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Demographic performance of European tree species at their hot and cold climatic edges

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Abstract

1. Species range limits are thought to result from a decline in demographic performance at range edges. However, recent studies reporting contradictory patterns in species demographic performance at their edges cast doubt on our ability to predict climate change demographic impacts. To understand these inconsistent demographic responses, we need to shift the focus from geographic to climatic edges and analyse how species responses vary with climatic constraints at the edge and species' ecological strategy.

2. Here we parameterised integral projection models with climate and competition effects for 27 tree species using forest inventory data from over 90,000 plots across Europe. Our models estimate size-dependent climatic responses and evaluate their effects on two life trajectory metrics: life span and passage time—the time to grow to a large size. Then we predicted growth, survival, life span and passage time at the hot and dry or cold and wet edges and compared them to their values at the species climatic centre to derive indices of demographic response at the edge. Using these indices, we investigated whether differences in species demographic response between hot and cold edges could be explained by their position along the climate gradient and functional traits related to their climate stress tolerance.

3. We found that at cold and wet edges of European tree species, growth and passage time were constrained, whereas at their hot and dry edges, survival and life span were constrained. Demographic constraints at the edge were stronger for species occurring in extreme conditions, that is, in hot edges of hot-distributed species and cold edges of cold-distributed species. Species leaf nitrogen content was strongly linked to their demographic responses at the edge. In contrast,

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we found only weak links with wood density, leaf size and xylem vulnerability to embolism.

4. *Synthesis*. Our study presents a more complicated picture than previously thought with demographic responses that differ between hot and cold edges. Predictions of climate change impacts should be refined to include edge and species characteristics.

1 | INTRODUCTION

In the face of climate change, there are increasing concerns about the future redistribution of plant species ranges (Zimmermann et al., 2013). Range shifts are thought to be directly related to changes in population dynamics. The classical view of the link between population dynamics and species ranges comes from a long-standing hypothesis in biogeography known as the 'abundant-centre hypothesis' (hereafter ACH, Brown, 1984; Pironon et al., 2017), which proposes that demographic performance decline at the range edge results in a decrease in abundance, occupancy and genetic diversity (Pironon et al., 2017). This is directly related to the hypothesis that at equilibrium, a species' range edge should occur where the mean population growth rate (λ) drops below one ($\lambda < 1$) due to changes in one or more vital rates (i.e. survival, growth or reproduction; Case et al., 2005; Holt & Keitt, 2005).

Understanding the demographic pathways of population response at range edges is crucial for forecasting climate change impacts. However, existing studies comparing population growth rates or vital rates in the periphery versus the centre of species geographic range provide weak support for the ACH (Pironon et al., 2017). Transplant experiments have shown a decline in population growth rate or vital rates beyond the geographic edge but not necessarily right at the edge (Hargreaves et al., 2014; Lee-Yaw et al., 2016). Similarly, model-based analyses of natural population monitoring data have found no clear evidence of a decrease in demographic performance at the geographic edge (Csergo et al., 2017; Purves, 2009).

Recent reviews have highlighted the difficulties of synthesising existing results because most studies explored performance of geographically peripheral populations without a clear understanding of the local climatic or environmental constrains (Pironon et al., 2017; Vilà-Cabrera et al., 2019). Changes in demographic performance are, however, likely to vary depending on the type of biophysical constraints at the edge (Gaston, 2009) and therefore, demographic performance at the edge should be analysed in relation to the local main climatic contraints (the 'central-marginal' hypothesis in Pironon et al., 2017). Firstly, demographic constraints could differ between drought- and cold-limited edges because tolerance to different abiotic stresses requires different adaptative strategies (Niinemets & Valladares, 2006) resulting in different vital rates being constraint at these edges (Gaston, 2009; Hargreaves et al., 2014). Secondly, it

has been proposed that biotic interactions (*e.g.* competition) could be key constraints of demographic performance at the edge and that this effect would be stronger for edges in productive environments than in unproductive environments. However, support for this hypothesis is limited (see Cahill et al., 2014; Hargreaves et al., 2014; Louthan et al., 2015). Thirdly, constraints on the demographic performance at a climatic edge are likely to vary with species' physiological strategy (Anderegg et al., 2019). These physiological differences can be captured by species' climatic optimum and by functional traits related to species physiological climate response, such as wood (Chave et al., 2009) or leaf characteristics (Wright et al., 2017). Finally, an additional difficulty arises in long-lived organisms such as trees because the response of their vital rates to climatic constraints at the edge might vary depending on the size of the individual (Tredennick et al., 2018). This size-dependent response to climate can be crucial for size-structured populations (De Roos et al., 2003; Tredennick et al., 2018) and can affect the population performance at the edge. We thus need to analyse the performance at the edge with size-structured models translating size-dependent climatic responses and the demographic compensation effect that may occur between sizes and vital rates into life trajectory metrics.

Here, we explored these questions in European forests, which play a crucial role for multiple ecosystem services such as sheltering a significant proportion of biodiversity and carbon stocks and contributing to the livelihoods of local populations (van der Plas et al., 2018). We used size-structured models fitted to forest inventory data documenting survival and growth of more than one million adult trees across the continent covering Mediterranean, temperate and boreal biomes. Firstly, we fitted survival and growth models for 27 species to capture size-dependent climate and competition responses of these vital rates. Secondly, we built size-structured population models using integral projection models (IPM; Ellner et al., 2016) to evaluate how size-dependent responses to climatic constraints at the edge translate into two life trajectory metrics—mean life span and passage time (time to grow from small to large size). Mean life span and passage time scale up individual level measurements into metrics summarising the population dynamics and they facilitate the comparison of the demographic structure between populations (Cochran & Ellner, 1992). These metrics are also key elements of population growth rate and the rate of population turnover. We then used these models to compare

species' predicted demographic performance at the 'hot and dry' or 'cold and wet' climatic edges with their performance at the climatic centre. Using these metrics we tested the following hypotheses: (1) vital rates and IPM-derived performance metrics are reduced at the climatic edge compared to the climatic centre but the demographic metrics affected differ between cold and hot edges; (2) the decline in demographic performance at the climatic edges is stronger in the presence of local competition than without; and (3) demographic performance at the climatic edge depends on species' median climate and functional traits related to species' climate stresses tolerance (testing the effect of wood density, leaf economic spectrum traits, leaf size and xylem vulnerability to embolism).

2 | MATERIALS AND METHODS

In this section we present: (a) the development of climate-dependent IPMs based on growth and survival models and the data used to fit them; (b) the development of species distribution models used to select climatic edges corresponding to a species distribution limits; (c) the derivation of metrics of demographic performance at the climatic edge versus the climatic centre of the species distribution from the IPMs and (d) the methodology to test our three hypotheses.

2.1 | Forest inventory

We used the European forest inventory (NFI) data compiled in the FunDivEUROPE project (Baeten et al., 2013; Ratcliffe et al., 2015, 2020). The data cover 91,528 plots and more than one million trees in Spain, France, Germany, Sweden and Finland. NFIs record information on individual trees in each plot, including species identity, diameter at breast height (dbh) and status (alive, dead, harvested or ingrowth). Plot design varies between countries but generally plots are circular with variable radii depending on tree size (see Appendix S1). The minimum dbh of trees included in the dataset was 10 cm and plots were remeasured over time allowing estimations of individual growth and survival. The time between two surveys varied from 4 to 16 years. Only the French NFI is based on a single measurement but provides a measurement of radial growth from cores (over 5 years) and an estimation of time since death. We selected species with >2,000 individuals and >500 plots, to ensure a good coverage of their range, growth and survival. We excluded exotic species for which the distribution is mainly controlled by plantation operations. For the demographic analyses, we also excluded all plots with records of harvesting operations or disturbances between the two surveys, which would otherwise influence our estimation of local competition.

2.2 | Climate variables

We used two bioclimatic variables known to control tree demography (Kunstler et al., 2011): (a) the sum of degree days above 5.5°C

(*sgdd*), and (b) the water availability index (*wai*). *sgdd* is the cumulative day-by-day sum of the number of degrees >5.5°C and is related to the mean annual temperature and the length of the growing season. It was extracted from E-OBS, a high resolution (1 km²) downscaled climate dataset (Moreno & Hasenauer, 2016) for the years between the two surveys plus 2 years before the first survey. In preliminary analyses, we also explored the number of frost days but it was too correlated with *sgdd* to be included in the models. *wai* was computed using precipitation (*P*, extracted from E-OBS) and potential evapotranspiration (*PET*) from the Climatic Research Unit (Harris et al., 2014) dataset, as (*P* − *PET*)/*PET* (see Ratcliffe et al., 2017) and is related to the water availability. We also explored other water stress indices but they did not improve the demographic models so we decided to use *wai*.

2.3 | Integral projection models

An IPM predicts the size distribution, $n(z', t + 1)$, of a population at time $t + 1$ from its size distribution at t, $n(z, t)$, where z the size at t and z' the size at $t + 1$, based on the following equation (Easterling et al., 2000; Ellner et al., 2016):

$$n\left(z', t + 1\right) = \int_{L}^{U} K\left(z', z\right) n\left(z, t\right) dz. \quad (1)$$

The kernel $K(z', z)$ can be split into the survival and growth kernel ($P(z', z)$) and the fecundity kernel ($F(z', z)$), as follows $K(z', z) = P(z', z) + F(z', z)$. $P(z', z)$ is defined as $P(z', z) = s(z)G(z', z)$ and represents the probability that an individual of size z survives between t and $t + 1$ and reaches the size z'. The size of the individual z can range between L and U, the lower and upper bounds of size, respectively. NFI data do not provide direct information on tree fecundity, thus our models describe the fate of a cohort (a cohort IPM for individuals with dbh ≥10 cm) by focusing only on $P(z', z)$. Even without covering the full life cycle, cohort IPMs are useful to estimate demographic performance because they allow to predict life trajectory metrics while accounting for size-dependent climate responses and compensatory effect between vital rates.

