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Colonization and extinction dynamics and their link to the distribution of European trees at continental scale

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Abstract

- **Aim:** Processes driving current tree species distribution are still largely debated. Attempts to relate species distribution and population demography metrics have shown mixed results. In this context, we would like to test the hypotheses that the metapopulation processes of colonization and extinction are linked to species distribution models. - **Location:** Europe: Spain, France, Germany, Finland, and Sweden. - **Taxon:** Angiosperms and Gymnosperms. - **Methods:** For the 17 tree species analyzed we fitted species distribution model (SDM) relating environmental variables to presence absence data across Europe. Then using independent data from national forest inventories across Europe we tested whether colonization and extinction probabilities are related to occurrence probability estimated by the SDMs. Finally, we tested how colonization and extinction respectively drive probability of presence at the metapopulation equilibrium. - **Results:** We found that for most species at least one process (colonization/extinction) is related to the occurrence probability, but rarely both. - **Main conclusions:** Our study supports the view that metapopulation dynamics are partly related to SDM occurrence probability through one of the metapopulation probabilities. However these links are relatively weak and the metapopulation models tend to overestimate the occurrence probability. Our results call for caution in model extrapolating SDM models to metapopulation dynamics.

34 1 Introduction

35 The vast majority of species have restricted geographical ranges (Holt & Keitt, 2000). Understanding
36 the factors determining these ranges is fundamental to have insights on future species redistribution
37 in the face of climate change. Species distribution is thought to be tightly connected to its ability
38 to cope with local abiotic conditions and thus species' niche (Pulliam, 2000; Thuiller, Lavorel, &
39 Araújo, 2005; Soberón, 2007). This view underpins most statistical species distribution models
40 (SDM) that relate species presence with local environmental conditions. These models have been
41 extensively used in recent years and provided very detailed descriptions of species environmental
42 requirements based on occurrence data. They provide, however, very little indication on how species
43 distribution arises from population dynamics. This is surprising because we start to have a rich
44 theoretical understanding of how population dynamics control species distribution and species range
45 limits (Holt, Keitt, Lewis, Maurer, & Taper, 2005).

46 There are numerous routes through which range limits can arise (Holt et al., 2005). The first class
47 of mechanisms consider only the local population dynamics when there is little effect of dispersal.
48 The most classical view of this approach is that species ranges match the environmental conditions
49 where birth rates exceed mortality rates (i.e. where the rate of population growth is above 1, Brown
50 (1984)). Most existing field studies did not support this assumption, see the review by Pironon
51 et al. (2017). For instance, Thuiller et al. (2014) demonstrated that major demographic parameters
52 of European tree species were not strongly correlated to the occurrence probability derived from
53 SDMs. Generally, only transplant experiments beyond species range have shown a tendency of a
54 decrease in population growth rate or some demographic rates (Hargreaves, Samis, & Eckert, 2013;
55 Lee-Yaw et al., 2016).

56 Then Holt et al. (2005) proposed two other classes of mechanisms based on local population
57 dynamics: demographic stochasticity and temporal variability. Demographic stochasticity could
58 increase the risk of extinction at the range limits (Boyce, Haridas, Lee, & Group, 2006; Ovaskainen
59 & Meerson, 2010). For instance, this could be due to a lower absolute density leading to an increase
60 in risk of extinction solely due to stochastic variability. However, several studies did not find strong
61 support for the abundance center hypothesis which propose that abundance should be higher in the
62 center of the distribution (Murphy, VanDerWal, & Lovett-Doust, 2006; Sagarin, Gaines, & Gaylord,
63 2006). Temporal variability could also increase the risk of local extinction even if the average
64 growth rate and the average population size are not limiting factors. Rare extreme conditions or
65 highly unstable environmental conditions might control the extinction risk at the range limits. Field
66 tests of these mechanisms are extremely rare, and show weak support for this hypothesis. Csergő
67 et al. (2017), using detailed demographic data for plant species (including trees), found no clear link
68 between climate suitability, derived from a SDM, and several detailed population metrics including
69 time to quasi-extinction, stochastic population growth rate or transient population dynamics.

70 Another class of mechanisms underlying species ranges is a regional equilibrium between colo-
71 nization/extinction dynamics of populations connected by dispersal (Holt et al., 2005; Holt & Keitt,
72 2000). This last class relates to the metapopulation paradigm and proposes that species ranges
73 arise from the gradient of three variables: the extinction rate, the colonization rate, and the habitat
74 structure (i.e. the availability of suitable area for settlement). In this model, the dynamic of the

75 local population is ignored when compared to regional dynamics (Drechsler & Wissel, 19997). Such
76 models thus ignore the details of the population dynamics but rather focus on patch occupancy
77 dynamics (extinction and colonization events). Few studies have focused on these processes for
78 tree species (see Purves, Zavala, Ogle, Prieto, & Benayas, 2007; García-Valdes, Zavala, Araújo, &
79 Purves, 2013; García-Valdés, Gotelli, Zavala, Purves, & Araújo, 2015; Talluto, Boulangeat, Vissault,
80 Thuiller, & Gravel, 2017). Results in North America (Talluto et al., 2017) showed that metapopula-
81 tion processes captured potential future range shifts for most tree species. García-Valdes et al. (2013)
82 inferred potential changes in species distribution in Spain, but we still lack studies that explore this
83 mechanism at the European scale, covering a larger portion of different species distribution.

84 Here we propose to analyze how local species extinction and colonization probabilities vary within
85 the range of the the main European tree species across the entire continent using more than 80 000
86 plots of national forest inventories. Species distributions are summarized by occurrence probability
87 estimated with an ensemble SDM fitted to independent data extracted from the EU-Forest data
88 set (Mauri, Strona, & San-Miguel-Ayanz, 2017). We use presence/absence from NFI data to get
89 observations of extinction and colonization events, a colonization event being then separated into
90 an outcome of a seed input and a successful recruitment.

91 We then analyze the relationship between the occurrence probability derived from SDMs and
92 the extinction/recruitment probabilities derived from NFI data to test the following hypotheses:

- 93 • Extinction probabilities increase when the SDM occurrence probability decreases.
- 94 • Recruitment probabilities decrease when the SDM occurrence probability decreases.
- 95 • Finally we analyse how the equilibrium occurrence probability, predicted by metapopulation
96 models using estimates of extinction/recruitment probability, match the current SDM oc-
97 currence probability. This allows us to evaluate the relative importance of extinction and
98 recruitment in driving the distribution of each species.

99 2 Materials and Methods

100 Our objective is first to test how extinction and recruitment probabilities vary as a function of the
101 SDM derived occurrence probability for the dominant European tree species. Then we derived a
102 potential equilibrium and compared it to current SDM occurrence probability to analyse the relative
103 importance of each of the two processes, and whether it under or overestimates the current occurrence
104 probability.

