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Characterizing biotic interactions within the Order Lagomorpha using Joint Species Distribution Models at 3 different spatial scales

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1 Running heading: Biotic interactions between lagomorphs

2

3 **Characterizing biotic interactions within the Order Lagomorpha using Joint Species**

4 **Distribution Models at three different spatial scales**

5 Katie Leach*, W. Ian Montgomery, and Neil Reid

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7 *School of Biological Sciences, Queen's University Belfast, Belfast, BT9 7BL. Northern*

8 *Ireland, United Kingdom (KL, WIM, NR)*

9 *Institute for Global Food Security (IGFS), Queen's University Belfast, Belfast, BT9 5BN.*

10 *Northern Ireland, United Kingdom (WIM, NR)*

11

12 * Correspondent: kleach01@qub.ac.uk

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16 Species Distribution Models (SDMs) rarely incorporate biotic interactions, even though the
17 latter may have great impacts on biogeographical patterns, because interactions can be
18 difficult to model in time and space. In addition, the resolution of input data can have
19 dramatic effects on results, with coarser resolutions unlikely to capture climatic variation at
20 small scales, particularly in mountainous regions. Joint SDMs can be used to explore
21 distributions of multiple, coexisting species and characterize modelled biotic interactions;
22 however, the influence of scale on predictions is yet to be tested. We produced Joint SDMs
23 for European lagomorph species at 3 hierarchical resolutions and calculated residual and
24 environmental correlations that could explain why species may or may not co-occur, thereby
25 suggesting biotic interactions. European lagomorph species exhibited similar environmental
26 and biotic responses at all 3 resolutions (50 km, 25 km, and 10 km), with models at finer
27 resolutions producing more precise estimates but requiring considerable computing time. The
28 majority of pairwise residual responses were negative, indicating that European lagomorph
29 species co-occur less than expected given their similarity in environmental responses, and
30 suggesting modelled biotic interactions consistent with those reported in the literature. Fine-
31 scale data and models offer greater precision but are not always necessary for multi-species
32 models. However, caution is advised when inferring biotic interactions using data and models
33 based on a coarser scale.

34

35 Key words: competition, co-occurrence, Europe, hare, MCMC, lagomorph, probit regression,
36 rabbit, species interactions

37

38 Species Distribution Models (SDMs) are a widely used analytical approach in modern
39 ecology, particularly with respect to predicting the impacts of climate change; however,
40 SDMs have well known limitations (see Elith and Leathwick 2009). Spatial scale, in terms of
41 resolution, is a major concern when using SDMs, with large-scale environmental data likely
42 too coarse to capture the effects of local climatic variation, especially in areas with large
43 topographical variation (Dobrowski et al. 2009). Environmental and distributional data may
44 be characterized by their extent, referring to the geographical area covered (for example,
45 global, continental, or national), or by their resolution (or grain), which refers to the size of
46 the grid cells in which data are sampled (Wiens 1989; Nystrom Sandman et al. 2013; Wisz et
47 al. 2013). Extent and resolution may be linked, although a greater extent will not always lead
48 to coarser resolution, but an increase in extent is likely to be associated with a decrease in
49 resolution (Pearson and Dawson 2003).

50 Conducting studies at different spatial scales can lead to very diverse results (Wiens 1989;
51 Hamer and Hill 2000). For example, change in biodiversity may be different in strength and
52 direction using data collected at different scales (Keil et al. 2011) due to differential impacts
53 of natural and anthropogenic drivers of ecological change (Moorcroft et al. 2001). Further, in
54 using SDMs to project distributions under future climate scenarios, fine-scale climate
55 projections have been shown to provide very different estimates of climate change impacts
56 compared to their coarse-scale equivalents (Franklin et al. 2013). Notwithstanding, SDMs are
57 often used without regard for the effect of scale (Elith and Leathwick 2009), even though
58 differences among scales are frequently acknowledged. Bradter et al. (2013) advocated
59 studies identifying the appropriate spatial scale of predictors in order to produce more
60 accurate species distribution projections. However, how this identification is undertaken will
61 most likely vary depending on the species and environmental variables in question.

62 Macroclimate is suggested to be one of the main drivers of distribution patterns at
63 continental and global scales, whereas biotic interactions and microclimate may control
64 distributions at community and landscape scales, with human impacts one of the factors
65 explaining ranges at intermediate scales (Whittaker 1975; Pearson and Dawson 2003;
66 Thuiller et al. 2003). However, there is growing evidence of a role for biotic interactions in
67 shaping species distributions at the global scale (Jablonski 2008; Wiens 2011; Wisz et al.
68 2013). Historically, distributional studies have focused on interspecific competition
69 (MacArthur 1972; Amarasekare 2003), but facilitation (mutualism), predation, parasitism,
70 and disease, are now recognized as additional factors in species distribution patterns (Araujo
71 and Rozenfeld 2014). Biotic interactions within trophic levels, such as competition and
72 facilitation, are much harder to observe than interactions between trophic levels, for example
73 predation, but are well known to produce sharp boundaries in species distributions with little
74 or no overlap (Flux 2008).

