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8 *Running head:* Intra-species variability in pines' allometry

9 **Adaptation and plasticity in aboveground allometry variation of four pine species**  
10 **along environmental gradients**

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### 33 **Abstract**

34 Plant species aboveground allometry can be viewed as a functional trait that reflects the  
35 evolutionary trade-off between above- and belowground resources. In forest trees,  
36 allometry is related to productivity and resilience in different environments, and it is  
37 tightly connected with a compromise between efficiency-safety and competitive ability.  
38 A better understanding on how this trait varies within and across species is critical to  
39 determine the potential of a species/population to perform along environmental  
40 gradients. We followed a hierarchical framework to assess tree height-diameter  
41 allometry variation within and across four common European *Pinus* species. Tree  
42 height-diameter allometry variation was a function of solely genetic components –  
43 approximated by either population effects or clinal geographic responses of the  
44 population's site of origin–, and differential genetic plastic responses –approximated by  
45 the interaction between populations and two climatic variables of the growing sites,  
46 (temperature and precipitation)–. Our results suggest that, at the species level, climate of  
47 the growing sites set the tree height-diameter allometry of xeric and mesic species  
48 (*Pinus halepensis*, *P. pinaster* and *P. nigra*) apart from the boreal species (*P. sylvestris*),  
49 suggesting a weak signal of their phylogenies in the tree height-diameter allometry  
50 variation. Moreover, accounting for inter-population variability within species for the  
51 four pine species aided to: (i) detect genetic differences among populations in allometry  
52 variation, which in *P. nigra* and *P. pinaster* were linked to gene pools –genetic diversity  
53 measurements–; (ii) reveal the presence of differential genetic variation in plastic  
54 responses along two climatic gradients in tree allometry variation. In *P. sylvestris* and *P.*  
55 *nigra*, genetic variation was the result of adaptive patterns to climate, whilst in *P.*  
56 *pinaster* and *P. halepensis*, this signal was either weaker or absent, respectively; and  
57 (iii) detect a local adaptation response in the exponent of the tree height-diameter  
58 allometry relationship in two out the four species (*P. sylvestris* and *P. nigra*), as it was a  
59 function of populations' latitude and altitude variables. Our findings suggest that the  
60 four species have been subjected to different historical and climatic constraints that  
61 might have driven their aboveground allometry and promoted different life strategies.

62

63 **Keywords:** Bayesian modeling; Climatic and geographical clines, Environmental  
64 gradients; Functional trait; Iberian Peninsula; Intra-species variability; Provenance tests.

## 65 **Introduction**

66 Aboveground allometry is considered a functional trait that links the changes in total  
67 height to those in stem diameter and reflects the evolutionary outcome in plant species  
68 dynamics for above and belowground resources (Hallé, Oldeman & Tomlinson 1978;  
69 King 1996). Both height and stem diameter are tightly associated with species foraging  
70 and resource allocation strategy (Tilman 1988; Poorter *et al.* 2012): while tree height  
71 reflects a strategy for securing carbon profit via light capture (Moles *et al.* 2009), stem  
72 diameter is closely related to mechanical support and water-absorbing capacity  
73 (McMahon 1973; Niklas 1993; Bullock 2000). A finite set of allometric outcomes is  
74 then expectable, due to trade-offs in plant allocation strategies along resource gradients  
75 (*sensu* Tilman (1988)) or biomechanical and hydraulic constraints (e.g. Ryan & Yoder  
76 1997; Chave *et al.* 2005; Mäkelä & Valentine 2006).

77 Tree height-diameter allometry has profound effects on species fitness and  
78 consequently on ecosystem structure. It correlates with bioclimatic variables (e.g. Aiba  
79 & Kohyama 1996; López-Serrano *et al.* 2005; King *et al.* 2006), and can change along  
80 biotic and abiotic gradients such as those for temperature, aridity and competition (e.g.  
81 Banin *et al.*, 2012; Lines *et al.*, 2012). However, intraspecific variation in allometry has  
82 usually been neglected and most studies have focused either on the species level or on  
83 the broad geographical scales (López-Serrano *et al.* 2005; Chave *et al.* 2005; Dietze,  
84 Wolosin & Clark 2008; Poorter *et al.* 2012; Lines *et al.* 2012; but see Vieilledent *et al.*  
85 2010; Pretzsch & Dieler 2012, that considered individual variability). The extent and  
86 patterns of variation in inter-population genetic in tree height-diameter allometry still  
87 remains unclear. Those patterns could be as a result of adaptive or neutral genetic  
88 processes, such as past events e.g. migration, bottlenecks, drift, etc., at different scales  
89 (species, population or individual), of plastic responses to the environment, or by any  
90 combinations of them. Consequently, aboveground allometry emerges as a  
91 comprehensive and integrative trait in which the pattern of allocation variation within  
92 species could be driven by climate and inter-population genetic variation. A deep  
93 understanding of these interconnected levels of variability (species and populations) in

94 tree height-diameter allometry is necessary to forecast the full potential of tree species  
95 to adapt and/or evolve under climate change conditions (e.g. Benito-Garzón *et al.* 2011;  
96 Valladares *et al.* 2014).

97 Common gardens experiments are established for testing genetic differences  
98 among populations grown under similar environmental conditions and generate valuable  
99 information for the study of intraspecific genetic variation (e.g. Matyas 1996; Alberto *et*  
100 *al.* 2013). Multi-locality common gardens, additionally, allow studying phenotypic  
101 plastic responses along environmental gradients and to identify genetic variation on  
102 them (i.e. population-environment interaction) and the adaptive value of those responses  
103 as well (i.e. correlations between the growing environments and local environments of  
104 population's origin).

105 In this study, we used total height and stem diameter –over bark– measured in  
106 multi-locality common garden tests to assess allometry relationships for the four most  
107 planted European pine species: *Pinus sylvestris* L., *P. nigra* Arnold, *P. pinaster* Aiton  
108 and *P. halepensis* Miller. The first two species (*P. sylvestris* and *P. nigra*) belong to the  
109 *Pinus* subsection that corresponds to Eurasian pines; and the last two species to the  
110 *Pinaster* subsection which relates to Mediterranean pines. Accordingly, these species  
111 display differentiated demographic backgrounds and genetic compositions resulting in a  
112 predictable pattern along temperature and water availability gradients across Europe  
113 (Richardson 1998; Tapias *et al.* 2004; Soto *et al.* 2010). We implemented a flexible log-  
114 linear model taking into consideration each species' population origin and associated  
115 geographic characteristics (to account for intraspecific genetic variation), and the  
116 climatic characteristics of the growing site (to account for the among-site variation).  
117 Correspondingly, we tested three hypotheses: (i) the patterns of height-diameter  
118 allometry variation in pines are driven by both the species and the inter-population  
119 variation; (ii) at the species level, tree allometry will vary depending on the climatic  
120 characteristics of the species, and specially with marked differences among Eurasian an  
121 Mediterranean species; (iii) inter-population variation in allometry could be the result of  
122 adaptation to local environments –namely climate and/or geographical variables of  
123 populations' site of origin– or historical events that took place in the past of the species.  
124 Testing these hypotheses will allow us to understand the underlying abiotic drivers that  
125 shape allometry variation at two interconnected levels, species and populations, and to  
126 identify the class of adaptive responses if existent. Understanding phenotypic  
127 integration of tree species responses along abiotic conditions could then assist in

128 forecasting the performance of forest species and populations in the context of global  
129 warming.

