Title: Disentangling the relative importance of climate, size and competition on tree growth in Iberian forests: implications for forest management under global change

Running title: neighborhood models and tree growth drivers

Lorena Gómez-Aparicio<sup>1\*</sup>, Raúl García-Valdés<sup>2</sup>, Paloma Ruíz-Benito<sup>2,3</sup> and Miguel A. Zavala<sup>2,3</sup>

<sup>1</sup>Instituto de Recursos Naturales y Agrobiología (IRNAS, CSIC), PO Box 1052, Seville E-41080, Spain

<sup>2</sup>CIFOR, INIA, Ctra. de la Coruña Km. 7, Madrid E-28040, Spain
 <sup>3</sup>Dpto. Ecología, Universidad de Alcalá de Henares, Alcalá de Henares, Madrid E-28040,
 Spain

\*Corresponding author: Lorena Gómez Aparicio, phone: +34 954 624 711, fax: +34 954 624 002, e-mail: lorenag@irnase.csic.es

Manuscript type: primary research article

*Keywords:* climate change, competitive effect, competitive response, Iberian Peninsula, maximum likelihood, Mediterranean forests, neighhorbood models, species rank reversals, tree coexistence

# 1 Abstract

2 Most large-scale multi-species studies of tree growth have been conducted in tropical and 3 cool temperate forests, whereas Mediterranean water-limited ecosystems have received much 4 less attention. This limits our understanding of how growth of coexisting tree species varies 5 along environmental gradients in these forests, and the implications for species interactions 6 and community assembly under current and future climatic conditions. Here, we quantify the 7 absolute effect and relative importance of climate, tree size and competition as determinants 8 of tree growth patterns in Iberian forests, and explore inter-specific differences in the two 9 components of competitive ability (competitive response and effect) along climatic and size 10 gradients. Spatially-explicit neighborhood models were developed to predict tree growth for 11 the 15 most abundant Iberian tree species using permanent-plot data from the Spanish Second 12 and Third National Forest Inventory (IFN). Our neighborhood analyses showed a climatic 13 and size effect on tree growth, but also revealed that competition from neighbors has a 14 comparatively much larger impact on growth in Iberian forests. Moreover, the sensitivity to 15 competition (i.e. competitive response) of target trees varied markedly along climatic 16 gradients causing significant rank reversals in species performance, particularly under xeric 17 conditions. We also found compelling evidence for strong species-specific competitive 18 effects in these forests. Altogether, these results constitute critical new information which not 19 only furthers our understanding of important theoretical questions about the assembly of 20 Mediterranean forests, but will also be of help in developing new guidelines for adapting 21 forests in this climatic boundary to global change. If we consider the climatic gradients of this 22 study as a surrogate for future climatic conditions, then we should expect absolute growth rates to decrease and sensitivity to competition to increase in most forests of the Iberian 23 24 Peninsula (in all but the northern Atlantic forests), making these management considerations 25 even more important in the future.

# 1 Introduction

2 Regeneration, growth and mortality are the main processes driving plant community dynamics (Shugart 1984; Kohyama 1992; Oliver & Larson 1996). In plants, birth and death 3 4 are mediated by growth, since growth is usually negatively related to survival and positively related to reproduction (Nakashizuka 2001; Wyckoff & Clark 2002; Pérez-Ramos et al. 5 6 2010). This renders understanding growth variation within and among populations a central aspect of plant population and community ecology (Hara 1984; Stoll et al. 1994; Zavala et al. 7 8 2007; Coomes et al. 2009). Although the study of growth variation in multispecific forest 9 communities is not new, important advances have recently been made, as accessibility to 10 long-term forest inventories has facilitated the formulation of individual tree-based models at large spatial scales and for many co-existing species (Uriarte et al. 2004; Canham et al. 2006; 11 12 Sánchez-Gómez et al. 2008; Purves 2009).

Most studies of growth variation have been conducted in tropical and cool temperate 13 forests, usually with the aim of understanding patterns and mechanisms of species 14 15 coexistence and biodiversity maintenance in the context of the neutral theory debate (e.g. 16 interspecific variation; Hubbell 2001; Uriarte et al. 2004; Coates et al. 2009), the mechanistic plant competition theory (e.g. assembly trade-offs; Pacala et al. 1994; Lin et al. 2002) or the 17 18 metabolic theory (i.e. scaling of growth with other processes; Muller-Landau et al. 2006; Russo et al. 2007). Mediterranean water-limited forests, however, have received much less 19 attention, resulting in a poor understanding of how different tree species respond along 20 21 environmental gradients in these forests and the implications for community assembly. 22 Moreover, understanding interspecific tree growth variation in Mediterranean ecosystems may be crucial in a global change context, since these systems are among those most 23 24 threatened by global warming and land-use changes, and are therefore highly susceptible to suffer dramatic changes in species range and abundance (Bakkenes *et al.* 2002; Schröter *et al.* 2005; Benito-Garzón *et al.* 2008).

3 As a result of climate change, temperature in the Mediterranean basin is expected to 4 increase while precipitation may decrease (Christensen et al. 2007). These changes could 5 have important direct effects on tree growth due to abiotic constraints (e.g. hydric stress) and 6 indirect effects due to increases in resource competition, as shown in a number of single-7 species studies (Jump et al. 2006; Martín-Benito et al. 2008; Martinez-Vilalta et al. 2008; 8 Gea-Izquierdo et al. 2009; Linares et al. 2009; Vicente-Serrano et al. 2010). Multi-species 9 studies are, however, extremely rare, and tend to focus on a limited subset of species or a 10 given region (Sabaté et al. 2002; Andreu et al. 2007; Sánchez-Gómez et al. 2008). This 11 hampers our understanding of the likely effects of climate change not only on tree growth, 12 but also on plant-plant interactions, including competitive reciprocal effects and shifts in rank hierarchies. 13

14 In recent decades, Mediterranean countries have also suffered large land use shifts including agricultural land abandonment, reforestation of extensive areas with high tree 15 16 densities, and strict protection of many natural areas (Zavala & Oria 1995; Gómez-Aparicio et al. 2009). All these practices have favored forest expansion and stand densification, and 17 18 have had a striking effect on tree growth patterns. For example, Linares et al. (2010) showed 19 that excessive protection of the endangered species Abies pinsapo in southern Spain increased 20 tree-to-tree competition and reduced the adaptive capacity of the species to drier climatic 21 conditions. All this suggests the need to explore competitive and tree growth responses along 22 tree density gradients, so that proper adaptation measures to cope with climatic changes can be undertaken. 23

In this study, we used forest inventory data for continental Spain to analyze the effect of climate, size and competition on tree growth of the 15 most abundant Iberian canopy

1 species using a spatially-explicit neighborhood scale approach. The main advantage of the 2 neighborhood approach is that it captures the local nature of plant interactions and 3 acknowledges that the frequency with which individuals interact with other conspecific or 4 heterospecific neighbors may depend less on their relative abundance than on their particular 5 spatial pattern (Silander & Pacala 1985; Wagner & Radosevich 1998; Stoll & Newbery 2005). 6 Specifically, we aimed to: 1) quantify the absolute effect and relative importance of climate, 7 size and competition on tree growth patterns in Iberian forests; 2) analyze how the sensitivity 8 to competition (i.e. *competitive response* or ability to withstand growth suppression by 9 neighbors, sensu Goldberg 1990) of the different species varied along climatic and size 10 gradients with potential consequences for the ranking of species performance; and 3) explore 11 whether different species of neighboring trees were functionally equivalent competitors (i.e. have similar *competitive effects* or abilities to suppress neighbor growth, sensu Goldberg 12 13 1990) using inter-specific competition coefficients (i.e. the per capita effect of one species on 14 another, Freckleton & Watkinson 2001). By combining in the same analysis the three main 15 drivers of tree growth for the 15 most important tree species of the Iberian Peninsula, we 16 aimed to provide detailed quantitative information about how these factors interact to 17 determine growth of coexisting species at large spatial scales. This information will, in turn, 18 be crucial for predicting potential community-level consequences (e.g. rank reversals in 19 species performance) of growth alterations due to climate warming and human land use 20 change.

#### 1 Material and Methods

# 2 Data set

Our analyses of tree growth were based on the Spanish Second and Third National Forest 3 Inventories (IFN2 and IFN3) conducted between 1986-1996 and 1997-2007, respectively 4 5 (Villaescusa & Díaz 1998; Villanueva 2004). The IFNs are extensive datasets consisting of 6 circular sample plots distributed across the forested surface of Spain, with an approximate density of 1 plot Km<sup>-2</sup> across a large altitudinal gradient (sea level to 2330 m). Plots are of 7 8 various concentric radii (i.e. the minimum tree diameter measured varied with the radius of 9 the plot): all trees with diameter at breast height (d.b.h.) > 7.5 cm are measured within a 5-m 10 radius plot, trees with d.b.h. > 12.5 cm within a 10-m radius plot, trees with d.b.h. > 22.5 cm 11 within a 15-m radius plot, and trees with d.b.h. > 42.5 cm within a 25-m radius plot. We 12 limited our growth analyses to target trees within the smallest 5-m radius plot, considering as neighbors those trees located within the 10-m radius plots. To compensate for the fact that 13 14 small trees (7.5 cm < d.b.h. < 12.5 cm) were measured only in the smallest 5-m radius plots, 15 we assumed that their density and size structure in these plots could be extrapolated to the 16 whole 10-m radius plot, and therefore generated "artificial" neighbor trees that were assigned 17 random positions within the outer 5 m of the 10-m radius plots. Thus, we obtained a full 18 census of all neighbors > 7.5 cm d.b.h. present at the beginning of the census interval (IFN2) 19 in a minimum radius of 5 m around each target tree. This radius size was consistent with that 20 used in other tree competition studies (Stoll et al. 1994; He & Duncan 2000; Boyden et al. 21 2005). Diameter growth of each target tree (in millimeters per year) between the two 22 censuses was calculated by dividing the total increase in d.b.h. (in millimeters) by the number 23 of years between the two census dates.