For each of the 27 species, we fitted growth and survival functions depending on tree size, the two climatic variables (*sggd* and *wai*) and local competition estimated as the sum of basal area of competitors (following Kunstler et al., 2011). The shape of the climatic response curves and the type of interaction between climate and tree size and climate and competition (which represents a size-dependent response) can have a large impact on vital rate predictions and IPM-derived life trajectory metrics. To account for uncertainties in climatic response curves shape and the interactions, we re-sampled 100 times 70% of the data to fit the growth and survival models and selected the best type of climatic response curves and interactions based on the Akaike information criteria (i.e. lowest AIC; Burnham & Anderson, 2002). Because there were fewer plots in extreme

climatic conditions, we re-sampled the data with a higher probability of sampling plots in extreme climatic conditions for the given species (see details in Appendix S1). Then we used the remaining 30% of the data to evaluate the goodness of fit of the growth and survival models. Goodness of fit and response curves of growth and survival models are presented in Appendix S1 (respectively normalised root-mean-square deviation—NRMSD, Figure S4 and area under the curve—AUC, Figure S5).

2.3.1 | Growth model

After preliminary exploration, we selected two alternative shapes of the climatic response curves: asymptotic or quadratic polynomial corresponding respectively to the Equation 2 and 3. These equations are flexible and allow for increasing, decreasing, bell or U-shape responses. These two equations allow to represent two alternative biological models: (a) either all species have their optimum at high water availability and sum of degree days; or (b) species have bell-shaped climate response curves with different optima along the climatic variables.

$$
\begin{aligned}
\log\left(G_{i,p}\right) &= a_{0,c} + a_{0,p} + a_1 D_i + a_2 \log\left(D_i\right) \\
&\quad + a_3 BA_i + a_4 \frac{1}{sgdd_p} + a_5 \frac{1}{wai_p} + \varepsilon_i,
\end{aligned}
\tag{2}
$$

$$
\begin{aligned}
\log\left(G_{i,p}\right) &= a_{0,c} + a_{0,p} + a_1 D_i + a_2 \log\left(D_i\right) + a_3 BA_i \\
&\quad + a_4 sgdd_p + a_5 sgdd_p^2 + a_6 wai_p + a_7 wai_p^2 + \varepsilon_i,
\end{aligned}
\tag{3}
$$

where $G_{i,p}$ is the annual diameter growth of tree i in plot p, D_i is the dbh of tree i, BA_i is the sum of basal area of local competitors of tree i per ha (sum basal area of both conspecific and heterospecific trees in the plot in a single local competition index), $sgdd_p$ is the sum of growing degree days, wai_p is the water aridity index, a_0 to a_7 are estimated parameters, and $a_{0,p}$ is a normal random plot effect accounting for unexplained variation at the plot level. The intercept $a_{0,c}$ is country specific to account for differences in sampling protocol between the NFIs (plot size and difference in mean survey time) and ε_i is the unexplained tree level variability following a normal distribution. We also tested models with interactions between the climatic variables—$1/sgdd_p$ and $1/wai_p$ for model (2) and $sgdd_p$ and wai_p for model (3)—and size (D_i and $\log(D_i)$) and the climatic variables and competition. We fitted the models in R-cran separately for each species (R Core Team, 2019) using the 'lmer' function ('LME4' package; Bates et al., 2015).

2.3.2 | Survival model

For survival, we also used the same two basic alternative models as for growth with different shapes of the climatic response curves (see Equations S1 and S2 in Appendix S1). The predictors and the interactions explored were the same as in the growth model. Survival models were fitted with a generalised linear model with a binomial

error. To account for variable survey times between plots we used the complementary log-log link with an offset representing the number of years between the two surveys (y_p; Morris et al., 2013). We fitted the model in R-cran using the 'glm' function. We did not include a random plot intercept because in most plots no individuals died between the surveys, making the estimation of the random plot effect challenging.

2.3.3 | Tree harvesting

Although we excluded plots with evidence of harvesting between the two surveys to fit the survival functions, most European forests are subject to management, which has a strong impact on population dynamics (Schelhaas et al., 2018). Preferential harvesting of dying or damaged trees before their death probably results in an underestimation of the natural mortality rate. To make sensible predictions with our IPMs, it was necessary to incorporate a harvesting rate to prevent an overestimation of tree life span. We set the individual tree harvesting rate, as the mean harvesting annual probability across all species and countries. The estimate was 0.5% per year. We did not model species, size and climate dependence of the harvesting rate, as we focused on climatic and not anthropogenic constraints on tree demography.

2.4 | Prediction of demographic metrics at the climatic edges and centre of species range

2.4.1 | Species distribution

To identify the climatic edge of a species range, a simple representation of its distribution in climate space is necessary. Across Europe, *sgdd* and *wai* are strongly negatively correlated, and the first axis of the PCA of *sgdd* and *wai* explains more than 84% of the variance (Figure S3, only the first axis was retained by Horn's parallel analysis). This allowed us to describe species ranges along a single climatic axis corresponding to the first axis (PC1) of this PCA. Species showed a clear segregation along this climatic axis in Europe (Figure 1). Based on the coordinates on PC1 of the plots where the species was present, we identified the median climate as their median value of PC1 (which we used as an index of species position along the climate gradient), the hot and dry edge (hereafter hot edge) and the cold and wet edge (hereafter cold edge), respectively, as their 5% and 95% quantiles. These quantiles represent two extreme climatic conditions experienced by the species. By focusing on climatically marginal populations, our approach differs from most tests of the ACH reviewed in Pironon et al. (2017) that studied populations at the periphery of the species geographic range.

To evaluate which species' edges corresponded to an actual limit in the species distribution and not just to limits in data coverage, we fitted species distribution models with BIOMOD2

FIGURE 1 Species distribution along the first axis of the PCA of the two climatic variables *sgdd* and *wai*. The median of the species distribution along this axis is represented by a black circle and the hot and dry edge and the cold and wet edge by the red and blue circles, respectively. Filled circles represent edges selected for the analysis, corresponding to edges where the species distribution models predicted at least a 10% drop in the probability of presence of the species (see Appendix S1 for details on the models)
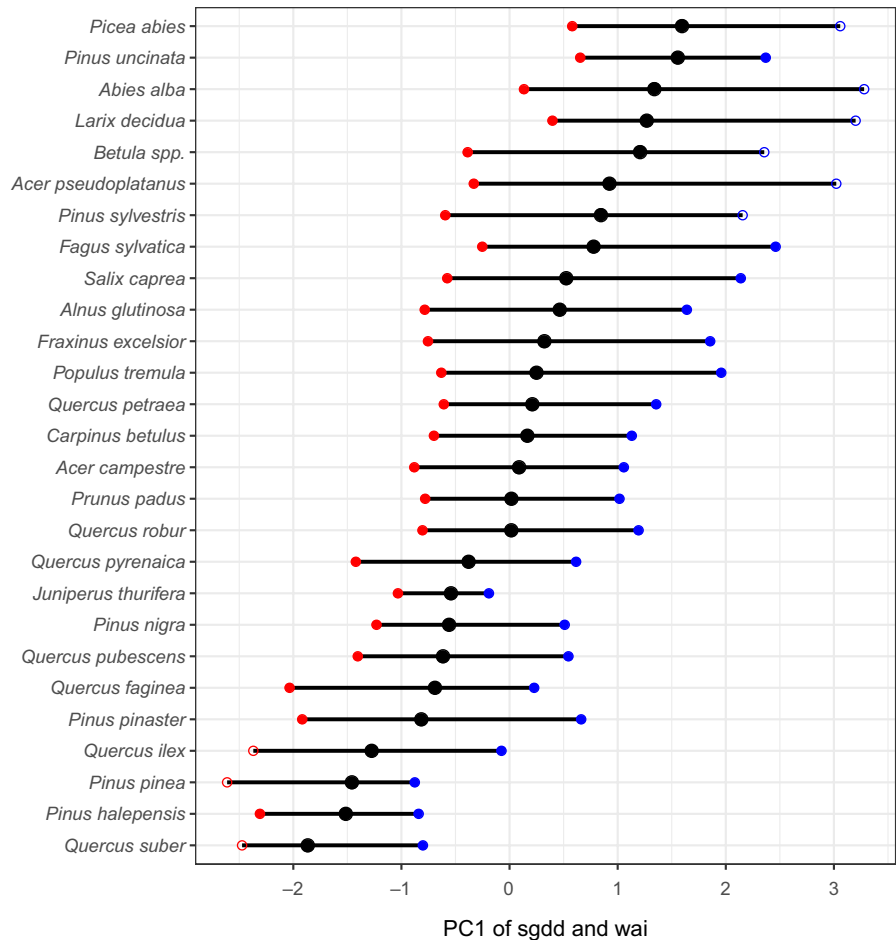
(Thuiller et al., 2009) using presence/absence data covering all Europe (Mauri et al., 2017; see Appendix S1). For comparison of the demographic performance at the edge versus the centre of the distribution, we retained only the edges where the SDM predicted at least a 10% drop in the probability of presence of the species (Figure 1).

