

1 **Refining predictions of population decline at species' rear edges**

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26 **Abstract**

27 According to broad-scale application of biogeographical theory, widespread retractions of
28 species' rear edges should be seen in response to ongoing climate change. This prediction
29 rests on the assumption that rear edge populations are 'marginal' since they occur at the limit
30 of the species' ecological tolerance and are expected to decline in performance as climate
31 warming pushes them to extirpation. However, conflicts between observations and
32 predictions are increasingly accumulating and little progress has been made in explaining this
33 disparity. We argue that a revision of the concept of marginality is necessary, together with
34 explicit testing of population decline, which is increasingly possible as data availability
35 improves. Such action should be based on taking the population perspective across a species'
36 rear edge, encompassing the ecological, geographical and genetic dimensions of marginality.
37 Refining our understanding of rear edge populations is essential to advance our ability to
38 monitor, predict and plan for the impacts of environmental change on species range
39 dynamics.

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51 **Introduction**

52 Climate change impacts species performance and distribution across the globe (Parmesan &
53 Yohe, 2003). Biogeographical theory suggests that rising global temperatures should drive
54 species to move poleward and upward in elevation as they track the climates to which they
55 are adapted. Therefore, it is reasonable to expect that population loss and range retractions
56 should be seen in the most low-latitude, drought-prone areas of a species' distribution (the
57 rear edge, Hampe & Petit, 2005), given that widespread climate-driven extinction has been
58 predicted (Thomas et al., 2004; Urban, 2015). However, assumptions of declining rear edge
59 population performance are a long-lasting legacy of uncritical application of the centre-
60 periphery hypothesis (Brown, 1984; Safriel, Volis, & Kark, 1994). This prediction assumes
61 that rear edge populations are fundamentally at higher risk of extinction than those
62 populations at the core of the species' range. This elevated extinction risk is attributed to the
63 expectation that they occur in less favourable climates (or habitats) and are more at risk from
64 demographic stochasticity because of lower and highly variable population sizes.
65 Consequently, widespread 'marginality' is predicted at the species' rear edge, i.e. decreased
66 population performance because populations occur at the limits of the species' physiological
67 and ecological tolerance.

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69 The assumption of rear edge population decline in response to climate change appears well
70 supported in the literature (e.g. Allen et al., 2010; Carnicer et al., 2011; Feeley et al., 2011;
71 Lesica & Crone, 2016; Marqués, Camarero, Gazol, & Zavala, 2016; Reich et al., 2015).
72 However, such support is often derived from an amalgamation of case-studies of decline,
73 risking inaccurate predictions when attempting to extrapolate regionally across the rear edge
74 of a species distribution. 'Marginality' at the population level is determined by the interaction
75 of a variety of constraints, including climate and local-scale environmental conditions, habitat

76 fragmentation, species traits, physiology and biotic interactions, as well as population
77 demography and genetics. At the same time, anthropogenic land-use changes shape how
78 species are distributed, and their legacies strongly influence population dynamics. All
79 together result in ecological and evolutionary mechanisms that are dependent upon far more
80 than the biogeographical location of a population (Hampe & Petit, 2005; Pironon et al., 2016;
81 Sexton, McIntyre, Angert, & Rice, 2009). Consequently, conflicts between predictions and
82 observed population responses are increasingly accumulating (e.g. Bertrand et al., 2011;
83 Cavin & Jump, 2017; Doak & Morris, 2010; Granda et al., 2018; Rabasa et al., 2013;
84 Rapacciuolo et al., 2014). Here we examine the potential reasons for this disparity by
85 decomposing the causes of marginality and discuss why simplifying assumptions on
86 marginality have implications for predicting species' range shifts. We propose a generally
87 applicable rationale for research design and analysis to better integrate population-level
88 responses into a biogeographical context of species decline. Our focus is on plant – and
89 especially tree – species because of the abundance of data available and the key roles forests
90 play in global carbon and hydrological cycles and maintaining biodiversity. We argue that, as
91 data availability increases, greater emphasis should be placed on recognising the scale-
92 dependency of the factors determining population dynamics, which is fundamental in highly
93 heterogeneous regions like the rear edges, where global change is strongly altering the
94 structure and function of forest ecosystems.

95

96 **Empirical evidence in agreement with biogeographical theory**

97 A broad range of studies in the literature provides empirical evidence of declining rear edge
98 populations relative to those of the range-core or across low-altitude relative to high-altitude
99 areas in concordance with biogeographical predictions. For example, sudden population
100 mortality associated with elevated drought stress at species rear edges has been observed in

101 forest ecosystems across the globe (Allen et al., 2010). Equally, evidence of population
102 decline that heralds range retractions is often provided by dendroecological approaches. For
103 example, Scots pine (*Pinus sylvestris*) forests in the Gúdar range (southern Iberian Range,
104 Iberian Peninsula) are representative populations of the species' rear edge. The species occurs
105 in a mountainous orography, where low-altitude, dry-edge populations coexist with a more
106 drought-tolerant pine species, the black pine (*Pinus nigra* subsp. *salzmannii*). In accordance
107 with biogeographical predictions, Scots pine growth is enhanced by temperature at mid- and
108 upper elevations, and constrained because of enhanced drought stress at low-elevations. In
109 these low-altitude areas, where both species co-occur, black pine is more resilient than Scots
110 pine to extreme drought events, suggesting that future changes in species composition are
111 likely (Marqués et al., 2016). Experimental evidence of species' responses to climate
112 manipulation also supports biogeographical predictions. For example, *in situ* experimental
113 warming in northern Minnesota, North America, showed reductions in photosynthesis and
114 growth near warm range limits and increases near cold range limits in juvenile trees of 11
115 boreal and temperate forest species (Reich et al., 2015). Species' range shifts predicted by
116 biogeographical theory have been observed in biodiversity hotspots like the Tropical Andes.
117 Elevational shifts during a 4-year period were assessed for 38 tree genera across an
118 elevational gradient from 950 to 3400 m in Manu National Park in south-eastern Peru. Mean
119 migration rate was 2.5–3.5 vertical metres upslope per year and low-elevation genera also
120 increased in abundance in most of the study plots. However, the rate of elevational migration
121 was lower than predicted according to the temperature increase in the region, suggesting a
122 lagged response to climate change of primary tropical montane forests (Feeley et al., 2011).