Each of the plots was characterized with 12 topographic and climatic variables. The Iberian Peninsula is an area of transition between the Euro-Siberian, or temperate, and the

1 Saharo-Sindic climatic regions, and therefore contains a mosaic of different climates, from 2 humid Atlantic in the north to Mediterranean and semiarid in the center and south. Topographic variables were calculated from a digital elevation model (COP 1998), and 3 4 climatic variables obtained through ordinary co-kriging from 5426 weather stations (series 1951-1999; Gonzalo 2008). Raster maps and plot locations (UTM coordinates) were 5 6 combined to determine the values of each abiotic variable in each plot using ArcView Gis 9.2 7 (ESRI Inc., Redlands, USA, 2000). The 12 abiotic variables were: altitude; slope; annual and 8 seasonal (i.e., spring, summer, fall, and winter) precipitation; mean annual temperature, 9 mean temperature of the hottest month and mean temperature of the coldest month; drought 10 length, taken as the number of months in which potential evapotranspiration exceeded 11 precipitation; and *drought intensity*, calculated on the basis of the quotient  $A_d/A_h$ , where  $A_d$  is the dry area of the climodiagram (precipitation curve below the temperature curve) and A<sub>h</sub> is 12 the humid area of the climodiagram (precipitation curve above the temperature curve, Allué-13 14 Andrade 1990). The relationships between the 12 abiotic variables were explored with 15 Principal Component Analyses (PCA) based on correlations with varimax rotation. The first 16 axis of the PCA (explaining 36.1% of the variance) was strongly correlated (r > 0.9) with mean annual temperature and mean temperature of the coldest month. The second axis 17 18 (explaining 36.0% of the variance) was strongly correlated with annual, spring and fall 19 precipitation. Therefore, for our modeling analyses we chose the two variables most 20 representative of each of the two axes: mean annual temperature and annual precipitation.

21

#### 22 A maximum-likelihood analysis of tree growth

We used likelihood methods and model selection as an alternative to traditional hypothesis
testing for analysis of our data (Johnson & Omland 2004; Canham & Uriarte 2006).
Following the principles of likelihood estimation, we estimated model parameters that

maximized the likelihood of observing the growth data measured in the field given a suite of
alternative models.

3 We conducted separate analyses for the 15 most common canopy tree species. These 4 species present a wide range of functional traits, from deciduous temperate species typical of 5 northern cool Atlantic forests to evergreen Mediterranean oak and pine species from central 6 and southern Iberia (Table 1). We focused on abundant species (> 1000 individuals) because 7 small sample sizes can make it difficult to obtain competition coefficients for all neighbor 8 species (Coates et al. 2009). Our analyses of diameter growth of target trees estimated four 9 terms: 1) average potential diameter growth (PDG, in mm/year), and three sets of scalar 10 modifiers ranging from 0 to 1 that quantified the effects on average potential growth of 2) 11 local climatic conditions (expressed in terms of mean annual temperature and annual precipitation), 3) initial target tree size (d.b.h., in cm), and 4) the characteristics of the 12 neighborhood. Our *full model* had the following form: 13

14 Growth = PDG x Climatic effect x Size effect x Neighborhood effect 
$$(1)$$

15 *Potential diameter growth* (PDG) represents the expected target growth when the 16 other factors are at optimal values. The *climatic effect* was modeled using a bivariate 17 Gaussian function:

18 Climatic effect = exp
$$\left[-\frac{1}{2}\left(\frac{T-X1_0}{X1_b}\right)^2\right] * exp\left[-\frac{1}{2}\left(\frac{P-X2_0}{X2_b}\right)^2\right]$$
 (2)

where  $X1_0$  and  $X2_0$  are the mean annual temperature (T) and annual precipitation (P) values, respectively, at which maximum potential growth occurs; and  $X1_b$  and  $X2_b$  are estimated parameters that control the breadth of the function (i.e. the variance of the normal distribution). Equation 2 produces the classic Gaussian distribution of species performance along an environmental axis usually assumed to describe vegetation-environment relationships (e.g. Curtis 1959; Whittaker 1975; Gauch 1982), but can also produce sigmoidal,
 monotonic curves within restricted ranges of either axis. We also tested univariate functions
 in which terms for one of the two axes were dropped from the analysis.

Following recent studies (Uriarte *et al.* 2004; Canham *et al.* 2006; Coates *et al.* 2009),
the *size effect* was modeled using a lognormal function:

6 Size effect = exp 
$$\left[ -\frac{1}{2} \left( \frac{\ln(dbh/X3_0)}{X3_b} \right)^2 \right]$$
 (3)

where X3<sub>0</sub> represents the target d.b.h. (cm) at which maximum growth occurs, and X3<sub>b</sub>
controls the breadth of the function. Depending on the value of X3<sub>0</sub>, this functional form can
be hump-shaped or monotonically increasing/decreasing.

10 The *neighborhood effect* was modeled as a function of a *neighborhood competition*11 *index* (NCI) using a Weibull function:

12 Neighborhood effect = exp
$$\left[-a(NCI)^{b}\right]$$
 (4)

13 The Weibull function assumes that the neighborhood effect (and therefore target tree growth) declines monotonically as a function of NCI. A positive neighborhood effect was not 14 15 considered in our models because an initial exploration of the data revealed that target growth 16 always declined with increasing number or basal area of neighbors. The absence of a positive 17 neighbor effect on target growth might be due to the fact that our analyses focused on adult 18 trees and large saplings (d.b.h. > 7.5 cm), whereas facilitation decreases with ontogeny and it 19 is mostly found for seedlings and small saplings (Miriti 2006; Quero et al. 2008). The NCI 20 quantifies the net effect of i=1,...,n neighboring trees of i=1,...,s species on the growth of a 21 target tree z. Following the long tradition of distance-dependence analysis of competition (e.g. 22 Bella 1971; Daniels 1976), NCI was assumed to vary as a direct function of the size (d.b.h.) 23 and an inverse function of the distance to neighbors following the form:

Gómez-Aparicio et al. 10

1 
$$\operatorname{NCI} = \sum_{i=1}^{s} \sum_{j=1}^{n} \lambda_{iz} \ db h_{ij}^{\alpha} \exp\left(-\gamma \ \operatorname{distance}_{ij}^{\beta}\right)$$
(5)

where  $\alpha$ ,  $\beta$  and  $\gamma$  are parameters estimated by the analyses and determine the shape of the effect of the d.b.h. and the distance to the neighbor on NCI. In order to facilitate comparisons of neighborhood effects across target species, NCI values for each species were scaled relative to the maximum NCI found for that given species. This created a 0 to 1 index where 0 and 1 represented the minimum and maximum neighborhood interactions observed for any target species.

8 We were particularly interested in exploring whether neighbor effects on a target 9 species z varied between species or groups of neighbors. In order to do this, we multiplied the 10 net effect of an individual neighbor by a per-capita competition coefficient ( $\lambda$ ) that ranged 11 from 0 to 1 and allowed for differences between species in their competitive effect on a target 12 tree. Competition coefficients were only estimated for those species of competitors for which there were at least 100 neighbors. All remaining species of neighbors for that target species 13 14 were then grouped into a "rare species" group. We tested four different groupings of neighboring species in Eqn. 5: 1) a model that calculated a different  $\lambda_{i,z}$  value for each 15 16 neighbor species, 2) a model that calculated two separate  $\lambda$ , one for conspecifics and another for heterospecifics, 3) a model that calculated two separate  $\lambda$ , one for conifers and one for 17 18 angiosperms, and 4) a model in which all species were considered equivalent (i.e. fixing  $\lambda =$ 19 1).

We also tested variants of Eqn. 4, in which sensitivity to competition varied as a function of climatic conditions (temperature and precipitation) and target tree size. For this, the exponential decay term (a) was allowed to vary as a function of mean annual temperature, annual rainfall and target d.b.h.:

24 
$$a = a' * Temperature^{\delta} * Precipitation^{\sigma} * dbh^{\varphi}$$
 (6)

1 If either  $\delta$ ,  $\sigma$ , or  $\phi$  are = 0, then sensitivity to competition does not vary as a function of 2 temperature, precipitation or d.b.h. (respectively). If either  $\delta$ ,  $\sigma$  or  $\phi$  are < 0, then sensitivity to competition declines with temperature, precipitation or d.b.h. (i.e. at a given level of 3 4 competition, trees at lower temperatures and precipitation levels and with smaller d.b.h. will 5 suffer greater reductions in growth than trees at higher temperatures and precipitation levels 6 and with larger d.b.h.), and if either  $\delta$ ,  $\sigma$ , or  $\phi$  are > 0 then sensitivity to competition increases 7 with climatic conditions and d.b.h.. These effects were assumed to be independent of the 8 underlying effect of climate or target tree size on potential growth (i.e. *climatic effect* or *size* 9 effect, in the absence of a neighborhood effect).

10

# 11 Parameter estimation and model selection

12 We used information theoretics (the Akaike Information Criterion, AIC) to select the best 13 growth model. Following the principle of parsimony, we employed the strategy of 14 systematically reducing the number of different parameters to the simplest model that was not a significantly worse fit than any more complicated model. Thus, the full model was 15 16 compared to models that ignored the effect of climate, size, competition, or the three of them 17 (i.e. null, or intercept-only model), with lower AIC values indicating stronger empirical 18 support for a model (Burnham & Anderson 2002). Growth values were modeled using a 19 gamma error distribution defined by a shape parameter *n*, which varies from exponential-like 20 to bellshaped but left-skewed forms. We used simulated annealing, a global optimization 21 procedure, to determine the most likely parameters (i.e., the parameters that maximize the 22 log-likelihood) given our observed data (Goffe et al. 1994). The slope of the regression (with 23 a zero intercept) of observed radial growth on predicted radial growth was used to measure bias (with an unbiased model having a slope of 1) and the  $R^2$  of the regression was used as a 24 measure of goodness-of-fit. We used asymptotic two-unit support intervals to assess the 25

strength of evidence for individual maximum likelihood parameter estimates (Edwards 1992).
A support interval is defined as the range of the parameter value that results in less than a
two-unit difference in AIC. It is roughly equivalent to a 95% support limit defined using a
likelihood ratio test (Hilborn & Mangel 1997). All analyses were performed using software
written specifically for this study using Java (Java SE Runtime Environment v6, Sun
Microsystems Inc., California, USA, 2010).