130

## 131 **Methods**

### 132 *Plant material and common garden provenance tests*

133 Aboveground allometry was measured in multi-locality common garden provenance  
134 tests located in Spain for four pines species: *Pinus sylvestris*, *P. nigra*, *P. pinaster* and  
135 *P. halepensis*. Populations from the distribution range of the species, mostly from the  
136 Iberian Peninsula (Spain and Portugal), were sampled by collecting seed lots from at  
137 least 25 mother trees with a 50-meter separation distance. Plants originating from the  
138 seed lots were collected in different populations (22 for *P. sylvestris*, 23 for *P. nigra*, 52  
139 for *P. pinaster* and 56 for *P. halepensis*), and established in comparative common  
140 garden provenance tests for each species (Fig. 1 and see Table S1 in Supporting  
141 Information).

142 Measurements were collected at  $11 \pm 1$  years of age, depending on the common  
143 garden tests, for two variables: *height* (total height in cm, measured with a pole) and  
144 *dbh* (diameter at breast height [130 cm] in mm, measured with a caliper). A common  
145 age was chosen to minimize species differences along their ontogenies (López-Serrano  
146 et al. 2005), and avoid confusion of allometric changes due to size, known as ‘passive  
147 or apparent plasticity’ (Wright & McConnaughay 2002). We also selected a young age  
148 to minimize inter-population competition effects in the experimental design. A previous  
149 study using the same experiment setup as in the present research did not find either inter  
150 or intra population competition effects in any of the two variables measured (*height* and  
151 *dbh*) in 32-year-old *P. pinaster* individuals (Alía, Moro & Denis 2001).

152 In total, we used data from 4,853 *P. sylvestris* trees from 22 populations planted  
153 in 6 sites; 3,644 *P. nigra* trees from 23 populations in 8 sites; 9,976 *P. pinaster* trees  
154 from 54 populations in 4 sites; and 1,928 *P. halepensis* trees from 56 populations in 3  
155 sites.

156

### 157 *Climatic and geographical data*

158 Each site was characterized by a set of 47 climatic variables: minimum, average and  
159 maximum mean monthly temperature, minimum and maximum average temperature of  
160 the coldest and warmest months and total and seasonal precipitation. As we lacked real

161 climate data from weather stations, we estimated these variables based upon Gonzalo-  
162 Jiménez's (2010) climatic model for the Iberian Peninsula, with a 1-km<sup>2</sup> spatial  
163 resolution, from climate data gathered between 1951 and 1999 (see Appendix text S1  
164 for further information).

165 According to both literature and exploratory analyses, we selected the subset of  
166 climatic variables at the growing sites most relevant to plant allometry for the four  
167 species. The selected variables were *MMT* (minimum average temperature of coldest  
168 month, °C), and *AP* (annual precipitation, mm). Both *MMT* and *AP* affect physiological  
169 and growth processes of plant species in the Mediterranean region (Thompson 2005),  
170 and have been consistently used in previous studies (e.g. Wang *et al.* 2006; O'Neill &  
171 Nigh 2011; Leites *et al.* 2012). Moreover, these variables presented substantial  
172 correlation with *height* and *dbh* variables (see Table S3). Geographical variables of the  
173 populations' site of origin, such as latitude, longitude and altitude, are surrogates for  
174 environmental conditions, e.g. the amount of heat energy received relative to the sun  
175 angle, temperature, humidity, and solar radiation; and they can usually reflect adaptation  
176 patterns to local conditions (see Alberto *et al.* 2013). Climatic variables of the growing  
177 sites and geographic variables of the populations' site of origin were then standardized  
178 before analyses to ease comparison among variables in the model.

179 Although the number of growing sites is low (ranging between 3 and 8), they  
180 cover most of the natural climatic range associated with each species distributional  
181 range, including contrasting climates (Ruiz-Benito *et al.* 2013), (see Fig. S1).

### 182 183 *Statistical models*

184 We estimated tree height as a function of diameter by using three classic allometric  
185 functions (Linear, Power, and Gompertz), and two link functions (normal and log-  
186 normal) and implemented generalized linear models (GLMs). The best allometric model  
187 fitting the data was selected using the Deviance Information Criteria, DIC (Spiegelhalter  
188 *et al.* 2002). A power function with a log-normal link function was the best model for  
189 two out of the four species, and the second best model for the other two species (see  
190 Table S2). We selected a common allometric model, power function with a log-normal  
191 link, for the four species to facilitate parameter comparisons.

192 Based on this allometric model, we constructed a hierarchical model (Clark  
193 2005, 2007). These models are more appropriate to connect and represent the biological  
194 hierarchy of the data, e.g. populations within species. To build the best final model, we

195 considered several variations of the basic model (i.e. in Eq. 1), where  $a$  and  $c$  scaling  
 196 parameters were constant, and they were estimated with different combinations of the  
 197 variables associated with the growing sites and the origin of populations. The best final  
 198 model structure was selected based on both biological relevance and the DIC criterion.

199 The final model estimated tree height allometry as a combination of climate at  
 200 the growing site ( $s$ ) and geographic characteristics at the origin site of the population  
 201 ( $p$ ).

202 Considering an individual  $i$ , from population  $p$  growing in growing site  $s$ , its height-  
 203 diameter allometry was modeled as:

204 Likelihood:  $\text{height}_i \sim \log \text{Normal}(H_i, \sigma^2)$

205 and the following process model:

$$206 \quad H_i = \ln(a_{p(i),s(i)}) + c_{p(i)} \times dbh_i \quad \text{Eq. 1}$$

207 where the scaling coefficient  $\ln(a_{p(i),s(i)})$  was estimated as:

$$208 \quad \ln(a_{p(i),s(i)}) = \alpha_{1p} + \alpha_{2p} \times MMT_s + \alpha_{3p} \times AP_s \quad \text{Eq. 2}$$

209 and the scaling exponent,  $c_{p(i)}$ , was estimated as:

$$210 \quad c_{p(i)} = \beta_1 + \beta_2 \times LAT_p + \beta_3 \times ALT_p \quad \text{Eq. 3}$$

211 Tree height-diameter allometry, therefore, is the outcome result of population  
 212 genetic effects on the basal height, parameterized in  $\alpha_{1p}$ ; plus a genetic (population)  
 213 clinal geographical pattern of the scaling exponent on latitude and altitude ( $\beta_2 \times LAT_p$   
 214 ,  $\beta_3 \times ALT_p$ ), and of genetic differential plastic responses along temperature and or  
 215 precipitation gradients of the growing site ( $\alpha_{2p} \times MMT_s$ ,  $\alpha_{3p} \times AP_s$ ). Because all  
 216 explanatory variables were standardized, parameter  $\alpha_{1p}$  was the allometric curve's  
 217 intercept at average climate conditions of across all growing sites. A summary of model  
 218 parameters, significance, and insights that can be assessed on each one is shown in  
 219 Figure 2.