75 European lagomorphs exhibit strong competitive interactions and occupy a wide range of
76 environmental conditions (Leach et al. 2015a). They occupy extreme elevations in the Alps,
77 and are found across all European latitudes, from the Arctic Circle to the Mediterranean
78 (Chapman and Flux 2008). In addition to the European rabbit, *Oryctolagus cuniculus*, there
79 are 5 species of hare: the Apennine hare, *Lepus corsicanus*, and broom hare, *Lepus*
80 *castroviejoi*, have highly restricted ranges, whereas the European hare, *Lepus europaeus*,
81 mountain hare, *Lepus timidus*, and Iberian hare, *Lepus granatensis*, have much wider ranges.
82 Competition between the latter 3 species is asymmetrical and in most cases, the ranges are
83 parapatric (Acevedo et al. 2012a). For example, in the Iberian Peninsula, European hare
84 densities decrease in areas where they contact Iberian hares (Gortázar et al. 2007; Acevedo et
85 al. 2012a). In mainland Italy, the Apennine hare is decreasing as a result of multiple
86 pressures, including habitat degradation, and probable competition with introduced European

87 hares (Angelici et al. 2008). The Apennine hare competes with the introduced European hare,
88 which is larger and has a higher reproductive rate than the Apennine hare (Angelici et al.
89 2010). When the 2 species occur in sympatry, the Apennine hare is found at higher altitudes,
90 whilst in allopatry they occur in the same altitudinal range (Angelici and Luiselli, 2007).
91 Mountain hare populations typically decline in contact with expanding European hare
92 populations usually with upslope range contraction (Thulin 2003; Reid 2011).

93 In most of the European hare's native range, the mountain hare seems to be restricted to
94 high elevations and forests, as it is driven away from lowland grassland plains (Thulin 2003,
95 Flux 2008), but in Ireland, Finland, Russia, and Sweden, the European hare is found in
96 sympatry with the mountain hare (Flux 2008). In Ireland, introduced European hares and
97 endemic Irish hares, *Lepus timidus hibernicus*, occupy similar habitats in sympatry (Reid and
98 Montgomery 2007). They would probably show strong interspecific competition if resources
99 were limiting (Reid 2011), but this is highly unlikely as the majority of available habitat is
100 grassland and thus optimal for both species. Nevertheless, the European hare has actively
101 displaced the Irish hare within its core invasive range presumed related to competition for
102 space and hybridization (Caravaggi et al. 2015, 2016a).

103 Hares and rabbits frequently co-occur but rarely interact. The European hare and rabbit
104 form one of the most commonly studied and observed systems with respect to competition.
105 Before anthropogenic introductions, the European hare was restricted to central Europe and
106 the Asian steppes, and the European rabbit to the Iberian Peninsula (Flux 1994), but overlap
107 in the ranges of these 2 species is now widespread, and coexistence occurs in many
108 introduced populations (Flux 2008). In most areas of their range they graze side by side,
109 showing significant dietary overlap (e.g., Katona et al. 2004).

110 Here, we produce Joint SDMs for European lagomorph species at 3 hierarchical
111 resolutions: 50 km, 25 km, and 10 km grid cell resolutions. Although home ranges of

112 European hares and rabbits span up to 1 km² (Jones et al. 2009), these resolutions were
113 chosen because environmental or species data are often collected at these levels for atlases
114 and, therefore, these resolutions are frequently used to model species distributions. Model
115 outputs were used to calculate residual and environmental correlations that can explain why
116 species may or may not co-occur, and thus suggest modelled biotic interactions. We
117 hypothesized that the strength of modelled biotic interactions varies with scale due to
118 differential impacts of natural and anthropogenic drivers of ecological change at varying
119 scales (Moorcroft et al. 2001). Modelled biotic interactions are likely to play a greater role at
120 finer resolutions on a community and landscape scale, i.e., 10 km grid cell resolution
121 (Whittaker 1975; Pearson and Dawson 2003; Thuiller et al. 2003).

122

123

MATERIALS AND METHODS

124 *Species and environmental data.*— International Union for Conservation of Nature
125 (IUCN) geographic range polygons for each European lagomorph species (Fig. 1) were
126 rasterized in R v.3.1.1 at 3 hierarchical resolutions: 50 x 50 km ($n = 6,255$ cells), 25 x 25 km
127 ($n = 23,118$ cells), and 10 x 10 km ($n = 224,691$ cells), with a value of 1 for species presence
128 and 0 for absence. IUCN polygons have been used in a number of SDM studies to date (e.g.,
129 Lawler et al. 2009; Visconti et al. 2015), and whilst they may have higher commission errors
130 (Graham and Hijmans 2006), the detailed construction of the polygons together with the
131 internal review process and expert assessments by the IUCN (see
132 <http://www.iucnredlist.org/technical-documents/red-list-training/iucnspatialresources> for
133 further information) can lead to the production of more realistic SDMs (Fourcade 2016). To
134 illustrate the consequences in using different input data for lagomorph species distributions,
135 Leach et al. (2016) compared models built with IUCN polygons to those built with point

136 occurrence data. Predicted probabilities of presence were found to vary substantially between
137 models. Although using IUCN polygons may result in false positives, in this case, point
138 occurrence data resulted in false positive and false negative predictions of occurrence. For
139 example, the Iberian hare is restricted to the Iberian Peninsula, yet models utilizing point
140 occurrence incorrectly predicted areas in northern Europe to be suitable. In addition, the
141 European hare and rabbit are distributed throughout central Europe extending into eastern
142 Europe, yet models using point occurrence data predicted distributions skewed to western
143 Europe. This reflects the sparse and biased nature of point occurrence data, whilst suggesting
144 that IUCN polygons, at least for European lagomorphs, lead to more realistic species
145 distribution models.