## 1 Results

2 Effects of climate, size and competition on potential tree growth

A full model that included the effect on growth of all the factors explored (climate, size and competition) was the best model for the 15 species analyzed (Table 2). However, the competitive effect was always much stronger than the climatic or size effect, as indicated by the larger increase in AIC when the competition term was dropped from the full models (Table 2). All of the models produced unbiased estimates of growth (i.e. slopes of predicted versus observed growth were all close to 1) and explained a percentage of variance in the data  $(R^2)$  that ranged from 13% for *Quercus suber* to 34% for *Pinus nigra* (Table 2).

10 The effect of mean annual temperature on growth (Eqn. 2, see parameter values in 11 Appendix S1 of Supporting Information) varied widely among species, having a positive 12 effect on the 8 broadleaved species (particularly *F. sylvatica*), a negative effect on 5 conifers 13 (particularly *P. sylvestris*), and a neutral effect on 2 pine species (*P. pinea* and *P. uncinata*; 14 Fig. 1a). The effect of annual precipitation on growth (Eqn. 2, Appendix S1) was positive for 15 all species except *F. sylvatica*, presenting increased growth with precipitation either through 16 the entire precipitation range or reaching a peak at a lower precipitation level (Fig. 1b).

The 15 species showed 3 different patterns of variation in potential growth as a function of target tree size (Eqn. 3, Appendix S1). Most pines showed a growth peak in the 15-25 cm size class range and a sharp decline in larger size classes (Fig. 1c). *Juniperus thurifera* and broadleaved species reached a maximum growth peak at larger sizes than pines (30-50 cm) and then showed moderate or slight declines in growth in larger size classes. Finally, *Pinus pinea* was the only species for which size had virtually no effect on growth.

The magnitude of the negative effect of competition (calculated for a tree of mean size under mean climatic conditions; Eqn. 4) was largest for conifers, with reductions close to 100% of potential growth under the maximum observed relative competition (i.e. NCI = 1; 1 Fig. 1d). Mediterranean evergreen broadleaves were the least sensitive to competition, with 2 reductions of 40-60% in potential growth at maximum competition levels. The  $\alpha$  parameter (Eqn. 5), which controls the effect of neighboring tree size on NCI, had values close to 2 for 3 4 conifers (range 1.76-1.96) and close to 1 for the rest of the species (range 0.76-1.40; 5 Appendix S1). These results indicate that competitive effects scale approximately linearly with biomass (i.e. DBH<sup>2</sup>) for conifers and with DBH for broadleaved species in Iberian 6 7 forests. The exponents  $\beta$  and  $\gamma$  (Eqn. 5) control the shape and the steepness (respectively) of 8 the effect of neighbor distance on NCI, and therefore on target tree growth. Conifers showed 9 faster declines in neighbor effects with distance than broadleaved species (Fig. 2).

10

# 11 Variation in the competitive response of target species along climatic and size gradients

Our analyses indicate that, for all species, sensitivity to competition varied markedly along one or two of the climatic gradients explored (i.e.  $\delta$  and/or  $\sigma \neq 0$ ; Eqn. 6, Table 1). Atlantic deciduous broadleaves had a lower competitive response ability at lower temperatures ( $\delta < 0$ ; Appendix S1), whereas the remaining species showed the opposite pattern ( $\delta > 0$ ; Appendix S1). For all species, trees at lower precipitation levels were more sensitive to competition than trees located at wetter sites (-0.22 <  $\sigma$  < -1.13; Appendix S1). In all cases, smaller trees were much more sensitive to competition than larger trees (-0.35 <  $\phi$  < -1.50; Appendix S1).

In order to explore the community-level consequences of the variation in sensitivity to competition with climate and size, we analyzed shifts in growth hierarchies along the competition gradient under two contrasting climatic scenarios, xeric ( $T = 14^{\circ}C$ , PP = 600 mm) vs. mesic ( $T = 8^{\circ}C$ , PP = 1500 mm), and for two different size classes, small trees (d.b.h. = 10 cm) vs. large trees (d.b.h. = 40 cm). We found that larger shifts occurred in xeric rather than in mesic habitats, and for small trees than for larger ones (Fig. 3). Thus, for 10-cm d.b.h. trees in xeric habitats, conifers had the largest growth rates without competition, but were displaced to the last positions of the ranking by broadleaved species at high competition
levels (Fig. 3a). Within each functional group (conifers vs. broadleaved), Mediterranean
species occupied the first positions of the ranking. On the other hand, basically no shifts
occurred for large trees in mesic habitats, with conifers (and particularly Mediterranean pines)
occupying the first places of the growth ranking along the whole competition gradient (Fig. 3d).

7

# 8 Comparison of species competitive effects on tree growth

9 For 11 of the 15 target tree species, the model that considered species-specific effects was the 10 most parsimonious of the four alternative models that differed in the grouping of neighbor 11 species (Table 2). For the remaining 4 species, the best model was the one that discriminated 12 between intra- and inter-specific competitors.

13 The competition indices ( $\lambda$ ) estimated in the species-specific full models were used to assemble a matrix of pairwise competition coefficients for the 15 tree species studied (Table 14 15 3). The magnitude of the competition coefficients was very large ( $\lambda \ge 0.75$ ) in about 40% of the cases, medium (0.25 <  $\lambda$  < 0.75) in 30% of the cases, and low ( $\lambda \le 0.25$ ) in the remaining 16 30%. Competition among conspecifics was generally strong (diagonal of  $\lambda$  with high values, 17 18 including maximum  $\lambda$  values [i.e.  $\lambda = 1$ ] for more than half of the species). We explored the existence of asymmetric pairwise interactions in the matrix (i.e. effect of species i on  $i [\lambda_{ij}]$ 19 much smaller/larger than the effect of species j on i  $[\lambda_{ii}]$ , but found no evidence for this 20 21 (Spearman rank correlation = -0.24, p = 0.09, n = 48).

#### 1 Discussion

2 The use of nation-wide forest inventory data, spatially-explicit neighborhood models, and 3 information theoretics allowed us to quantify the effect of climate, tree size and competition 4 on the growth of the 15 most common tree species of the Iberian Peninsula. Our 5 neighborhood analyses revealed that, in Iberian forests, competition from neighbors is an 6 important driver of tree growth, causing much larger reductions in potential growth than 7 climate or tree size. Moreover, the sensitivity to competition (i.e. competitive response) of 8 target trees varied markedly along environmental gradients causing significant rank reversals 9 in species performance, particularly under xeric conditions. We also found compelling 10 evidence for strong species-specific competitive effects in these forests. Altogether, these 11 results constitute critical new information which not only furthers our understanding of the 12 assembly of Mediterranean forests, but will also be of help in developing new guidelines for adapting forests in this climatic boundary to global change. 13

14

# 15 *Effects of climate and size on tree growth*

16 The response of growth to climate varied widely among species, particularly the response to 17 mean annual temperature. We found that rising temperatures generally had a positive effect 18 on growth of broadleaved species, particularly deciduous Atlantic species, but neutral or 19 negative effects on conifers, particularly on mountain pines (Fig. 1a). The positive effects of 20 temperature on broadleaved species' growth might arise from the fact that higher 21 temperatures may indicate longer growing seasons (Menzel & Fabian 1999; Peñuelas et al. 22 2002; Gordo & Sanz 2009). This would be especially true in cold sites such as those occupied by Atlantic deciduous species, where the absence of frosts could favor early and late season 23 24 photosynthesis, increasing carbohydrate storage for future growth (Miyazawa & Kikuzawa 25 2005; Seynave et al. 2008; Vitasse et al. 2009). On the other hand, the negative effect of increasing temperatures on conifer growth could be related to an increase in the atmospheric
water demand, causing greater drought stress at the sites occupied by these water-demanding
species (Barberó *et al.* 1998; Ferrio *et al.* 2003; Bogino & Bravo 2008). This indirect effect
of higher temperatures would be particularly apparent for pines typical of northern wetter
climates such as *P. sylvestris*, which persists at its southernmost distribution limit often at
high-altitude refuges (Castro *et al.* 2004).

7 The effect of precipitation on growth was much more uniform across tree species than 8 that of temperature. With the unexpected exception of F. sylvatica, all species showed higher 9 growth with increasing annual precipitation. This generalized positive effect of precipitation 10 is consistent with the limiting role that water plays in Mediterranean systems (e.g. Pigott & 11 Pigott, 1993), and with the recent evidence showing that Iberian tree species are becoming 12 increasingly water-stressed in summer (Andreu et al. 2007; Gea-Izquierdo et al. 2009). 13 Moreover, for several of the tree species analyzed here (e.g. P. sylvestris, P. uncinata, F. 14 sylvatica, Q. robur, Q. petraea) the Iberian Peninsula represents the edge of their southern 15 range, and thus they are already living at the limit of their physiological tolerances (Jump et 16 al. 2006; Reich & Oleksyn 2008; Martínez-Vilalta et al. 2010). An interesting prediction that 17 arises when putting together the species' responses to temperature and precipitation is that 18 alterations in temperature regimes due to climate change would be more likely to modify 19 inter-specific growth differences in forest communities than changes in precipitation. Thus, 20 whereas a reduction in precipitation could translate into a generalized decrease in growth, an 21 increase in temperature might cause a performance disadvantage of conifers compared to 22 broadleaves, particularly in northern Atlantic and montane sites where low temperatures are a limiting factor for hardwoods expansion (Prentice et al. 1992; Terradas & Savé 1992; 23 24 Seynave et al. 2008).

25

The effect of target tree size on growth was quite variable among species. This result

1 agrees with previous studies for tropical and temperate forests, which have shown that the 2 size-growth relationship does not seem to follow a single rule (Muller-Landau et al. 2006; 3 Russo et al. 2007). The most common pattern found in Iberian tree species was a peak of 4 maximum growth at some intermediate DBH and subsequent decrease with size. This is also 5 probably the pattern of tree growth most frequently reported in other forest communities 6 (Ryan et al. 2004; Coates et al. 2009). However, this peak was generally reached at lower 7 sizes in conifers, which also showed a much steeper decrease with size than broadleaved 8 species (Fig. 1c). Therefore, despite the high variability found, we could conclude that the 9 magnitude of the size effect on growth in Iberian forests was larger for conifers than for 10 broadleaved species.