123

124 **Why disparities between biogeographical theory and population ecology matter**

125 Four complementary explanations drawn from empirical evidence clarify why rear edge
126 population performance can deviate from biogeographical predictions:

127 *(i) Geographical and ecological edges do not always overlap at the population scale*

128 Assuming a complete overlap of geographical and ecological range limits at the rear edge of
129 a species' distribution may explain counterintuitive population responses. For example,
130 decline in the abundance of plant species with an arctic-alpine and boreal distribution across
131 western North America has been observed across rear edge populations occurring in the
132 northern Rocky Mountains. Although the overall trend of species' abundance decline is in
133 agreement with biogeographical predictions, 50% of monitored populations remained stable
134 or even increased in abundance (Lesica & Crone, 2016). Therefore, decreased population
135 performance at rear edges cannot be assumed because ecological and geographical range
136 margins do not always overlap.

137 *(ii) Interactions among ecological factors determine population dynamics*

138 Species distributions and population dynamics are determined by complex interactions of
139 ecological factors (Harper, 1977). For example, soil phosphorus strongly limits tropical tree
140 distributions along a gradient of dry-season moisture along the Panama Canal (Condit,
141 Engelbrecht, Pino, Pérez, & Turner, 2013) and, in Mediterranean communities, several plant
142 species only survive at the drier edge of their ranges in communities beneath the facilitative
143 effects of the shrub "retama amarilla" (*Retama sphaerocarpa*) (Armas, Rodríguez-
144 Echeverría, & Pugnaire, 2011). However, such complexity is typically simplified in large-
145 scale studies because of methodological limitations when trying to represent population-level
146 processes over broader spatial scales. Consequently, disparities between population responses
147 and biogeographical predictions are likely to be common. For example, elevational range
148 shifts inferred from adult and juvenile abundance in Mediterranean, temperate and boreal tree
149 species in Europe are idiosyncratic rather than consistent with temperature-based predictions

150 (Rabasa et al., 2013). Similarly, downslope shifts in elevation are as common as upslope
151 shifts across a broad range of taxa in California (Rapacciuolo et al., 2014). Common
152 explanations for these unexpected responses are factors such as human land-use, water
153 balance or soil quality, species physiological and dispersal traits, demographic dynamics and
154 biotic interactions (Rabasa et al., 2013; Rapacciuolo et al., 2014).

155 *(iii) Decoupling between microclimates and macroclimates*

156 Large-scale predictions from bioclimatic models are generally derived from coarse gridded
157 climatic data because fine-resolution or microclimatic data are rarely available over large
158 spatial scales. Organisms, however, respond to their local environment. For instance,
159 microclimatic variation due to topographic factors is generally not captured by the resolution
160 of interpolated climatic data while differences between regional free-air and local
161 temperatures may amount to several degrees (Dobrowski, 2011). At finer scales, biophysical
162 processes have impressive effects. For example, structural characteristics of old-growth
163 forests may provide microclimates cooler by as much as 2.5°C across forest stands (Frey et
164 al., 2016). Therefore, it is not surprising that climate at resolution of 100 or more meters
165 poorly explains variation of leaf and wood traits across populations of temperate and
166 Mediterranean trees (Vilà-Cabrera, Martínez-Vilalta, & Retana, 2015). In the context of
167 marginality, a highly illustrative example of mismatch between micro- and macroclimates is
168 the persistence of rear edge populations such as the stands of pedunculated oak (*Quercus*
169 *robur* L.) in Jerte valley, western Iberian Peninsula (Moracho, Moreno, Jordano, & Hampe,
170 2016) which has a regional climate significantly hotter and dryer than that tolerated by this
171 species. Consequently, a decoupling between micro- and macroclimates has strong
172 implications for climate-based predictions on population decline (Hampe & Jump, 2011).

173 *(iv) Evolutionary processes*

174 Populations (or genotypes) are adapted to a specific range of ecological conditions and,
175 consequently, each individual within a species may experience stress from climate change
176 (Harte, Ostling, Green, & Kinzig, 2004). Therefore, the existence (or lack) of genetic
177 adaptations to climatic stress may also explain some of the former unexpected responses. For
178 example, greenhouse experiments show that dry-edge populations of the spurge olive
179 (*Cneorum tricoccon*), a Mediterranean evergreen shrub with a narrow distribution, exhibit
180 more drought-tolerant phenotypes, and growth of individuals inhabiting drier habitats is less
181 affected by drought stress (Lázaro-Nogal et al., 2016). However, most empirical evidence on
182 spatial variation of key species traits comes from observations across broad latitudinal
183 gradients. For example, rear edge populations of the European beech tree show higher
184 resistance to xylem embolism relative to mid-latitude, range-core populations (Stojnić et al.,
185 2018). Yet, a proper understanding on whether variation in this and other traits relevant for
186 species persistence occurs across rear edge populations is lacking.

187

188 The former explanations point to two subtly interrelated aspects that, if not acknowledged,
189 strongly limit our understanding of marginality, and our ability to predict population loss.
190 First, marginality is a multidimensional property of populations that encompasses ecological,
191 geographical, and genetic components. Second, methodological limitations and lack of data
192 restrict our capacity to link population ecology with biogeography (but see SDMs accounting
193 for phenotypic plasticity and local adaptation in Benito Garzón, Robson, & Hampe, 2019).
194 Consequently, local predictions of rear edge decline only based on distribution patterns at the
195 regional scale become unrealistic (Thuiller et al., 2008). Overcoming such limitations is
196 essential to reconcile population ecology with biogeographical theory at species' rear edges
197 to enable a predictive understanding of their dynamics, function and management (Mouquet
198 et al., 2015).