146 Current climate variables (~1950-2000) were downloaded from WorldClim
147 (www.worldclim.org) and resampled to the same resolution as the species data.
148 Evapotranspiration was calculated using the Hargreaves equation (see Leach et al. 2015b for
149 more details) and annual water balance was calculated by subtracting annual
150 evapotranspiration from mean annual precipitation. The number of months with a Positive
151 Water Balance (PWB) was calculated by subtracting each monthly evapotranspiration from
152 its corresponding monthly precipitation, then converting into a binary format, where a value
153 greater than 0 was given a value of 1 and a value less than 0 was kept at 0, and finally
154 summing the 12 binary scores (Kremen et al. 2008). Mean annual Normalized Difference
155 Vegetation Index (NDVI) was calculated from monthly values which were downloaded from
156 the European Distributed Institute of Taxonomy (EDIT) Geoplatform ([http://edit.csic.es/Soil-
157 Vegetation-LandCover.html](http://edit.csic.es/Soil-Vegetation-LandCover.html)). Hilliness, an index of surface roughness, was calculated by
158 finding the difference between maximum and minimum gradient values, based on a global
159 Digital Elevation Model at 30 arc-minute resolution (Newton-Cross et al. 2007). Human
160 Influence Index data were downloaded from the NASA Socioeconomic Data and

161 Applications Centre (SEDAC) website (<http://sedac.ciesin.columbia.edu/>; WCS CIESIN
162 2005). Subsequently, correlated environmental variables (minimum precipitation, minimum
163 temperature, mean annual precipitation, mean annual temperature, solar radiation, annual
164 water balance, and annual evapotranspiration) were removed, leaving the following:
165 maximum temperature, temperature seasonality, maximum precipitation, precipitation
166 seasonality, PWB, NDVI, Hilliness, and Human Influence Index. Environmental variables
167 were centered on 0 and scaled by their standard deviations.

168 The environmental variables chosen ultimately for modelling were known to determine
169 distributions of European lagomorph species. Leach et al. (2015b) found the following
170 variables were important in describing the distribution of more than 1 European lagomorph:
171 Hilliness, Human Influence Index, maximum temperature, NDVI, precipitation seasonality,
172 temperature seasonality, and water balance. Altitude, maximum precipitation, and
173 precipitation seasonality were significantly important in describing the distribution of the
174 Iberian hare; precipitation and temperature seasonality in describing the distribution of the
175 European hare; and maximum temperature in describing the distribution of the mountain hare
176 (Acevedo et al. 2012a, b). In addition, temperature seasonality was the most influential
177 environmental variable for predicting the distributions of European and mountain hares
178 (Caravaggi et al. 2016a, b).

179

180 *Model structure.*— We used the code provided in Pollock et al. (2014) to produce Joint
181 SDMs at the 3 hierarchical resolutions. Joint SDMs simultaneously estimate the ranges of
182 multiple coexisting species producing mixtures of possible species assemblages (Pollock et
183 al. 2014; Harris 2014). Pollock et al. (2014) used a hierarchical, multivariate, probit
184 regression model to include multiple species into a single SDM, with 1 model run per spatial
185 scale. The model response is species occurrence represented by a matrix with dimensions of

186 sites by species. The response is predicted by a data matrix with dimensions of sites by
187 environmental variables. The number of dependent and independent variables did not vary
188 across model runs. Interactions between species will cause un-modelled (i.e., unaccounted
189 for) dependence in the residuals of the model, but these residual correlations can provide
190 insight into the abiotic and biotic factors driving species co-occurrence patterns.

191 Models were fitted using the MCMC Bayesian modelling software JAGSv3.4.0 run
192 through Rv3.1.1 via the R2jags packagev0.5-6. For all 3 resolutions, we ran 2 chains for
193 850,000 generations with the first 150,000 discarded as burn-in in order to reach an
194 asymptote and with the remaining samples thinned by a factor of 1,000 meaning we retained
195 985 samples per chain for post-processing. We used vague priors for all model parameters
196 and considered models to be converged once all elements of the parameter and correlation
197 matrices had potential scale reduction factor values close to 1. This convergence diagnostic
198 value suggests that each of the sets of simulated observations is close to the target distribution
199 (Brooks and Gelman 1998).

