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# Where will species on the move go? Insights from climate connectivity modelling across European terrestrial habitats

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## 16 Abstract

17 Climate change can induce species range shifts. However, the intensity of climate change, the 18 intrinsic dispersal ability of species and the anthropization of landscapes are impeding species 19 movements in most cases. In this context, preserving and promoting climate corridors for 20 species to migrate from their current habitats to their future climatically similar habitats is an 21 important strategy for preventing species extinction. Climate connectivity modelling is a tool 22 that can identify these potential movement pathways. Here, we aimed to model connectivity 23 between climate analogues across Europe under various ecological assumptions and climate 24 change scenarios, in order to identify areas of high potential connectivity and to quantify 25 variation in connectivity across a range of hypotheses. We also overlapped connectivity maps 26 with protected areas to determine whether climate connectivity was sufficiently protected. We 27 showed that climatic connectivity did not differ much between different scenarios of climate 28 change, but was strongly dependent on species' dispersal assumptions. It was also relatively 29 similar to a scenario of non-climatic connectivity. Therefore, it may be feasible to anticipate 30 the effect of climate change on species movements regardless of the future trajectory of 31 climate, but the implementation of protection strategies for multiple species will certainly 32 prove complex. Overall, protected areas were located in the regions of high and stable 33 connectivity, but some countries lack the appropriate protection schemes, especially regarding 34 strong protections. Our results have the potential to serve in the construction of land cover 35 change scenarios to identify the best strategies to improve climate connectivity. 36

## 37 Keywords

38 Climate change, climate connectivity, climate corridors, protected areas, movement.

## 39 1. Introduction

Knowing where species are distributed in space and how these distributions change under 40 human pressure are key knowledge to acquire for conserving biodiversity (Santini et al., 41 2021). Species ranges are determined by both biotic (e.g. predation, pathogens, competition, 42 resource availability) and abiotic factors (e.g. climate, soil properties, light availability) to 43 which species are adapted and which determine their ecological niches. However, the large-44 45 scale distribution of species is generally thought to be mainly driven by their Grinnellian niche, which describes species' fundamental niche through the physicochemical properties of 46 habitats (Soberón, 2007, Soberón & Nakamura, 2009). One of the main factors of the 47 Grinnellian niche is climate, which largely determines the abiotic conditions that species 48 49 experience at macro-ecological scale. Any long-term change in climate will thus modify the geographical areas that are suitable for a given species. However, abiotic factors are often not 50 51 sufficient to describe species' realized niche, representing the conditions actually occupied by 52 species. Indeed, dispersal also control the areas that can effectively be reached, independently of whether they are suitable otherwise (Soberón & Nakamura 2009). Therefore, area-based 53 conservation strategies must take into account the interaction between physical environmental 54 55 conditions, including climate, and dispersal opportunities of species with various intrinsic movement abilities, in a complex landscape of contrasted land use intensities. 56

Climate change has been recognized since the 1970s as one of the most severe 57 anthropogenic environmental changes. This includes a global increase in the Earth's 58 temperature, which could reach 1.5°C on average by 2040 (IPCC, 2018). In addition to the 59 increase in average temperature, climate change may also lead to significant changes in the 60 structure of habitats, for example by altering moisture regimes and causing initially wet areas 61 62 to become dry (Raymond et al. 2019). Climate change is already having significant effects on terrestrial species, which have been widely observed in recent years. Among them are 63 morphological changes of organisms through body-size shrinking (Sheridan & Bickford, 64 65 2011) and appendage lengthening (Ryding et al., 2021), changes in individuals' metabolic rate 66 (Williams et al., 2015) and phenology (Macgregor et al., 2019), or acceleration of senescence (Le Roux et al. 2005). At the ecosystem-scale, climate change is expected to result into 67 68 changes in species richness (Currie, 2001), modification of community composition and functionality (Mokany et al., 2015), shifts in biome distribution (Boonman et al., 2021), and 69 to an overall increase in net primary production (Bergh et al. 2003). As a consequence of this 70 71 profound alteration of their habitats, species may need to change their distributional areas to remain in suitable environmental conditions in order to limit the short and long term effects of 72 climate change and to avoid extinction (Soberón and Nakamura, 2009; Caplat et al., 2016; 73 74 Estrada et al., 2016). When species are able to tolerate temperature changes, climate change 75 will likely lead to range expansion towards newly suitable habitat even if no range retraction occur at the trailing edge. In this context, predicting where and how species will carry out 76 such range shifts appears as an important challenge for biodiversity conservation in a 77 78 changing climate.