220

### 221 *Model parameters estimation and posthoc comparisons*

222 Parameters were estimated following a Bayesian approach highly suited for hierarchical  
 223 analyses (Gelman & Hill 2007). Each of the population level parameters,  $\alpha_{*p}$ , was  
 224 estimated from a species-level prior normal distribution, with hyperparameters  $\mu^*$  and  
 225  $\sigma_{\alpha^*}^2$ ,  $\alpha_p^* \sim N(\mu^*, \sigma_{\alpha^*}^2)$ , estimated from non-informative prior distributions  $\mu^* \sim N(0,$

226 1000) and  $\sigma_{\alpha^*} \sim \text{Uniform}(0, 100)$ . These species-level parameters  $\mu^*$  and  $\sigma_{\alpha^*}^2$  would  
 227 correspond *sensu stricto* to an inter-population average among the studied populations.  
 228 However, we refer to these parameters *sensu lato* as representative of a species proxy  
 229 response. Note that for the scaling coefficient we have referred in Eq. 2 to populations'  
 230 parameters to enhance comprehension of the full scope of the relationship, instead of  
 231 including the species' parameters.

232 Parameters  $\beta^*$  were also estimated from non-informative prior distributions,  $\beta^* \sim N(0, 1000)$ . Variance associated with the individual random effects was estimated as  
 233  $1/\sigma^2 \sim \text{Gamma}(0.01, 0.01)$ . As standard deviation of residual errors around a fitted  
 234 power function might increase with diameter, we tested whether the residuals were a  
 235 linear function of diameter, as recommended in Lines *et al.* (2012). However, our  
 236 residuals did not show this trend, so we considered unnecessary to account for diameter  
 237 size in the estimation of the variance.  
 238

239 We formally tested marginal significant intraspecific genetic differences in  $\alpha_p^*$   
 240 for each species by computing all pairwise combinations of population differences  
 241 accounting for the 95% credible interval, CI, of the estimated parameter distribution  
 242 (e.g. intraspecific genetic differences in  $\alpha_p^* = \alpha_{p_A}^* - \alpha_{p_B}^*$ , being A and B two  
 243 populations of a specific pine species), while the rest of variables were kept to their  
 244 mean values in the range, that is why we refer to these differences as 'marginal'. Two  
 245 populations were significantly different if zero was not included in the credible interval  
 246 around their difference. Additionally, we quantified the level of marginal intraspecific  
 247 genetic differences as the percentage of the total number of significant pairwise  
 248 comparisons relative to the total number of pairwise comparisons within species.  
 249 Finally, to end the characterization of intraspecific genetic variability within species, we  
 250 provided the range of variability among populations within species as the standard  
 251 deviation of  $\alpha_p^*$ , i.e. the set of parameters estimated for each population.

252 Models were run in OpenBUGs (version 3.2.2 rev 1063) (Thomas *et al.* 2006).  
 253 Three chains were run for ~50,000 iterations and parameters convergence was reached  
 254 after ~25,000 iterations. After the burn-in period, chains were thinned (every 100) to  
 255 reduce autocorrelation, then posterior parameter values (mean and 95% credible  
 256 intervals) were calculated. Plots of predicted vs. observed values were also used to  
 257 evaluate model fit (unbiased models having a slope of one and  $R^2$  values indicating  
 258 goodness-of-fit). A slope parameter was considered to be statistically significant when



259 the 95% credible interval (CI) did not include zero. Population level parameters were  
 260 considered significantly different when their 95% CI did not overlap (or the 95% CI  
 261 around their difference did not include zero).

262

263

264 *Species' adaptive patterns in height-diameter allometry variation.*

265 We tested whether variation in the  $\alpha_p^*$  parameters was the result of neutral and/or  
 266 adaptive responses to local environments, by using two different approaches. First, to  
 267 assess the influence of neutral responses on  $\alpha_p^*$  parameters of allometry, we tested the  
 268 influence of previously defined gene pools groups for each species on  $\alpha_p^*$  parameters.  
 269 Gene pools are proxies of genetic relationships among populations, indicating common  
 270 demographic or evolutionary factors. In *P. sylvestris*, *P. pinaster* and *P. halepensis*,  
 271 gene pool groups were defined using molecular markers in Robledo-Arnuncio et al.  
 272 (2005) and in Bucci et al. (2007). In *P. nigra*, we lacked information based on  
 273 molecular markers; hence populations were grouped by sub-species. One-way ANOVA  
 274 was used to detect the existence of association between  $\alpha_p^*$  parameters and groups, and  
 275 post-hoc comparisons with a HSD Tukey's test were employed. When homogeneity and  
 276 normality assumptions were not reached, non-parametric Kruskal-Wallis test and post-  
 277 hoc comparisons with a Nemenyi test, corrected for ties if necessary, were used.  
 278 Second, to identify the influence of local environments –namely climate and/or  
 279 geographical position of populations' site of origin– on tree allometry variation; we  
 280 tested the existence of climatic or geographical clines due to local adaptations in  
 281 phenotypic plasticity. More specifically, we computed Pearson correlation coefficients  
 282 ( $\rho$ ) between  $\alpha_p^*$  parameters and climate and geographical characteristics of the  
 283 populations' site of origin.

## 284 **Results**

### 285 *Patterns of height-diameter allometry variation across and within species*

286 The four final models produced unbiased estimates of *height* with high  $R^2$  of observed  
 287 vs predicted values (0.90 in *P. sylvestris*, 0.91 in *P. nigra*, 0.85 in *P. pinaster* and 0.89  
 288 in *P. halepensis*).

289 *Pinus pinaster* had the lowest intercept value, measured by hyperparameter  $\mu_1$ ,  
 290 and it did not overlap with the other three species. *P. nigra* and *P. sylvestris* had

291 intermediate but overlapping values, while *P. halepensis* had the highest value and did  
292 not overlap with any of the other pine species (Table 1). The intraspecific variability,  
293 standard deviation of  $\alpha_{1p}$ , also varied among species. *P. pinaster* displayed the greatest  
294 value, followed by *P. nigra*, *P. sylvestris* and *P. halepensis* (Table 1). This intraspecific  
295 variability can also be visualized in Figure S2. Moreover, there was significant  
296 intraspecific genetic variation in  $\alpha_{1p}$  in all species, measured by the number of  
297 significant pairwise comparisons within species relative to the total number: *P. pinaster*  
298 was the species with the highest percentage of significant differences (50.24 %);  
299 followed by *P. nigra* (40.32%), *P. sylvestris* (31.17%) and *P. halepensis* (17.21%).

300 Temperature (*MMT*) of the growing site influenced tree height allometry, being  
301 this hyperparameter,  $\mu_2$ , statistically significant and positive in three out of the four  
302 species, and significant but negative in *P. sylvestris* (Table 1). Moreover, we found  
303 evidence of inter-population genetic differentiation in phenotypic plasticity to  
304 temperature (*MMT*) in three out of the four species (except *P. halepensis*). The four  
305 species showed some degree of intraspecific genetic variability, *P. sylvestris* having the  
306 greatest standard deviation, followed by *P. pinaster*, *P. nigra* and *P. halepensis*.  
307 Specifically, the level of significant intraspecific genetic variation varied according to  
308 each species. Thus, *P. sylvestris* displayed the greatest level of genetic differences in  
309 plasticity in response to *MMT* (38.10 %) among the populations tested, followed by the  
310 other two species: *P. nigra* (9.88%) and *P. pinaster* (9.57%). All these results should be  
311 considered based on the total of populations tested. This intraspecific genetic variability  
312 can also be visualized in Figure S3.