105 To do this, we first gathered data on tree local extinction and colonization events from national
106 forest inventory plots. Then, we estimated the occurrence probability with SDM models fitted to
107 independent data extracted from the EU-Forest database. Subsequently, we modelled extinction
108 and recruitment probabilities in function of SDM occurrence probability with two observations of
109 occupancy data via a spatially inhomogeneous Markov chain. Because national forest inventories
110 provide little information on the local seed source around the plots, we used estimation of species
111 local frequency in 1 km grid directly from JRC maps (see section 2.2) as a surrogate of seed source.
112 Finally, we derived the probability of presence at equilibrium based on the estimated extinction and

113 recruitment considering two alternative formulations: one in which communities are considered as
114 closed systems, the other in which communities are open to external seed sources.

115 2.1 NFI datasets

116 To calibrate our model, we required information on the presence/absence of each species at two dif-
117 ferent dates over a large geographical area to cover, as far as possible, the entire species distributions.
118 We used a database of tree data from the National Forest Inventories (NFI) of Finland, France, Ger-
119 many, Spain, Sweden, compiled as part of the FunDivEurope project (<http://www.fundiveurope.eu>,
120 Baeten et al. (2013)).

121 Inventory protocols differs between NFIs (see Supplementary materials Section 1 for a detailed
122 description of each survey protocol). Surveys were conducted within a circular plot with a fixed
123 radius or in concentric subplots with different radius and minimum diameter at breast height (DBH)
124 for all NFIs except Germany, where an angle-count method (basal area factor of $4 \text{ m}^2 \text{ ha}^{-1}$) was
125 used. Because the DBH thresholds for trees to be recorded varied between the inventories, we only
126 included trees with a DBH of 10 cm or greater. For each NFI, except France, two inventory surveys
127 were conducted with a variable time interval (from 4 to 16 years, see Figure 1 b). The French
128 inventory is based on single surveys where the growth of alive trees (based on a short core) and
129 approximate the date of death of dead trees are estimated and can be used to reconstruct the stand
130 structure five years before the census, making it comparable with revisited plots data.

131 To avoid influences of management on the extinction and colonization events, we discarded plots
132 where a management effect was reported between surveys. This led to a selection of 80 157 plots
133 with 173 species. Among these species we selected the most abundant species (the cumulative basal
134 area of species retained represented more than 95% of the total basal area) and excluded exotic
135 species as well as species for which JRC maps (see below) were not available.

136 For each plot, a species was considered present when at least one tree was observed. The
137 succession of two surveys allowed then to deduce state transitions ($0 \rightarrow 1$ for local colonization, 1
138 $\rightarrow 0$ for local extinction). Since several protocols are based on concentric circular plots with varying
139 DBH thresholds, a newly observed tree might not be a recruited tree, i.e. its DBH during the first
140 census was above 10 cm, but it was not recorded due to the larger DBH threshold for its subplot.
141 We used a species specific growth model to estimate the probability that a new tree (present only
142 in the second census) had a former DBH below 10 cm. The growth model was built as a generalized
143 linear model using an aridity index, the sum growing degree days, and tree DBH as explanatory
144 variables (see Supplementary Materials Section 2). We thus considered a plot as colonized if the
145 probability that the largest newly observed tree had a former DBH below 10 cm greater or equal
146 to 0.5, otherwise the species was considered as present at both censuses ($1 \rightarrow 1$). This correction
147 had a strong impact on the Spanish and German inventories, significantly reducing the number of
148 colonization events. We decided to exclude from further analysis species with less than 10 events
149 for extinction or colonization (i.e. *Quercus suber*, *Pinus pinea* and *Acer pseudoplatanus*), resulting
150 in a final selection of 17 species.

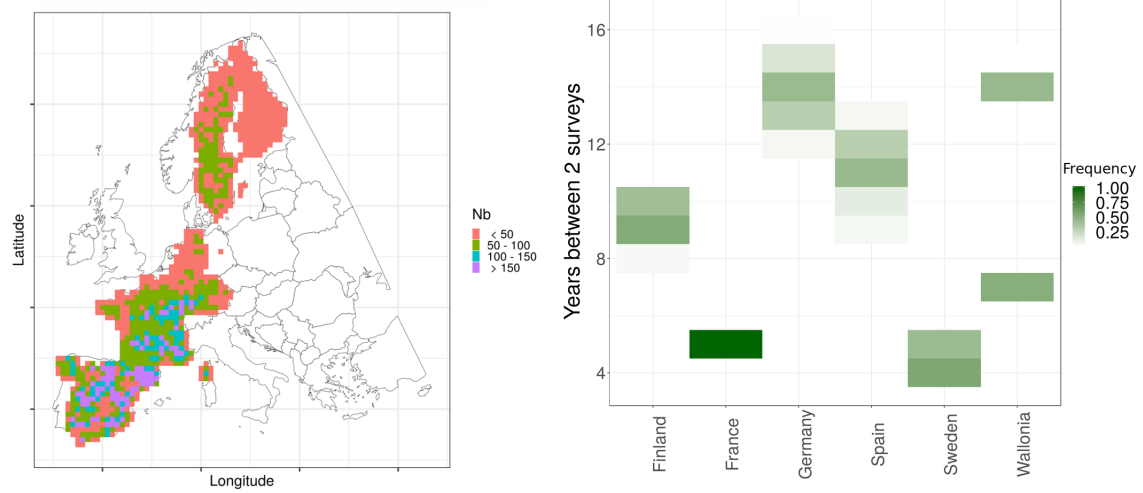


Figure 1: (a) Density map of NFI plots (grid of 50 km x 50 km) used to estimate colonization and extinction events.

(b) Distribution of number of years between subsequent surveys by country

151 2.2 Joint research center - species local frequency

152 The density of NFI plots is too low to accurately describe the local abundance of trees that
153 can disperse seeds into a given plot. Distances between NFI plots are about 1 km or above
154 whereas most dispersal events occur in less than 100 m from the seed source (see for example
155 Nathan, Safriel, and Imanuel (2001), Bullock et al. (2017)). To represent seed inputs into a
156 plot, we thus used the species' local frequency (hereafter F_{JRC}) in the corresponding 1 km cell
157 produced by the Joint Research Center (RPP - Relative Probability of Presence on JRC web-
158 site (<https://forest.jrc.ec.europa.eu/en/european-atlas/atlas-data-and-metadata/>), see San-Miguel-
159 Ayanz, de Rigo, Caudullo, Houston Durrant, and Mauri (2016). Each map estimates the relative
160 frequency of the species based on datasets of field observations as represented in the Forest In-
161 formation System for Europe (FISE), which integrates National Forest Inventories, BioSoil and
162 Forest Focus data sets. The presence/absence data are assimilated at a spatial resolution of 1
163 km based on multiple smoothing kernels of varying dimension. Independent estimations of forest
164 cover extracted from the Pan-European Forest Type Map 2006 (FTYP2006, <http://forest.jrc.ec.europa.eu/forest-mapping/forest-type-map>) are used to rescale the species frequency by the cover
165 of broadleaved forest, coniferous forest or other non-forest categories based on 25 m x 25 m pix-
166 els (San-Miguel-Ayanz et al., 2016). We chose this variable because it summarized a very large
167 amount of data on a European scale and can be considered as a strong indicator of the proportion
168 of adjacent plots in which the species is present within a 1 km patch. An explicit representation of
169 seed availability via dispersal mechanisms was beyond the scope of this work because available data
170 do not provide a detailed description of the seed source in the plot surroundings from where most
171 dispersal events occur (Nathan et al., 2001). However, because the JRC species local frequency data
172 is based on the spatial integration of presence-absence observations, our seed source estimate can
173 be influenced by data beyond the 1 km grid. The long-distance seed dispersal events are thus not
174 excluded, even if there is no observation of presence in the 1 km cell.
175