11

# *Effect of competition on tree growth: variation in the competitive response of target species along climatic and size gradients*

14 Although our results support a climate and size effect on tree growth in Iberian forests, 15 competition from neighbors had a comparatively much larger impact on growth for the 15 16 species analyzed (Fig. 1). Interesting differences emerged in the competitive response of the 17 studied species. Conifers in general, and mountain pines in particular, were the most sensitive 18 tree species to competition, with reductions close to 100% of potential growth under the 19 maximum observed relative competition. The high sensitivity of conifers to competition was 20 also reflected in the form of the competition kernel, which scaled faster with size (larger  $\alpha$ 21 parameter) and decreased more steeply with distance (larger  $\gamma$  parameter; Fig. 2) than in 22 broadleaved species. At the other extreme, Mediterranean evergreen oaks were the most resistant to competition, probably favored by their strategy of resource storage and 23 24 conservative use, which allows them to maintain very low growth rates (see low PDG values 25 in Appendix S1; Zavala et al. 2000; Ogava et al. 2003). These results coincide with previous studies conducted in Mediterranean forests which have suggested that pines are in general
 less shade-tolerant than oaks and consequently more sensitive to competition from neighbors
 (Zavala & Zea 2004; Sánchez-Gómez *et al.* 2006). In general, shade-intolerant species have
 been found to require more "growing space" than shade-tolerant species (Burton 1993;
 Simard & Sachs 2004).

6 Our results indicate that species' competitive responses are not constant, but vary along climatic and size gradients. We found a more homogeneous interspecific response 7 8 along the precipitation gradient than along the temperature gradient. Thus, sensitivity to 9 competition increased with decreasing precipitation for all species, especially in the smallest 10 size classes. This result agrees with the idea that competition increases when resources are 11 more limiting (Tilman 1988), and specifically with the prediction that climate warming 12 would increase competition for water in Mediterranean forests (Cotillas et al. 2009; Linares 13 et al. 2010). Also, and in accordance with this prediction, sensitivity to competition increased 14 with increasing temperature for Mediterranean and sub-Mediterranean broadleaved species, 15 which could be linked to a higher transpirational demand and a decrease in water supply. 16 Atlantic deciduous broadleaved were once more the exception to the rule, their sensitivity to 17 competition increasing with decreasing temperature. This indicates that warmer temperatures 18 due to climate change could benefit these species not only by the direct climatic effect on 19 growth already discussed, but also indirectly by causing a release from competition.

A profound community-level consequence of the variation in sensitivity to competition along climatic and size gradients was the occurrence of rank reversals in species performance along the competition gradient under different environmental conditions. The most abundant and clearest rank reversals occurred for early ontogenetic stages under xeric conditions, with pines showing the highest growth under low competition intensity but the lowest growth under high competition intensity (Fig. 3). In mesic environments, however, the much higher absolute growth rates of pines than of broadleaved species compensated for their higher sensitivity to competition, so they occupied the first places of the growth ranking at all competition levels. Based on these results, an aridification of the Mediterranean climate due to climate warming could cause rank reversals in multispecific forests, with conifers being at a clear disadvantage due to their poorer ability to deal with resource shortages both above and belowground in a warmer and more competitive environment.

7

### 8 Inter-specific differences in competitive effects of neighbors on tree growth

9 Our results offer strong support not only for a high variability of competitive responses 10 among coexisting tree species in Iberian forests, but also for species-specific competitive 11 effects of neighbors on target trees. Thus, on a per unit size basis, tree species cannot be considered equivalent in Iberian forests, neither from a response nor an effect perspective. 12 13 The question of whether response and effect competitive abilities are linked in plant species 14 has received much attention due to its relevance for understanding the structure of plant 15 interactions (see Wang et al. 2010 and references therein). In this study, we did not find a 16 clear correlation between the competitive response and effect of tree species (Goldberg & 17 Landa 1991; Cahill et al. 2005; Fraser & Miletti 2008). For example, P. halepensis and F. 18 sylvatica had by far the largest competitive effects ( $\lambda \sim 0.90$ ; Table 3), whereas they were not 19 characterized by particularly high or low competitive responses (Fig. 1d). The target-20 neighbor interactions in these forests might therefore be the result of different combinations 21 of competitive abilities to tolerate and suppress neighbors.

We found that competitive interactions were in general of large magnitude, particularly among conspecifics. To explore whether this result could be affected by the spatial segregation of the species (i.e. that heterospecific neighbors were distinctively further away from targets than conspecific neighbors) we examined the mean distance to both groups

1 of neighbors for the different species. Mean distances happened to be remarkly similar (data 2 not shown), so we did not find evidence for an influence of spatial segretation on the strong negative neighbor effects on conspecifics. An alternative and more plausible explanation for 3 4 this result could lie in the resource partitioning hypothesis, which predicts more intense 5 competition among closely related individuals due to a higher similarity in the use of 6 resources (Tilman 1982; Argyres & Schmitt 1992; Uriarte et al. 2004). Previous studies have 7 also found density-dependent effects where nearby conspecifics decrease individual 8 performance to a much larger extent that heterospecifics (Stoll & Newbery, 2005; Zhao et al. 9 2006). It has been suggested that these types of negative density-dependent effects constitute 10 a mechanism for maintaining diversity in tropical forests (Janzen 1970; Wright 2002), and 11 they could also be acting in the much less diverse Mediterranean forests to promote species 12 coexistence at the neighborhood scale.

13

# 14 Implications for management of Mediterranean forests under climate change

15 Several of the key questions that scientists and managers currently face regarding the future 16 of Mediterranean forests concern their response to predicted climate change and the 17 mitigation tools available to increase their resilience to these changes. Our results clearly 18 indicate that although climate exercises considerable direct control over tree species 19 performance, its effect is modulated by biotic factors such as competition from neighbors. 20 Because competitive effects in these forests can suppress growth by almost 100%, species 21 can have a very limited capacity to respond to climate at high competition levels (Fig. 4). 22 Three main corollaries arise directly from these results. First, studies of climate change effects on tree performance must necessarily consider stand structural characteristics, 23 24 something that has seldom been done so far (but see Hurteau et al. 2007; Gea-Izquierdo et al. 25 2009; Linares et al. 2009). Second, the reduction of competition by thinning should be

1 considered as an adaptation measure that could counteract the negative effects of climate 2 warming on tree performance (Cescatti & Piutti 1998; Gracia et al. 1999; Martín-Benito et al. 3 2010). Third, any management measure that promotes forest densification (e.g. abandonment 4 of thinning operations in reforestations, severe protection in natural areas) should be 5 evaluated carefully, since it could cause an increase in competition for resources (particularly 6 water) enhancing species vulnerability to a drier climate (Linares et al. 2010). If the climatic 7 gradients of this study can be considered as a surrogate for future climatic conditions, then we 8 should expect absolute growth rates to decrease and sensitivity to competition to increase in 9 most forests of the Iberian Peninsula (in all but the northern Atlantic forests), making these 10 management considerations even more important in the future.

11 Our neighborhood analyses also provide detailed recommendations to guide silvicultural activities aimed at reducing competition. These activities should focus on 12 neighbors around small trees, which are by far the most sensitive to competition, and 13 14 prioritize the removal of close conspecifics in order to decrease negative density-dependent 15 effects. Neighborhoods of species (and particularly conifers) at their southern limit, such as P. 16 sylvestris, should also be a priority, since these species are the most sensitive to both climate and competition. In summary, our results show that adapting forests to climate change can 17 18 greatly benefit from spatially-explicit information that indicates how competitive effects vary with distance, size and identity of neighbors. Specifically, this knowledge will be essential to 19 20 evaluate which species mixtures, tree densities and size structures will perform better under a 21 given set of climatic conditions.

# 1 Acknowledgements

We thank the Spanish Ministry of the Environment and Rural and Marine Affairs for granting access to the forest inventory data. LGA was supported by the Spanish Ministry of Science and Innovation grant INTERBOS1 (CGL2008-04503-C03-01); MAZ was supported by grants INTERBOS3-CGL2008-04503-C03-03 and the INIA project SUM2008-00004-C03-01; and RGV and PRB were supported by FPI-MCI (BES-2007-14404) and FPU-MEC (AP2008-01325) fellowships, respectively. This research is part of the GLOBIMED (www.globimed.net) network on forest ecology.

#### References

- Allué-Andrade JL (1990) Atlas fitoclimático de España. Taxonomías. Madrid, Spain, Ministerio de Agricultura, Pesca y Alimentación, Instituto Nacional de Investigaciones Agrarias.
- Andreu L, Gutiérrez E, Macias M, Ribas M, Bosch O, Camarero JJ (2007) Climate increases regional tree-growth variability in Iberian pine forests. *Global Change Biology*, 13, 804-815.
- Argyres AZ, Schmitt J (1992) Neighbor relatedness and competitive performance and impatiens-capensis (Balsaminaceae): a test of the resource partitioning hyphotesis. *American Journal of Botany*, **79**, 181-185.
- Bakkenes M, Alkemade JRM, Ihle F, Leemans R, Latour JB (2002) Assessing effects of forecasted climate change on the diversity and distribution of European higher plants for 2050. *Global Change Biology*, **8**, 390-407.
- Barberó M, Loisel R, Quézel P, Richardson DM, Romane F (1998) Pines of the Mediterranean basin. In: *Ecology and Biogeography of Pinus*. (ed Richardson DM) pp. 153-170. Cambridge, UK, Cambridge University Press.
- Bella IE (1971) New competition model for individuals trees. Forest Science, 17, 364-&.
- Benito-Garzón M, de Dios RS, Ollero HS (2008) Effects of climate change on the distribution of Iberian tree species. *Applied Vegetation Science*, **11**, 169-178.
- Bogino SM, Bravo F (2008) Growth response of *Pinus pinaster* Ait. to climatic variables in central Spanish forests. *Annals of Forest Science*, **65**.
- Boyden S, Binkley D, Senock R (2005) Competition and facilitation between *Eucalyptus* and nitrogen-fixing Falcataria in relation to soil fertility. *Ecology*, **86**, 992-1001.
- Burnham KP, Anderson DR (2002) *Model selection and multimodel inference: a practical information-theoretic approach*, New York, USA, Springer-Verlag.