313 Annual precipitation (*AP*) also influenced tree height allometry. Values for  
314 hyperparameter  $\mu_3$  were statistically significant and negative in three out the four  
315 species, but positive in *P. sylvestris*. The estimated values for *P. sylvestris* and *P.*  
316 *halepensis* did not overlap, but the pairs composed by *P. nigra* and *P. pinaster*, and *P.*  
317 *pinaster* and *P. halepensis* did (Table 1). Similarly, we found intraspecific genetic  
318 differences in phenotypic plasticity to rainfall (*AP*) in three out of four species, the  
319 exception again being *P. halepensis*. Furthermore, the four pine species presented some  
320 degree of intraspecific variability. *P. sylvestris* and *P. pinaster* presented similar  
321 degrees, followed by *P. nigra* and *P. halepensis*, (Table 1). The level of significant  
322 intraspecific genetic variation was greatest in *P. sylvestris* (29.87%), followed by *P.*  
323 *pinaster* (3.60%) and *P. nigra* (2.77%). This intraspecific variability can also be  
324 visualized in Figure S4.

325 In three out the four species, the effects of *AP* on the tree height-diameter  
 326 relationship and also the intraspecific genetic variability were smaller than those  
 327 reported in response to *MMT*; specifically, between ca. 2 and 3.5-folds greater for *P.*  
 328 *nigra*, *P. halepensis* and *P. pinaster* –in an increasing order–. Interestingly, the opposite  
 329 effect was found in *P. sylvestris* –the effect of *AP* was ca. 2.5 folds greater than *MMT*–.

330

### 331 *Species' adaptive patterns in height-diameter allometry variation*

332 Overall, we found that tree height allometry variation was the result of adaptive  
 333 responses to either local environments –climate and geographical sites of origin– or to  
 334 past historical events in the demography of species. First, we found a significant  
 335 geographical cline, i.e. an association between the scaling exponent parameter ( $c_p$ ) and  
 336 the latitude of origin for two of the four pine species (*P. sylvestris* and *P. nigra*), but not  
 337 for the other two, more xeric, species (*P. pinaster* and *P. halepensis*) (Table 1). Second,  
 338 gene pool groups were significantly associated with  $\alpha_{1p}$  values just in *P. pinaster* and *P.*  
 339 *nigra* ( $p < 0.001$  and  $p < 0.05$ , respectively), but not in the two others (Table 2).

340 Third, we found chiefly stronger local environment –namely climate and  
 341 geographical position of populations' site of origin– associations with  $\alpha_{1p}$  parameters  
 342 compared to  $\alpha_{2p}$  and  $\alpha_{3p}$ . Interestingly, *P. halepensis* was the only species that lacked  
 343 any type of relationship, suggesting the inexistence of climate adaptive responses in tree  
 344 allometry variation. Specifically,  $\alpha_{1p}$  values were significantly correlated ( $p < 0.05$ ) to  
 345 different climatic variables of population's site of origin (Table 3). *P. sylvestris* with  
 346 altitude ( $\rho = 0.56$ ) and annual precipitation ( $\rho = 0.54$ ); *P. nigra*, in general, with  
 347 minimum average monthly temperature (ranging from  $\rho = 0.45$  to  $0.60$ ); and, weaker  
 348 than the previous two, *P. pinaster* with spring precipitation ( $\rho = 0.28$ ) and mean  
 349 temperature of the warmest month ( $\rho = -0.28$ ). Parameters  $\alpha_{2p}$  and  $\alpha_{3p}$  were  
 350 significantly correlated to climatic variables of populations' sites of origin ( $p < 0.05$ ); *P.*  
 351 *sylvestris*, *P. nigra* and *P. pinaster* displaying significant correlations between  $\alpha_{2p}$  and  
 352 climate, although the associations were weaker in the last species. Finally, we only  
 353 found significant and positive correlations between  $\alpha_{3p}$  and related temperature  
 354 variables in *P. sylvestris* (Table 3).

355 **Discussion**

356 We addressed inter-population tree height-diameter allometry variation across  
 357 precipitation and temperature gradients of the four most planted pine species in Spain,  
 358 i.e. tree height-diameter allometry variation was accounted at two interconnected levels,  
 359 species and population. Additionally, we found that local adaptation and past historical  
 360 events of the species were associated to inter-population allometry variation, except for  
 361 *P. halepensis*, the most xeric species among the four. We employed a hierarchical  
 362 approach to better understand the contributions of the species' genetic variation, their  
 363 demographic genetic background and their phenotypic plasticity, in their responses to  
 364 environmental variability.

365

366 *Patterns of height-diameter allometry variation across and within species*

367 This is the first time, to our knowledge, that the patterns of aboveground allometry  
 368 across climatic gradients have been described including intraspecific variation from a  
 369 genetic perspective. The species-level parameters related to allometry (hyperparameters  
 370  $\mu_*$  and  $\beta_*$ , Figure 2) did not overlap among species in most comparisons, reflecting the  
 371 existence of evolutionary species-specific allocation strategies to cope with the current  
 372 environment, although revealing an unclear association with their phylogeny, because  
 373 *P. nigra* behaved more like *P. halepensis* and *P. pinaster* than like *P. sylvestris*.

374 In general, *P. sylvestris* showed the opposite pattern in regards to aboveground  
 375 allometry variation across precipitation and temperature gradients with respect to the  
 376 other three species. Aboveground variation was also more responsive to changes in the  
 377 minimum average temperature of the coldest month, *MMT*, than to changes in annual  
 378 precipitation, *AP* (excluding again *P. sylvestris*). This is contrary to expectation as  
 379 Iberian forests are strongly constrained by water availability (Gómez-Aparicio *et al.*  
 380 2011; Ruiz-Benito *et al.* 2013). We hypothesize that mesic and xeric pine species could  
 381 be more limited by low temperatures than by water shortage, since they may have  
 382 developed adaptive mechanisms to cope with drought stress, such as tight stomatal  
 383 control, or specific wood anatomy traits such as thick cell walls, thick pit membrane,  
 384 narrow lumens or different root hydraulic resistance (Yastrebov 1996; Tyree &  
 385 Zimmermann 2002; García Esteban *et al.* 2009; Zuccarini *et al.* 2015). Yet our data did  
 386 not allow us to explore all potential interactions, e.g. too cool-too wet; too warm-too  
 387 wet that are likely to shape evolutionary responses in these species and populations.

388 Consistently with previous studies, in three out of the four pine species, taller  
389 heights at a given size are found under warmer conditions (Lines *et al.* 2012), except for  
390 *P. sylvestris*. Warmer conditions during the cold season might allow a higher  
391 photosynthetic capacity, resulting in a higher rate of carbon assimilation (Way & Oren  
392 2010), and hence taller heights. Yet *P. sylvestris* showed the opposite trend, suggesting  
393 a lack of responsiveness to warmer winters. A similar result was reported by Reich &  
394 Oleksyn (2008) in a regional study in Northern Europe (latitude from 46° to 68° N).  
395 They observed that *P. sylvestris* responses to climate differed between northern and  
396 southern populations: while in southern populations height decreased as temperature  
397 increased, the opposite was observed in northern populations. It was suggested that, at  
398 least for this temperate-boreal species, warmer temperatures –at its warmer range–  
399 might rather enhance heat stress and heat-induced moisture stress than alleviate cold  
400 stress.