176 2.3 SDM

177 We estimated species occurrence probability (hereafter P_{occ}) on each NFI plot with ensemble species
178 distribution models fitted to the EU-Forest data set (Mauri et al., 2017) which provides species
179 presence/absence on a 1 km grid. The initial EU-Forest data set includes more than 250 000 plots
180 across Europe including countries not present in FUNDIV. We excluded all NFI observations from
181 the EU-Forest to avoid using the same data to estimate both extinction/colonization and the SDM
182 probability of presence. This exclusion was performed to avoid any potential circularity arising from
183 the use of the same data in both analyses. After excluding NFI observations, we retained 9600 data
184 points across Europe, coming from ForestFocus and BioSoil campaigns. For each grid point, we
185 extracted mean annual temperature, precipitation of wettest quarter, temperature and precipitation
186 seasonality from CHELSA climatologies (Karger et al., 2017), pH measured in water solution (5 cm
187 depth) from SoilGrid (Hengl et al., 2017), and aridity index (the mean annual precipitation divided
188 by the mean annual potential evapotranspiration) and actual evapo-transpiration from CGIAR-CSI
189 (Trabucco & Zomer, 2010). We verified that correlation coefficients between variables were always

190 lower than 0.7 (Dormann et al., 2013), except for pH and mean annual temperature which had a
 191 correlation coefficient of 0.72. We nevertheless decided to keep both variables, as the correlation
 192 was only slightly higher than 0.7, and both soil information and mean temperature variables may be
 193 important drivers for species distribution. Then we fitted ensemble SDM models with BIOMOD2
 194 (Thuiller, Lafourcade, Engler, & Araújo, 2009) using four different models (GAM, GLM, GBM, and
 195 Random Forest). Based on this ensemble model we estimated species occurrence probability on each
 196 NFI plot for all species. Details on the evaluation on the predictive power of the SDM are provided
 197 in the Supplementary materials Section 3 (see Figure 1 with performance scores of SDM for each
 198 species based on True Skill Statistic, TSS and Area Under the Curve of the Receiver Operating
 199 Characteristic, AUC).

200 2.4 Patch occupancy transition model

201 The patch occupancy model is a spatially inhomogeneous Markov chain, the state vector being the
 202 patch occupancy of the N plots $X(t)$ at time t . The probability of transition between the two time
 203 successive patch occupancy patterns is:

$$P[\mathbf{X}(t+1)|\mathbf{X}(t)] = \prod_{i=1}^N \begin{cases} 1 - C_i(t) & \text{if } X_i(t) = 0 \text{ and } X_i(t+1) = 0 \\ C_i(t) & \text{if } X_i(t) = 0 \text{ and } X_i(t+1) = 1 \\ E_i(t) & \text{if } X_i(t) = 1 \text{ and } X_i(t+1) = 0 \\ 1 - E_i(t) & \text{if } X_i(t) = 1 \text{ and } X_i(t+1) = 1 \end{cases} \quad (1)$$

204 where N is the total number of plots observed, E_i the extinction probability in plot i , and C_i the
 205 colonization probability in plot i .

206 The extinction probability (E) of a species in a plot only depends on the local environmental
 207 conditions, i.e. the occurrence probability derived from the SDM (P_{occ}). The colonization probability
 208 (C) is divided into two contributions: recruitment probability (R) which depends on P_{occ} , and seed
 209 source (S). The recruitment probability R is the probability of at least one tree reaching 10 cm
 210 between two protocols. Colonization probability is simply expressed as the product of R and S ,
 211 where the seed source S is estimated by the JRC as presented above. Colonization events can occur
 212 in any plot with a non-null seed source.

213 Recruitment (R) and extinction (E) probabilities were related to the SDM occurrence probability
 214 P_{occ} and the species local frequency F_{JRC} as follows:

$$\begin{aligned} \text{logit}(E_i) &= \alpha + \beta * P_{occ_i} \\ \text{logit}(R_i) &= \gamma + \delta * P_{occ_i} \\ C_i &= R_i * S_i, \text{ with } S_i = F_{JRC_i} \end{aligned} \quad (2)$$

215 Differences in protocols between countries can influence the probability of observing extinction and
 216 colonization events. To account for this protocol effect in our analysis we used fixed country specific
 217 intercept parameters (α and γ).

218 Because the time interval between two censuses may vary across plots (between 4 and 15 years),
 219 we standardized the parameters to a 5 years sampling interval as done in Talluto et al. (2017), the

220 probability of an recruitment/extinction was computed as:

$$P(Event) = 1 - (1 - P(Event_{5years}))^{\frac{n}{5}} \quad (3)$$

221 with n being the number of years between the two censuses.

222 2.5 Calibration of the model

223 For each species, extinction and recruitment parameters were estimated separately using a Metropo-
224 lis Hastings Monte Carlo sampling algorithm, with priors following a Cauchy distribution (Gelman,
225 Jakulin, Pittau, & Su, 2008; Ghosh, Li, & Mitra, 2018) using JAGS (Plummer, 2003). Conver-
226 gence was checked by evaluating whether the Gelman-Rubin convergence statistic was below 1.1, as
227 recommended by Brooks and Gelman (1998), using 4 chains.

228 2.6 Probability of presence at equilibrium

229 Finding a link between either recruitment or extinction and the SDM P_{occ} does not necessarily
230 mean that the equilibrium model would yield the same occurrence probability as the SDM. To
231 evaluate this, we derived from the estimates of recruitment and extinction a probability of presence
232 at equilibrium (hereafter P_{eq}). We explored the match with P_{occ} and the relative contribution of
233 extinction and recruitment probabilities. The equilibrium can be defined in two ways: (1) We can
234 assume that grid cells are open systems with a fixed seed source S , where the probability of presence
235 in the grid cell is a function of extinction, recruitment, and the value of seed source. In this case
236 there is no feedback of the colonization and extinction on the seed source. (2) We can assume that
237 grid cells are closed systems of interconnected suitable patches, with a feedback of the colonization
238 and extinction processes on the seed source. In this case, an extinction probability exceeding the
239 colonization probability would lead to a species absence.

240 Both types of equilibrium can be derived from the same equation:

$$\frac{dp}{dt} = C * (1 - p) - E * p = 0 \quad (4)$$

241 with p the proportion of suitable patches occupied.

242 The difference between the two types of equilibrium corresponds to different formulations of C .
243 In the first formulation, S is constant over time and $C = R \times S$, while in the second formulation, S
244 varies with P_{occ} and $C = p * R$. These two alternative formulations lead to the following equilibria:

- 245 • (1) when we consider a fixed seed source, and compute the equilibrium state for each plot:
246 $P_{eq} = \frac{R \times S}{R \times S + E}$,
- 247 • (2) when we consider that the seed source is linked to the proportion of occupied patches
248 within each 1 km grid cell, then the proportion of suitable occupied patches is $P_{eq} = (1 - \frac{E}{R})$.

