower productivity vs mid-elevation lower productivity	Additive	-0.23	-0.45	-0.02
Low-elevation higher productivity vs mid-elevation lower productivity	Additive	-0.16	-0.34	-0.01
Drought frequency	Additive	0.08	0.01	0.16
Resilience				
Pre-drought growth	Additive	-0.18	-0.28	-0.07
Pre-drought growth variability	Additive	-0.22	-0.43	-0.01

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Figure 1. Location of the 30 Scots pine study sites distributed along a latitudinal gradient that ranged from southern Spain to northern Germany. Where, LEHP: low-elevation
 sites with high productivity, MELP: mid-elevation sites with low productivity, and HELP: high-elevation sites with low productivity. The grey shade used as a background
 within the map represents the natural distribution of Scots pine adapted from Mátyás *et al.* (2004).



Figure 2. Seasonal correlations between Scots pine tree-ring width indices and the Standardized Precipitation-Evapotranspiration Index (SPEI) for the period of approximately 1960-2011 across the latitudinal gradient. Only
the seasons that exhibited the strongest effect on tree-ring width indices are plotted (see materials and methods).
Note. 'previous' refers to the year previous to tree ring formation, while 'current' refers to the current year of ring formation, summer: June, July, and August, spring: March, April, and May, autumn: September, October, and November. LEHP: low-elevation sites with high productivity; MELP: mid-elevation sites with low
productivity; and HELP: high-elevation sites with low productivity. Pearson's product-moment correlation with

a threshold <0.05 was used for statistical significance. Correlation magnitude: the larger the circles, the stronger
the correlations. See Table S7 for correlation scores that displayed in this figure.



Figure 3. Mean annual radial growth in drought and non-drought years for the period of approximately 1980 2011 across the three site types (i.e., LEHP: low-elevation sites with high productivity; MELP: mid-elevation sites with low productivity; and HELP: high-elevation sites with low productivity). The error bars represent the
 mean ± standard error (n=615). The letters on top of the bars show the results (a<b<c) of the post hoc Tukey
 multiple comparison test with a threshold <0.05 for statistical significance indicating the differences among the three site types and between non-drought years and drought years within each site type.





Figure 4. Tree-level resistance (A), recovery (B), and resilience (C) to the most extreme drought during 1980-1999 and during 2000-2011 for three site types. The error bars represent the mean ± standard error (n=615). The letters on top of the bars show the results (a<b<c) of the post hoc Tukey multiple comparison test with a threshold <0.05 for statistical significance indicating the differences among the three site types and between the two periods within each site type. LEHP: low- elevation sites with high productivity, MELP: mid- elevation sites with low productivity, and HELP: high-elevation sites with low productivity.



Figure 5. Tree growth resistance, recovery, and resilience to the most extreme drought during 1980-2011 with
 95% confidence intervals. Note. Average pre-drought growth (ring width indices) were quantified from tree growth performances during the 10 consecutive years prior to drought excluding the years considered as pre drought period for quantifying the three indices (resistance, recovery, and resilience). LEHP: low-elevation sites
 with high productivity, MELP: mid-elevation sites with low productivity, and HELP: high-elevation sites with low productivity. See statistics for the fitted line in Table 1.



1141 frequency (number of droughts occurred within 10 years preceding to extreme drought) over the two 1142 periods across the three site types. The error bars represent the mean±standard error (n=30). The 1143 letters on top of the bars show the results of the post hoc Tukey multiple comparison test with a 1143 threshold <0.05 for statistical significance. LEHP: low-elevation sites with high productivity, MELP:</p>

- 1144 mid-elevation sites with low productivity, and HELP: high-elevation sites with low productivity.
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