401 Tree height allometry variation across the precipitation gradient resulted species-  
402 specific and diverse. *P. sylvestris* is expected to decrease its height at given size under  
403 drier conditions, a common pattern found in many parts of the world, e.g. Méndez-  
404 Alonzo *et al.* (2008). This variation has often been attributed to the changing hydraulic  
405 structure of vessels in drought-prone areas. The opposite, however, was observed for *P.*  
406 *halepensis*. Periods of soil moisture saturation and flooding may act as stressors in arid-  
407 climate forests by reducing tree height (Rodríguez-González *et al.* 2010). Also higher  
408 precipitation levels in some regions could imply poorer soil quality, because of  
409 increased runoff and nutrient leaching. However, as we do not have these  
410 measurements, we cannot confirm its potential influence. Intermediate patterns in tree  
411 allometry variation were shown for *P. pinaster* and *P. nigra*, which displayed negligible  
412 variation along the precipitation gradient tested. Lines *et al.* (2012) found a clear pattern  
413 of allometric variation across species along the studied precipitation gradient, although  
414 not within species. That finding together with ours suggest that tree height-allometry  
415 variation could have a very conservative performance across precipitation gradients.  
416 This latter would be in agreement the results presented in Table 1. Here, the estimated  
417 credible intervals for the hyperparameters in *P. pinaster* and *P. nigra* were very close to  
418 containing zero. This is somehow reflecting the almost lack of influence of precipitation  
419 on tree height-diameter allometry variation.

420

421 *Species' adaptive patterns in height-diameter allometry variation*

422 According to our findings, inter-population tree height allometry variation was the  
 423 result of local adaptation (Table 3). In addition, for *Pinus nigra* and *P. pinaster*, the  
 424 demographic history of the species associated to distinct neutral gene pools was also  
 425 important (as they reflect past events along the species' history, such as genetic  
 426 bottlenecks, founder effects, drift, etc.; Bucci *et al.* 2007; Soto *et al.* 2010; Jaramillo-  
 427 Correa *et al.* 2015) (Table 2). Thus, in these two species, gene pool signals correlate  
 428 with its allometry – its phenotype–, and therefore, such gene pools could be further used  
 429 for the study of different evolutionary processes on phenotype variation, although the  
 430 delineated groups were appreciably different compared to those based on DNA data  
 431 (Afzal-Rafii & Dodd 2007; Bucci *et al.* 2007). Absence of this signal in *P. sylvestris*  
 432 could be partly explained by a greater influence from local environments relative to  
 433 species historical background, as it is reflected by significant correlations between  $\alpha_p$   
 434 and local population climate. In contrast, *P. pinaster* and *P. nigra* presented weak  
 435 signals of adaptation to climate, specifically in plastic responses to temperature. Finally,  
 436 *P. halepensis* represented a different case; its null degree of genetic variation, in any of  
 437 the parameters of the model, agrees with the fact that the species' European western  
 438 populations are genetically uniform (Soto *et al.* 2010), due to a relatively recent long-  
 439 range colonization from its ancestral range in the eastern Mediterranean Basin (Grivet *et*  
 440 *al.* 2009).

441 The clear latitudinal variation in the scaling exponent parameter in *P. sylvestris*  
 442 and *P. nigra* reveals a consistent regional correlation in tree allometry and photoperiod.  
 443 Previous studies along latitudinal gradients have also found a genetic cline of adaptation  
 444 (e.g. northern populations set buds and hardened earlier, and presented lower growth  
 445 rates than the southern ones; see Alberto *et al.* 2013 and references therein). In any case,  
 446 the lack of latitudinal clines in *P. halepensis* and *P. pinaster* could be explained to either  
 447 insufficient span in our data or to a real lack of latitudinal variability. Interestingly, this  
 448 is the first time that adaptive patterns have been shown for a composite trait such as tree  
 449 height allometry. Our results confirm that this trait and its confined variation may be  
 450 under natural selection control and consequentially play an important role in both the  
 451 adaptation and acclimation potential of tree species to future conditions.

452 In conclusion, these four pine species are a heterogeneous group with a  
 453 recognized ability to adapt to extremely variable environments. Our findings support the

454 eco-evolutionary knowledge we already have about them, but nonetheless reveal that  
 455 height-diameter allometry variation patterns have developed under different natural  
 456 selection pressures, despite the study species sharing a sizeable part of their distribution  
 457 area in the studied region. This might have resulted in species, such as *P. halepensis*,  
 458 where phenotypic plasticity is more important than genetic variation; while for others,  
 459 e.g. *P. pinaster*, genetic variation and local adaptation might be more relevant.  
 460 Together, local environments – at the origin– and current growing conditions outline the  
 461 likely possible outcomes of integrated phenotypes.

462 The full potential of forest resilience and resistance along new temperature and  
 463 aridity gradients, i.e. climate-change driven, would depend on local adaptation and  
 464 levels of phenotypic plasticity of the populations. Our results point that considering both  
 465 the species specific and population ecological and historical background is key for  
 466 assessing likely population responses to environmental variation

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- 620

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628

629 **Conflict of interest:** None declared.

630 **Author contributions**

631 N.V.P., M.A.Z., S.C.G.M. and R.A. conceived this study; N.V.P. and I.I. analysed the  
 632 data; all authors contributed to the writing of the manuscript.

633 **Data accessibility**

634 All the data used is available in the Database on Genetic trials GENFORED  
 635 ([www.genfored.es](http://www.genfored.es)) upon request, and also it is in process of submitting into the  
 636 CitaREA repository (<http://citarea.cita-aragon.es/citarea/?locale=en>). Identificators for  
 637 the data will be available before publication.

638 **Table 1.** Parameter estimates from the selected best model. The table gathers  
 639 information of two interconnected levels of hierarchy, species and populations. The  
 640 parameters  $\mu_*$  and  $\beta_*$  make reference to the species, and  $\alpha_{*p}$  to populations within  
 641 species. The overall species response, i.e.  $\mu_*$  and  $\beta_*$  posterior mean estimates and 95%  
 642 credible intervals in square brackets [ , ] are given. The range of parameter values among  
 643 populations within species, i.e. posterior mean estimates of  $\alpha_{*p}$ , are shown in square  
 644 brackets, the lowest value is separated from the highest one by three dots [ ... ]. **Bold**  
 645 numbers indicate that fixed-effect coefficients were statistically significant (i.e. 95% CI  
 646 does not include zero). Letters indicate comparison and different letters indicate  
 647 differences among species for each of the parameters when statistically significant.