2.4.2 | Demographic metrics

To evaluate how individual tree performance varied between the species' median climate and the climatic edges, we derived four metrics representing key dimensions of population performance. The first two metrics were related to individual vital rates, and were defined by the growth and survival of 15-cm dbh individuals. We decided to focus on a size corresponding to a small individual because small individuals have a large effect on population dynamics (Grubb, 1977; Figures S6 and S10 show the size response curves for growth and survival). The last two metrics were life trajectory metrics integrating the vital rates and size-dependent responses to climate in the IPM over the full size range of the species, and were defined by the mean life span of a 10-cm dbh individual and the passage time of a 10-cm dbh individual to 60 cm (corresponding to smaller upper size limit **U** across all species). The details of the numerical methods used

to compute life span and passage time from the IPM are provided in Appendix S1. Model diagnostics showed that our numerical approach was not sensitive to the number of bins used to discretised the size in the IPM when the number of bins was >800 (Figure S14). We thus retained a number of bins of 800.

We predicted the four demographic metrics at the centre and the hot and cold climatic edges of the species using their positions on the climatic axis. The median, and 5% and 95% quantiles on the PC1 correspond to the projection of a unique combination of *sggd* and *wai* for which we predicted the metrics. We integrated uncertainty into our estimates by deriving each demographic metric for all 100 re-sampled growth and survival models (see above). Because competitive interactions may also be important in controlling species demography at the edge of the range (Louthan et al., 2015), we made these predictions either without local competition (by setting *BA* to 0) or with a high level of local competition (by setting *BA* to 30 m^2/ha, corresponding to a closed forest).

2.4.3 | Analysis of the relative demographic performance at the climatic edges

For each demographic metric (*m*) we computed the relative difference in the metric at the edge (hot or cold) versus the centre as:

$$\Omega_{\text{edge}}^{m} = \left(m_{\text{edge}} - m_{\text{centre}} \right) / m_{\text{centre}}. \qquad (4)$$

We integrated uncertainty by deriving estimates of Ω_{edge}^{m} for each of the 100 re-sampled growth and survival models. Then we used Ω_{edge}^{m} to evaluate our three hypotheses.

Firstly, for each metric, we tested whether species demographic performance declined at the climatic edge compared to the climate centre (Hypothesis 1) by fitting a mixed model to test whether m was function of the range position type (edge vs. centre) using the function *lmer* in LME4. We included a random species effect to account for the non-independence of the 100 re-sampled estimates per species. We ran this analysis separately for hot and cold edges to see how demographic responses differed between them. Secondly, we tested whether the effects were different without or with competition (Hypothesis 2).

Thirdly, we explored whether Ω_{edge}^{m} was dependent on species median climate and functional traits related to species' climatic response (Hypothesis 3). We used Phylogenetic generalised least squares regression (PGLS; Symonds & Blomberg, 2014) using a phylogeny extracted from Zanne et al. (2014) to account for phylogenetic dependence between species. We accounted for the uncertainty in the demographic response by including a weight proportional to the inverse of the variance of Ω_{edge}^{m} (estimated over the 100 re-sampled growth and survival models). The PGLS regression with maximum likelihood estimation of Pagel's lambda (a measure of the phylogenetic signal ranging between 0 and 1) did not always converge (Pagel, 1999). In those cases we fitted a PGLS model with a Brownian model (Pagel's lambda set at 1). We retained only the regressions that were both significant (after a Bonferroni correction to account for multiple comparisons) and had a non-negligible magnitude of the effect (Camp et al., 2008). The magnitude of the effect was considered negligible when the confidence interval of the effect size intercepted the interval −0.10 and 0.10 (Camp et al., 2008). Effect sizes were computed as the standardised slope (Schielzeth, 2010).

To test the link between Ω_{edge}^{m} and species median climate, we ran the PGLS regression between Ω_{edge}^{m} and the species median position on PC1. To test the links between Ω_{edge}^{m} and functional traits, we ran the same type of PGLS regressions with four functional traits that are known to influence tree response to climate. We selected the following traits: (a) wood density, because of its links with drought and temperature response (Chave et al., 2009; Stahl et al., 2014); (b) the leaf economic spectrum (LES) because species at the conservative end of the spectrum are thought to be more tolerant to extreme climate (Reich, 2014); (c) leaf size, because of its links with water stress and frost response (Wright et al., 2017) and (d) xylem vulnerability to embolism measured by the water potential leading to 50% loss of xylem conductivity, Ψ_{50}, because of its link with drought-induced mortality (Anderegg et al., 2016). LES is based on the covariance of specific leaf area, leaf life span and leaf nitrogen content per mass (Wright et al., 2004). We used leaf nitrogen content per mass (N_{mass}), as it was the LES trait with the best coverage across our species. Trait data were sourced from open databases (Chave et al., 2009; Choat et al., 2012; Maire et al., 2015; Wright et al., 2004, 2017).

3 | RESULTS

3.1 | Growth and survival size-dependent responses to climate

For most species the growth and survival models showed evidence of interactions between climate and tree size and for a smaller subset of species also between climate and competition (see Tables S2 and S3). This indicates that size-dependent climatic responses were common. Model selection over the 100 re-sampled data showed that for 23 of the 27 species the most frequently selected growth model included interactions between climate variables and tree size (see Table S2). Selection of the best survival model was more variable between the 100 data re-sampling than for the growth models. For 17 of the 27 species the most frequently selected survival models included interactions between climatic variables and tree size (see Table S3). For both growth and survival several species also showed evidence of interactions between climate variables and competition (respectively 12 and 11 species out of 27, see Tables S2 and S3).

3.2 | Demographic responses differ between edge types and metrics

Across the 27 species, we found evidence of a significant decrease in growth of 15-cm dbh individuals and increase in passage time (longer time needed to grow from 10 to 60 cm) at the cold edge in comparison with the median climate but no effect at the hot edge (Figure 2). In contrast, at the hot edge, we found evidence of a significant decrease in both tree survival and life span (Figure 2). This result is consistent with the hypothesis that at least one metric will decline in performance at the edge, and that different metrics are affected depending on the edge type. In contrast, we found that life span was significantly longer at the cold edge than at the median climate (Figure 2). Generally, these patterns were unaffected by local competition (Figure 3). It is, however, important to note that the relative decrease in survival at the hot edge and the increase of passage time at the cold edge became non-significant at high levels of competition (Figure 3).

Despite the overall demographic response at the edge, there were large variations between species. For each metric and edge type we found species showing a decrease and species showing an increase in performance (Figures S16–S19).

3.3 | Demographic responses vary with species median climate

Growth response at the hot and cold edges was related to the median climate of the species; species associated with hot climates were more

FIGURE 2 Differences in the demographic metrics at climatic edge versus the median climate of the species distribution $(\Omega_{edge}^{m} = (m_{edge} - m_{centre})/m_{centre})$. The box plots represent the relative difference of the demographic metrics between the climatic edge and the median climate computed over the 100 data resampling and the 27 species for the four demographic metrics (annual diameter growth and survival for an individual 15 cm in diameter, passage time from 10 cm in diameter to 60 cm in diameter and life span of tree 15 cm in diameter). The two edge types are the hot and dry edge in red (short name 'hot edge') and the cold and wet edge in blue (short name 'cold edge'). The p-value of the test for the difference in each demographic metric between the edge and the median climate is presented at the top of the box plot (ns: nonsignificant, *p-value < 0.05, **p-value < 0.01, ***p-value < 0.001). The p-value was computed with a mixed model with species as a random effect (see Section 2 for details)
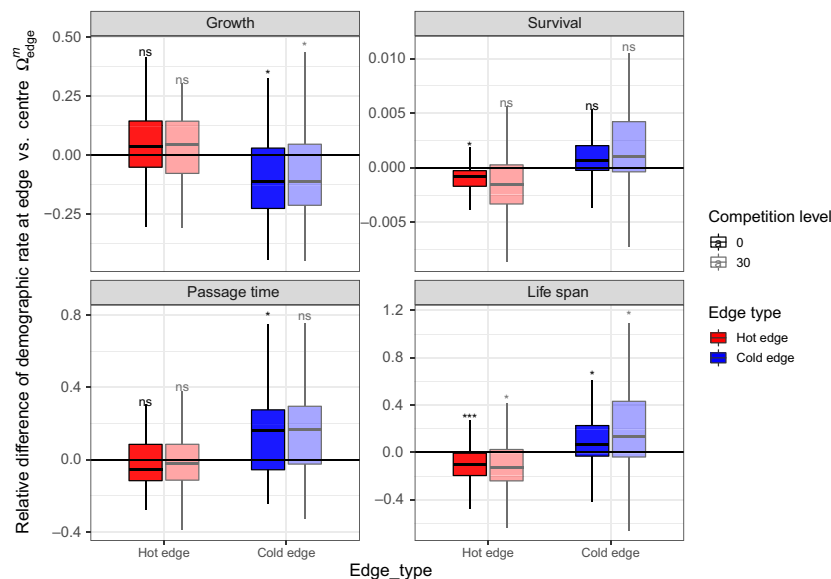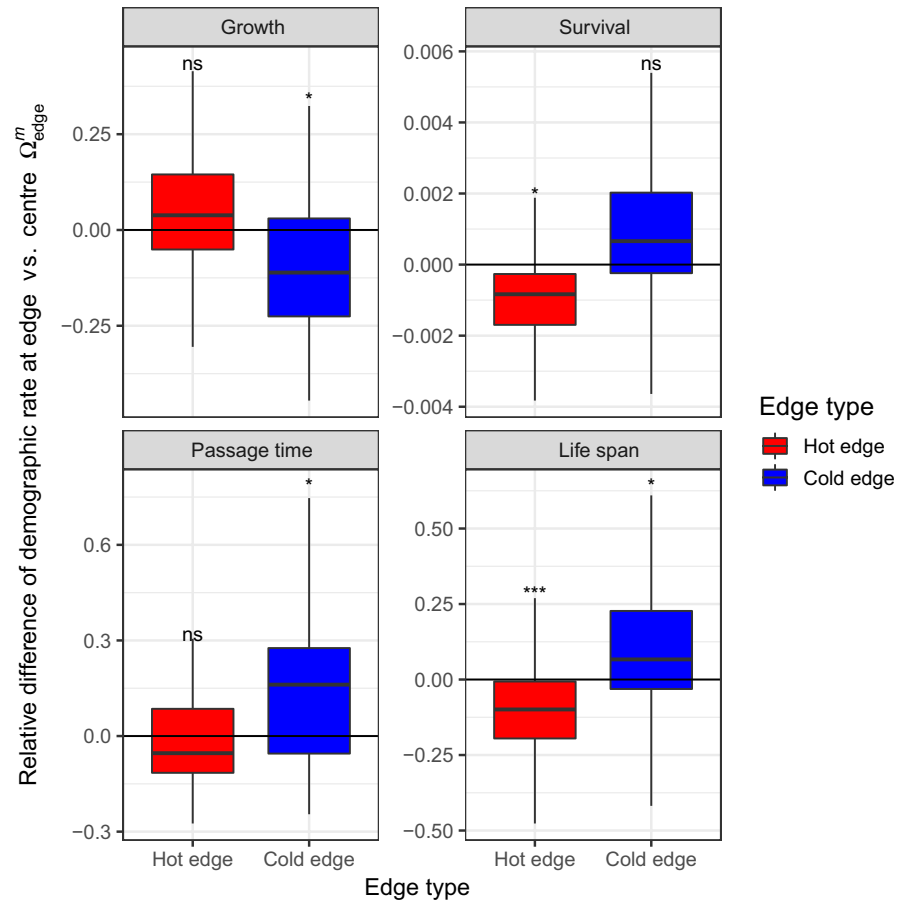