199

200 **Refining our predictive understanding of rear edge population decline**

201 We propose a rationale that integrates the ecological, geographical and genetic dimensions of
202 marginality to determine the regional- and local-scale mechanisms shaping the probability of
203 persistence (or extinction) of rear edge populations (Figure 1). Importantly, the scale-
204 dependency of ecological mechanisms influencing the persistence probability of populations
205 may result in contrasting predictions between the regional and local scales. Consequently, we
206 argue that a hypothesis-driven approach is necessary, with population decline tested rather
207 than assumed according to predicted marginality. At the core of the rationale lies a data-
208 driven methodology that permits the incorporation of increasingly available data sources into
209 experimental study design. Essentially, each marginality dimension can be inferred from
210 multiple ecological components (e.g. climatic range, landscape connectivity, community
211 composition, human-driven habitat degradation, etc.) across the species' rear edge. The
212 distribution and edges of these components and their interactions can be identified and
213 populations categorized across marginality types (Figure 2A) ensuring that, at the regional
214 scale, the entire rear edge structure is represented (Figure 1). At the same time, population
215 and individual parameters need to be measured with replication within- and compared across
216 marginality types to ensure a balanced sampling and accurate parameter assessment (Figure
217 2B). Observed population responses are then contrasted with regional-level predictions and,
218 if disparities arise, local-scale mechanisms need to be considered (Figure 2B). We
219 demonstrate how application of this rationale improves understanding of marginality and
220 highlights the need to consider the scale-dependency of ecological suitability.

221

222 **(i) Conceptualising the dimensions of marginality**

223 Our understanding of marginality as a multidimensional concept, the rear edge structure, as
224 well as the regional- and local-level hypotheses of population decline are illustrated in Figure
225 1. In analogy with the limits of the realized niche (Hutchinson, 1957), abiotic and biotic
226 factors define ecological marginality at the regional and local scales. The regional climate (or
227 macroclimate) of the population location relative to the edge of the species' climatic
228 distribution (or the threshold of species' climatic tolerance) is used to infer ecological
229 marginality at the regional scale, while the range of population-scale habitat characteristics
230 (e.g. microclimate, soil quality, land-use history) is used to derive local ecological
231 marginality. Population decline is thus predicted to occur at the extremes of these factors, e.g.
232 drier climates, poor soils or intense disturbance. Rear edge populations occur along
233 bioclimatic transition zones (Jump, Mátyás, & Peñuelas, 2009), where species climatic
234 suitability decreases and habitat heterogeneity is high over small spatial scales. Consequently,
235 changes in the composition of communities can occur abruptly with shifts in habitat quality
236 such that community composition can be used alongside abiotic conditions to infer ecological
237 marginality. At the landscape scale, the composition of communities surrounding the focal
238 rear edge population is used to infer regional-scale ecological marginality, which increases
239 approaching the transition between bioclimatic zones. At the local scale, the community
240 composition is used to infer interactions among organisms – within or across trophic levels –
241 potentially determining ecological marginality. If co-occurring species, relative to the focal
242 one, are competitors under an ecological advantage (e.g. drought-tolerant) or antagonists (e.g.
243 biotic agents), such biotic interactions result in increased local ecological marginality.
244 Contrary, biotic interactions result in decreased local ecological marginality if beneficial
245 effects can emerge from species coexistence (e.g. facilitation, mutualism, or
246 complementarity).

247

248 The rear edge is typically made up of populations of variable size and connectivity, defining a
249 fragmented landscape (Hampe & Petit, 2005; Jump et al., 2009). Therefore, the spatial
250 distribution, size and connectivity of populations (i.e. habitat configuration) are used to infer
251 regional-scale geographical (and genetic) marginality. Increased fragmentation and isolation
252 as a consequence of either natural processes or anthropogenic impacts, result in decreased
253 population performance. This detrimental effect is associated with an altered habitat leading
254 to edge effects (Murcia, 1995), increased metapopulation dynamics due to dispersal
255 limitation (Hanski, 1991), disrupted biotic networks and novel interactions or invasion
256 (Hagen et al., 2012), and the loss of genetic variation and individual fitness because of
257 increased chance of genetic drift and inbreeding (Templeton, Shaw, Routman, & Davis,
258 1990). However, in parallel with deviation of local ecological conditions from the regional
259 scale, population responses that are the product of local-scale mechanisms (e.g. local
260 adaptation) or biotic interactions (e.g. mutualistic symbioses) may contradict predicted
261 marginality based on habitat configuration alone.

262

263 **(ii) Quantifying marginality and testing regional-scale hypotheses of population decline**

264 Marginality can be quantified along multiple axes at the regional scale using existing data
265 sources, allowing hypothesis-testing on the regional mechanisms determining population
266 decline (Figure 1). Climatic and geographic range-edges may not completely overlap (Cavin
267 & Jump, 2017; Chardon, Cornwell, Flint, Flint, & Ackerly, 2015). Consequently, while
268 geographical ranges frequently correlate with climate at the continental scale, it cannot be
269 assumed that all rear edge populations are climatically limited. This idea can be understood,
270 for example, from the variable relationship between the climatic characteristics and
271 geographical location of populations of the European beech (*Fagus sylvatica* L.) tree from the
272 Iberian Peninsula to Northern Scotland. Populations inhabiting dry and wet sites relative to

273 the species' climatic distribution can be found at the rear edge with contrasting implications
274 for population performance (Cavin & Jump, 2017). Large-scale forest inventories or remotely
275 sensed data layers such as land-cover maps can be used to determine geographical
276 marginality, with gridded climate data used to infer ecological marginality relative to the
277 climatic distribution of the species (Figure 2A). The interaction between both types of
278 marginality results in variable predicted extinction risk across the rear edge (Figure 1).