- Burton P (1993) Some limitations inherent to static indexes of plant competittion. *Canadian* Journal of Forest Research, 23, 2141-2152.
- Cahill JF, Kembel SW, Gustafson DJ (2005) Differential genetic influences on competitive effect and response in *Arabidopsis thaliana*. *Journal of Ecology*, **93**, 958-967.
- Canham CD, Papaik MJ, Uriarte M, McWilliams WH, Jenkins JC, Twery MJ (2006) Neighborhood analyses of canopy tree competition along environmental gradients in New England forests. *Ecological Applications*, **16**, 540-554.
- Canham CD, Uriarte M (2006) Analysis of neighborhood dynamics of forest ecosystems using likelihood methods and modeling. *Ecological Applications*, **16**, 62-73.
- Castro J, Zamora R, Hodar JA, Gomez JM (2004) Seedling establishment of a boreal tree species (*Pinus sylvestris*) at its southernmost distribution limit: consequences of being in a marginal Mediterranean habitat. *Journal of Ecology*, **92**, 266-277.
- Cescatti A, Piutti E (1998) Silvicultural alternatives, competition regime and sensitivity to climate in a European beech forest. *Forest Ecology and Management*, **102**, 213-223.
- Coates KD, Canham CD, LePage PT (2009) Above- versus below-ground competitive effects and responses of a guild of temperate tree species. *Journal of Ecology*, **97**, 118-130.
- Coomes DA, Kunstler G, Canham CD, Wright E (2009) A greater range of shade-tolerance niches in nutrient-rich forests: an explanation for positive richness–productivity relationships? *Journal of Ecology*, **97**, 705-717.
- Cotillas M, Sabate S, Gracia C, Espelta JM (2009) Growth response of mixed mediterranean oak coppices to rainfall reduction: Could selective thinning have any influence on it? *Forest Ecology and Management*, **258**, 1677-1683.
- Christensen JH, Hewitson B, Busuioc A et al. 2007. Regional Climate Projections. Contribution of working group I to the fourth assessment report of the IPCC. In: *Climate Change 2007: The Physical Science Basis* (eds Solomon S, Qin D, Manning

M et al.), pp. 847-943. Cambridge, UK, Cambridge University Press.

- Comité Oleícola Permanente (1998) *Modelo Digital del Terreno del SIG Oleícola Español*, Madrid, Spain, Ministerio de Agricultura, Pesca y Alimentación.
- Curtis JT (1959) *The vegetation of Wisconsin: an ordination of plant communities*, Madison, Wisconsin, USA, University of Wisconsin Press.
- Daniels RF (1976) Simple competition indexes and their correlation with annual loblolly-pine tree growth. *Forest Science*, **22**, 454-456.
- Edwards MB (1992) Likelihood, Baltimore, Maryland, USA, Johns Hopkins University Press.
- Ferrio JP, Florit A, Vega A, Serrano L, Voltas J (2003) D13C and tree-ring width reflect different drought responses in *Quercus ilex* and *Pinus halepensis*. *Oecologia*, 442, 512-518.
- Fraser LH, Miletti TE (2008) Effect of minor water depth treatments on competitive effect and response of eight wetland plants. *Plant Ecology*, **195**, 33-43.
- Freckleton RP, Watkinson AR (2001) Predicting competition coefficients for plant mixtures: reciprocity, transitivity and correlations with life-history traits. *Ecology Letters*, 4, 348-357.
- Gauch HG (1982) *Multivariate analysis and community structure*, Cambridge, UK, Cambridge University Press.
- Gea-Izquierdo G, Martín-Benito D, Cherubini P, Isabel C (2009) Climate-growth variability in *Quercus ilex* L. west Iberian open woodlands of different stand density. *Ann. For. Sci.*, 66, 802.
- Goffe WL, Ferrier GD, Rogers J (1994) Global optimization of statistical functions with simulated annealing. *Journal of Econometrics*, **60**, 65-99.
- Goldberg DE (1990) Components of resource competition in plant communities. In: *Perspectives in Plant Competition* (eds Grace J, Tilman D) pp. 27-49. New York,

USA, Academic Press.

- Goldberg DE, Landa K (1991) Competitive effect and response: hierarchies and correlated traits in the early stages of competition. *Journal of Ecology*, **79**, 1013-1030.
- Gomez-Aparicio L, Zavala MA, Bonet FJ, Zamora R (2009) Are pine plantations valid tools for restoring Mediterranean forests? An assessment along abiotic and biotic gradients. *Ecological Applications*, **19**, 2124-2141.
- Gonzalo J (2008) Diagnosis fitoclimática de la España peninsular. Actualización y análisis geoestadístico aplicado. Unpublished PhD thesis, Universidad Politécnica de Madrid.
   Escuela Técnica Superior de Ingenieros de Montes, Madrid, 559 pp.
- Gordo O, Sanz JJ (2009) Long-term temporal changes of plant phenology in the Western Mediterranean. *Global Change Biology*, **15**, 1930-1948.
- Gracia CA, Sabate S, Martinez JM, Albeza E (1999) Functional responses to thinning. In:
   *Ecology of the Mediterranean Evergreen Oak Forests*. (eds Roda F, Retana J, Gracia CA, Bellot J) pp. 329–337. Berlin Heidelberg, Springer-Verlag.
- Hara T (1984) Dynamics of stand structure in plant monocultures. *Journal of Theoretical Biology*, **110**, 223-239.
- He FL, Duncan RP (2000) Density-dependent effects on tree survival in an old-growth Douglas fir forest. *Journal of Ecology*, **88**, 676-688.
- Hilborn R, Mangel M (1997) *The ecological detective: confronting models with data,* Princeton, NJ, USA, Princeton University Press.
- Hubbell SP (2001) *The unified theory of biodiversity and biogeography*, Princeton, USA, Princeton University Press.
- Hurteau M, Zald H, North M (2007) Species-specific response to climate reconstruction in upper-elevation mixed-conifer forests of the western Sierra Nevada, California. *Canadian Journal of Forest Research*, **37**, 1681-1691.

- Janzen DH (1970) Herbivores and the number of tree species in tropical forests. *The American Naturalist*, **104**, 501-528.
- Johnson JB, Omland KS (2004) Model selection in ecology and evolution. *Trends in Ecology* & *Evolution*, **19**, 101-108.
- Jump AS, Hunt JM, Peñuelas J (2006) Rapid climate change-related growth decline at the southern range edge of *Fagus sylvatica*. *Global Change Biology*, **12**, 2163-2174.
- Kohyama T (1992) Size-structured multi-species model of rain forest trees. *Functional Ecology*, **6**, 206-212.
- Lin J, Harcombe P, Fulton M, Hall R (2002) Sapling growth and survivorship as a function of light in a mesic forest of southeast Texas, USA. *Oecologia*, **132**, 428-435.
- Linares JC, Camarero JJ, Carreira JA (2009) Interacting effects of changes in climate and forest cover on mortality and growth of the southernmost European fir forests. *Global Ecology and Biogeography*, **18**, 485-497.
- Linares JC, Camarero JJ, Carreira JA (2010) Competition modulates the adaptation capacity of forests to climatic stress: insights from recent growth decline and death in relict stands of the Mediterranean fir *Abies pinsapo*. *Journal of Ecology*, **98**, 592-603.
- Martín-Benito D, Cherubini P, del Río M, Cañellas I (2008) Growth response to climate and drought in *Pinus nigra* Arn. trees of different crown classes. *Trees - Structure and Function*, 22, 363-373.
- Martín-Benito D, Del Rio M, Heinrich I, Helle G, Canellas I (2010) Response of climategrowth relationships and water use efficiency to thinning in a *Pinus nigra* afforestation. *Forest Ecology and Management*, **259**, 967-975.
- Martinez-Vilalta J, Lopez BC, Adell N, Badiella L, Ninyerola M (2008) Twentieth century increase of Scots pine radial growth in NE Spain shows strong climate interactions. *Global Change Biology*, 14, 2868-2881.

Martínez-Vilalta J, Mencuccini M, Vayreda J, Retana J (2010) Interspecific variation in functional traits, not climatic differences among species ranges, determines demographic rates across 44 temperate and Mediterranean tree species. *Journal of Ecology*, no-no.

Menzel A, Fabian P (1999) Growing season extended in Europe. Nature, 397, 659-659.

- Miriti MN (2006) Ontogenetic shift from facilitation to competition in a desert shrub. *Journal of Ecology*, **94**, 973-979.
- Miyazawa Y, Kikuzawa K (2005) Winter photosynthesis by saplings of evergreen broadleaved trees in a deciduous temperate forest. *New Phytologist*, **165**, 857-866.
- Muller-Landau HC, Condit RS, Chave J et al. (2006) Testing metabolic ecology theory for allometric scaling of tree size, growth and mortality in tropical forests. *Ecology Letters*, **9**, 575-588.
- Nakashizuka T (2001) Species coexistence in temperate, mixed deciduous forests. *Trends in Ecology & Evolution*, **16**, 205-210.
- Ogaya R, Penuelas J, Martinez-Vilalta J, Mangiron M (2003) Effect of drought on diameter increment of *Quercus ilex*, *Phillyrea latifolia*, and *Arbutus unedo* in a holm oak forest of NE Spain. *Forest Ecology and Management*, **180**, 175-184.
- Oliver CD, Larson BC (1996) Forest stand dynamics, New York, USA, John Wiley & Sons.
- Pacala SW, Canham CD, Silander JA, Kobe RK (1994) Sapling growth as a function of resources in a north temperate forest. *Canadian Journal of Forest Research*, 23, 1980-1988.
- Peñuelas J, Filella I, Comas P (2002) Changed plant and animal life cycles from 1952 to 2000 in the Mediterranean region. *Global Change Biology*, **8**, 531-544.
- Pérez-Ramos IM, Ourcival JM, Limousin JM, Rambal S (2010) Mast seeding under increasing drought: results from a long-term data set and from a rainfall exclusion

experiment. Ecology, 91, 3057-3068.