79 Empirical evidence shows that terrestrial species are moving primarily poleward and toward higher elevations in response to climate change (Parmesan, 2006, Chen et al., 2011). 80 For example, a review of range shifts across several terrestrial taxa concluded that ca. 70% of 81 species have migrated towards the poles over the last few decades (Thomas, 2010). However, 82 not all species shift their range at the same pace, which can disrupt biological interactions. In 83 some cases, mutualistic interactions can be altered because one of two interacting species is 84 unable to shift its range (Pelini et al., 2009); alternatively, the arrival of range-shifting 85 86 predators or competitors can disturb native ecosystems (Alexander et al., 2015; Kubelka et al.,

2018). Species that are more sensitive to environmental changes are also more prone to move 87 into new environments (Gilman et al., 2010). Importantly, species intrinsic dispersal ability 88 89 influences range shifts opportunities, actively mobile species being generally favoured compared to species whose dispersal is achieved by the passive transport of propagules 90 (Estrada et al., 2016). Trees' distributions, for example, may show a strong and growing 91 92 disequilibrium with climate because their slow dispersal make them unable to respond as fast as climate changes (Vissault et al., 2020; Seliger et al., 2021). Therefore, fast and/or high 93 changes in temperature compared to species migration ability may cause lags between 94

95 temperature increase and community responses (Devictor et al., 2012).

96 Species' range shifts in response to climate change may be influenced by anthropogenic 97 changes that impact species' ability to move. Occasionally, human activity can promote species' dispersal through the transport of individuals or the creation of new habitats. Still, 98 99 habitat loss and fragmentation caused by anthropogenic land use often reduce the ability of species to disperse among habitats patches (Schtickzell et al. 2006). For example, 100 infrastructures such as dams or roads frequently act as barriers for dispersal, preventing the 101 102 movement of aquatic organisms along the river, and increasing mortality during crossing, respectively (Caplat et al. 2016). In addition to having a direct effect on the dispersal of non-103 synanthropic species, several studies have shown that habitat fragmentation can affect certain 104 105 environmental variables, which can in return alter species' dispersal abilities. This includes changes in microclimatic conditions, with generally leads to an increase in mean temperature 106 in fragmented habitats (Kuczynski et al., 2018). Changes in precipitation, on the other hand, 107 was found to be context-dependent and could either increase (Kuczynski et al., 2018) or 108 decrease (Platts et al., 2019) with fragmentation. Several studies conducted on bird, fish, and 109 insect species have demonstrated that range shifts induced by changes in temperature can be 110 impaired in highly fragmented areas (Jarzyna et al., 2015; Kuczynski et al., 2018; Platts et al., 111 2019; Fourcade et al., 2021). Enabling species' movement to allow them to track climate 112 change is thus an important strategy for preventing species extinction. 113

Implementing conservation actions for promoting range shifts implies that we must be 114 115 able to anticipate species' movement from their current habitats towards their future habitats (Krosby et al. 2010; Littlefield et al., 2019). In this context, the identification of climate 116 corridors that species will use would allow targeting specific areas where connectivity 117 networks could be strengthened to facilitate the movement of species in the landscape (Caplat 118 et al., 2016). Several attempts have been made to identify the impact of land use, habitat 119 fragmentation and climate on species movement (see Littlefield et al., 2019 for a review of 120 121 climate connectivity modelling approaches). The simplest models assume that species movement in response to climate change occur based on temperature gradients (Nuñez et al., 122 2013; McGuire et al., 2016). However, considering more variables, including land cover and 123 landscape fragmentation, may be more realistic (e.g. Littlefield et al., 2017). Climatic 124 connectivity, corresponding to the capacity of a landscape to permit the movement of species 125 that carry out range shifts in response to climate change, can be modelled explicitly between 126 the current distribution of a focal species of conservation concern and its predicted 127 128 distribution in future climate (Fitzpatrick and Dunn, 2019). In a broader context of landscape planning where all existing communities are to be protected against the deleterious impacts of 129 climate change, connectivity can be alternatively modelled by identifying paths between 130 climatic analogues, i.e., between two habitats where climatic conditions in future scenarios of 131 climate change are similar to those in current climate (Littlefield et al., 2019). The 132 identification of that path is a complex process that depends on various hypotheses. For 133 134 example, it could be defined as the shortest distance between current habitat and its climatic analogue, or may be based on the least cost path following the geophysical properties of the 135

environment (Carroll et al., 2018). Recent developments in connectivity modelling that are 136 based on the analogy with electrical current provide an unvaluable tool to model climate 137 connectivity at a large spatial scale (McRae et al. 2016; Leonard et al., 2017). Although 138 several studies on the connectivity between climate analogues have already been conducted in 139 the United States (Cushman et al., 2013; Littlefield et al., 2017; Carroll et al., 2018; 140 141 Fitzpatrick and Dunn, 2019), few studies are available on a European scale. Moreover, whether the resulting models are stable to a range of hypotheses remains unknown. This 142 information is crucial because there are many uncertainties regarding the way species will 143 shift their range in the future, which could potentially modify the areas predicted to be used as 144 climate corridors. For instance, climate-induced species movements will likely depend on the 145 intensity of climate change (Nuñez et al., 2013); species with different ecological traits may 146 also use different climate corridors (Littlefield et al., 2019). This uncertainty may hamper the 147 implementation of efficient actions aiming at protecting climate corridors. 148