279

280 At rear edges, abrupt bioclimatic transitions may not be explained by climate alone. For
281 example, the pine–cloud forest ecotone on the windward slopes of the Cordillera Central,
282 Dominican Republic, is primarily a result of high-elevation fire regimes. Declining
283 temperature and precipitation with elevation together with trade wind inversion, and small-
284 scale variation in topography and vegetation determine fire occurrence and ecotone formation
285 (Martin, Sherman, & Fahey, 2007). Existing data sources that incorporate species
286 composition data (e.g. inventories and land-cover maps) can be used to infer bioclimatic
287 transitions at the landscape scale, and thus refine predictions on ecological marginality based
288 on climate alone (Figure 2A; Figure 1). This idea can be exemplified by the exceptional range
289 retraction of ponderosa pine (*Pinus ponderosa*) after a severe drought in mid-1950s at the
290 ecotone between this species and piñon–juniper woodland (*Pinus edulis* and *Juniperus*
291 *monosperma*) in northern New Mexico (Allen & Breshears, 1998). Forest dieback
292 predominantly concentrated in low-altitude, drought-prone populations, but more climatically
293 favourable areas along the entire altitudinal gradient were also affected likely because of a
294 competitive disadvantage relative to more drought-tolerant species. The interaction between
295 climate and community composition at the regional scale reflects a mosaic of ecological
296 conditions at rear edges not only dependent on climate (Figure 1), and should, therefore, be
297 incorporated into empirical study design (Figure 2A).

298

299 Populations at similar levels of ecological marginality are at higher risk of extinction with
300 increasing geographical (and genetic) marginality at the regional-scale (Figure 1). Spatial-
301 pattern and landscape-connectivity GIS analyses (e.g. Wegmann et al., 2018) on land-cover
302 maps and other remote-sensing derived-sources can be used to accurately infer habitat
303 configuration and test predictions of decreased population performance (Figure 2A).
304 Population fragmentation is associated with ecological edge effects (Murcia, 1995). For
305 example, in tropical montane forests in the Bolivian Andes, temperature gradients from the
306 edge to the interior of forest patches are equivalent to a 100-m shift in elevation. Higher
307 temperatures at forest edges cause warmer and drier habitats with corresponding elevation of
308 drought stress, changes in species composition and increased fire risk (Lippok et al., 2014).
309 Fragmentation may also strongly decrease individual fitness and alter population dynamics
310 through rapid genetic changes. For example, loss of large-vertebrate dispersers because of
311 human-driven habitat fragmentation across Brazilian Atlantic rainforests is associated with a
312 rapid (< 100 years) evolutionary seed size reduction in a keystone palm species (*Euterpe*
313 *edulis*). Seed size reduction results in increased seed vulnerability to desiccation and
314 decreased seedling growth. At the same time, genetic diversity among seedlings in
315 fragmented (defaunated) sites is lower than in non-fragmented sites. Altogether, these
316 impacts have strong implications for population dynamics under predicted drier conditions in
317 the studied forests (Carvalho, Galetti, Colevatti, & Jordano, 2016; Galetti et al., 2013).

318

319 **Shifting to the population perspective: refocusing on local-scale hypotheses**

320 Framing hypotheses of population decline based on marginality predicted at the regional-
321 scale can result in disparities between regional predictions and observed population
322 responses. Such disparities demonstrate the need to refocus studies exploring rear edge

323 performance on local-scale hypotheses (Figure 1; Figure 2B). Below we first address the
324 strong influence that anthropogenic land-uses and their legacies have on our understanding of
325 marginality and their likely prominent role to explain the mismatch between predictions and
326 observations. Thereafter, we illustrate with selected examples from the literature how rapidly
327 increasing data availability can be harnessed for the evaluation of local-scale mechanisms
328 across marginality-types (Figure 2B), thereby refining our predictive understanding of rear
329 edges.

330 **(i) Anthropogenic land-uses and their legacies**

331 Anthropogenic land-use during the last few hundred years has altered the realised niche of
332 species and consequently their contemporary distribution is often not in equilibrium with the
333 range of ecological conditions they are able to exploit. For example, using ‘pre-settlement’
334 vegetation estimations inferred from survey records (1830–1910), and historical climate and
335 contemporary data, Goring & Williams (2017) demonstrated that human land conversion
336 shifted the past distribution of some tree genera in Midwestern United States, from drier and
337 warmer climates in the past to wetter and cooler conditions today. Land-use changes and
338 associated habitat modifications, therefore, complicate the identification of ‘ecological edges’
339 of a species’ distribution (Figure 1). Anthropogenic land-use also interacts with climate
340 change impacts on population dynamics. For example, human-driven forest loss prevails in
341 warmer (low-latitude or altitude) regions and, rather than climate change, recent habitat loss –
342 quantified from ~30-m resolution data generated from Landsat image analysis – explains the
343 biotic attrition observed in these areas (Guo, Lenoir, & Bonebrake, 2018). On the other hand,
344 tree species plantations for wood or food production and fire suppression can contribute to
345 species expansion beyond their climatic limits, but increase the risk of dieback episodes and
346 wildfires during extreme dry years (Maranz, 2009; Nowacki & Abrams, 2015; Sánchez-
347 Salguero, Navarro-Cerrillo, Swetnam, & Zavala, 2012). At the same time, socioeconomic

348 changes can lead to widespread forest expansion over abandoned land (Meyfroidt & Lambin,
349 2011). For example, the combination of forest inventory data with historical and modern
350 land-cover maps generated from aerial images shows that the ~25% of current forests in the
351 Iberian Peninsula, the rear edge of several temperate and boreal tree species, are growing on
352 former agricultural and grazing land abandoned after the 1950s (Vilà-Cabrera, Espelta,
353 Vayreda, & Pino, 2017). Consequently, anthropogenic habitat modification and its legacies
354 represent a critical dimension of marginality as they may intensify, confound or delay
355 climate-driven population decline at rear edges.