200 Species pairs were then examined after the models were fitted. Residual and
201 environmental correlations for species pairs were decomposed from model outputs and used
202 to explain why species may or may not co-occur. The model outputs include predicted
203 probabilities of presence for each species in each grid cell, regression coefficients for the
204 response of each species to each environmental variable, and species-by-species grids with
205 correlation due to similar environmental responses and residual correlations. Environmental
206 correlations between species are a function of those species' scaled regression coefficients
207 and the covariance's of the environmental variables. Positive environmental correlations
208 suggest shared environmental responses, with strong negative or positive residual correlations
209 potentially suggesting evidence for biotic interactions (Fig. 2; see right quadrants).

RESULTS

210

211 For a particular species pair, the direction of environmental and residual correlation
212 coefficients were largely similar at all spatial resolutions examined (Fig. 3). The majority of
213 species pairs shared environmental responses; however, the mountain hare's environmental
214 correlation coefficients were negatively related to those of the European rabbit and the
215 European hare, suggesting that the mountain hare has strikingly different environmental
216 responses. In addition, most species pairs had negative residual correlations, indicating that
217 species co-occurred less than expected given the similarity in environmental responses (Table
218 1). Nevertheless, models at finer resolutions took considerably longer to run using a high
219 performance desktop computer (64-bit, two 3.10GHz processors and 192GB RAM); the 50
220 km model took ~3 days, 25 km took ~3 months, and 10 km took ~6 months. Regression
221 coefficients to show which environmental variables were driving the positive and negative
222 correlations between species are given in Supplementary Data S1.

223 A variety of pairwise responses were evident from the models. The broom hare co-
224 occurred more than expected with the European hare and rabbit at all spatial scales given
225 shared environmental responses and suggesting the potential for facilitative interactions. The
226 Apennine and Iberian hares co-occurred less than expected with the European hare and rabbit
227 given shared environmental responses, suggesting the potential for competitive interactions.
228 European hares and rabbits co-occurred more than expected given their shared environmental
229 responses, whereas European hares and mountain hares occupy very different environments
230 and were less likely to co-occur than expected. The European rabbit and mountain hare also
231 occupy very different environments and were less likely to co-occur than expected (Fig. 4).
232 No species pairs occupied the upper left quadrant of Fig. 4, i.e., species with distinct
233 environments did not co-occur more than expected. The strength of environmental and
234 residual correlations was similar across different scales, although credible intervals were

235 substantially narrower at finer resolutions, i.e., 10 km grid cells (by 75.8% on average when
236 compared to those associated with the 50km resolution) and, therefore, provided greater
237 precision (Fig. 4).

238 Co-occurrence patterns varied substantially between spatial scales (Fig. 5). There was no
239 evidence for co-occurrence between mountain and European hares at the 50 km and 25 km
240 scales, but models at the 10 km scale predict co-occurrence between these species with
241 greater accuracy. In probit regression models, the mean of the normal distribution is an
242 analogue of the linear predictor; therefore, a large positive value indicates high probability of
243 presence and a large negative value indicates a low probability of presence. Therefore,
244 patterns extending into the upper right quadrant of Fig. 5 indicate co-occurrence between
245 those species, for example the mountain hare and European rabbit.

246

247

DISCUSSION

248 Lagomorphs occupy a considerable range of environmental conditions (Chapman and Flux
249 2008), from the Arctic Circle, Scandinavia, and the mountains of northern Scotland where
250 cold temperatures and high precipitation are common, to the Iberian Peninsula and the
251 Mediterranean with semi-arid environments. So initially, it may be surprising that most
252 species shared environmental responses. However, *within*-species variation can be large due
253 to the huge range of environments each occupies. The Iberian hare occupies the whole of the
254 Iberian Peninsula experiencing concomitant variation in climate from lowland coastal regions
255 to high elevation arid regions inland (Acevedo et al. 2012b). In contrast, *between*-species
256 variation can also be large with some species occupying distinct environmental conditions,
257 for example, the mountain hare and the European rabbit and hare. The former has a high
258 latitudinal and elevational range, and occurs in areas with lower temperatures, compared to

259 the European rabbit and hare; therefore, we would expect the mountain hare to demonstrate
260 very different environmental responses (Thulin 2003).

261 Our analysis suggested that the majority of residual correlation coefficients were negative,
262 and thus, species co-occurred less than expected given their shared environmental responses.
263 Strong negative residual correlations indicate the possibility of competitive interactions for
264 lagomorphs in Europe consistent with published sources, specifically: Iberian and European
265 hares (Gortázar et al. 2007; Acevedo et al. 2012a), Apennine and European hares (Angelici
266 and Luiselli, 2007; Angelici et al. 2008, 2010), and mountain and European hares (Thulin
267 2003; Reid 2011; Caravaggi et al. 2015). Thirty-three lagomorph species are known to have
268 competitive interactions reported in the literature, with closely related, large-bodied, similarly
269 sized species, occurring in regions of human-modified, typically agricultural landscapes or at
270 high elevations, such as Apennine, European, Iberian, and mountain hares, significantly more
271 likely to have reported competitive interactions than other lagomorph species (Leach et al.
272 2015a). In addition, the models suggest a facilitative interaction between European hares and
273 rabbits. Evidence for biotic interactions between these 2 species has been debated, but the
274 current general consensus is that they co-occur without competition (Flux 2008), comparable
275 to our results. It should be noted that these are hypotheses of species interactions that need to
276 be tested empirically and confirmed using natural history data. Unexplained residual variance
277 between some species pairs, however, may not be explained by modelled biotic interactions if
278 key determinants of the extent of their ranges have been left out of our models (i.e., other
279 environmental variables not included could account for the unexplained residual variation).
280 For example, minimum temperature is known to be a key determinant of distributions of
281 European lagomorph species (Leach et al. 2015b) but was left out of the models in this study
282 due to high multicollinearity with other environmental variables, notably maximum
283 temperature, causing undue model leverage.