- Pigott CD, Pigott S (1993) Water as a determinant of the distribution of trees at the boundary of the Mediterranean zone. *Journal of Ecology*, **81**, 557-566.
- Prentice IC, Cramer W, Harrison SP, Leemans R, Monserud RA, Solomon AM (1992) A global biome model based on plant physiology and dominance, soil properties and climate. *Journal of Biogeography*, **19**, 117-134.
- Purves DW (2009) The demography of range boundaries versus range cores in eastern US tree species. *Proceedings of the Royal Society B: Biological Sciences*, **276**, 1477-1484.
- Quero JL, Gómez-Aparicio L, Zamora R, Maestre FT (2008) Shifts in the regeneration niche of an endangered tree (*Acer opalus* ssp *granatense*) during ontogeny: Using an ecological concept for application. *Basic and Applied Ecology*, **9**, 635-644.
- Ryan MG, Binkley D, Fownes JH, Giardina CP, Senock RS (2004) An experimental test of the causes of forest growth decline with stand age. *Ecological Monographs*, **74**, 393-414.
- Reich PB, Oleksyn J (2008) Climate warming will reduce growth and survival of Scots pine except in the far north. *Ecology Letters*, **11**, 588-597.
- Russo SE, Wiser SK, Coomes DA (2007) Growth–size scaling relationships of woody plant species differ from predictions of the Metabolic Ecology Model. *Ecology Letters*, **10**, 889-901.
- Sabaté S, Gracia CA, Sanchez A (2002) Likely effects of climate change on growth of *Quercus ilex, Pinus halepensis, Pinus pinaster, Pinus sylvestris* and *Fagus sylvatica* forests in the Mediterranean region. *Forest Ecology and Management*, **162**, 23-37.
- Sánchez-Gómez D, Valladares F, Zavala MA (2006) Performance of seedlings of Mediterranean woody species under experimental gradients of irradiance and water

availability: trade-offs and evidence for niche differentiation. *New Phytologist*, **170**, 795-806.

- Sánchez-Gómez D, Zavala MA, Van Schalkwijk DB, Urbieta IR, Valladares F (2008) Rank reversals in tree growth along tree size, competition and climatic gradients for four forest canopy dominant species in Central Spain. *Annals of Forest Science*, **65**, Article Number 605.
- Schroter D, Cramer W, Leemans R *et al.* (2005) Ecosystem service supply and vulnerability to global change in Europe. *Science*, **310**, 1333-1337.
- Seynave I, Gegout JC, Herve JC, Dhote JF (2008) Is the spatial distribution of European beech (*Fagus sylvatica* L.) limited by its potential height growth? *Journal of Biogeography*, 35, 1851-1862.
- Shugart HH (1984) A theory of forest dynamics: the ecological implications of forest succession models, New York, USA, Springer-Verlag.
- Silander JA, Pacala SW (1985) Neighborhood predictors of plant performance. *Oecologia*, **66**, 256-263.
- Simard SW, Sachs DL (2004) Assessment of interspecific competition using relative height and distance indices in an age sequence of seral interior cedar-hemlock forests in British Columbia. *Canadian Journal of Forest Research*, **34**, 1228-1240.
- Stoll P, Newbery DM (2005) Evidence of species-specific neighborhood effects in the dipterocarpaceae of a Bornean rain forest. *Ecology*, 86, 3048-3062.
- Stoll P, Weiner J, Schmid B (1994) Growth variation in a natural established population of *Pinus sylvestris. Ecology*, **75**, 660-670.
- Terradas J & Savé R (1992) The influence of summer and winter stress and water relationships on the distribution of *Quercus ilex* L. *Vegetatio*, **100**, 137-145.

Tilman D (1982) Resource competition and community structure, Princeton, New Jersey,

USA, Princeton University Press.

- Tilman D (1988) Plant strategies and the dynamics and structure of plant communities, Princeton, New Jersey, USA, Princeton University Press.
- Uriarte M, Canham CD, Thompson J, Zimmerman JK (2004) A neighborhood analysis of tree growth and survival in a hurricane-driven tropical forest. *Ecological Monographs*, 74, 591-614.
- Vicente-Serrano SM, Lasanta T, Gracia C (2010) Aridification determines changes in forest growth in *Pinus halepensis* forests under semiarid Mediterranean climate conditions. *Agricultural and Forest Meteorology*, **150**, 614-628.
- Villaescusa R, Diaz R (1998) Segundo Inventario Forestal Nacional (1986-1996). España, Madrid, Spain, Ministerio de Medio Ambiente, ICONA.
- Villanueva JA (2004) Tercer Inventario Forestal Nacional (1997-2007). Comunidad de Madrid, Madrid, Spain, Ministerio de Medio Ambiente.
- Vitasse Y, Porte AJ, Kremer A, Michalet R, Delzon S (2009) Responses of canopy duration to temperature changes in four temperate tree species: relative contributions of spring and autumn leaf phenology. *Oecologia*, **161**, 187-198.
- Wagner RG, Radosevich SR (1998) Neighborhood approach for quantifying interspecific competition in coastal Oregon forests. *Ecological Applications*, **8**, 779-794.
- Wang P, Stieglitz T, Zhou DW, Cahill JF (2010) Are competitive effect and response two sides of the same coin, or fundamentally different? *Functional Ecology*, **24**, 196-207.

Whittaker RH (1975) Communities and ecosystems., New York, USA, MacMillan New York.

Wright SJ (2002) Plant diversity in tropical forests: a review of mechanisms of species coexistence. Oecologia, 130, 1-14.

- Wyckoff PH, Clark JS (2002) The relationship between growth and mortality for seven cooccurring tree species in the southern appalachian mountains. *Journal of Ecology*, **90**, 604-615.
- Zavala MA, Angulo O, de la Parra RB, Lopez-Mareos JC (2007) An analytical model of stand dynamics as a function of tree growth, mortality and recruitment: The shade tolerance-stand structure hypothesis revisited. *Journal of Theoretical Biology*, 244, 440-450.
- Zavala MA, Espelta JM, Retana J (2000) Constraints and trade-offs in Mediterranean plant communities: The case of holm oak-Aleppo pine forests. *Botanical Review*, 66, 119-149.
- Zavala MA, Oria JA (1995) Preserving biological diversity in managed forest: a meeting point for ecology and forestry. *Landscape and Urban Planning*, **31**, 363-378.
- Zavala MA, Zea E (2004) Mechanisms maintaining biodiversity in Mediterranean pine-oak forests: insights from a spatial simulation model. *Plant Ecology*, **171**, 197-207.
- Zhao DH, Borders B, Wilson M, Rathbun SL (2006) Modeling neighborhood effects on the growth and survival of individual trees in a natural temperate species-rich forest. *Ecological Modelling*, **196**, 90-102.

Table 1.	Sample sizes and	mean [minimum	- maximum]	annual	temperature	(°C), annua	l rainfall (	mm), tai	rget d.b.h.	(cm) a	at the start
of the gr	owth period, numb	per of neighbors in	n a 5-m radius	s plot, a	nd basal area	of neighbor	$rs (m^2/ha)$	for the 1	15 study spo	ecies.	

Species	Acronym	Functional group	N° target	N°	Mean annual	Annual	d.b.h.	N°	Basal
			trees	plots	temperature	rainfall		Neighbors	area
Juniperus	JUTH	Mediterranean conifer	2015	844	10.1	617	13.2	3	0.71
thurifera					[8.1 - 14.6]	[366 - 1347]	[7.5 - 75.4]	[0 - 25]	[0 - 15.56]
Pinus	PIHA	Mediterranean conifer	15403	5139	13.8	517	13.9	4	0.99
halepensis					[10.0 - 19.2]	[170 - 1484]	[7.5 - 70.]	[0 - 27]	[0 - 7.50]
Pinus	PIPINE	Mediterranean conifer	3426	1435	14.9	642	18.0	4	1.13
pinea					[10.9 - 18.5]	[304 - 1682]	[7.5 - 96.1]	[0 - 46]	[0 - 9.48]
Pinus	PIPINA	Mediterranean conifer	16730	4867	12.2	742	17.9	6	2.12
pinaster					[8.1 - 18.5]	[322 - 3058]	[7.5 - 71.6]	[0 - 46]	[0 - 14.57]
Pinus	PINI	Mountain conifer	14409	3754	11.0	717	13.4	7	1.70
nigra					[6.3 - 18.3]	[360 - 2240]	[7.5 - 95.5]	[0 - 47]	[0 - 15.56]
Pinus	PISY	Mountain conifer	22311	5063	9.2	920	15.1	8	2.26
sylvestris					[3.2 - 14.6]	[428 - 1977]	[7.5 - 128.3]	[0 - 45]	[0 - 17.96]
Pinus	PIUN	Mountain conifer	2286	546	6.0	1191	16.0	7	2.55
uncinata					[2.4 - 14.2]	[667 - 1975]	[7.5 - 78.0]	[0 - 32]	[0 - 14.99]
Quercus	QUIL	Mediterranean	21700	7075	12.1	723	10.5	7	1.13
ilex		evergreen broadleaved			[6.6 - 18.6]	[268 - 1812]	[7.5 - 152.2]	[0 - 53]	[0 - 24.19]
Quercus	QUSU	Mediterranean	3352	1360	15.1	799	16.4	5	1.41
suber		evergreen broadleaved			[10.1 - 18.5]	[480 - 1862]	[7.5 - 106.6]	[0 - 30]	[0 - 15.42]
Quercus	QUPY	Sub-Mediterranean	9004	2164	10.3	876	11.1	9	1.56
pyrenaica		deciduous broadleaved			[5.9 - 16.4]	[460 - 2304]	[7.5 - 129.9]	[0 - 40]	[0 - 26.45]
Quercus	QUFA	Sub-Mediterranean	6123	2009	11.1	735	10.	8	1.41
faginea		deciduous broadleaved			[5.8 - 18.3]	[409 - 1574]	[7.5 - 103.5]	[0 - 50]	[0 - 11.88]

 Table 1. Extended

Species	Acronym	Functional group	N <sup>o</sup> target	N°	Mean annual	Annual	d.b.h.	N°	Basal
-	2	<b>C</b> 1	trees	plots	temperature	rainfall		Neighbors	area
Quercus	QUPE	Atlantic deciduous	1614	513	10.3	991	12.4	8	1.70
petraea		broadleaved			[4.0 - 16.2]	[540 - 2066]	[7.5 - 146.5]	[0 - 33]	[0 - 24.33]
Quercus	QURO	Atlantic deciduous	1928	754	11.7	1273	15.0	5	1.70
robur		broadleaved			[6.4 - 14.7]	[461 - 2236]	[7.5 - 127.3]	[0 - 30]	[0 - 17.82]
Fagus	FASY	Atlantic deciduous	4391	1032	9.0	1128	14.4	8	2.40
sylvatica		broadleaved			[4.8 - 13.7]	[587 - 2150]	[7.5 - 112.0]	[0 - 43]	[0 - 27.02]
Castanea	CASA	Atlantic deciduous	1138	436	12.3	1142	14.3	7	1.84
sativa		broadleaved			[8.3 - 16.2]	[637 - 2275]	[7.5 - 219.6]	[0 - 34]	[0 - 18.53]

**Table 2.** Comparison of the alternative models for the 15 target tree species using AIC. The four full models varied in their treatment of competition, considering species-specific competition, intra- vs. inter-specific competition, conifer vs. angiosperm competition, and equivalent competition. The "No climate", "No size", and "No comp." models ignore the effect of climate, size, and competition, respectively. The best fitting model is given a  $\Delta$ AIC value of zero (in bold), and the difference in AIC between this model and all others is reported.