249 For both formulations, we studied the relative impact of extinction and colonization (including
250 seed source and recruitment probability) on the equilibrium state by fixing one of the probabilities

251 to its mean and letting the other vary based on our estimated slope of response to the SDM oc-
252 currence probability. We also computed the probability at equilibrium, letting both extinction and
253 colonization vary with P_{occ} . In the first model we can also set the fixed seed source to one (no
254 dispersal limitation) or let the fixed seed source vary with P_{occ} based on their observed links.

255 3 Results

256 3.1 Recruitment/Extinction dependencies

257 Results show that at least one of the estimated probabilities (recruitment or extinction) is signif-
258 icantly related to the SDM occurrence probability for all species, with the exception of *Fraxinus*
259 *excelsior*.

260 Overall, recruitment probability increases with the SDM occurrence probability (δ is positive,
261 Figure 2 left). The slope for the recruitment model is positive when considering all species posteriors,
262 and all species but *Abies alba* have a positive mean slope value. However, the effect is significant
263 (at the 5% level) for only nine species out of 17.

264 Extinction probability is not significantly influenced by occurrence probability when considering
265 all species. *Pinus nigra* and *Pinus halepensis* both present a positive slope, and seven species have
266 a significant negative slope.

267 *Populus tremula* and *Alnus glutinosa* exhibits a very broad posterior for the slope parameters
268 which can be related to the small range of probabilities of occurrence and the relative low discrimi-
269 native power of their SDM (see Supplementary materials Section 4).

270 Model performances according to the True Skill Statistics (TSS, see (Allouche, Tsoar, & Kadmon,
271 2006)) varied from good ($TSS > 0.5$), average ($0.3 < TSS < 0.5$), to poor ($TSS < 0.3$) depending
272 on the species and process (Tables 1 and 2).

273 Recruitment models showed average to good performance for all species. Extinction models
274 showed average to good performance for seven (41 %) species. Model scores were not related to
275 the number of observations (p-values of 0.69 and 0.48 for extinction and recruitment respectively).
276 We also computed ΔDIC for each process and species as a complementary quantification of model
277 performance (Spiegelhalter, Best, Carlin, & Van Der Linde, 2002). DIC helps compare the relative
278 fit of models, and in our case a negative value support the inclusion of P_{occ} dependency in the model.

279 Given the scarcity of colonization or extinction events, we also tested the robustness of our slope
280 estimates to the proportion of zeros by refitting the model after controlling the proportion of zeros
281 in the data (see Supplementary materials Section 5).

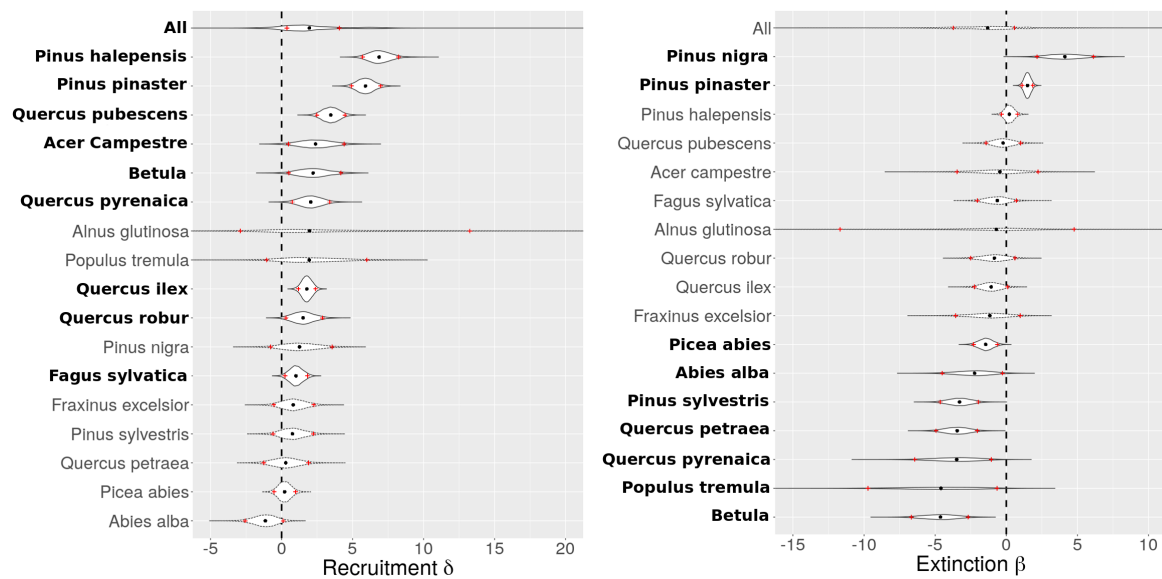


Figure 2: Posterior distribution of the slope of response of recruitment (left, δ) and extinction (right, β) to P_{occ} . Black points are posterior medians, red crosses indicate the 5th-95th percentile intervals. Species with name in bold have their interval not crossing 0.

Species	Nb of events	Median δ	90% Interval δ	TSS	Δ DIC
<i>Pinus sylvestris</i>	149	0.8	-0.6/2.2	0.61	0.3
<i>Picea abies</i>	231	0.2	-0.5/1.0	0.70	1.7
<i>Fagus sylvatica</i>	160	1.0	0.2/1.8	0.52	-2.9
<i>Quercus robur</i>	103	1.5	0.3/2.9	0.58	-3.2
<i>Quercus petraea</i>	48	0.3	-1.3/1.9	0.51	1.3
<i>Pinus pinaster</i>	66	5.9	4.9/7.0	0.76	-107.5
<i>Quercus ilex</i>	351	1.8	1.2/2.4	0.58	-23.0
<i>Pinus nigra</i>	48	1.3	-0.8/3.6	0.67	0.2
<i>Abies alba</i>	53	-1.1	-2.6/0.10	0.60	-0.5
<i>Pinus halepensis</i>	50	6.9	5.7/8.2	0.83	-120.0
<i>Quercus pubescens</i>	87	3.5	2.5/4.5	0.73	-35.0
<i>Betula</i>	265	1.3	0.5/4.2	0.63	-2.9
<i>Fraxinus excelsior</i>	121	0.8	-0.6/2.3	0.35	0.7
<i>Quercus pyrenaica</i>	85	2.0	0.8/3.4	0.79	-6.1
<i>Alnus glutinosa</i>	47	2.0	-2.9/13.3	0.54	-0.2
<i>Populus tremula</i>	73	1.9	-1.1/6.0	0.48	-0.5
<i>Acer campestre</i>	97	2.4	0.5/4.4	0.67	-2.9

Table 1: Estimates of δ , the slope of recruitment response to P_{occ} per species and their 90% confidence interval (see Materials and Methods for details on the model). Δ DIC is the difference of deviance information criterion – DIC – between the model and a null model (a model including only fixed country effects). TSS is the True Skill Statistics. Nb of events is the number of colonization events. TSS are calculated using the median of parameter posterior distributions.