284 Co-occurrence in terms of overlapping species presence can change substantially at finer
285 resolutions. Mountain and European hares, and European hares and rabbits, exhibited more
286 overlap in their ranges at finer resolutions, whereas Apennine hares and European rabbits,
287 and mountain hares and European rabbits, showed less overlap at finer resolutions. Species
288 exhibiting less overlap occupied high elevational ranges in the Alps, Apennines, and Scottish
289 Highlands, indicating again that finer resolutions capture small changes in microclimatic
290 variation in mountainous regions (Dobrowski et al. 2009), and suggesting that models at finer
291 resolutions may be more appropriate for species found in these areas.

292 Using rasterized IUCN geographic range polygons to build SDMs may lead to outputs
293 particularly vulnerable to false positives (Murray et al. 2011), and potentially influence our
294 interpretation of ‘interaction’; 2 species with identical range extents may never meet because
295 of habitat partitioning, especially when separated by elevational gradients. Another
296 potentially confounding effect is that models built with point-occurrence data will have been
297 downloaded at a specific time and, therefore, may not reflect ecology based on long-term
298 climate trends. To the best of our knowledge this has not yet been addressed within the field
299 of Species Distribution Modelling. However, neither the use of range maps nor point
300 occurrence data is without error (Pineda and Lobo 2012), and the relationship with scale may
301 in fact be an artefact of coarse input data, regardless of resolution. In this study, we preferred
302 to accept the risk of omission errors over commission errors because only the interactions
303 with most confidence are likely to be captured by the models. Nonetheless, we suggest that
304 when deciding what input data are to be used, the purpose of the study and quality of the data
305 available should be considered.

306 Joint SDMs run at fine-scale resolutions had extremely long processing times using a high
307 performance desktop computer, and although they produced estimates of residual and
308 environmental correlation coefficients with greater precision, the strength and direction of

309 correlations were similar, and in most cases identical, at all scales. This contradicts our
310 hypothesis that the strength of modelled biotic interactions varies with scale. Predicted
311 probabilities of occurrence were more precise at finer resolutions for some species, agreeing
312 with our hypothesis that modelled biotic interactions play a greater role at finer resolutions,
313 but for others an increase in spatial resolution resulted in little change to these values. If the
314 aim is to accurately infer biotic interactions, modelling at finer resolutions is recommended.
315 However, if only the strength and direction of environmental and residual correlations is of
316 interest, then a coarser resolution may be adequate in the interest of saving processing time.
317 Coarse resolution data may be just as useful in terms of accuracy (not precision), so it may
318 not always be necessary to collect fine-resolution species occurrence data that could require
319 considerable effort.

320 Scale is highly important when modelling multi-species distributions, but will nearly
321 always result in a compromise between processing time and precision of results. The strength
322 and directions of estimated correlations from joint SDMs were similar across scales, but with
323 greater precision at finer resolutions, especially with respect to predicted probabilities of
324 occurrence. Fine-scale models and data collection may not always be necessary for multi-
325 species models; however, caution is advised when seeking to accurately infer biotic
326 interactions using coarse data, especially when the species in question occupies mountainous
327 regions.

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333

334 **SUPPLEMENTARY DATA**

335 **Supplementary Data S1.** Regression coefficients between European lagomorph species and
336 environmental variables at 3 hierarchical resolutions.

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474

475 **FIGURE LEGENDS**

476 **Figure 1.** IUCN geographic range polygons for European lagomorph species.

477

478 **Figure 2.** Diagrammatic interpretation of negative and positive residual and environmental
479 correlations. These interpretations may become less reliable with weaker correlation
480 coefficients.

481

482 **Figure 3.** Network diagrams representing modelled environmental and residual correlation
483 between European lagomorph species at 3 different scales: (a) 50 km, (b) 25 km and (c) 10
484 km. Black edges indicate positive correlations between species and red edges indicate
485 negative correlations. Each edge is labelled with its correlation coefficient. Only significant
486 correlations, i.e., those for which the credible intervals do not cross 0, are shown. Species
487 pairs without connecting edges do not have spatially overlapping ranges.

488

489 **Figure 3.** Modelled environmental and residual correlations between European lagomorph
490 species pairs at 3 different scales: (a) 50 km, (b) 25 km and (c) 10 km. Error bars represent
491 95% credible intervals.