FIGURE 3 Differences in the demographic metrics at climatic edge versus the median climate of the species distribution $(\Omega_{edge}^{m} = (m_{edge} - m_{centre})/m_{centre})$ without and with a high level of competition. The box plots represent the relative difference the demographic metrics between the climatic edge and the median climate over the 100 data resampling and the 27 species for the four demographic metrics (annual diameter growth and survival for an individual 15 cm in diameter, passage time from 10 cm in diameter to 60 cm in diameter and life span of tree 15 cm in diameter), and the two levels of competition (without competition: basal area of competitors, $BA = 0$, no transparency, with a high level of competition: basal area of competitors, $BA = 30$ m²/ha colour transparency). The two edge types are the hot and dry edge in red (short name 'hot edge') and the cold and wet edge in blue (short name 'cold edge'). The p-value of the test for the difference in each demographic metric between the edge and the median climate is presented at the top of the box plot (ns: nonsignificant, *p-value < 0.05, **p-value < 0.01, ***p-value < 0.001). The p-value was computed with a mixed model with species as a random effect (see Section 2 for details)
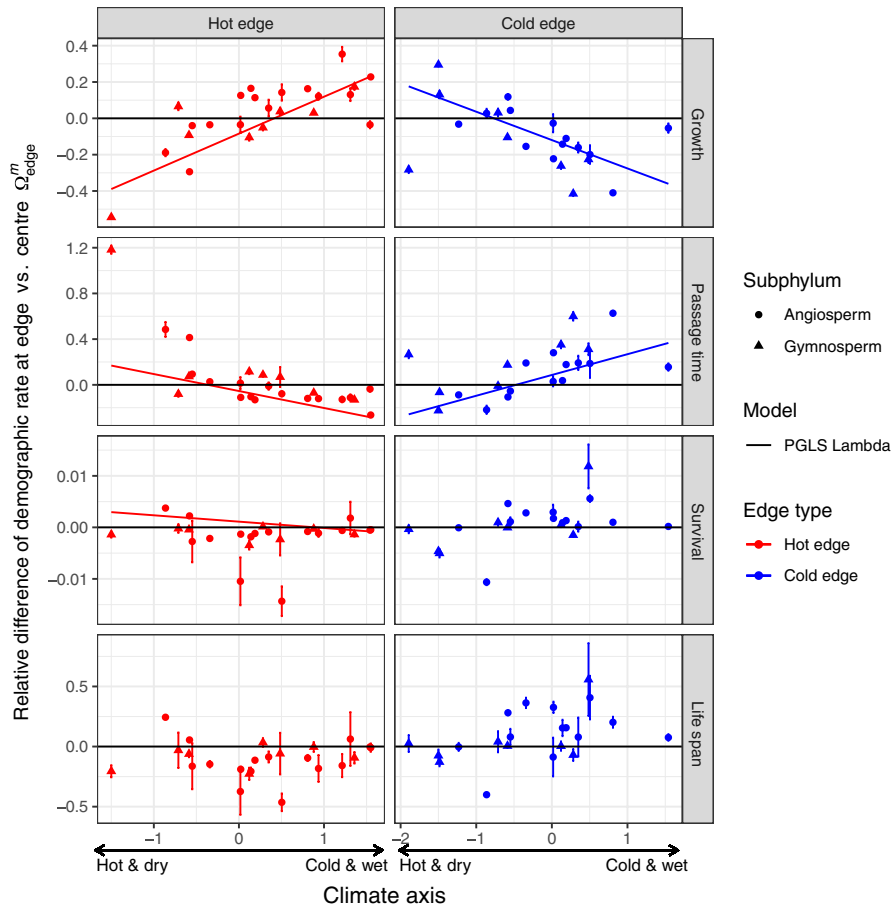
FIGURE 4 Changes in demographic responses at the edge – Ω^m_{edge} – in function of species median position on the climatic axis. Species demographic response at the edge ($\Omega^m_{edge} = (m_{edge} - m_{centre})/m_{centre}$) as a function of the median position of the species on the first axis of the climate PCA. Negative values on the climatic axis correspond to a hot and dry climate and positive values to a cold and wet climate. For each species the mean (point) and the 95% quantiles (error bar) of the demographic response over the 100 data resampling is represented for the hot and dry edge in red (short name 'hot edge') and the cold and wet edge in blue (short name 'cold edge'). Phylogenetic generalised least squares (PGLS Lambda) regressions are represented only for significant relationship with a non-negligible magnitude of the effect. Gymnosperm and angiosperm species are represented with different symbols
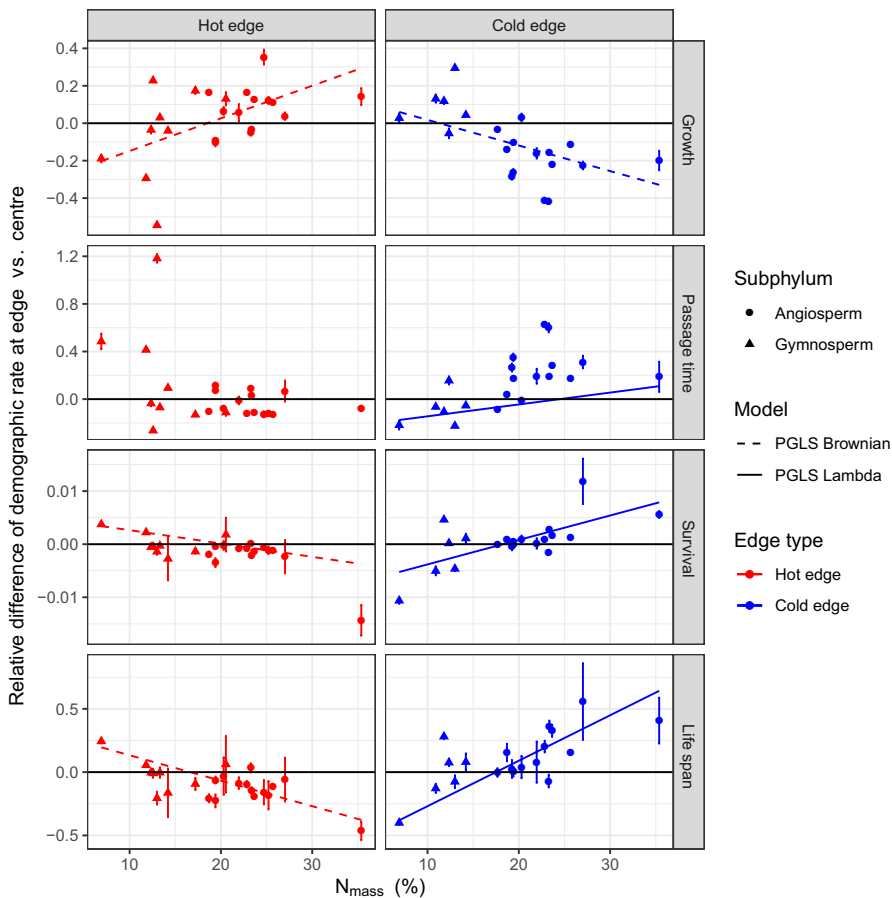

FIGURE 5 Changes in demographic responses at the edge – Ω^m_{edge} – in function of species leaf nitrogen content per mass. Species demographic response at the edge ($\Omega^m_{edge} = (m_{edge} - m_{centre})/m_{centre}$) as a function of species leaf nitrogen content per mass. For each species, the mean (point) and 95% quantiles (error bar) of the demographic response over the 100 data resampling is represented for both the hot and dry edge in red (short name 'hot edge') and the cold and wet edge in blue (short name 'cold edge'). Phylogenetic generalised least squares (PGLS) regressions are represented only for significant relationship with a non-negligible magnitude of the effect (see details in caption of Figure 4)

constrained at their hot edge while species associated with cold climates were more constrained at their cold edge. This result is depicted in Figure 4 by a positive relationship between the median climate of the species and $\Omega_{\text{hot edge}}^{\text{growth}}$ and a negative relationship with $\Omega_{\text{cold edge}}^{\text{growth}}$. The same pattern is visible for passage time, but in the opposite direction, because passage time is longer when growth is slower (Figure 4). The responses of Ω_{edge}^{m} for survival and life span were much weaker or null. We found a negative relationship for $\Omega_{\text{hot edge}}^{\text{survival}}$, which was largely related to a few extreme species, and no effect for life span (Figure 4).