356 **(ii) Population demography and structure**

357 Forest inventory networks are very useful for assessing recent demographic dynamics over
358 large geographical scales. However, the spatiotemporal resolution and the quantity of data are
359 limited and need to be complemented with more detailed data and studies. Long-term
360 population responses can be better understood taking advantage of the increasing availability
361 of dendroecological data over large geographical areas (e.g. Sánchez-salguero et al., 2017),
362 while field-based investigations can inform on particular persistence mechanisms such as
363 compensatory changes in demographic rates (Doak & Morris, 2010) or stabilising processes
364 (e.g. competition release) after extreme drought events (Lloret, Escudero, Iriondo, Martínez-
365 Vilalta, & Valladares, 2012). However, detailed information on population structural
366 characteristics including human uses needs to be assessed using inventory data and, together
367 with observed population demography, explicitly placed in the context of past management
368 and its legacy. Such characterisation of population structure is essential given that, for
369 example, regular forest management (e.g. thinning) can assist a species to persist under
370 chronic climatic stress (Linares, Camarero, & Carreira, 2009), delaying or even concealing
371 the decline of the species if the less vigorous individuals are removed. However, when forest
372 use is abandoned and the stand matures this beneficial effect can reverse due to greater

373 physiological constraints associated with larger trees (D'Amato, Bradford, Fraver, & Palik,
374 2013). If coupled with long-term acclimation to favourable water availability, such structural
375 shifts (i.e. bigger stems and higher leaf area) may lead to greater demand of water resources
376 that are not available during extreme drought (Jump et al., 2017), resulting in increased
377 population decline even across better-quality habitats (Figure 1).

378 **(iii) Local-scale environmental conditions**

379 Rear edges mostly occur within areas of high habitat heterogeneity at small spatial scales
380 (Hampe & Petit, 2005). Micro-topography is an important driver of small-scale variation in
381 habitat quality, and it can be modelled from existing data such as high-resolution digital
382 elevation models (DEM) derived from remote sensing. For example, Adams et al. (2014)
383 used 1-m resolution DEM to show how micro-topographic control on moisture conditions
384 mediates tree growth and water-use responses to drought near the elevational range-limits of
385 lodgepole pine (*Pinus contorta*) and ponderosa pine (*Pinus ponderosa*) in the Gordon Gulch
386 catchment, Colorado. Such topographic variability together with a range of other physical
387 (e.g. lithology, edaphic characteristics) and biophysical factors (e.g. vegetation structure and
388 traits) facilitates the existence of microrefugia (Figure 1; McLaughlin et al., 2017). For
389 instance, rock outcrops and associated habitat can create microclimates 4.9 °C cooler, 12%
390 wetter, and less variable than the climate of the surrounding habitat. This microclimate is
391 associated to the persistence of a rear edge population of *Podocarpus lambertii* at the species'
392 drier range-edge located in a semiarid region in Brazil (Locosselli, Cardim, & Ceccantini,
393 2016). Microclimate data can be derived from local networks of climate data loggers and
394 combined with remotely sensed topographical and vegetation structural data. Improvements
395 in data resolution are essential in highly variable regions in terms of habitat conditions, where
396 the potential for microclimatic buffering strongly relies on microrefugia occurrence and
397 human impacts on habitat structure. For example, along a land-use intensity gradient in

398 Borneo, from unlogged old-growth forests to mature oil-palm plantations, canopy structure
399 and topography are strong drivers of small-scale variation in understory temperature and
400 vapour pressure deficit. Assessing and modelling variation in microclimatic conditions is
401 critical in regions like the lowland tropics, where many species reach their thermal tolerance
402 limits (Jucker et al., 2018).

403 **(iv) Biotic interactions**

404 Alterations to species coexistence can reflect an altered habitat, for example, such that more
405 drought- and shade-tolerant species gain a competitive advantage. For example, the local
406 coexistence between the boreal pine species Scots pine (*Pinus sylvestris*) and Mediterranean
407 oak species (e.g. *Quercus ilex* and *Q. pubescens*) can be observed along altitudinal gradients
408 in many European mountain systems, such as the Pyrenees. Oak seedling abundance and
409 performance are higher under drought-induced Scots pine decline but this association is not
410 only restricted to the most drought-prone stands at low-altitudes. Habitat deterioration and
411 past species-selective management explain observed community dynamics at the local scale
412 (Galiano, Martínez-Vilalta, Eugenio, Granzow-de la Cerda, & Lloret, 2013). The local
413 community composition can be directly obtained from inventory data or field-based
414 sampling, directly informing on ecological marginality, supporting a better understanding of
415 marginality-type (Figure 1; Figure 2A).

416

417 Large-scale inventories are useful to assess how variation in biotic interactions scale-up over
418 broad geographical areas, for example, those involving antagonistic interactions such as
419 insect and fungal damage on trees (e.g. Carnicer et al., 2011). Although these large-scale
420 analyses are often based on categorical data or species relative abundance, they provide a first
421 identification of the spatial variation in species assemblages and should be used for setting
422 more detailed experiments and studies on relevant biotic interactions. For example,

423 uncommon or novel interactions can be established if climate change or anthropogenic land-
424 uses, like fire suppression, shift the identity of coexisting species. Experimental evidence
425 demonstrates that the performance of populations failing to migrate as temperature increases
426 will be strongly reduced by novel competitors migrating upwards in elevation (Alexander,
427 Diez, & Levine, 2015). Other more complex situations, e.g. coevolution in mutualistic
428 symbioses, need specific approaches but existing information can support hypothesis
429 development and experimental design. For example, the structural characteristics of drought-
430 tolerant, moth-susceptible pinyon pine (*Pinus edulis*) individuals differ from drought-
431 intolerant, moth-resistant ones at the edge of the pine species' physiological tolerance in
432 Northern Arizona. This information supported Gehring et al. (2017) to demonstrate that under
433 drought stress, interactions between plant genotype, resistance to herbivory and mutualistic
434 fungi operate differentially among individuals, providing an interpretation for landscape-scale
435 patterns of population decline. Drought-tolerant, moth-susceptible trees have higher growth
436 and survival than drought-intolerant, moth-resistant ones, and this differential performance
437 correlates with distinct, genetically-based ectomycorrhizal communities.