492

493 **Figure 5.** Co-occurrence patterns for all combinations of European lagomorph species using
494 predicted probabilities of co-occurrence from Joint SDMs at 3 different scales: 50 km (grey),
495 25 km (blue) and 10 km (red). Large positive values indicate high probability of presence and
496 large negative values low probability of presence – for further explanation please see the
497 Results section.

498 **TABLES**

499

500 **Table 1.** Modelled environmental and residual correlations between pairs of European lagomorph species at 3 hierarchical resolutions. NS

501 indicates a species pair with credible intervals overlapping 0, i.e., non-significant. Interpretations are based on Fig. 2.

502

Species 1	Species 2	Environmental correlation			Residual correlation			Interpretation
		50 km	25 km	10 km	50 km	25 km	10 km	
Apennine hare	European hare	0.406	0.435	0.664	0.773	0.380	0.470	Potential facilitative interaction
Apennine hare	European rabbit	0.575	NS	0.711	0.496	NS	0.145	Potential facilitative interaction
Corsican hare	European hare	0.268	0.199	0.273	-0.611	-0.460	-0.629	Potential competitive interaction
Corsican hare	European rabbit	0.469	NS	0.463	-0.140	NS	-0.073	Potential competitive interaction
European hare	Mountain hare	-0.911	-0.899	-0.890	-0.208	-0.331	-0.175	Distinct environments, co-occur less than expected
European hare	European rabbit	0.799	0.810	0.806	0.612	0.554	0.641	Potential facilitative interaction
Iberian hare	European hare	0.657	0.722	0.775	-0.361	-0.416	-0.332	Potential competitive interaction
Iberian hare	European rabbit	0.815	0.856	0.894	-0.370	-0.456	-0.304	Potential competitive interaction
Mountain hare	European rabbit	-0.613	-0.737	-0.616	-0.304	-0.381	-0.301	Distinct environments, co-occur less than expected

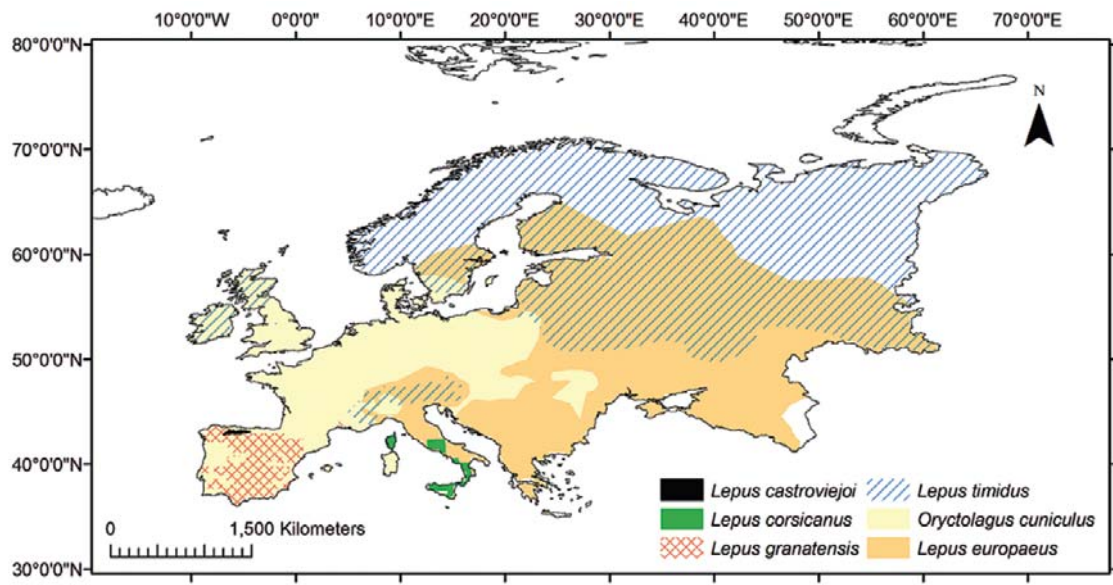


Figure 4. IUCN geographic range polygons for European lagomorph species.