3.4 | Weak links between demographic response and species traits

Regardless of the demographic metric m, N_{mass} had the strongest relationship with Ω_{edge}^{m} of all the traits we tested. At the hot edge, species with high N_{mass} experienced a stronger decrease in their survival and life span than species with a low N_{mass} (Figure 5). In contrast, at the cold edge, species with low N_{mass} experienced a stronger decrease in their survival and life span than species with high N_{mass} (Figure 5). In addition, species with high N_{mass} had less limitation of their growth at the hot edge than species with low N_{mass} (Figure 5). In contrast, species with high N_{mass} had stronger limitation of their growth at the cold edge (Figure 5).

Relationships between Ω_{edge}^{m} and wood density, leaf size and xylem vulnerability to embolism (Ψ_{50}) were generally weak (Figures S21–S23). Most of these relationships were driven by only a few species (Figures S21–S23). Species with small leaf area had better survival and life span at the hot edge and better passage time at the cold edge than large leafed species (Figure S23). Species with high Ψ_{50} experienced a stronger decrease in their growth at the hot edge than species with low Ψ_{50} (Figure S22).

4 | DISCUSSION

Our analysis based on pan-European forest inventory data and integral projection models of 27 tree species, found weak support for the ACH prediction that demographic performance is lower at the climatic edge than at the centre of the species range. Instead, decline in demographic performance was strikingly different between the cold and the hot edges. At cold and wet edges, growth and passage time were constrained, whereas at hot and dry edges, survival and life span were constrained. Beyond these general patterns, we found important variability between species in their demographic performance at the edge, which was partially explained by species' median climate and traits.

4.1 | Different demographic responses at the hot and the cold edge

We found mixed support for the ACH; not all the demographic metrics were limited at the two edges and patterns were variable between species. This is consistent with observational studies that

found limited evidence of a relationship between species demography and their distribution. For instance, both Thuiller et al. (2014) and Csergo et al. (2017) found limited correlation between plants demographic performance and probability of presence. In addition, Purves (2009) reported mixed evidence of a decrease in demographic performance at the south and north edges of North American tree species.

Growth and passage time were constrained at the cold edge in comparison with the centre of the species climatic range. This is consistent with studies on North American tree species that found a decrease in growth at the cold edge in adult trees (Purves, 2009) and juveniles (Ettinger & HilleRisLambers, 2017; Putnam & Reich, 2017). In contrast with the ACH, we found a tendency for a slightly faster growth at the hot edge than at the centre, which has also been reported in North American trees (Ettinger & HilleRisLambers, 2017; Purves, 2009; Putnam & Reich, 2017). Interestingly, studies on *Fagus sylvatica* radial growth in Europe found a higher drought resistance at the hot edge than at the core of the range (Cavin & Jump, 2017).

At the hot and dry edge, tree survival and life span were lower than at the centre of the climatic range. The same decrease in survival at the hot edge was also found by Archambeau et al. (2020) for *F. sylvatica* and *Pinus sylvestris* in Europe. In contrast, Purves (2009) found no such decrease in survival at the hot edge of eastern North American species. This difference could be explained by the fact that the hot edge of most European species corresponds to both a hot and a dry climate, whereas in eastern America the hot edge is less constrained by drought (Zhu et al., 2014). We found that life span was longer at the cold edge than at the centre of the distribution, which contradicts the classical view that survival is constrained in cold climates and the results of Purves (2009). Given that tree diameter growth is constrained at the cold edge, this longer life span could be explained by a trade-off between tree growth rate and tree longevity (see Black et al., 2008; Di Filippo et al., 2015) and the observation that survival rate correlates negatively with site productivity (Stephenson et al., 2011).

We found strong evidence of size dependence of growth and survival responses to climatic constraints. Our results agree with previous studies which found that tree growth or survival responses to climate varied with ontogeny (Canham & Murphy, 2017; Trouillier et al., 2019). For instance, Canham and Murphy (2017) found a displacement of the climatic optimum of growth and survival between seedlings, saplings and canopy trees. These size-dependent climatic responses, however, did not strongly influence the life trajectory metrics derived with IPMs as the response of life span at the edge was closely connected to the survival of a 15-cm dbh tree and the passage time was closely related to the growth of a 15-cm dbh tree. This means that these size-dependent responses were either of small magnitude or led to few compensation effects between size classes. Tredennick et al. (2018) also found that the size dependence of vital rate responses to exogenous environmental fluctuations had limited effect on the population growth rate of perennial plant species.

4.2 | Lack of competition effect

Numerous studies have proposed that competitive interactions could be crucial in setting demographic limits, particularly when site productivity is high (see Alexander et al., 2016; Ettinger & HilleRisLambers, 2017; Hargreaves et al., 2014; HilleRisLambers et al., 2013; Louthan et al., 2015). In our analyses, we explored the effect of competition by comparing the relative demographic performance at the edge in comparison with the centre (Ω) without local competition or with a high level of local competition. Despite the strong direct effects of competition on both growth and survival and interactions between competition and climate (see the variables importance reported in Tables S4 and S5), the relative demographic responses at the edges versus the centre (measured by Ω) were not strongly influenced by the degree of local competition. Competition is thus a strong determinant of demographic rates but its effect is not stronger at the climatic edge than at the climatic centre (rejecting Hypothesis 2). Rather competition blurs demographic constraints at the edge. Indeed, limitations of survival at the hot edge and passage time at the cold edge were significant without competition but not with a high level of competition.

Three main reasons could explain the lack of competition effect on the demographic response at the edges in our study. Firstly, properly estimating competition effect with observational data is notoriously difficult (Tuck et al., 2018). Secondly, we did not differentiate between intra- and inter-specific competition, whereas inter-specific competition might have the strongest impact at the edge (Alexander et al., 2016). Thirdly, as our cohort IPMs do not cover the full life cycle it was not possible to evaluate whether competitive exclusion—the final effect of competition (Chesson, 2018)—occurs at the edge.

4.3 | Strong effect of species median climate on growth response at the edge

We found that the hotter the centre of the species range, the greater were the constraints on growth and passage time at its hot edge. The same pattern was found with the cold edge and the species median climate proximity to cold extreme. This is in agreement with the general observation that, in Europe, vegetation productivity in Europe is at its maximum in temperate climates where both drought and cold stress are limited (Jung et al., 2007).

4.4 | Weak trait effect on species demographic response at the edge

Part of the variation in the demographic response at the edge between species was related to N_{mass}, a key dimension of the leaf economic spectrum. An important difficulty in the interpretation of these results is that our understanding of the link between leaf economic traits and climate is limited. Multiple mechanisms, some of them contradictory, have been proposed to explain the link between leaf nitrogen content and climate. For instance, it is generally considered that species with low N_{mass} have a more conservative strategy of resource use and perform better in stressful conditions than species with high N_{mass} (Reich, 2014). In agreement with this finding, we found that species with low N_{mass} had a better survival and life span at the hot edge. In contrast, high leaf nitrogen content has been linked with photosynthesis tolerance to drought and low temperatures because of higher enzyme activities (Reich & Oleksyn, 2004; Wright et al., 2003). Consistent with this mechanism, we found that species with high N_{mass} had a higher growth rates at the hot edge and better survival and life span at the cold edge.

We found limited relationships between wood density, leaf size or xylem vulnerability to embolism and demographic responses at the climatic edge, which was surprising as the mechanisms related to climate response are better understood for these traits. Smaller leaves were related to a longer life span and a better survival at the hot edge and a better passage time at the cold edge. This in agreement with Wright et al. (2017) who proposed that large leaves are disadvantaged in hot and dry climates because their transpiration rate during the day is too high and are disadvantaged in cold climate because they have greater risks of reaching critical low temperatures during the night. Anderegg et al. (2019) also reported weak links between traits and drought-related mortality at the edge, with only an effect for xylem vulnerability to embolism. The effect was, however, that drought-adapted species experienced higher drought mortality at the edge (Anderegg et al., 2019). In this study, we found no link between xylem vulnerability to embolism and survival response at the edge. In contrast, a low xylem vulnerability to embolism (drought-adapted species) was related to better growth at the hot edge (Figure S22).

Finally, our traits analysis might underestimate the role of traits because we ignore intraspecific traits variability. Traits phenotypic plasticity and local adaptation might however be large for species with a broad distribution (see for instance results for *P. sylvestris* in Reich et al., 1996).

4.5 | On the challenge of connecting population dynamics and species ranges

A key limitation of our analysis is that it did not include the regeneration phase, which is typically considered a bottleneck in tree population dynamics and is key to cover the full life cycle to estimate population growth rate (Grubb, 1977). Because of this limitation, we could not estimate the population growth rate (λ), but just vital rates and life trajectory metrics based only on adult growth and survival. In Appendix S1, we provide an evaluation of the relative importance of the regeneration phase for tree population growth rate with an elasticity analysis of matrix population models extracted from the COMPADRE Plant matrix database

(Salguero-Gómez et al., 2015). The elasticity analysis showed that the regeneration and adult phases were equally important (see Figure S25). Our IPMs analysis thus captures an important part of a tree's life cycle for the population growth rate. However, we cannot rule out the possibility that the regeneration phase has a disproportional importance for the dynamics at the edge, as several studies have shown that this phase is extremely sensitive to climate (Canham & Murphy, 2016; Clark et al., 2014; Defossez et al., 2016). Integrating fecundity and juvenile life stages in tree IPMs is challenging because we have much less data on them (Needham et al., 2018; Ruiz-Benito et al., 2020; but see Lines et al., 2019).