438 **(v) Population genetics matters but within a context of ecological change**

439 The putative long-term stability of relict populations during Quaternary climatic oscillations
440 – the result of microrefugia occurrence and evolutionary processes (Hampe & Jump, 2011;
441 Hampe & Petit, 2005; Woolbright, Whitham, Gehring, Allan, & Bailey, 2014) – is an
442 excellent example of the mismatch between predictions and observed responses at rear edges
443 (Figure 1). Relict populations reinforce the idea that species' extinction risk depends on the
444 interaction between population genetics and ecology. However, it has long been recognised
445 that negative ecological impacts (e.g. demographic decline, restriction to dispersal, disruption
446 of community dynamics) can often outweigh genetic factors in a context of rapid
447 environmental change (Lande, 1988). Studies addressing questions of genetic marginality

448 primarily need to account for species-specific ecological requirements and demography. For
449 example, along fragmented forests in southern Australia, decreased pollen diversity and
450 increased selfing associate with fragmentation for two insect-pollinated eucalypt tree species,
451 but not for a bird-pollinated one (Breed et al., 2015). Moreover, where fragmentation drives
452 decreased genetic diversity and increased risk of inbreeding, population performance is not
453 necessary reduced if, for instance, functional genetic variation is not altered (Reed &
454 Frankham, 2001), genotypes are adapted to the local habitat (Kawecki, 2008) or the mating
455 system evolves to ensure population viability (Ouayjan & Hampe, 2018). Furthermore, the
456 amount of genetic variation (functional or neutral) and the degree of evolutionary adaptation
457 to a marginal habitat may not matter when rapid environmental change drives abrupt shifts in
458 population demography and increases species' regional extinction risk (Lande, 1988) (Figure
459 1). Consequently, while population genetics can contribute toward refining predictions of rear
460 edge population decline, it should be considered in the context of population ecology, with
461 the focus on variation of functionally relevant phenotypic traits and demographic
462 performance.

463

464 **A population-focused study at the species' rear edge**

465 The European beech (*Fagus sylvatica* L.) tree is drought-sensitive and it is expected to be
466 particularly vulnerable to deteriorating water balance across rear edge populations occurring
467 in the north-eastern Iberian Peninsula. To highlight this approach to experimental design we
468 used different existing data sources: (i) three regional forest inventories (the Ecological and
469 Forest Inventory of Catalonia, the Spanish National Forest Inventory, and the Catalan
470 Inventory of Singular Forests); (ii) an 8 m² resolution land-cover map (Land Cover Map of
471 Catalonia); and (iii) 1 km² resolution gridded layer of the ratio of annual precipitation to
472 potential evapotranspiration derived from the WorldClim database. Using these data, we

473 selected 40 beech populations classified into four main population types according to
474 ecological marginality, based on climate and community composition, and geographical
475 (genetic) marginality, based on plot spatial distribution (Figure 1 and 2). At each location we
476 assessed population decline parameters, i.e. adult mortality and canopy defoliation based on
477 measurements in one point in time (see Supporting information Appendix S1) and tested
478 regional hypotheses on population decline (Figure 1). The direct comparison among
479 marginality types provides evidence on two fundamental aspects. First, population decline
480 seems to be occurring regionally but especially across ecologically marginal areas within the
481 continuous range (Figure 3A), rather than at geographical edges where population extinction
482 is first predicted to occur. Second, isolated populations inhabiting marginal habitats show
483 lower levels of mortality and canopy decline than expected, which also are comparable to
484 those observed in populations occurring across better-quality habitats. This mismatch
485 between predictions and local observation is consistent with recent evidence showing high
486 stability of rear edge beech populations (Cavin & Jump, 2017; Hackett-Pain & Friend, 2017;
487 Stojnić et al., 2018).

488

489 We also show that differences across populations are mediated by the variability of decline
490 along gradients resulting from interactions among marginality dimensions (Figure 3B). First,
491 fragmentation and climate interact to explain patterns of population decline, evidencing
492 regional population loss and local population retention. Second, climate and landscape-scale
493 community composition interact to explain trends in population decline that might seem
494 counterintuitive based on the effects of the dimensions separately. Broadly, mortality
495 increases while approaching the transition area between bioclimates (i.e. from temperate to
496 Mediterranean) across populations located in relatively wet habitats and, to the contrary, it
497 decreases while approaching the transition area between bioclimates across populations

498 located in dry habitats, with a trend from continuous-range to isolated populations (Figure
499 3B). All together, these results provide evidence on three main aspects. First, the mosaic of
500 ecological conditions at the species' rear edge where climate alone cannot explain population
501 responses. Second, the putative persistence of some relict populations across the species' rear
502 edge. Third, the uneven but predictable pattern of population decline across populations, that
503 can occur also in better-quality habitats.

504

505 This simple study-case application demonstrates that some disparities between predictions
506 and observations can be reconciled accounting for simple interactions among marginality
507 components, and that the potential scale-dependency of the mechanisms involved in
508 population decline is a critical issue for modelling species distributions and regional
509 biodiversity patterns at rear edges (Figure 1). By incorporating existing data sources to better
510 infer the ecological structure of species rear edges through marginality-type classification and
511 taking a hypothesis-driven approach, the rationale provided is flexible enough to be
512 applicable to field-based approaches, *in situ* or controlled-condition experimentation,
513 population genetic studies and approaches accounting for land-use changes, and allows better
514 integration of population ecology and biogeography.