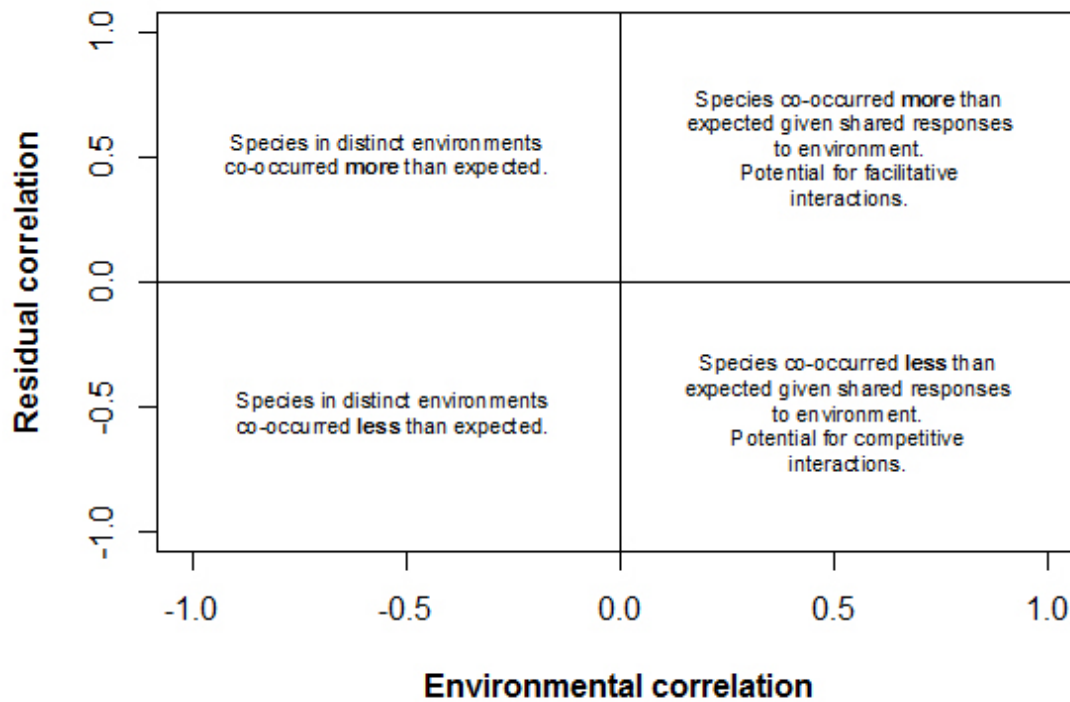


Figure 5. Diagrammatic interpretation of negative and positive residual and environmental correlations. These interpretations may become less reliable with weaker correlation coefficients.

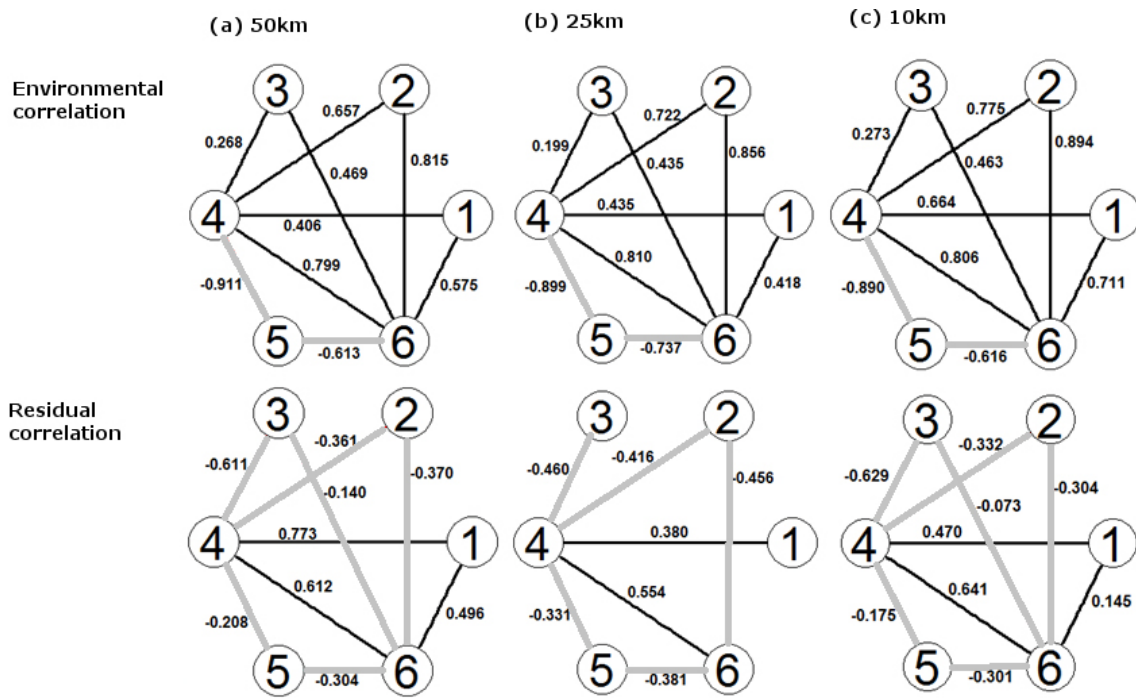


Figure 3. Network diagrams representing modelled environmental and residual correlation between European lagomorph species at 3 different scales: (a) 50 km, (b) 25 km and (c) 10 km. Black edges indicate positive correlations between species and red edges indicate negative correlations. Each edge is labelled with its correlation coefficient. Only significant correlations, i.e., those for which the credible intervals do not cross 0, are shown. Species pairs without connecting edges do not have spatially overlapping ranges.