It is also important to keep in mind that species ranges are not necessarily only related to the mean population growth rate but could also be related to other processes controlling extinction risk. For instance, the temporal variability of population growth rate and the population resilience to disturbances could be crucial at the edge (Holt et al., 2005) but it was not possible to evaluate these processes in our study with the NFI data. Another explanation is that suitable habitats where population growth rates are unaffected might exist up to the edge due to the presence of suitable microsites (Cavin & Jump, 2017). In this case, the species edges arise because the fraction of suitable habitats available to the metapopulation decreases (Holt & Keitt, 2000).

Finally, tree species distributions might not be in equilibrium with the current climate. This could be because species are either still in the process of recolonising from their ice age refugia (Svenning & Skov, 2004) or already affected by climate change. Such disequilibrium should however be visible by better performance at the cold edge (Talluto et al., 2017) and we found no evidence for this in our results.

4.6 | Synthesis

Our study shows that trees' demographic responses at range edges are more complex than predicted by the ACH. Here, the patterns of demographic response of the 27 European tree species differed between their hot and cold edges. We only found strong evidence of demographic limits for edges occurring in extreme conditions (hot edges of hot-distributed species and cold edges of cold-distributed species). Our findings open an important perspective, as they show that one should not expect the same demographic response at the hot versus the cold edge and that we need to refine predictions of climate change impacts as a function of the edge and species characteristics.

AUTHORS' CONTRIBUTIONS

G.K. conceived the ideas and designed methodology with the help of S.R., N.R., R.S.-G. and D.Z.C.; S.R. formatted the forest inventory and climatic data with the help of P.R.-B., M.A.Z., G.K., A.L. and C.W.; G.K. and A.G. analysed the data and developed the IPM models with the help of R.S.-G., D.Z.C. and N.R.; R.S.-G. analysed the COMPADRE matrix population models; W.T. helped with SDM models development; G.K. led the writing of the manuscript with important inputs from A.G., S.R., N.R. and R.S.-G. All authors contributed critically to the drafts and gave final approval for publication.

PEER REVIEW

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DATA AVAILABILITY STATEMENT

Data are available from Spanish National Forest Inventory (https://www.miteco.gob.es/en/biodiversidad/servicios/banco-datos-naturaleza/informacion-disponible/ifn3_bbdd_descargas.htm.aspx and https://www.miteco.gob.es/en/biodiversidad/servicios/banco-datos-naturaleza/informacion-disponible/ifn2_descargas.aspx), French National Forest Inventory (https://inventaire-forestier.ign.fr/spip.php?article532) and German National Forest Inventory (https://bwi.info/Download/de/BWI-Basisdaten/ACCESS2003/). Further details about these datasets can be found in Appendix S1. The Swedish and Finnish data are not public on the NFI websites, but data with blurred geographic coordinates and the climatic data used in the analysis are available from the Dryad Digital Repository https://doi.org/10.5061/dryad.wm37pvmkw (Ratcliffe et al., 2020).

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REFERENCES

Alexander, J. M., Diez, J. M., Hart, S. P., & Levine, J. M. (2016). When climate reshuffles competitors: A call for experimental macroecology. *Trends in Ecology & Evolution*, *31*, 831–841. https://doi.org/10.1016/j.tree.2016.08.003

Anderegg, W. R. L., Anderegg, L. D. L., Kerr, K. L., & Trugman, A. T. (2019). Widespread drought-induced tree mortality at dry range edges indicates that climate stress exceeds species' compensating mechanisms. *Global Change Biology*, *25*(11), 3793–3802. https://doi.org/10.1111/gcb.14771

Anderegg, W. R., Klein, T., Bartlett, M., Sack, L., Pellegrini, A. F., Choat, B., & Jansen, S. (2016). Meta-analysis reveals that hydraulic traits explain cross-species patterns of drought-induced tree mortality across the globe. *Proceedings of the National Academy of Sciences of the United States of America*, *113*(18), 5024–5029. https://doi.org/10.1073/pnas.1525678113

Archambeau, J., Ruiz-Benito, P., Ratcliffe, S., Fréjaville, T., Changenet, A., Muñoz Castañeda, J. M., Lehtonen, A., Dahlgren, J., Zavala, M. A., & Benito Garzón, M. (2020). Similar patterns of background mortality across Europe are mostly driven by drought in European beech and a combination of drought and competition in Scots pine. *Agricultural and Forest Meteorology*, *280*, 107772. https://doi.org/10.1016/j.agrformet.2019.107772

Baeten, L., Verheyen, K., Wirth, C., Bruelheide, H., Bussotti, F., Finér, L., Jaroszewicz, B., Selvi, F., Valladares, F., Allan, E., Ampoorter, E., Auge, H., Avăcăriei, D., Barbaro, L., Bărnoaiea, I., Bastias, C. C., Bauhus, J., Beinhoff, C., Benavides, R., ... Scherer-Lorenzen, M. (2013). A novel comparative research platform designed to determine the functional significance of tree species diversity in European forests. *Perspectives in Plant Ecology, Evolution and Systematics*, *15*(5), 281–291. https://doi.org/10.1016/j.ppees.2013.07.002

Bates, D., Mächler, M., Bolker, B., & Walker, S. (2015). Fitting linear mixed-effects models using lme4. *Journal of Statistical Software*, *67*(1), 1–48. https://doi.org/10.18637/jss.v067.i01

Black, B. A., Colbert, J. J., & Pederson, N. (2008). Relationships between radial growth rates and lifespan within North American tree species. *Écoscience*, *15*(3), 349–357. https://doi.org/10.2980/15-3-3149

Brown, J. H. (1984). On the relationship between abundance and distribution of species. *The American Naturalist*, *124*, 255–279. https://doi.org/10.1086/284267

Burnham, K. P., & Anderson, D. R. (2002). *Model selection and multimodel inference: A practical information-theoretic approach*. Springer-Verlag.

Cahill, A. E., Aiello-Lammens, M. E., Caitlin Fisher-Reid, M., Hua, X., Karanewsky, C. J., Ryu, H. Y., Sbeglia, G. C., Spagnolo, F., Waldron, J. B., & Wiens, J. J. (2014). Causes of warm-edge range limits: Systematic review, proximate factors and implications for climate change. *Journal of Biogeography*, *41*(3), 429–442. https://doi.org/10.1111/jbi.12231

Camp, R. J., Seavy, N. E., Gorresen, P. M., & Reynolds, M. H. (2008). A statistical test to show negligible trend: Comment. *Ecology*, *89*(5), 1469–1472. https://doi.org/10.1890/07-0462.1

Canham, C. D., & Murphy, L. (2016). The demography of tree species response to climate: Seedling recruitment and survival. *Ecosphere*, *7*(8), e01424. https://doi.org/10.1002/ecs2.1424

Canham, C. D., & Murphy, L. (2017). The demography of tree species response to climate: Sapling and canopy tree survival. *Ecosphere*, *8*(2), e01701. https://doi.org/10.1002/ecs2.1701

Case, T. J., Holt, R. D., McPeek, M. A., & Keitt, T. H. (2005). The community context of species' borders: Ecological and evolutionary perspectives. *Oikos*, *108*(1), 28–46. https://doi.org/10.1111/j.0030-1299.2005.13148.x

Cavin, L., & Jump, A. S. (2017). Highest drought sensitivity and lowest resistance to growth suppression are found in the range core of the tree *Fagus Sylvatica* L. Not the equatorial range edge. *Global Change Biology*, *23*(1), 362–379. https://doi.org/10.1111/gcb.13366

Chave, J., Coomes, D. A., Jansen, S., Lewis, S. L., Swenson, N. G., & Zanne, A. E. (2009). Towards a worldwide wood economics spectrum. *Ecology Letters*, *12*(4), 351–366. https://doi.org/10.1111/j.1461-0248.2009.01285.x

Chesson, P. (2018). Updates on mechanisms of maintenance of species diversity. *Journal of Ecology*, *106*(5), 1773–1794. https://doi.org/10.1111/1365-2745.13035

Choat, B., Jansen, S., Brodribb, T. J., Cochard, H., Delzon, S., Bhaskar, R., Bucci, S. J., Field, T. S., Gleason, S. M., Hacke, U. G., Jacobsen, A. L., Lens, F., Maherali, H., Martínez-Vilalta, J., Mayr, S., Mencuccini, M., Mitchell, P. J., Nardini, A., Pittermann, J., ... Zanne, A. E. (2012). Global convergence in the vulnerability of forests to drought. *Nature*, *491*(7426), 752–755. https://doi.org/10.1038/nature11688

Clark, J. S., Bell, D. M., Kwit, M. C., & Zhu, K. (2014). Competition-interaction landscapes for the joint response of forests to climate change. *Global Change Biology*, *20*(6), 1979–1991. https://doi.org/10.1111/gcb.12425

Cochran, M. E., & Ellner, S. (1992). Simple methods for calculating age-based life history parameters for stage-structured populations. *Ecological Monographs*, *62*(3), 345–364. https://doi.org/10.2307/2937115

Csergő, A. M., Salguero-Gómez, R., Broennimann, O., Coutts, S. R., Guisan, A., Angert, A. L., Welk, E., Stott, I., Enquist, B. J., McGill, B., Svenning, J.-C., Violle, C., & Buckley, Y. M. (2017). Less favourable climates constrain demographic strategies in plants. *Ecology Letters*, *20*, 969–980. https://doi.org/10.1111/ele.12794