515

516 **Conclusions**

517 Taking the population perspective on marginality is challenging for empirical studies yet it is
518 both possible and essential for our understanding of rear edge dynamics. It is of primary
519 importance to determine interactions among ecological mechanisms driving population
520 decline and the influence of anthropogenic land-use. Similarly, scaling-up the complexity of
521 marginality to broader scales presents a critical challenge for biogeographical studies. The
522 problem of data resolution driving a mismatch between regional predictions and local

523 observations can be improved as data availability increases, which is critical to plan for
524 climate change impacts. For example, if management and conservation decisions are to be
525 based on predictions and the actions implemented ‘locally’, we must know the spatial
526 resolution of data that is needed to accurately predict rear edge dynamics. At the same time,
527 data availability is distributed unevenly across spatial scales, systems and world regions, with
528 regional scales, plant species and the Northern Hemisphere over-represented. Local
529 environmental monitoring is essential to avoid scale-dependent hazards, and large-scale and
530 systematic sampling protocols in the Southern Hemisphere and across taxa other than plants
531 are needed. Increasingly, application of remote sensing methodologies and modelling can
532 help fill data gaps, although ground truth data are still required. Importantly, the rationale
533 presented allows the incorporation of other marginality dimensions not considered here. For
534 example, it is critical to account for biological invasions, including novel competitors and
535 pathogens, or nitrogen deposition and nutrient limitation. Such progress is essential to better
536 understand and predict the impacts of a warming climate and how it interacts with other
537 environmental changes to drive population retention or loss at species’ rear edges.

538

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548

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798 **Figure captions**

799 **Figure 1. Conceptual representation of the structure of species' rear edges and**
800 **persistence probability of populations.** Marginality and the interactions among its
801 dimensions, together with the regional- and local-level hypotheses on population decline are
802 represented. Regional-level predictions: (i) the geographical edge (horizontal dashed line)
803 represents the threshold between continuous range and isolated populations. Geographical
804 (and genetic) marginality are higher with increasing fragmentation and population isolation;
805 (ii) the climatic edge (vertical continuous line) represents the threshold of species' climatic
806 tolerance. Ecological marginality is higher below this threshold. The direction of the line
807 (bottom-right to top-left) represents higher abundance below the climatic edge in isolated
808 populations relative to continuous range populations; (iii) the ecological edge (vertical dashed
809 line) represents the threshold of species' ecological tolerance and a bioclimatic transition. It is
810 defined by the interaction between the climatic edge and the community composition at the
811 regional and/or local scale. Ecological marginality is higher below this threshold. The
812 direction of the line (bottom-right to top-left) represents higher population abundance below
813 the ecological edge in isolated populations relative to continuous range populations. Local-
814 level predictions: the persistence probability may be higher or lower than expected at the
815 regional scale because of population-level mechanisms. For a detailed description of
816 mechanisms and examples, see section *Shifting to the population perspective: refocusing on*
817 *local-scale hypotheses*.

818 **Figure 2. Guidelines for empirical study design.** (A) The distribution of marginality
819 dimensions can be inferred from existing data sources (e.g. macroclimate, habitat
820 configuration, community composition). The position of populations relative to the
821 geographical, climatic and ecological edges is used to classify them into marginality-types
822 according to the criteria of the flow diagram shown. The ecological edge results from the

823 interaction between the climatic edge and the community composition at the regional and/or
824 local scale. The interaction between ecological marginality and geographical (and genetic)
825 marginality results in four main marginality-types (see also Figure 1). (B) Population decline
826 can be tested according to the predicted marginality-types, based on a balanced experimental
827 design. Population/individual parameters need to be measured and regional-level hypotheses
828 tested. Disparities between observed population responses and regional-scale predictions
829 indicate that local-scale hypotheses need to be considered. For a practical application of this
830 guidelines see section *A population-focused study at the species' rear edge*.

831 **Figure 3. Population decline of the European beech tree across marginality types and**
832 **gradients.** (A) Tree mortality and canopy decline as a function of the four population
833 marginality-types that result from the interaction between geographical (genetic) marginality
834 (isolated/continuous range) and ecological marginality (ecologically-marginal/non-
835 ecologically marginal); (B) population mortality across the gradients related to interactions
836 between (i) climate (water balance expressed as the ratio of annual precipitation to potential
837 evapotranspiration, P/PET) and geographical isolation (number of beech plots within a radius
838 of 5 km around each sampled beech population), and (ii) climate (P/PET) and regional
839 community composition surrounding sampled populations (% of Mediterranean communities
840 relative to the total number of plots within a radius of 1.7 km around each beech population).
841 Geographical, climatic and ecological edges (see Fig. 1 and 2) were derived from plot-level
842 data of the Ecological and Forest Inventory of Catalonia and the Spanish National Forest
843 Inventory, and 1-km² resolution interpolated climate derived from the WorldClim database
844 (see supplementary material).

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848 **Supporting information**

849 **Appendix S1.** List and details of studies assessing rear edge population decline.

850 **Appendix S2.** Methodology used in the population-focused study presented.

851

852 **Graphical abstract**

853 Climate change is expected to drive population loss at the species' rear edge, however,
854 disparities between predictions and observations are accumulating. We argue for a revision of
855 the concept of marginality together with an explicit testing of population decline across the
856 species' rear edge, given the scale-dependency of the ecological mechanisms determining
857 population dynamics. Such progress is possible as data availability improves and essential to
858 better predict the consequences of species range shifts.