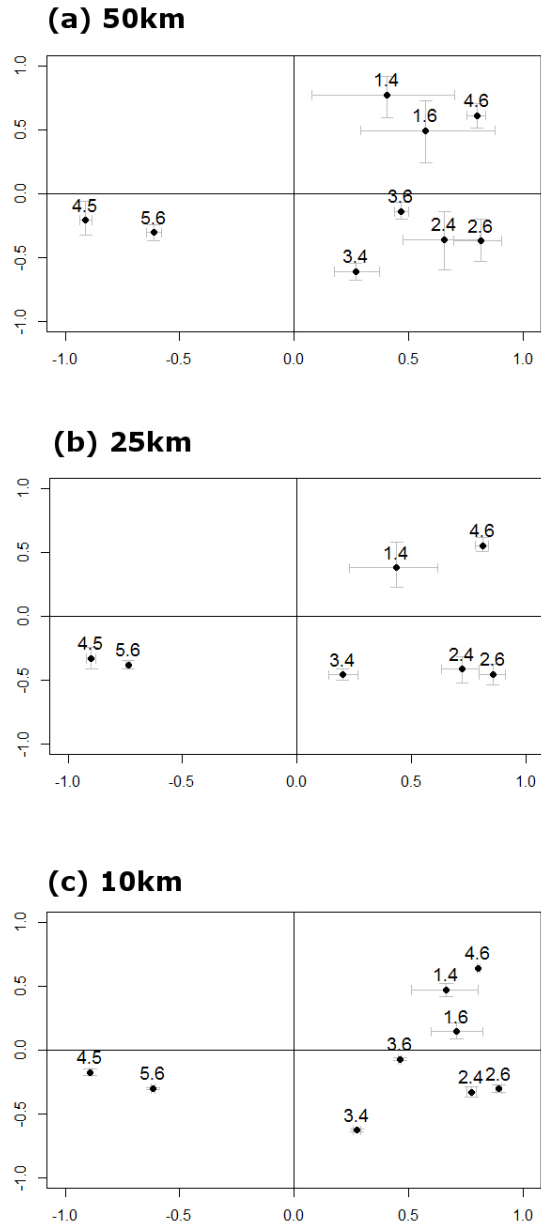


Figure 6. Modelled environmental and residual correlations between European lagomorph species pairs at 3 different scales: (a) 50 km, (b) 25 km and (c) 10 km. Error bars represent 95% credible intervals.

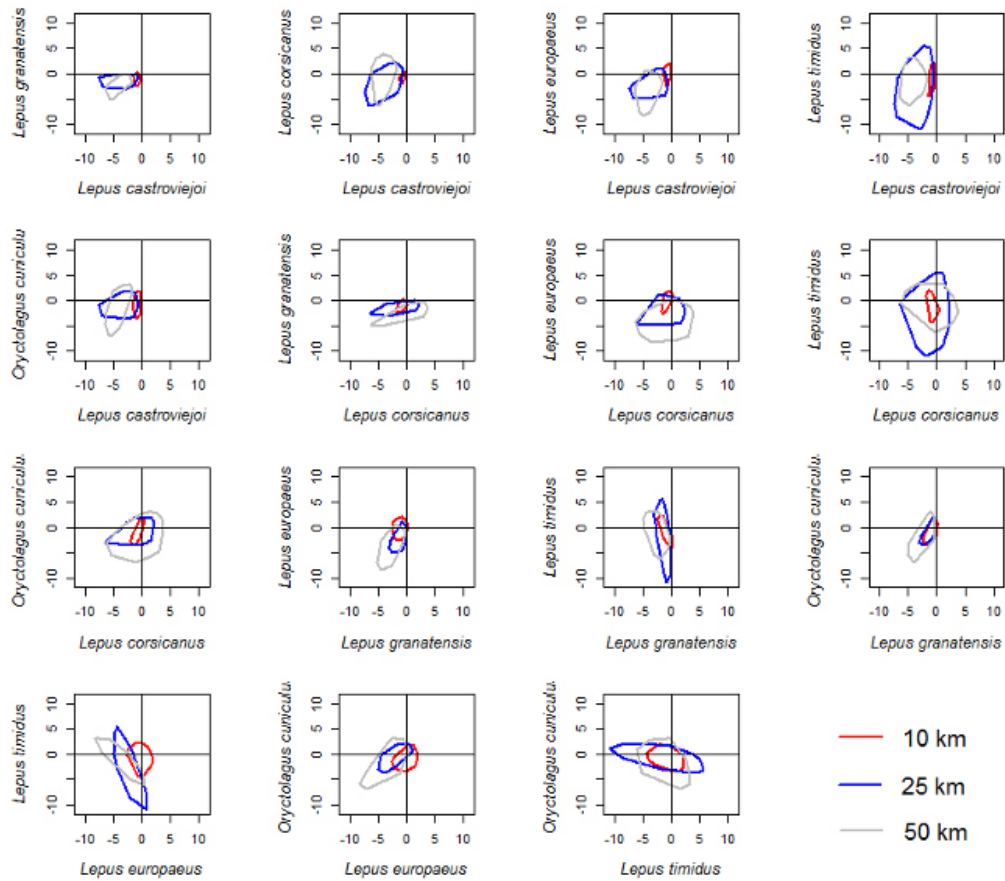


Figure 5. Co-occurrence patterns for all combinations of European lagomorph species using predicted probabilities of co-occurrence from Joint SDMs at 3 different scales: 50 km (grey), 25 km (blue) and 10 km (red). Large positive values indicate high probability of presence and large negative values low probability of presence – for further explanation please see the Results section.