De Roos, A. M., Persson, L., & McCauley, E. (2003). The influence of size-dependent life-history traits on the structure and dynamics of populations and communities. *Ecology Letters*, *6*(5), 473–487. https://doi.org/10.1046/j.1461-0248.2003.00458.x

Defossez, E., Courbaud, B., Lasbouygues, O., Schiffers, K., & Kunstler, G. (2016). Are variations of direct and indirect plant interactions along a climatic gradient dependent on species' strategies? An experiment on tree seedlings. *Oikos*, *125*(5), 708–717. https://doi.org/10.1111/oik.02331

Di Filippo, A., Pederson, N., Baliva, M., Brunetti, M., Dinella, A., Kitamura, K., Knapp, H. D., Schirone, B., & Piovesan, G. (2015). The longevity of broadleaf deciduous trees in Northern Hemisphere temperate forests: Insights from tree-ring series. *Frontiers in Ecology and Evolution*, *3*, 46. https://doi.org/10.3389/fevo.2015.00046

Easterling, M. R., Ellner, S. P., & Dixon, P. M. (2000). Size-specific sensitivity: Applying a new structured population model. *Ecology*, *81*(3), 694–708. https://doi.org/10.1890/0012-9658(2000)081[0694:SSAAN]2.0.CO;2

Ellner, S. P., Childs, D. Z., & Rees, M. (2016). *Data-driven modelling of structured populations*. Springer International Publishing.

Ettinger, A., & HilleRisLambers, J. (2017). Competition and facilitation may lead to asymmetric range shift dynamics with climate change. *Global Change Biology*, *23*(9), 3921–3933. https://doi.org/10.1111/gcb.13649

Gaston, K. J. (2009). Geographic range limits: Achieving synthesis. *Proceedings of the Royal Society B: Biological Sciences*, *276*(1661), 1395–1406. https://doi.org/10.1098/rspb.2008.1480

Grubb, P. J. (1977). The maintenance of species-richness in plant communities: The importance of the regeneration niche. *Biological Review*, *52*, 107–145. https://doi.org/10.1111/j.1469-185X.1977.tb01347.x

Hargreaves, A. L., Samis, K. E., & Eckert, C. G. (2014). Are species' range limits simply niche limits writ large? A review of transplant experiments beyond the range. *The American Naturalist*, *183*(2), 157–173. https://doi.org/10.1086/674525

Harris, I., Jones, P., Osborn, T., & Lister, D. (2014). Updated high-resolution grids of monthly climatic observations – The CRU TS3.10 Dataset: Updated high-resolution grids of monthly climatic observations. *International Journal of Climatology*, *34*(3), 623–642. https://doi.org/10.1002/joc.3711

HilleRisLambers, J., Harsch, M. A., Ettinger, A. K., Ford, K. R., & Theobald, E. J. (2013). How will biotic interactions influence climate change-induced range shifts?: Biotic interactions and range shifts. *Annals of the New York Academy of Sciences*, *1297*, 112–125. https://doi.org/10.1111/nyas.12182

Holt, R. D., & Keitt, T. H. (2000). Alternative causes for range limits: A metapopulation perspective. *Ecology Letters*, *3*(1), 41–47. https://doi.org/10.1046/j.1461-0248.2000.00116.x

Holt, R. D., & Keitt, T. H. (2005). Species' borders: A unifying theme in ecology. *Oikos*, *108*(1), 3–6. https://doi.org/10.1111/j.0030-1299.2005.13145.x

Holt, R. D., Keitt, T. H., Lewis, M. A., Maurer, B. A., & Taper, M. L. (2005). Theoretical models of species' borders: Single species approaches. *Oikos*, *108*(1), 18–27. https://doi.org/10.1111/j.0030-1299.2005.13147.x

Jung, M., Vetter, M., Herold, M., Churkina, G., Reichstein, M., Zaehle, S., Ciais, P., Viovy, N., Bondeau, A., Chen, Y., Trusilova, K., Feser, F., & Heimann, M. (2007). Uncertainties of modeling gross primary productivity over Europe: A systematic study on the effects of using different drivers and terrestrial biosphere models: Effects on simulated GPP. *Global Biogeochemical Cycles*, *21*(4), GB4021. https://doi.org/10.1029/2006GB002915

Kunstler, G., Albert, C. H., Courbaud, B., Lavergne, S., Thuiller, W., Vieilledent, G., Zimmermann, N. E., & Coomes, D. A. (2011). Effects of competition on tree radial-growth vary in importance but not in intensity along climatic gradients. *Journal of Ecology*, *99*, 300–312. https://doi.org/10.1111/j.1365-2745.2010.01751.x

Lee-Yaw, J. A., Kharouba, H. M., Bontrager, M., Mahony, C., Csergő, A. M., Noreen, A. M. E., Li, Q., Schuster, R., & Angert, A. L. (2016). A synthesis of transplant experiments and ecological niche models suggests that range limits are often niche limits. *Ecology Letters*, *19*(6), 710–722. https://doi.org/10.1111/ele.12604

Lines, E. R., Zavala, M. A., Ruiz-Benito, P., & Coomes, D. A. (2019). Capturing juvenile tree dynamics from count data using Approximate Bayesian Computation. *Ecography*, *43*, 406–418. https://doi.org/10.1111/ecog.04824

Louthan, A. M., Doak, D. F., & Angert, A. L. (2015). Where and when do species interactions set range limits? *Trends in Ecology & Evolution*, *30*(12), 780–792. https://doi.org/10.1016/j.tree.2015.09.011

Maire, V., Wright, I. J., Prentice, I. C., Batjes, N. H., Bhaskar, R., van Bodegom, P. M., Cornwell, W. K., Ellsworth, D., Niinemets, Ü., Ordonez, A., Reich, P. B., & Santiago, L. S. (2015). Global effects of soil and climate on leaf photosynthetic traits and rates: Effects of soil and climate on photosynthetic traits. *Global Ecology and Biogeography*, *24*(6), 706–717. https://doi.org/10.1111/geb.12296

Mauri, A., Strona, G., & San-Miguel-Ayanz, J. (2017). EU-Forest, a high-resolution tree occurrence dataset for Europe. *Scientific Data*, *4*, 160123. https://doi.org/10.1038/sdata.2016.123

Moreno, A., & Hasenauer, H. (2016). Spatial downscaling of European climate data: Spatial downscaling of European climate data. *International Journal of Climatology*, *36*(3), 1444–1458. https://doi.org/10.1002/joc.4436

Morris, W. K., Vesk, P. A., & McCarthy, M. A. (2013). Profiting from pilot studies: Analysing mortality using Bayesian models with informative priors. *Basic and Applied Ecology*, *14*(1), 81–89. https://doi.org/10.1016/j.baae.2012.11.003

Needham, J., Merow, C., Chang-Yang, C.-H., Caswell, H., & McMahon, S. M. (2018). Inferring forest fate from demographic data: From vital rates to population dynamic models. *Proceedings of the Royal Society B: Biological Sciences*, *285*(1874), 20172050. https://doi.org/10.1098/rspb.2017.2050

Niinemets, Ü., & Valladares, F. (2006). Tolerance to shade, drought, and waterlogging of temperate Northern Hemisphere trees and shrubs. *Ecological Monographs*, *76*(4), 521–547. https://doi.org/10.1890/0012-9615(2006)076[0521:TTSDAW]2.0.CO;2

Pagel, M. (1999). Inferring the historical patterns of biological evolution. *Nature*, *401*(6756), 877–884.

Pironon, S., Papuga, G., Villellas, J., Angert, A. L., García, M. B., & Thompson, J. D. (2017). Geographic variation in genetic and demographic performance: New insights from an old biogeographical paradigm: The centre-periphery hypothesis. *Biological Reviews*, *92*(4), 1877–1909. https://doi.org/10.1111/brv.12313

Plas, F., Ratcliffe, S., Ruiz-Benito, P., Scherer-Lorenzen, M., Verheyen, K., Wirth, C., Zavala, M. A., Ampoorter, E., Baeten, L., Barbaro, L., Bastias, C. C., Bauhus, J., Benavides, R., Benneter, A., Bonal, D., Bouriaud, O., Bruelheide, H., Bussotti, F., Carnol, M., … Allan, E. (2018). Continental mapping of forest ecosystem functions reveals a high but unrealised potential for forest multifunctionality. *Ecology Letters*, *21*(1), 31–42. https://doi.org/10.1111/ele.12868

Purves, D. W. (2009). The demography of range boundaries versus range cores in eastern US tree species. *Proceedings of the Royal Society B: Biological Sciences*, *276*(1661), 1477–1484. https://doi.org/10.1098/rspb.2008.1241

Putnam, R. C., & Reich, P. B. (2017). Climate and competition affect growth and survival of transplanted sugar maple seedlings along a 1700-km gradient. *Ecological Monographs*, *87*(1), 130–157. https://doi.org/10.1002/ecm.1237

R Core Team. (2019). *R: A language and environment for statistical computing*. R Foundation for Statistical Computing. Retrieved from https://www.R-project.org/

Ratcliffe, S., Dahlgren, J., Lehtonen, A., Adonteng, C., Ruiz-Benito, P., Zavala, M. A., Kaendler, G., Mäkipää, R., & Kunstler, G. (2020). Forest Inventory data from Finland and Sweden for: Demographic performance of European tree species at their hot and cold climatic edges, plus ancillary climate data. *Dryad Digital Repository*, https://doi.org/10.5061/dryad.wm37pvmkw