648

*P. sylvestris**P. nigra**P. pinaster**P. halepensis*

$\ln(a_{p,s})$	Intercept: $\mu_1$	<b>4.143 b</b> [4.123, 4.126]	<b>4.121 b</b> [4.093, 4.148]	<b>3.056 c</b> [3.023, 3.088]	<b>4.344 a</b> [4.302, 4.385]
	$\alpha_{1p}$	[4.107 ... 4.229]	[4.047 ... 4.198]	[2.907 ... 3.170]	[4.281 ... 4.396]
	$sd(\alpha_{1p})$	0.028	0.039	0.054	0.024
	MMT: $\mu_2$	<b>-0.030 d</b> [-0.049, -0.011]	<b>0.037 c</b> [0.026, 0.049]	<b>0.086 b</b> [0.072, 0.100]	<b>0.158 a</b> [0.143, 0.172]
	$\alpha_{2p}$	[-0.120 ... 0.026]	[0.014 ... 0.063]	[-0.002 ... 0.169]	[0.143 ... 0.174]
	$sd(\alpha_{2p})$	0.032	0.011	0.026	0.007
	AP: $\mu_3$	<b>0.080 a</b> [0.070, 0.095]	<b>-0.0184 b</b> [-0.028, -0.009]	<b>-0.023 bc</b> [-0.036, -0.009]	<b>-0.049 c</b> [-0.064, -0.035]
	$\alpha_{3p}$	[0.043 ... 0.128]	[-0.032 ... 0.000]	[-0.059 ... 0.032]	[-0.053 ... -0.043]
	$sd(\alpha_{3p})$	0.019	0.007	0.018	0.002
	$c_p$	Intercept: $\beta_0$	<b>0.426 b</b> [0.416, 0.431]	<b>0.412 bc</b> [0.405, 0.419]	<b>0.700 a</b> [0.691, 0.707]
LAT: $\beta_1$		<b>0.016 a</b> [0.007, 0.026]	<b>0.010 ab</b> [0.002, 0.017]	0.005 [-0.001, 0.012]	0.003 [-0.003, 0.008]
ALT: $\beta_2$		-0.006 [-0.015, 0.004]	-0.004 [-0.011, 0.004]	-0.007 [-0.013, 0.000]	-0.002 [-0.007, 0.004]

649

650 **Table 2.** A) Summary of one-way ANOVAs to test gene pool effects on  $\alpha_{*p}$ . When a  
651 non-parametric test was used, it is shown by the symbol  $\approx$ . B) Post-hoc comparisons  
652 among gene pools adjusted by Tukey's HSD for *P. nigra* and *P. pinaster*. Different  
653 letters indicate differences among gene pools.

654 A)				<i>P. pinaster</i>	$\alpha_{1p}$	12.43	***	
					$\alpha_{2p}$	14.23 $\approx$	n.s.	
					$\alpha_{3p}$	3.84 $\approx$	n.s.	
	<i>P. sylvestris</i>	$\alpha_{1p}$	0.60	n.s.	<i>P. halepensis</i>	$\alpha_{1p}$	1.07	n.s.
		$\alpha_{2p}$	0.57	n.s.		$\alpha_{2p}$	0.44	n.s.
		$\alpha_{3p}$	0.57	n.s.		$\alpha_{3p}$	0.08	n.s.
	<i>P. nigra</i>	$\alpha_{1p}$	6.95	**				
		$\alpha_{2p}$	7.20 $\approx$	655 n.s.				
		$\alpha_{3p}$	2.53 $\approx$	n.s.				

656 B)

Gene pools		$\alpha_{1p}$	sd	
<i>P. nigra</i>				
	spp. <i>laricio</i>	4.15	0.04	a
	spp. <i>salzmannii</i>	4.12	0.01	ab
	spp. <i>dalmatica</i>	4.08	*	ab
	spp. <i>nigra</i>	4.06	0.02	b
<i>P. pinaster</i>				
	Morocco	3.11	0.00	a
	Atlantic Iberian	3.10	0.02	a
	Eastern Spain	3.06	0.03	ab
	Southern Spain	3.06	0.05	abc
	Corsica	3.05	0.00	abc
	Central Spain	3.02	0.03	bc
	Italy	2.95	*	cd
	Eastern North Africa	2.91	0.00	d

657 \* only one datum, standard deviation was

658 not estimated.

659 **Table 3.** Heat map for Pearson's correlation coefficients,  $\rho$ , between  $\alpha_{*p}$  and climate  
660 variables from the populations' sites of origin. Dark grey indicates high positive  
661 correlation coefficients, light grey indicates high negative, and white color indicates  
662 low. Bold numbers mean significant correlations at  $p < 0.05$ .

663

	<i>P. sylvestris</i>			<i>P. nigra</i>			<i>P. pinaster</i>			<i>P. halepensis</i>		
	$\alpha_{1p}$	$\alpha_{2p}$	$\alpha_{3p}$	$\alpha_{1p}$	$\alpha_{2p}$	$\alpha_{3p}$	$\alpha_{1p}$	$\alpha_{2p}$	$\alpha_{3p}$	$\alpha_{1p}$	$\alpha_{2p}$	$\alpha_{3p}$
Latitude	<b>-0.45</b>	0.14	0.19	-0.05	-0.24	-0.09	<b>-0.48</b>	0.03	-0.07	0.20	0.17	-0.04
Longitude	<b>-0.45</b>	-0.02	0.11	-0.29	0.22	<b>0.37</b>	0.00	0.09	0.08	0.03	-0.04	0.00
Altitude	<b>0.56</b>	-0.17	-0.29	0.06	-0.23	-0.31	0.16	<b>-0.33</b>	0.01	0.01	0.02	0.07
MMTJan	-0.03	0.29	0.23	<b>0.61</b>	<b>0.42</b>	0.16	-0.07	<b>0.27</b>	0.01	0.03	-0.05	-0.17
MMTFeb	0.00	0.31	0.20	<b>0.54</b>	0.34	0.13	-0.05	0.25	0.02	0.02	-0.06	-0.17
MMTMar	-0.21	0.36	0.32	<b>0.45</b>	0.31	0.17	-0.06	0.25	0.00	0.02	-0.06	-0.15
MMTApr	<b>-0.48</b>	0.37	0.38	0.34	0.22	0.14	-0.07	0.20	-0.01	0.06	-0.01	-0.16
MMTMay	<b>-0.53</b>	0.40	0.41	0.32	0.18	0.14	-0.13	0.18	0.00	0.06	-0.06	-0.18
MMTJun	<b>-0.52</b>	<b>0.44</b>	<b>0.44</b>	0.33	0.21	0.12	-0.16	0.13	-0.03	0.06	-0.07	-0.22
MMTJul	<b>-0.48</b>	<b>0.53</b>	<b>0.47</b>	<b>0.45</b>	0.23	0.10	-0.20	0.05	-0.04	0.06	-0.11	-0.22
MMTAug	<b>-0.43</b>	<b>0.56</b>	<b>0.46</b>	<b>0.48</b>	0.25	0.11	-0.22	0.04	-0.07	0.09	-0.06	-0.20
MMTSep	-0.38	<b>0.50</b>	0.39	<b>0.51</b>	0.35	0.17	-0.19	0.16	-0.04	0.06	-0.04	-0.16
MMTOct	-0.30	<b>0.46</b>	0.40	<b>0.55</b>	0.36	0.19	-0.16	0.21	-0.04	0.07	-0.01	-0.14
MMTNov	-0.32	0.42	0.37	<b>0.57</b>	0.35	0.18	-0.09	0.24	0.00	0.09	-0.01	-0.15
MMTDec	-0.17	0.30	0.26	<b>0.59</b>	0.38	0.18	-0.06	<b>0.27</b>	0.01	0.06	-0.04	-0.18
MWTJan	-0.02	<b>0.44</b>	0.33	0.33	0.29	0.07	0.10	0.16	-0.02	-0.07	-0.13	-0.14
MWTFeb	-0.17	<b>0.46</b>	0.40	0.17	0.21	0.07	0.01	0.14	-0.04	-0.15	-0.16	-0.13
MWTMar	<b>-0.44</b>	<b>0.53</b>	<b>0.51</b>	-0.05	0.10	0.08	-0.06	0.06	0.01	-0.25	-0.20	-0.08
MWTApr	<b>-0.59</b>	<b>0.43</b>	<b>0.48</b>	-0.17	-0.01	0.05	-0.13	0.04	-0.02	-0.16	-0.10	-0.05
MWTMay	<b>-0.63</b>	<b>0.47</b>	<b>0.46</b>	-0.23	-0.13	-0.05	-0.20	-0.09	0.02	-0.18	-0.12	-0.10
MWTJun	<b>-0.57</b>	<b>0.54</b>	<b>0.52</b>	-0.21	-0.15	-0.09	-0.15	-0.14	0.01	-0.25	-0.11	-0.04
MWTJul	-0.23	<b>0.48</b>	0.37	-0.15	-0.15	-0.13	-0.15	-0.21	0.03	-0.25	-0.10	0.02
MWTAug	-0.20	<b>0.52</b>	0.39	-0.11	-0.17	-0.14	-0.14	-0.15	0.01	-0.25	-0.08	0.02
MWTSep	-0.37	<b>0.53</b>	<b>0.43</b>	-0.05	-0.07	-0.06	-0.14	-0.06	0.01	-0.24	-0.16	-0.06
MWTOct	-0.33	<b>0.55</b>	<b>0.50</b>	0.04	0.04	0.01	-0.15	0.11	0.00	-0.17	-0.08	-0.09
MWTNov	-0.16	<b>0.50</b>	0.42	0.28	0.16	0.05	-0.05	0.20	0.00	-0.06	-0.05	-0.11
MWTDec	-0.09	<b>0.47</b>	0.40	0.39	0.30	0.11	0.06	0.23	-0.03	-0.01	-0.13	-0.17
MTJan	-0.02	0.41	0.31	<b>0.50</b>	0.38	0.13	0.01	0.23	-0.01	-0.02	-0.09	-0.17
MTFeb	-0.10	<b>0.43</b>	0.34	0.37	0.29	0.10	-0.03	0.21	-0.01	-0.06	-0.12	-0.17
MTMar	-0.37	<b>0.49</b>	<b>0.46</b>	0.18	0.21	0.13	-0.07	0.18	0.00	-0.11	-0.13	-0.15