Species	Nb of events	Median β	90 % Interval β	TSS	Δ DIC
<i>Pinus sylvestris</i>	151	-3.3	-4.6/-2.0	0.37	-18.0
<i>Picea abies</i>	98	-1.4	-2.3/-0.5	0.30	-6.2
<i>Fagus sylvatica</i>	44	-0.6	-2.0/0.7	0.23	1.1
<i>Quercus robur</i>	70	-0.8	-2.5/0.6	0.38	0.8
<i>Quercus petraea</i>	101	-3.5	-4.9/-2.0	0.53	-16.2
<i>Pinus pinaster</i>	227	1.5	1.1/1.9	0.19	-34.7
<i>Quercus ilex</i>	50	-1.1	-2.2/0.1	0.27	-1.0
<i>Pinus nigra</i>	67	4.1	2.2/6.1	0.26	-15.5
<i>Abies alba</i>	18	-2.2	-4.5/-0.3	0.34	-2.7
<i>Pinus halepensis</i>	117	0.2	-0.3/0.8	0.14	1.5
<i>Quercus pubescens</i>	40	-0.2	-1.4/1.0	0.1	1.6
<i>Betula</i>	188	-3.2	-6.7/-2.7	0.36	-16.9
<i>Fraxinus excelsior</i>	40	-1.2	-3.6/1.0	0.16	-0.1
<i>Quercus pyrenaica</i>	36	-3.5	-6.4/-1.0	0.44	-6.8
<i>Alnus glutinosa</i>	21	-0.7	-11.7/4.8	0.26	-0.3
<i>Populus tremula</i>	75	-4.6	-9.7/-0.7	0.20	-2.5
<i>Acer campestre</i>	30	-0.4	-3.5/2.2	0.09	1.1

Table 2: Estimates of β , the slope of response of extinction to P_{occ} per species and their 90% confidence interval (see Materials and Methods for details on the model). Δ DIC is the difference of deviance information criterion – DIC – between the model and a null model (without P_{occ} dependency). TSS is the True Skill Statistics. Nb of events is the number of extinction events. TSS are calculated using the median of parameter posterior distributions.

282 Since the range of P_{occ} is different between species, and the link function is non linear, the slope
283 is not sufficient to evaluate the magnitude of recruitment and extinction variability. We thus also
284 computed the relative contribution of P_{occ} to extinction and recruitment (Figure 3) as the difference
285 between the probability of extinction (colonization) at the low *vs.* high end of P_{occ} (respectively 5
286 and 95 % percentiles).

287 For most species, the relative contribution was higher for recruitment than for extinction, i.e.
288 most species are above the diagonal in the Figure 3, particularly for *Quercus ilex*, *Quercus pubescens*
289 and *Pinus halepensis*. Only *Quercus petraea* and *Abies alba* were below the diagonal, with a higher
290 relative contribution of P_{occ} on the extinction than on the recruitment probability.

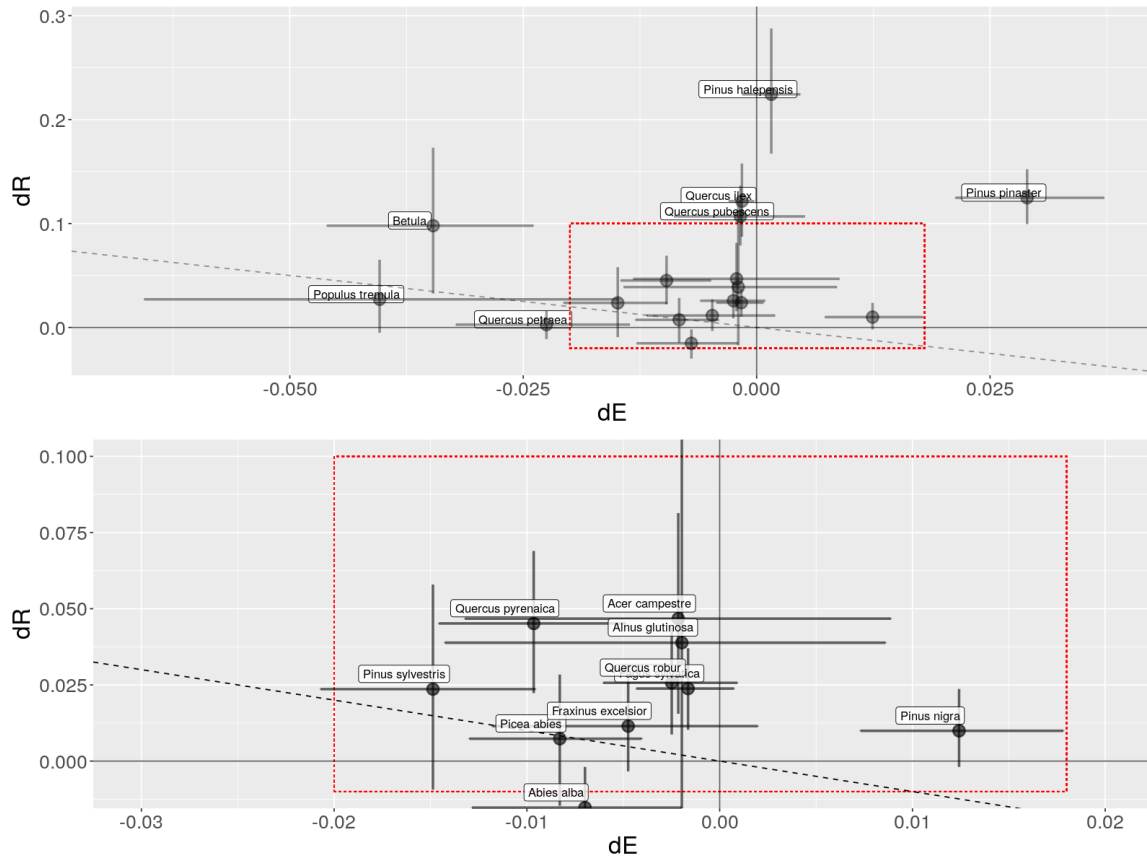


Figure 3: Relative contribution of P_{occ} to recruitment/extinction probabilities (dR and dE , respectively). For each species, dE and dR are calculated as the differences at high P_{occ} (95th centile) and low P_{occ} (5th centile). Bottom Figure is a zoom of the top Figure, indicated by a red square. Negative dE means a higher extinction rate at low occurrence probability; positive dR means a lower colonization at low occurrence probability. On both plots dashed line represents $dR = -dE$.

291 3.2 Equilibrium

292 The probabilities of both colonization and extinction depend on the SDM occurrence probability
293 P_{occ} . As a consequence, the probability of presence at equilibrium P_{eq} is directly a function of P_{occ} .
294 However, the shape of the function and the match between P_{eq} and P_{occ} depend on the estimates
295 of the slopes and intercepts of the colonization and extinction models (see Supplementary materials
296 section 6).

297 The relationship between P_{eq} and P_{occ} was positive for most species when we accounted for
298 the variation of both recruitment and extinction probability (green curve in figure 4). Only *Pinus*
299 *nigra* had a negative relationship. P_{eq} showed few variations and overall overestimated P_{occ} . When
300 dispersal limitation is not included, P_{eq} is above 0.5 when colonization probability exceeds extinction
301 probability, which is always the case for all species. If we included the seed approximation in the
302 formulation (black curves in figure 4), the match between P_{eq} and P_{occ} was stronger. Only *Quercus*
303 *ilex* exhibited systematically higher P_{eq} than P_{occ} , while for *Quercus petraea* P_{eq} tended to be lower
304 than P_{occ} . For all other species, P_{occ} stood within the range of P_{eq} .