	ΔΑΙC													
		Fi	ull											
Species	Species	Intra. vs	Conif.vs	Equiv.	No	No	No	Null	δ	σ	φ	NP	Slope	$R^2$
	specific	inter.	Ang.	comp.	climate	size	comp.							
JUTH	5.2	0	12.8	1.9	7.4	8.9	137.1	149.9	Yes	Yes	Yes	15	1.01	0.14
PIHA	29.9	0	12.3	204.2	520.4	404.9	804.9	2481.6	Yes	Yes	Yes	13	1.00	0.26
PIPINE	0	23.8	33.9	50.8	52.3	2.2	410.5	580.8	Yes	Yes	Yes	19	0.99	0.25
PIPINA	0	82.0	49.6	85.2	335.1	4.9	1848.5	2805.7	Yes	Yes	Yes	25	1.00	0.29
PINI	0	66.2	78.1	155.2	197.7	65.2	2029.1	2959.2	Yes	Yes	Yes	30	0.99	0.34
PISY	0	245.4	287.3	254.5	159.3	115.4	2723.7	4034.4	Yes	Yes	Yes	30	1.00	0.30
PIUN	0	55.4	198.6	22.8	4.1	16.9	207.2	311.0	Yes	No	Yes	16	1.04	0.25
QUIL	0	40.0	396.4	44.3	2815.6	3021.6	2847.2	4056.4	Yes	Yes	Yes	30	0.99	0.14
QUSU	0	8.3	162.9	2.0	54.4	145.3	249.2	293.1	No	Yes	Yes	13	1.01	0.13
QUPY	8.2	0	4.7	2.5	208.1	382.9	889.9	1450.3	Yes	Yes	Yes	15	1.01	0.20
QUFA	0	20.5	76.7	75.1	185.5	299.1	489.1	1035.1	Yes	Yes	Yes	23	1.00	0.17
QUPE	0	16.9	39.5	40.6	32.3	100.7	106.6	264.9	Yes	No	Yes	20	1.01	0.21
QURO	0	19.8	7.4	23.6	41.7	46.6	105.5	204.7	Yes	No	Yes	19	1.00	0.16
FASY	3.3	0	147.6	193.2	221.9	225.3	363.3	1019.5	Yes	Yes	Yes	15	1.00	0.30
CASA	0	58.1	14.9	26.1	12.4	31.4	137.2	172.3	Yes	Yes	Yes	19	1.01	0.15

*Notes*: Species are identified with acronyms; see Table 1 for full names. The  $\delta$ ,  $\sigma$  and  $\phi$  columns indicate whether (Y, yes; N, no) the best model included a term that allows sensitivity to competition to vary with mean annual temperature ( $\delta$ ), annual precipitation ( $\sigma$ ) or target size ( $\phi$ ). NP is the total number of parameters in the best model. The slope and R<sup>2</sup> for the relationship between predicted and observed growth are also given.

**Table 3.** Per capita competition coefficients  $(\lambda_{i,z})$  with 2-unit support intervals for the effects of a neighbor of species *i* on a target species *z* for the 15 most common tree species of the Iberian Peninsula.

						Effect of nei	ghbor					
On target	JUTH	PIHA	PIPINE	PIPINA	PINI	PISY	PIUN	QUIL	QUSU	QUPY	QUFA	QUPE
JUTH	1				0.92			0.92			0.67	
	[0.75-1]				[0.85-1]			[0.60-1]			[0.43-0.87]	
PIHA		1	0.13	0.01	0.04	0.84		0.04	0.01		0.22	
		[0.90-1]	[0-0.27]	[0-0.11]	[0-0.10]	[0.61-0.94]		[0-0.15]	[0-0.12]		[0-0.47]	
PIPINE		1	0.67	0.17				0.49	0.02			
		[0.85-1]	[0.56-0.71]	[0-0.34]				[0.23-0.67]	[0-0.11]			
PIPINA		1	0.69	0.40	0.48	0.36		0.31	0	0.01	0.70	
		[0.75-1]	[0.54-1]	[0.38-0.45]	[0.32-0.58]	[0.20-0.48]		[0.18-0.58]	[0-0.12]	[0-0.06]	[0.55-0.92]	
PINI	0.95	0.94		0.47	1	0.86		0.85		0.01	0.00	0.06
	[0.60-1]	[0.74-1]		[0.32-0.62]	[0.89-1]	[0.61-1]		[0.66-1]		[0-0.20]	[0-0.35]	[0-0.19]
PISY	0.80	0.38		0.31	1	0.63	0.25	0.65		0.00	0.39	0.01
	[0.50-1]	[0.10-0.50]		[0.20-0.39]	[0.92-1]	[0.57-0.67]	[0-0.40]	[0.45-0.82]		[0-0.05]	[0.24-0.46]	[0-0.15]
PIUN						1	0.61					
						[0.91-1]	[0.55-0.68]					
QUIL	0	0.98	0.34	0.02	1	1		0.75	0.06	0.23	0.31	0.95
	[0-0.50]	[0.58-1]	[0.14-0.60]	[0-0.32]	[0.85-1]	[0.75-1]		[0.65-0.90]	[0-0.26]	[0.09-0.32]	[0.12-0.56]	[0.80-1]
QUSU		1	0.84	0.11				0.75	0.61	0.28	0.80	
		[0.70-1]	[0.55-1]	[0-0.37]				[0.60-0.90]	[0.51-0.76]	[0.11-0.33]	[0.62-0.95]	
QUPY				0.87	0.71	0.92		1	0.80	1	1	0.98
				[0.62-1]	[0.55-0.88]	[0.67-1]		[0.68-1]	[0.70-1]	[0.90-1]	[0.80-1]	[0.73-1]
QUFA	0.16	1		0.39	0.26	0.80		0.25		0.11	1	
	[0-0.30]	[0.95-1]		[0.12-0.61]	[0.11-0.32]	[0.55-1]		[0.05-0.39]		[0-0.23]	[0.95-1]	
QUPE					0.52	0.96		0.23		0.29		1
					[0.31-0.70]	[0.56-1]		[0-0.51]		[0-0.35]		[0.80-1]
QURO				0.26				0.96		0.33		
				[0-0.39]				[0.82-1]		[0.22-0.48]		
FASY						0.06		0.02		0	0.76	0.01
~ . ~ .						[0-0.11]		[0-0.05]		[0-0.05]	[0.67-1]	[0-0.05]
CASA				0.01				0.55		1		
				[0-0.20]				[0.20-0.65]		[0.65-1]		
Mean CE	0.58	0.91	0.53	0.27	0.66	0.74	0.43	0.56	0.25	0.30	0.59	0.50

Gómez-Aparicio et al. 38

 Table 3. Extended

	Eff	ect of neighbor	
On target	QURO	FASY	CASA
JUTH			
PIHA			
PIPINE			
PIPINA	0		0
PINI	[0-0.13]		[0-0.36]
PISY	0.01	0.82	
PIUN	[0-0.09]	[0.60-1]	
QUIL	0.05	1	0.83
QUSU	[0-0.19]	[0.80-1]	[0.72-1]
QUPY	0.73	0.91	0.71
QUFA	[0.58-1]	[0.76-1] 0.94	[0.60-1]
QUPE		0.65	
QURO	1	[0.40-0.90] 	0.03
FASY	[ <b>U.85-1</b> ] 0.53	1	[0-0.25]
	[0.37-0.72]	[0.94-1]	
CASA	0.84		0.53
	[0.54-1]		[0.39-0.69]
Mean CE	0.35	0.90	0.44

*Notes:* Species are identified with acronyms; see Table 1 for full names. Coefficients were estimated only when there were at least 100 neighbors of that species (dashed lines otherwise). The competition coefficients are scaled so that the strongest competitor for each target tree species has a value of 1. Conspecific

interactions are shown in bold. Also shown is the mean competitive effect (CE) of each neighbor species on all target species for which it was a common neighbor.

#### **Figure legends**

**Figure 1.** Predicted effect of a) mean annual temperature, b) annual precipitation, c) target tree size, and d) NCI (neighborhood competition index) on growth for the 15 most common tree species of the Iberian Peninsula. The effects of climate and size are calculated in the absence of competition. The effect of competition is calculated for a tree of mean size at mean climatic conditions (see Table 1 for mean values). See Appendix 1 for the estimated parameters of the corresponding functions. See Table 1 for key to species abbreviations.

**Figure 2.** Predicted variation of NCI (neighborhood competition index) with distance to neighbors for the 15 target tree species studied. To simplify presentation of results, the parameter  $\alpha$  was set to zero in Eqn. 5. See Table 1 for key to species abbreviations.

**Figure 3.** Predicted variation in potential growth along a competition gradient for four different combinations of climatic and tree size conditions: a) small trees (d.b.h. = 10 cm) in xeric sites ( $T = 14^{\circ}C$ , P = 600 mm); b) large trees (d.b.h. = 40 cm) in xeric sites; c) small trees (d.b.h. = 10 cm) in mesic sites ( $T = 8^{\circ}C$ , P = 1500 mm); and c) large trees (d.b.h. = 40 cm) in mesic sites. See Appendix 1 for the estimated parameters of the corresponding functions. See Table 1 for key to species abbreviations.