MTApr	<b>-0.56</b>	0.42	<b>0.45</b>	0.06	0.10	0.09	-0.12	0.14	-0.02	-0.03	-0.05	-0.13
MTMay	<b>-0.61</b>	<b>0.45</b>	<b>0.45</b>	0.00	0.00	0.03	-0.20	0.06	0.01	-0.05	-0.10	-0.17
MTJun	<b>-0.57</b>	<b>0.51</b>	<b>0.50</b>	0.01	0.00	-0.01	-0.19	-0.02	0.00	-0.10	-0.11	-0.18
MTJul	-0.42	<b>0.59</b>	<b>0.50</b>	0.07	-0.01	-0.06	-0.21	-0.13	0.00	-0.15	-0.16	-0.16
MTAug	-0.35	<b>0.61</b>	<b>0.48</b>	0.13	-0.02	-0.06	-0.22	-0.08	-0.03	-0.11	-0.11	-0.15
MTSep	-0.41	<b>0.58</b>	<b>0.46</b>	0.22	0.12	0.05	-0.21	0.06	-0.02	-0.07	-0.11	-0.16
MTOct	-0.34	<b>0.56</b>	<b>0.50</b>	0.33	0.22	0.11	-0.18	0.18	-0.02	-0.02	-0.04	-0.13
MTNov	-0.26	<b>0.53</b>	<b>0.44</b>	<b>0.47</b>	0.28	0.12	-0.08	0.24	-0.01	0.03	-0.03	-0.14
MTDic	-0.14	<b>0.44</b>	0.37	<b>0.52</b>	0.36	0.15	0.00	0.26	-0.01	0.03	-0.08	-0.18
AP	<b>0.54</b>	-0.23	-0.20	-0.13	-0.17	-0.07	0.23	<b>0.40</b>	0.06	0.19	0.21	0.20
WintP	<b>0.45</b>	-0.33	-0.22	0.27	-0.20	-0.22	0.24	<b>0.40</b>	0.05	0.09	0.26	0.17
SpringP	<b>0.58</b>	-0.36	-0.31	-0.12	-0.17	-0.06	<b>0.28</b>	<b>0.35</b>	0.06	0.09	0.18	0.25
SumP	-0.17	0.00	0.11	<b>-0.52</b>	-0.04	0.17	0.18	0.07	0.03	0.12	0.01	0.08
AutP	<b>0.52</b>	-0.27	-0.14	0.16	0.04	-0.08	0.21	<b>0.34</b>	0.05	0.24	0.08	0.02
MAT	-0.39	<b>0.54</b>	<b>0.47</b>	0.25	0.17	0.07	-0.14	0.13	-0.01	-0.04	-0.11	-0.17
WT	<b>-0.43</b>	<b>0.57</b>	<b>0.47</b>	0.08	-0.02	-0.06	<b>-0.28</b>	-0.03	-0.01	-0.09	-0.15	-0.17
MWT	-0.21	<b>0.45</b>	0.35	-0.15	-0.16	-0.13	-0.14	-0.19	0.04	-0.22	-0.13	0.00
MT	0.00	0.40	0.30	<b>0.48</b>	0.37	0.12	-0.03	<b>0.32</b>	-0.02	-0.01	-0.12	-0.17
MMT	-0.01	0.30	0.24	<b>0.59</b>	0.41	0.15	-0.07	<b>0.27</b>	0.01	0.03	-0.07	-0.18

664

665

666 **MMT#** is the mean minimum temperature of the month #; **MWT#** is the mean  
667 maximum temperature of the month #; **MT#** is the mean temperature of the month #;  
668 **WintP** is total winter precipitation; **SpringP** is total spring precipitation; **SumP** is the  
669 total summer precipitation; **AutP** is total autumn precipitation; **AP** is the annual  
670 precipitation; **MAT** is the mean annual temperature; **WT** is the mean temperature of the  
671 warmest month; **MWT** is the mean maximum temperature of the warmest month; **MT**  
672 is the mean annual temperature; and **MMT** is the mean minimum temperature of the  
673 coldest month.