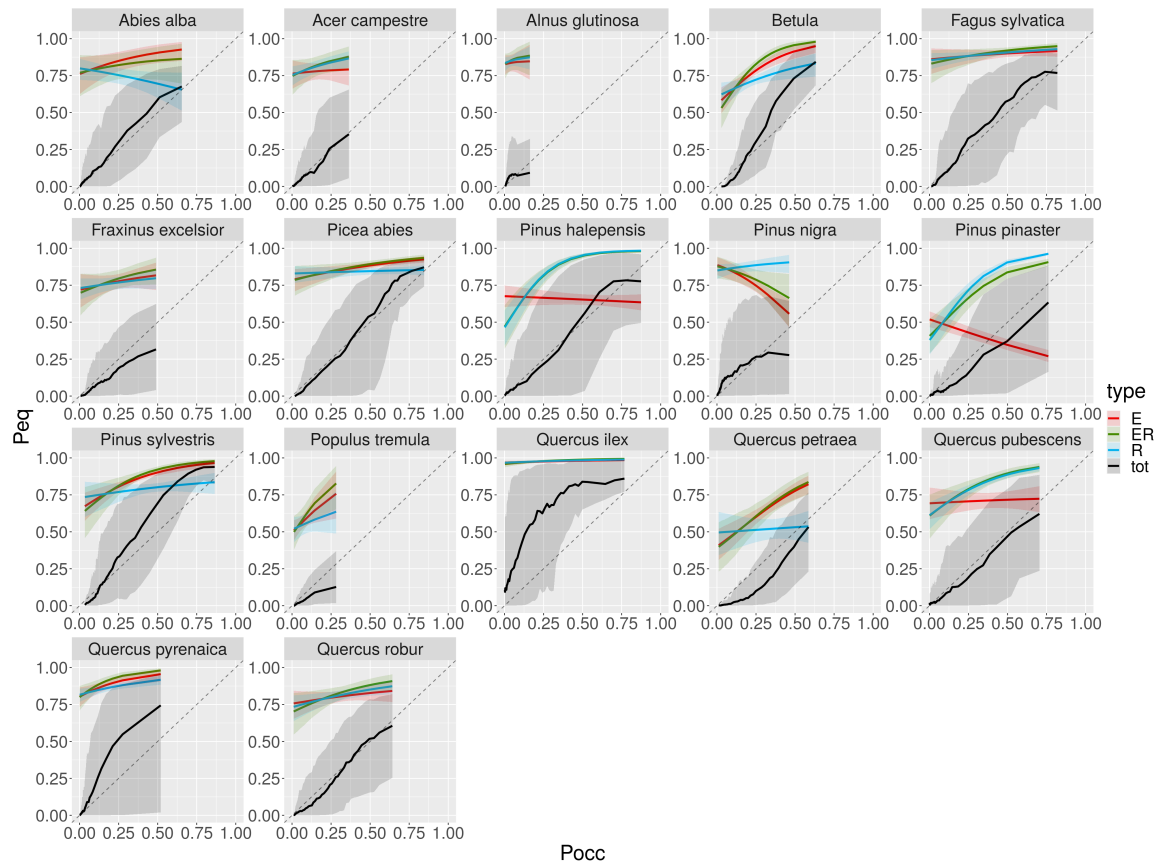


Figure 4: Equilibrium probabilities of presence (P_{eq}) against SDM occurrence probability (P_{occ}), calculated with open model with fixed seed source. Models either assume a seed source set to one and varying extinction (E in red), varying recruitment (R in blue), or both (ER in green) or extinction, recruitment, and seed source varying with P_{occ} (tot in back).

305 The second formulation of the equilibrium for a closed system leads also to a positive relationship
 306 between P_{eq} and P_{occ} (see green curves in Figure 5) with again the notable exception of *Pinus nigra*.
 307 In this case, an extinction probability higher than the recruitment probability would lead to a null
 308 value for P_{eq} . Overall, we also found that P_{eq} overestimated P_{occ} , and P_{eq} showed little variations
 309 along the P_{occ} gradient.

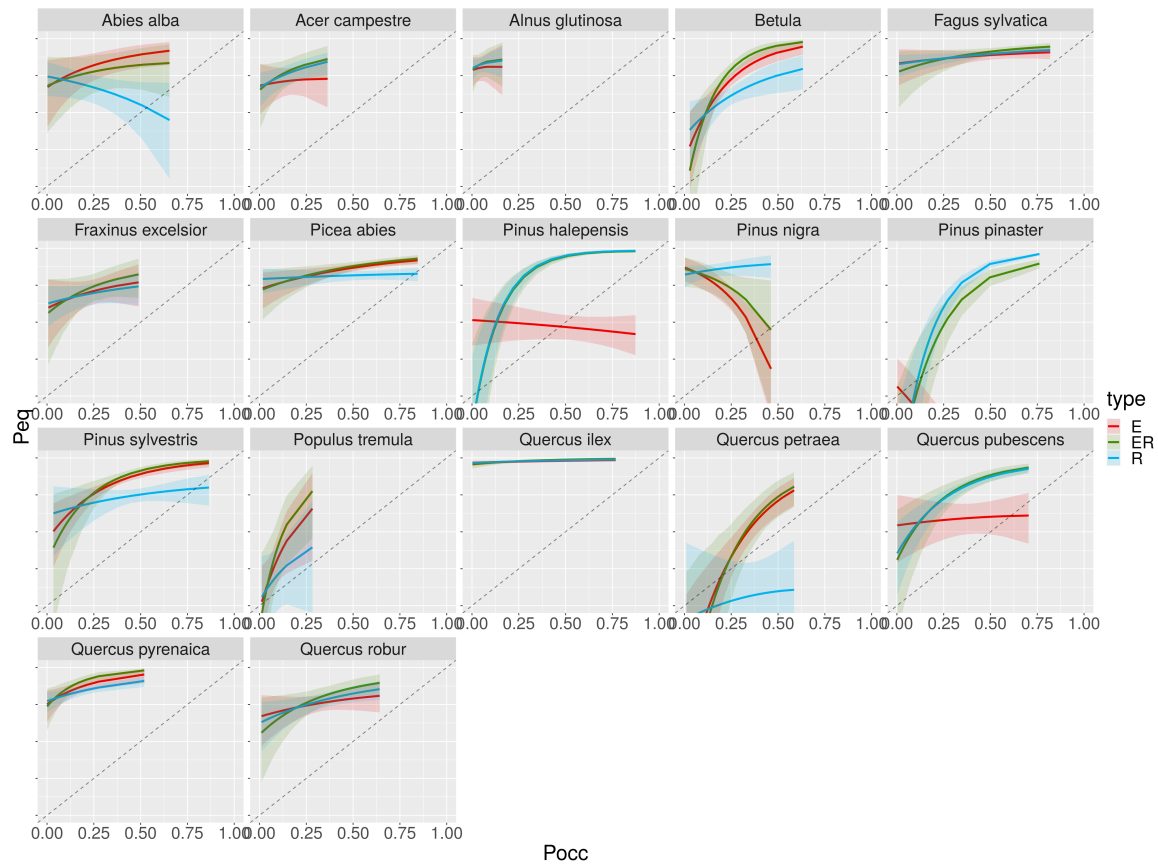


Figure 5: Equilibrium probability of presence (P_{eq}) against SDM occurrence probability (P_{occ}), calculated with closed formulation and varying extinction (E), varying recruitment (R), or both (ER).