Ratcliffe, S., Liebergesell, M., Ruiz-Benito, P., Madrigal González, J., Muñoz Castañeda, J. M., Kändler, G., Lehtonen, A., Dahlgren, J., Kattge, J., Peñuelas, J., Zavala, M. A., & Wirth, C. (2015). Modes of functional biodiversity control on tree productivity across the European continent. *Global Ecology and Biogeography*, *25*, 251–262. https://doi.org/10.1111/geb.12406

Ratcliffe, S., Wirth, C., Jucker, T., van der Plas, F., Scherer-Lorenzen, M., Verheyen, K., Allan, E., Benavides, R., Bruelheide, H., Ohse, B., Paquette, A., Ampoorter, E., Bastias, C. C., Bauhus, J., Bonal, D., Bouriaud, O., Bussotti, F., Carnol, M., Castagneyrol, B., … Baeten, L. (2017). Biodiversity and ecosystem functioning relations in European forests depend on environmental context. *Ecology Letters*, *20*(11), 1414–1426. https://doi.org/10.1111/ele.12849

Reich, P. B. (2014). The world-wide 'fast-slow' plant economics spectrum: A traits manifesto. *Journal of Ecology*, *102*(2), 275–301. https://doi.org/10.1111/1365-2745.12211

Reich, P. B., & Oleksyn, J. (2004). Global patterns of plant leaf N and P in relation to temperature and latitude. *Proceedings of the National Academy of Sciences of the United States of America*, *101*(30), 11001–11006. https://doi.org/10.1073/pnas.0403588101

Reich, P. B., Oleksyn, J., & Tjoelker, M. G. (1996). Needle respiration and nitrogen concentration in scots pine populations from a broad latitudinal range: A common garden test with field-grown trees. *Functional Ecology*, *10*(6), 768. https://doi.org/10.2307/2390512

Ruiz-Benito, P., Vacchiano, G., Lines, E. R., Reyer, C. P. O., Ratcliffe, S., Morin, X., Hartig, F., Mäkelä, A., Yousefpour, R., Chaves, J. E., Palacios-Orueta, A., Benito-Garzón, M., Morales-Molino,

C., Camarero, J. J., Jump, A. S., Kattge, J., Lehtonen, A., Ibrom, A., Owen, H. J. F., & Zavala, M. A. (2020). Available and missing data to model impact of climate change on European forests. *Ecological Modelling*, *416*, 108870. https://doi.org/10.1016/j.ecolmodel.2019.108870

Salguero-Gómez, R., Jones, O. R., Archer, C. R., Buckley, Y. M., Che-Castaldo, J., Caswell, H., Brinks, E., de Buhr, H., Farack, C., Gottschalk, F., Hartmann, A., Henning, A., Hoppe, G., Römer, G., Runge, J., Ruoff, T., Wille, J., Zeh, S., Davison, R., ... Vaupel, J. W. (2015). The compadrePlant Matrix Database: An open online repository for plant demography. *Journal of Ecology*, *103*(1), 202–218. https://doi.org/10.1111/1365-2745.12334

Schelhaas, M.-J., Fridman, J., Hengeveld, G. M., Henttonen, H. M., Lehtonen, A., Kies, U., Krajnc, N., Lerink, B., Ní Dhubháin, Á., Polley, H., Pugh, T. A. M., Redmond, J. J., Rohner, B., Temperli, C., Vayreda, J., & Nabuurs, G.-J. (2018). Actual European forest management by region, tree species and owner based on 714,000 re-measured trees in national forest inventories. *PLoS ONE*, *13*(11), e0207151. https://doi.org/10.1371/journal.pone.0207151

Schielzeth, H. (2010). Simple means to improve the interpretability of regression coefficients: Interpretation of regression coefficients. *Methods in Ecology and Evolution*, *1*(2), 103–113. https://doi.org/10.1111/j.2041-210X.2010.00012.x

Stahl, U., Reu, B., & Wirth, C. (2014). Predicting species' range limits from functional traits for the tree flora of North America. *Proceedings of the National Academy of Sciences of the United States of America*, *111*(38), 13739–13744. https://doi.org/10.1073/pnas.1300673111

Stephenson, N. L., van Mantgem, P. J., Bunn, A. G., Bruner, H., Harmon, M. E., O'Connell, K. B., Urban, D. L., & Franklin, J. F. (2011). Causes and implications of the correlation between forest productivity and tree mortality rates. *Ecological Monographs*, *81*(4), 527–555. https://doi.org/10.1890/10-1077.1

Svenning, J.-C., & Skov, F. (2004). Limited filling of the potential range in European tree species. *Ecology Letters*, *7*, 565–573. https://doi.org/10.1111/j.1461-0248.2004.00614.x

Symonds, M. R., & Blomberg, S. P. (2014). A primer on phylogenetic generalised least squares. In L. Z. Garamszegi (Ed.), *Modern phylogenetic comparative methods and their application in evolutionary biology* (pp. 105–130). Springer.

Talluto, M. V., Boulangeat, I., Vissault, S., Thuiller, W., & Gravel, D. (2017). Extinction debt and colonization credit delay range shifts of eastern North American trees. *Nature Ecology & Evolution*, *1*, 0182. https://doi.org/10.1038/s41559-017-0182

Thuiller, W., Lafourcade, B., Engler, R., & Araújo, M. B. (2009). BIOMOD – A platform for ensemble forecasting of species distributions. *Ecography*, *32*(3), 369–373. https://doi.org/10.1111/j.1600-0587.2008.05742.x

Thuiller, W., Münkemüller, T., Schiffers, K. H., Georges, D., Dullinger, S., Eckhart, V. M., Edwards Jr., T. C., Gravel, D., Kunstler, G., Merow, C., Moore, K., Piedallu, C., Vissault, S., Zimmermann, N. E., Zurell, D., & Schurr, F. M. (2014). Does probability of occurrence relate to population dynamics? *Ecography*, *37*(12), 1155–1166. https://doi.org/10.1111/ecog.00836

Tredennick, A. T., Teller, B. J., Adler, P. B., Hooker, G., & Ellner, S. P. (2018). Size-by-environment interactions: A neglected dimension of species' responses to environmental variation. *Ecology Letters*, *21*(12), 1757–1770. https://doi.org/10.1111/ele.13154

Trouillier, M., van der Maaten-Theunissen, M., Scharnweber, T., Würth, D., Burger, A., Schnittler, M., & Wilmking, M. (2019). Size matters—A comparison of three methods to assess age- and size-dependent climate sensitivity of trees. *Trees*, *33*(1), 183–192. https://doi.org/10.1007/s00468-018-1767-z

Tuck, S. L., Porter, J., Rees, M., & Turnbull, L. A. (2018). Strong responses from weakly interacting species. *Ecology Letters*, *21*(12), 1845–1852. https://doi.org/10.1111/ele.13163

Vilà-Cabrera, A., Premoli, A. C., & Jump, A. S. (2019). Refining predictions of population decline at species' rear edges. *Global Change Biology*, *25*(5), 1549–1560. https://doi.org/10.1111/gcb.1459710.1111/gcb.14597

Wright, I. J., Dong, N., Maire, V., Prentice, I. C., Westoby, M., Díaz, S., Gallagher, R. V., Jacobs, B. F., Kooyman, R., Law, E. A., Leishman, M. R., Niinemets, Ü., Reich, P. B., Sack, L., Villar, R., Wang, H., & Wilf, P. (2017). Global climatic drivers of leaf size. *Science*, *357*(6354), 917–921. https://doi.org/10.1126/science.aal4760

Wright, I. J., Reich, P. B., & Westoby, M. (2003). Least-cost input mixtures of water and nitrogen for photosynthesis. *The American Naturalist*, *161*(1), 98–111. https://doi.org/10.1086/344920

Wright, I. J., Reich, P. B., Westoby, M., Ackerly, D. D., Baruch, Z., Bongers, F., Cavender-Bares, J., Chapin, T., Cornelissen, J. H. C., Diemer, M., Flexas, J., Garnier, E., Groom, P. K., Gulias, J., Hikosaka, K., Lamont, B. B., Lee, T., Lee, W., Lusk, C., ... Villar, R. (2004). The worldwide leaf economics spectrum. *Nature*, *428*(6985), 821–827. https://doi.org/10.1038/nature02403

Zanne, A. E., Tank, D. C., Cornwell, W. K., Eastman, J. M., Smith, S. A., FitzJohn, R. G., McGlinn, D. J., O'Meara, B. C., Moles, A. T., Reich, P. B., Royer, D. L., Soltis, D. E., Stevens, P. F., Westoby, M., Wright, I. J., Aarssen, L., Bertin, R. I., Calaminus, A., Govaerts, R., ... Beaulieu, J. M. (2014). Three keys to the radiation of angiosperms into freezing environments. *Nature*, *506*, 89–92. https://doi.org/10.1038/nature12872

Zhu, K., Woodall, C. W., Ghosh, S., Gelfand, A. E., & Clark, J. S. (2014). Dual impacts of climate change: Forest migration and turnover through life history. *Global Change Biology*, *20*(1), 251–264. https://doi.org/10.1111/gcb.12382

Zimmermann, N. E., Jandl, R., Hanewinkel, M., Kunstler, G., Klling, C., Gasparini, P., Breznikar, A., Meier, E. S., Normand, S., Ulmer, U., Gschwandtner, T., Veit, H., Naumann, M., Falk, W., Mellert, K., Rizzo, M., Skudnik, M., & Psomas, A. (2013). Potential future ranges of tree species in the alps. In G. Cerbu, M. Hanewinkel, G. Giacomo, & R. Jandl (Eds.), *Management strategies to adapt alpine space forests to climate change risks* (pp. 37–48). InTech.

SUPPORTING INFORMATION

Additional supporting information may be found online in the Supporting Information section.

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