## Figure legends

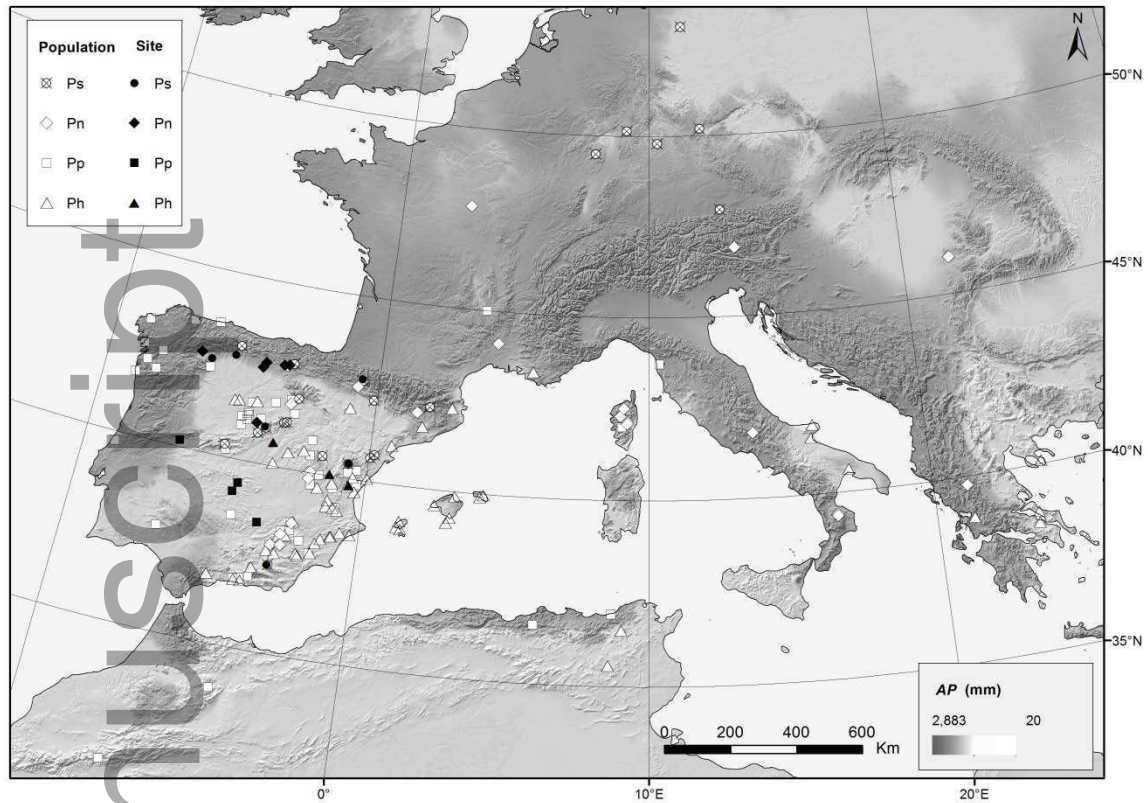
674

675

676 **Figure 1.** Common garden sites (Site, filled symbols), and population's sites of origin  
677 (Population, unfilled symbols) are represented in the map. Each pine species is  
678 represented in a different symbol Ps: *P. sylvestris* and it is represented by a circle; Pn:  
679 *P. nigra*, it is represented by a star; Pp: *P. pinaster* by a square, and Ph: *P. halepensis* by  
680 a triangle.

681

682 **Figure 2.** Summary information of the estimated parameters in the final tree height-  
683 diameter allometry model. We have described each parameter attending to its hierarchy,  
684 its significance and the set of research questions that can be addressed.



**Figure 1.** Location of common gardens (**Site**, filled symbols), and population origin (**Population**, unfilled symbols). Each pine species is represented by a different code and symbol: *P. sylvestris* (**Ps**, circle); *P. nigra* (**Pn**, star); *P. pinaster* (**Pp**, square), and *P. halepensis* (**Ph**, triangle).

Model	Hyperparameter	Model	Parameter	Insights from parameter
Scaling Coefficient  $\ln(a_{p(i),s(i)})$	$\mu_1$  <u>Hierarchy:</u> Species*  <u>Meaning:</u> -allometric curve's intercept at average growing conditions.	Scaling Coefficient  $\ln(a_{p(i),s(i)})$	$\alpha_{1p}$  <u>Hierarchy:</u> Population  <u>Meaning:</u> -allometric curve's intercept at average growing conditions.	-assessing genetic differences among populations on their basal height.  $\alpha_{1p} = \alpha_{1p_A} - \alpha_{1p_B}^e$ , A and B are any two populations  -quantify % of genetic differences based in $e$  -quantify the range of genetic variation, $sd(\alpha_{1p})$  -assessing adaptive values resulting from past demography of the species (gene pools allows for testing neutral patterns). One-way ANOVA  -assessing adaptive values resulting from climatic or geographical clines due to local adaptation. Pearson correlations ( $\rho$ )
Scaling Coefficient  $\ln(a_{p(i),s(i)})$	$\mu_2$  <u>Hierarchy:</u> Species*  <u>Meaning:</u> -responsiveness of a species to low temperatures where it grows	Scaling Coefficient  $\ln(a_{p(i),s(i)})$	$\alpha_{2p}$  <u>Hierarchy:</u> Population  <u>Meaning:</u> -responsiveness of each population to low temperatures where they grow	-assessing genetic differences among populations on their response to temperature.  $\alpha_{2p} = \alpha_{2p_A} - \alpha_{2p_B}^e$ , A and B are any two populations  -quantify % of genetic differences based in $e$  -quantify the range of genetic variation, $sd(\alpha_{2p})$  -assessing adaptive values resulting from past demography of the species (gene pools allows for testing neutral patterns). One-way ANOVA  -assessing adaptive values resulting from climatic or geographical clines due to local adaptation in plastic responses to temperature. Pearson correlations ( $\rho$ )
Scaling Coefficient  $\ln(a_{p(i),s(i)})$	$\mu_3$  <u>Hierarchy:</u> Species*  <u>Meaning:</u> -responsiveness of	Scaling Coefficient  $\ln(a_{p(i),s(i)})$	$\alpha_{3p}$  <u>Hierarchy:</u> Population  <u>Meaning:</u> -responsiveness of	-assessing genetic differences among populations on their response to precipitation  $\alpha_{3p} = \alpha_{3p_A} - \alpha_{3p_B}^e$ , A and B are any two populations  -quantify genetic differences % based in $e$  -quantify range of genetic variation, $sd(\alpha_{3p})$  -assessing adaptive values resulting from past demography of the species (gene pools allows for testing neutral patterns). One-way ANOVA  -assessing adaptive values resulting from

Model	Hyperparameter	Model	Parameter	Insights from parameter
	a species to annual precipitation where it grows		each population to precipitation where they grow	climatic or geographical clines due to local adaptation in plastic responses to precipitation. Pearson correlations ( $\rho$ )

Model	Parameter	Insights from parameter
Scaling exponent $c_{p(i)}$	$\beta_1$ <u>Hierarchy:</u> Species* <u>Meaning:</u> -intercept of $c_{p(i)}$ on average conditions on diameter $d$ ?	
Scaling exponent $c_{p(i)}$	$\beta_2$ <u>Hierarchy:</u> Species* <u>Meaning:</u> -responsiveness of the species to latitude	-if statistically significant, assessing the existence of a latitudinal cline This will suggest the existence of populations' adaptation to their local environment.
Scaling exponent $c_{p(i)}$	$\beta_3$ <u>Hierarchy:</u> Species* <u>Meaning:</u> -responsiveness of the species to altitude	-if statistically significant, assessing the existence of an altitudinal cline. This will suggest the existence of populations' adaptation to their local environment.

\*In this study, the species term is approximated as the average response from the set of populations assessed.

**Figure 2.** Summary information of the estimated parameters, hyperparameters and parameters, in the final tree height-diameter allometry model. We have described each one attending to its hierarchy, its significance and the set of research questions that can be addressed.