310 4 Discussion

311 There is a long history of analyzing the drivers of population distribution, but surprisingly few
312 studies have explored with field data the link between probability of presence and metapopulation
313 processes such as extinction and colonization. Here, using data from national forest inventories, we
314 explored the question at the scale of the European continent for 17 tree species. We found that for all
315 species but *Fraxinus excelsior*, at least one of the processes, extinction or recruitment, is related to
316 the SDM occurrence probability, but rarely both (only two species). When combining extinction and
317 colonization, we also found that the probabilities of presence at equilibrium, derived from recruit-
318 ment and extinction probabilities, were generally positively correlated with the observed occurrence
319 probability (with the exception of *Pinus nigra*). However, at equilibrium, the metapopulation model
320 generally overestimated the occurrence probability.

321 4.1 Variation of extinction and recruitment probability within species 322 ranges

323 Holt and Keitt (2000) showed with theoretical models that there are different routes in metapopula-
324 tion dynamics to range limits, via variations of colonization rates, variations of extinction rates, or
325 variations of habitat availability. These three mechanisms are not mutually exclusive but can all be
326 at play at the same time. Here, we directly explore the relative importance of the first two causes
327 through the variations of recruitment and extinction. Our results reveal that variations within each
328 species range were either through extinction or through recruitment but rarely both (only for *Quer-*
329 *cus pyrenaica* and *Betula*). Generally, the magnitude of the response was stronger for recruitment
330 than for extinction.

331 We found that for all species with a significant relationship between occurrence probability and
332 extinction the relationship was negative, as expected by theory, except for *Pinus nigra* and *Pinus*
333 *pinaster*. We found that for all species with a significant relationship between occurrence probability
334 and recruitment, the relationship was positive as expected by theory. Thus only *Pinus pinaster* and
335 *Pinus nigra* showed a significant response inverse to the theory for extinction, with an increase of the
336 extinction probability with the increase of the occurrence probability (*Pinus halepensis* also showed
337 a positive response but it was not significant). This opposite relationship for species belonging to the
338 genus *Pinus* might be related to their intensive management and frequent plantation outside their
339 native range (particularly in the case of *Pinus pinaster*). Current presence/absence data might be in
340 that case biased to include location outside suitable habitats. The relationships between recruitment
341 and occurrence probability were largely in agreement with the theoretical expectation as only *Abies*
342 *alba* showed a negative but non-significant response.

343 There is no obvious explanation for why species respond through extinction or through recruit-
344 ment. We found no clear explanation based on the species' ecological strategies. There was no link
345 between the slope of the response of recruitment or extinction and shade tolerance (using the shade
346 tolerance index of Niinemets and Valladares (2006)) or key functional traits, see Supplementary
347 Materials Section 7). Thuiller et al. (2014) proposed that shade tolerant species could show a closer
348 relationship between population growth rate and SDM occurrence probability, as this link would be

349 less blurred by competition, but they found, as do we, weak support for this hypothesis. We also
350 verified that the SDM discrimination scores had no direct impact on mean slope estimations.

351 Relatively few other studies have explored with field observation of extinction and colonization
352 if the metapopulation dynamics explain current species distribution (see for instance Talluto et al.
353 (2017), García-Valdes et al. (2013), Araújo, Williams, and Fuller (2002)). These studies generally
354 also supported the idea that metapopulation rates variations agreed with the species distribution,
355 even if there was evidence of extinction debt and colonization credit at the species range (Talluto
356 et al., 2017). Among these studies, the relative importance of extinction and colonization was not
357 explicitly considered. The uncertainty of the estimation seems larger for the response of extinction
358 to climate than for the response of colonization in Talluto et al. 2017, but it is not possible to
359 compare the relative role of these two rates based on their results. Garcia-Valdes et al. 2015 found
360 that climate had a stronger effect on extinction than colonization (whereas we found a stronger
361 response of colonization). Overall there is a lack of studies exploring the relative magnitude of the
362 variation of extinction and colonization within species ranges.

363 These previous studies on extinction-colonization probabilities of trees used patch occupancy
364 models (Talluto et al., 2017; Purves et al., 2007; García-Valdes et al., 2013) with polynomial functions
365 of climatic variables (such as temperature and aridity). A key difference with our model is that we did
366 not use climatic variables directly but instead used the SDM occurrence probability as a descriptor
367 of species environmental niches. Given the low number of colonization or extinction events in our
368 data, using an SDM to summarize the species climatic niche might be more powerful than fitting
369 complex multivariate responses to climatic variables. A similar approach has also been developed
370 for birds in Britain by Araújo et al. (2002), and highlighted a negative relationship between local
371 extinction probability and the occurrence probability.

372 More studies have focused on links between demographic rates and distribution. The links
373 between species distribution and demographic rates (growth rate and carrying capacity by Thuiller et
374 al. (2014), population growth rate, time to quasi-extinction, transient population dynamics by Csergő
375 et al. (2017)) seem weaker than with metapopulation rates. This might indicate that links between
376 population processes and species distribution are easier to capture with integrative metapopulation
377 metrics than with detailed population-level metrics, which could be related to an issue of observation
378 scale: colonization/extinction is a direct dynamical approach to presence/absence. In addition,
379 upscaling individual demographic dynamics to presence/absence is not easy, due to non-linearity of
380 demographic response to climate and temporal variability.

381 **4.2 Implication for the probability of presence at equilibrium**

382 Based on our analysis combining extinction and colonization to estimate the probability of presence
383 at equilibrium, we found that in general the probability of presence at equilibrium was positively
384 correlated to the occurrence probability estimated by the SDM, but with a strong overestimation.
385 This was the case with both equilibrium formulations (open and closed). Thus, our estimates of
386 extinction and colonization rates capture some drivers of the species environmental distribution but
387 were not able to represent the current observed distribution. This agrees well with the previous
388 patch occupancy model fitted to forest inventory data (Talluto et al., 2017; García-Valdes et al.,

389 2013; García-Valdés et al., 2015) who also found that models were capturing part of the species
390 range but with important deviations.

391 The two equilibrium formulations represent two extremes of the effect of seed dispersal. In the
392 case of the closed formulation, the seed source outside the cell is not taken into account in the
393 calculation. In that case, we see that our model would predict an increase in prevalence of most
394 species, which can be related to the higher probability of recruitment compared to extinction across
395 the occurrence gradient. In the open formulation, where seed input inside the cell is considered
396 fixed and not affected by the metapopulation dynamics, we also found a strong overestimation of
397 the occurrence probability from the SDMs. The most extreme case was *Quercus ilex* which showed
398 strong overestimation of the mismatch between current and equilibrium probability of presence and
399 very little variations with the open formulation. The only version of the model that did not strongly
400 overestimate the probability of presence was the open formulation where the seed source varied
401 according to the observed pattern within the species range. This latter formulation is strongly
402 constrained the model by the current geographical distribution and provided little understanding of
403 the mechanisms involved. This model might capture part of the last route to range limits proposed
404 by Holt (Holt et al., 2005) because it explicitly took into account the proportion of forest/non-
405 forest patches. However, a proper interpretation of these results would require to formally represent
406 dispersal processes which was not possible in this study (see the discussion on this issue in the section
407 on the limitations of patch occupancy models below).