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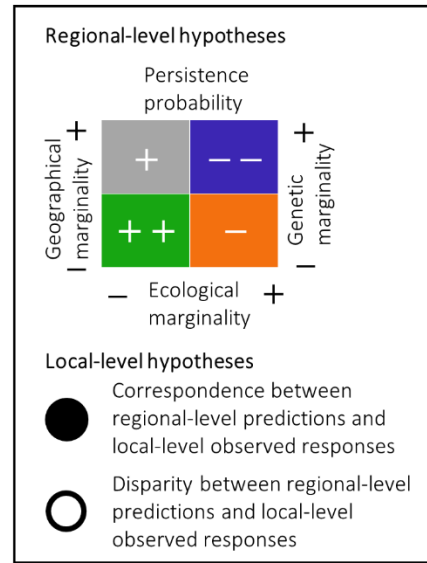
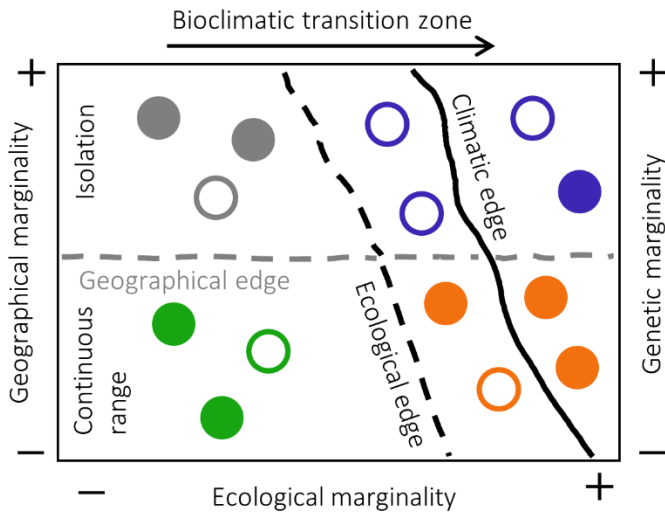
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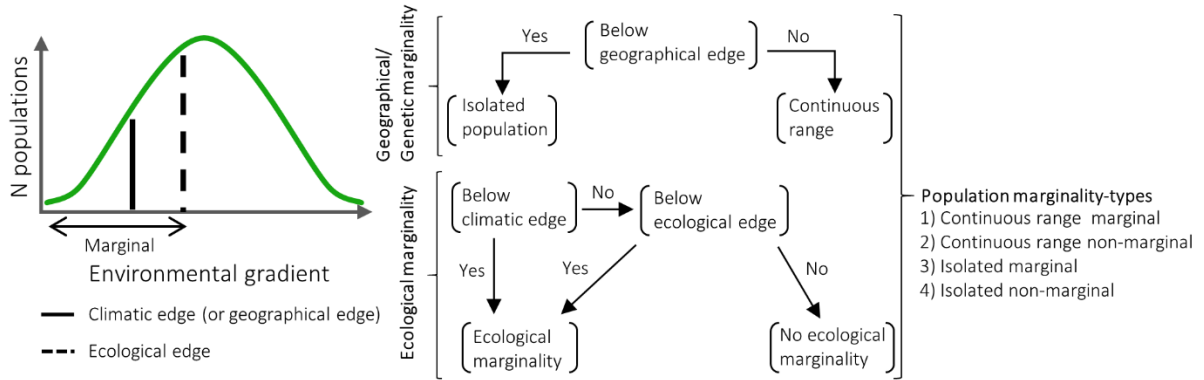
873 **Figure 1**



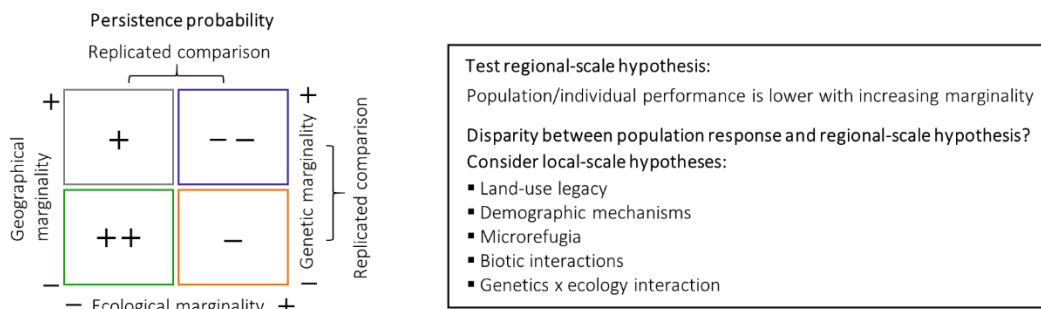
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894 **Figure 2**

(a) Example of classification of populations into population marginality-types



(b) Empirical study design for hypotheses testing on population decline at rear edges



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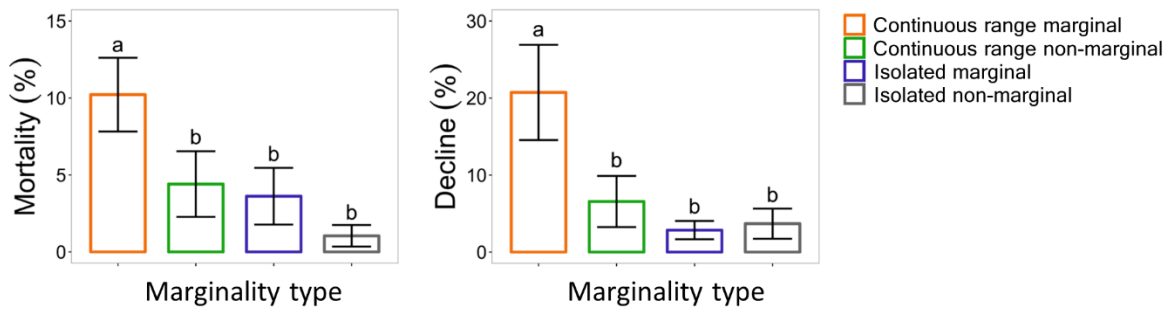
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(a) Mortality and canopy decline according to population marginality-type



(b) Mortality variation across population marginality-type