**Figure 4.** Predicted variation in potential growth of *Pinus sylvestris* along a temperature and competition gradient for two size classes: a) d.b.h. = 10 cm, and b) d.b.h. = 40 cm. Observe how the response to temperature is almost null at high competition levels (i.e. NCI = 1), particularly for small trees.

Appendix S1. Parameter estimates and 2-unit support intervals (in brackets) for the best model selected for each of the 15 target tree

Species	PDG	$X1_0$	X1 <sub>b</sub>	$X2_0$	X2 <sub>b</sub>	$X3_0$	X3 <sub>b</sub>
Juniperus thurifera	4.44	2.35	18.98	1994.38	1537.90	49.47	2.95
	[4.39-4.49]	[2.34-2.36]	[18.97-18.99]	[1994.37-1994.39]	[1537.88-1537.92]	[49.46-49.48]	[2.90-3.00]
Pinus halepensis	8.72	1.65	30.13	1002.05	729.77	24.88	1.33
	[8.62-8.82]	[1.64-1.66]	[30.12-30.14]	[1002.03-1002.06]	[727.87-730.97]	[24.33-25.68]	[1.28-1.38]
Pinus pinaster	12.40	2.60	18.20	1884.71	1577.19	23.30	1.45
	[12.39-12.41]	[2.59-2.61]	[18.19-18.21]	[1884.70-1884.72]	[1577.18-1577.20]	[23.29-23.31]	[1.44-1.46]
Pinus pinea	8.25	20.61	260.32	1100.45	760.64	26.68	5.63
	[8.20-8.30]	[20.60-20.62]	[260.31-260.33]	[1100.44-1100.46]	[760.63-760.65]	[26.63-26.73]	[5.58-5.68]
Pinus nigra	9.39	0.47	11.59	2569.75	2131.21	18.68	1.05
	[9.35-9.43]	[0.42-0.52]	[11.54-11.64]	[2569.70-2569.80]	[2131.16-2131.26]	[18.64-18.74]	[1.00-1.10]
Pinus sylvestris	9.13	1.09	11.75	2386.82	2660.83	19.92	1.11
	[9.12-9.14]	[1.08-1.10]	[11.74-11.76]	[2386.81-2386.83]	[2660.82-2660.84]	[19.91-19.93]	[1.10-1.12]
Pinus uncinata	6.52	71.59	230.38	1369.72	2999.98	15.57	1.44
	[6.47-6.65]	[71.58-71.60]	[230.37-230.39]	[1369.71-1369.73]	[2999.97-2999.99]	[13.86-17.17]	[1.29-1.96]
Quercus ilex	3.96	25.81	24.52	1262.49	1031.06	25.97	1.75
	[3.91-4.01]	[25.80-25.82]	[24.51-24.53]	[1262.48-1262.50]	[1031.05-1031.07]	[25.92-26.02]	[1.70-1.80]
Quercus suber	4.15	28.56	38.79	1614.59	1963.27	81.81	2.81
	[4.14-4.16]	[28.55-28.57]	[38.78-38.80]	[1614.58-1614.60]	[1963.26-1963.28]	[81.79-81.82]	[2.80-2.82]
Quercus pyrenaica	5.24	18.10	12.16	1172.17	1642.55	43.07	1.68
	[5.19-5.36]	[18.09-18.11]	[12.15-12.17]	[1172.16-1172.18]	[1642.54-1642.56]	[43.02-43.12]	[1.63-1.73]
Quercus faginea	5.27	37.21	26.16	1297.80	1147.10	52.99	1.70
	[5.22-5.32]	[37.20-37.22]	[26.15-26.17]	[1297.79-1297.81]	[1147.09-1147.11]	[52.94-53.04]	[1.65-1.75]
Quercus petraea	5.01	25.41	21.80	1021.65	2820.18	34.15	1.33
	[4.95-5.05]	[25.40-25.42]	[21.79-21.81]	[1021.64-1021.66]	[2820.17-2820.19]	[34.10-34.20]	[1.28-1.38]
Quercus robur	9.60	27.76	14.68	1316.30	977.04	47.54	1.84
	[9.55-9.65]	[27.75-27.77]	[14.67-14.69]	[1316.29-1316.31]	[977.03-977.05]	[47.49-47.59]	[1.79-1.89]
Fagus sylvatica	7.52	19.79	10.10	254.71	1670.40	43.41	1.38
	[7.47-7.57]	[19.78-19.80]	[10.09-10.11]	[254.70-254.72]	[1670.39-1670.41]	[43.36-43.46]	[1.33-1.43]
Castanea sativa	8.15	22.52	15.55	1303.24	1263.07	26.75	1.67
	[8.10-8.20]	[22.51-22.53]	[15.54-15.56]	[1303.23-1303.25]	[1263.06-1263.08]	[26.70-26.80]	[1.62-1.72]

species. See text for a description of the parameters.

# Appendix S1. Extended

Species	а	b	α	β	γ	δ	σ	φ	n
Juniperus thurifera	195.23	1.00	1.76	1.13	0.18	0.10	-0.56	-0.53	1.58
	[195.20-195.26]	[1-1.05]	[1.72-1.82]	[1.11-1.14]	[0.17-0.19]	[0.09-0.11]	[-0.57 to -0.55]	[-0.54 to -0.52]	[1.53-1.63]
Pinus halepensis	144.26	1.00	1.80	0.80	0.43	0.39	-0.23	-1.53	2.24
	[144.25-144.27]	[1-1.05]	[1.75-1.85]	[0.74-0.87]	[0.39-0.49]	[0.38-0.40]	[-0.24 to -0.22]	[-1.54 to -1.52]	[2.19-2.29]
Pinus pinaster	282.43	1.00	1.96	0.50	0.63	0.54	-0.41	-1.19	2.19
	[282.31-282.56]	[1-1.01]	[1.94-2.00]	[0.47-0.54]	[0.56-0.71]	[0.53-0.55]	[-0.42 to -0.40]	[-1.20 to -1.18]	[2.14-2.20]
Pinus pinea	25.88	1.00	1.82	0.91	0.13	0.19	-0.84	-0.57	2.25
	[24.78-26.66]	[1-1.05]	[1.77-1.86]	[0.79-1.16]	[0.08-0.25]	[0.18-0.20]	[-0.85 to -0.83]	[-0.58 to -0.56]	[2.24-2.26]
Pinus nigra	186.39	1.00	1.92	0.75	0.58	0.11	-0.19	-1.23	1.97
	[184.42-188.01]	[1-1.05]	[1.90-1.94]	[0.74-0.76]	[0.56-0.61]	[0.10-0.12]	[-0.20 to -0.18]	[-1.24 to -1.22]	[1.92-2.02]
Pinus sylvestris	78.56	1.00	1.79	0.61	0.81	0.32	-0.13	-1.11	2.24
	[78.55-78.57]	[1-1.01]	[1.76-1.80]	[0.60-0.62]	[0.79-0.83]	[0.31-0.33]	[-0.14 to -0.12]	[-1.12 to -1.10]	[2.22-2.26]
Pinus uncinata	22.49	1.00	1.80	1.13	0.32	0.27		-1.05	1.79
	[22.48-22.50]	[1-1.05]	[1.76-1.86]	[1.09-1.17]	[0.29-0.35]	[0.26-0.28]		[-1.03 to -1.06]	[1.78-1.80]
Quercus ilex	0.52	1.00	0.98	7.47	0.00	0.71	-0.05	-0.39	1.65
	[0.49-0.61]	[1-1.01]	[0.89-1.08]	[7.01-8.03]	[0-0.20]	[0.70 - 0.72]	[-0.06 to -0.04]	[-0.40 to -0.38]	[1.64-1.66]
Quercus suber	412.75	1.00	0.95	6.59	0.00		-0.73	-0.58	1.53
	[412.64-412.82]	[1-1.01]	[0.81-1.23]	[6.10-6.74]	[0-0.38]		[-0.75 to -0.70]	[-0.59 to -0.56]	[1.52-1.54]
Quercus pyrenaica	134.00	1.00	1.36	4.64	0.00	0.11	-0.12	-1.57	2.02
	[133.99-134.01]	[1-1.05]	[1.35-1.38]	[3.98-4.92]	[0-0.15]	[0.10-0.12]	[-0.13 to -0.11]	[-1.56 to -1.58]	[1.97-2.07]
Quercus faginea	717.19	1.00	1.25	6.92	0.00	-0.50	-0.26	-1.41	2.29
	[717.18-717.20]	[1-1.05]	[1.10-1.44]	[5.28-7.56]	[0-0.05]	[-0.54 to -0.44]	[-0.30 to -0.20]	[-1.42 to -1.40]	[2.24-2.34]
Quercus petraea	41.53	1.00	1.38	3.60	0.00	-0.25		-1.13	1.92
	[41.51-41.54]	[1-1.05]	[1.34-1.42]	[3.12-3.99]	[0-0.35]	[-0.26 to -0.24]		[-1.14 to -1.12]	[1.87-1.97]
Quercus robur	82.77	1.01	1.40	0.90	0.20	-0.95		-0.59	1.57
	[82.76-82.78]	[1-1.05]	[1.39-1.50]	[0.81-1.02]	[0.18-0.23]	[-0.98 to -0.91]		[-0.63 to -0.54]	[1.47-1.61]
Fagus sylvatica	1820.59	1.00	1.31	0.83	0.24	-0.90	-0.29	-1.07	1.39
	[1820.58-1820.60]	[1-1.05]	[1.28-1.36]	[0.69-1.03]	[0.19-0.36]	[-0.91 to -0.89]	[-0.30 to -0.28]	[-1.08 to -1.05]	[1.34-1.44]
Castanea sativa	392.61	1.00	0.76	4.56	0.00	-0.32	-0.33	-0.95	1.98
	[392.60-392.62]	[1-1.05]	[0.61-1.16]	[3.87-5.03]	[0-0.05]	[-0.33 to -0.31]	[-0.34 to -0.32]	[-0.98 to -0.92]	[1.97-1.99]