408 The overestimation of the equilibrium probability of presence can arise because (1) the metapop-
409 ulation processes are not in agreement with the current distribution and show some degree of non-
410 equilibrium, or (2) our estimation of metapopulation dynamics and the colonization and extinction
411 rate are not accurate enough. Below we discuss these two possible explanations.

412 4.3 Equilibrium vs. non-equilibrium of species distribution

413 If the distribution of a species was currently in equilibrium, we would expect a close match between
414 the SDM and the probability of presence computed at equilibrium (due to either extinction, colo-
415 nization, or both). It is important to note that equilibrium does not necessarily imply that both
416 extinction and colonization processes are strongly related to SDM (see section 8 in Supplementary
417 materials).

418 The fact that we are observing a positive correlation but not a perfect one to one relationship,
419 however, does not rule out that there may be some degree of non-equilibrium between the metapop-
420 ulation dynamics and the current distribution. The idea that each species is in current equilibrium
421 with the environment has been criticized by Svenning and Skov (2004), based on the idea that most
422 European tree species do not fully fill their potential ranges. This situation could be the result of
423 a post-glacial migration lag as illustrated in Svenning, Normand, and Kageyama (2008). The lag
424 would strongly affect *Abies alba*, the *Pinus* genus and the *Quercus* genus. This argument has how-
425 ever been partly contradicted by previous SDM results (Araújo & Pearson, 2005) and large dispersal
426 rates found based on pollen records (Giesecke, Brewer, Finsinger, Leydet, & Bradshaw, 2017). In-
427 terestingly, we found a weak response of recruitment and extinction to SDM occurrence probability
428 for *Abies alba*, a species with a recorded slow expansion rate (Giesecke et al., 2017). In Eastern

429 North America, results from a SPOM model (Talluto et al., 2017) identified species out of equi-
430 librium with climate at range margins. Their model formulation is close to our closed formulation
431 (a species is present at equilibrium when colonization probability exceeds extinction probability).
432 A SPOM developed by García-Valdes et al. (2013) in Spain also concluded on a non-equilibrium.
433 Their simulations based on a model with constant climatic conditions lead to an increased fraction
434 of occupied plots, but most species did not show range expansion.

435 It is important to stress that our analysis focused on testing whether metapopulation dynamics
436 (colonization and extinction) were related to the SDM occurrence probability. Because we consid-
437 ered only a single gradient of occurrence probability we can not distinguish between changes at
438 the southern or northern range and thus can not give an indication of directional range shift in
439 comparison to patch occupancy model fitted with climatic variables.

440 4.4 Limitations of patch occupancy models

441 Several factors might have contributed to limit our ability to estimate the links between SDM
442 and metapopulation dynamics and thus explain the mismatch between the equilibrium probability
443 of presence and the SDM. First, the NFI data do not provide perfect informations on the ab-
444 sence/presence at the plot scale. With protocols that are based on concentric circular subplots for
445 different size classes, we might miss the presence of trees larger than 10 cm DBH in one of the
446 subplots. We partially corrected this issue, by accounting for the probability that a tree was below
447 10 cm at the first census with a growth model. But this approach is not perfect and the data set
448 probably still contains colonization events that are not true colonization events but observation er-
449 rors. Conversely, we might have wrongfully excluded some colonization events for trees with extreme
450 growth. Using detailed recruitment data could improve our estimation, but they are not available
451 for all NFI.

452 Another limitation is that our model did not explicitly consider dispersal. Different studies on
453 patch occupancy models calibrated with NFI data have tried to formally include dispersal in the
454 model (Purves et al., 2007; García-Valdés et al., 2015). García-Valdes et al. (2013) tried to infer the
455 parameters of the dispersal kernel based on the Spanish forest inventory data. We considered that
456 available knowledge on the potential seed source surrounding a plot is insufficient to draw mechanistic
457 conclusions on seed dispersal. Field studies show that mean distances of seed dispersal are short for
458 most tree species (Nathan et al., 2001; Bullock et al., 2017; Cain, Milligan, & Strand, 2000), therefore
459 direct dispersal between plots should be restricted to extremely rare events (distance > 1 km) and
460 the tail of the kernel distribution. It is thus very unlikely that these models were really estimating a
461 dispersal kernel (as indicated by the very large mean dispersal distance inferred) but rather captured
462 a degree of spatial auto-correlation in the species distribution and the recruitment process. Here,
463 we use an estimate of local frequency which takes into account observed presence/absence and
464 smoothing kernels as well as fine scale forest cover maps (building on the approach of Talluto et al.
465 (2017)). We believe that if we want to include dispersal kernels in the model it is better to use
466 external information on the shape and parameters of the dispersal kernel and have more accurate
467 data on the seed source (see Schurr et al. (2007) or Schurr et al. (2012) for example).

468 Finally, our approach does not include biotic interactions and disturbances that might influence

469 population extinction and recruitment probabilities (Case, Holt, McPeck, & Keitt, 2005; Svenning
470 et al., 2014; Liang, Duveneck, Gustafson, Serra-Diaz, & Thompson, 2018). Trophic interactions
471 in a broader sense may have a potentially large impact on recruitment estimation. For instance
472 ungulate browsing may induce spatially varying limitations on recruitment for *Abies alba* (Kupfer-
473 schmid, 2018), and ungulate preferences could lead to limitation of certain species (see e.g. stronger
474 preference for *Pinus sylvestris* over *Pinus nigra* might reinforce *Pinus sylvestris* drought sensitivity,
475 Herrero, Zamora, Castro, and Hódar (2012)). Given the small number of colonization and extinction
476 events, a reliable estimate of tree species interactions with our data seems unrealistic.

477 5 Conclusion and perspectives

478 Several range dynamic models have already used SDMs to constrain metapopulation dynamics based
479 on the assumption that occurrence probabilities derived from SDMs can be used as predictors of
480 colonization and extinction rates (including range dynamics models and population viability analy-
481 sis). Based on this assumption, SDM outputs are used either directly to define which grid cells are
482 colonizable (see Engler and Guisan (2009)), or influence demographic information (Nenzén, Swab,
483 Keith, & Araújo, 2012). Here we test this core assumption for 17 European tree species and found
484 mixed support. At least one process, either colonization or extinction was related to the SDM, but
485 generally not both and the match was far from perfect. We thus caution that models cannot simply
486 assume that metapopulation dynamics is driven by SDM occurrence probability, but rather need to
487 test which processes are affected and at which magnitude. Data driven patch occupancy models have
488 the potential to go beyond criticized SDM correlative predictive approaches (Journé, Barnagaud,
489 Bernard, Crochet, & Morin, 2019).

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