In the present study, we were interested in identifying the stability of climate corridors in 149 Europe by modelling connectivity between climate analogues in relation to land use under 150 various scenarios. An important expected outcome was to evaluate how mapping the stability 151 of climate corridors in Europe would allow us to identify gaps in the current network of 152 protected areas, which could potentially provide a good protection against the effect of 153 climate change on biodiversity (Gillingham et al., 2015; Gaüzère et al., 2016; Haight and 154 Hammill, 2020). To answer this question, we tested hypotheses about the dispersal ability of 155 the species and their preferred habitat type (forests or semi-natural open habitats). Climate 156 analogues were also identified between the current period and two future time periods (2050 157 and 2070) using bioclimatic variables calculated from two scenarios of climate change (RCP 158 2.6 and RCP 8.5). This strategy allows to test a range of realistic scenarios that can serve as a 159 basis for biodiversity-friendly landscape planning at the European scale in a context of 160 climate change. When species' movements are impaired by fragmentation and poorly 161 managed land use (Jarzyna et al., 2015; Kuczynski et al., 2018; Platts et al., 2019; Fourcade et 162 al., 2021), mapping climate corridors may help identifying areas of high connectivity that 163 164 should be protected or restored, in such a way that species' and communities' climatic debt could be reduced (Devictor et al., 2012). 165

166

## 167 **2. Material and methods**

#### 168 2.1. Data on climate, land use and protected areas

We based our modelling of climate connectivity on 19 bioclimatic variables, which 169 correspond to biologically meaningful climate variables describing current and future climate. 170 These variables represent annual average and extreme values of precipitation and temperature, 171 172 as well as their seasonality, to which species are expected to respond. Current bioclimatic variables were produced by interpolating weather data between 1979 and 2013 from a global 173 network of weather station (Karger et al, 2017), while future variables were derived from 174 climate modelling under various scenarios (Karger et al, 2020). We selected future data from 175 the Model for Interdisciplinary Research On Climate version 5 (MIROC5), which is one of 176 the best atmospheric circulation models for representing the observed warming trend 177 178 (Altamirano del Carmen et al, 2021). Climate data were downloaded from the CHELSA website (https://chelsa-climate.org/), which compiles global climate variables from several 179 different models. Future bioclimatic variables were selected over two different time periods 180 181 (2041-2060 and 2061-2080, hereafter referred to as 2050 and 2070) and under two scenarios (RCP 2.6 and RCP 8.5). The RCP 2.6 (Representative Concentration Pathway) scenario 182

- 183 corresponds to a radiative forcing of  $+2.6 \text{ W/m}^2$  by 2100, resulting in an initial increase in the
- Earth's temperature followed by a decline. The second scenario (RCP 8.5) is the least optimistic and corresponds to an increase of  $+8.5 \text{ W/m}^2$  by 2100, associated to a strong
- optimistic and corresponds to an increase of  $+8.5 \text{ W/m}^2$  by 2100, associated to a strong temperature warming. Two contrasting scenarios were chosen to determine if large
- differences in movement pathways were observable depending on the severity of climate
- 188 change.

To account for landscape resistance to movement (see 'connectivity modelling' section), we used land cover data from the CORINE land cover project (version 2018), obtained from the Copernicus website (https://land.copernicus.eu). It compiles pan-European land cover maps classified into 48 land cover types, allowing the classification of habitat types and the discrimination of highly anthropized areas against natural areas. The maps provided by the Corine Land Cover inventory are mainly produced by visual interpretation of high-resolution satellite images.

In order to overlap climate connectivity and protected areas, we used data from the World Database on Protected Areas (WDPA, https://www.protectedplanet.net/), which provides a nearly comprehensive global compilation of protected areas. Here, we used both the whole set of terrestrial protected areas in Europe, as well as a reduced dataset comprising only highly protected areas (IUCN category Ia) (Appendix A, Figure A1). These areas correspond to strict nature reserves set aside for the protection of biodiversity and where human visits and land use are strictly controlled.

All spatial data were reprojected in the ETRS89/LAEA Europe coordinate system and resolution was coarsened to 500 m × 500 m in order to speed-up computations that would otherwise be intractable at finer resolutions.

Throughout the text, the term "species" will refer to theoretical terrestrial species living on the European continent and adapted to the current climatic conditions. We expect our approach to be realistic enough to be applicable to real species affected by climate change.

#### 209 2.2. Identifying climate analogues

To model habitat connectivity in a changing climate, we assumed that species movements will 210 occur between a set of starting points and their climate analogues in the future. Therefore, we 211 first identified climate analogues between current and future climates, i.e. locations that will 212 share similar climatic conditions now and in each climate change scenario. We selected 213 starting points in two different habitat types to simulate two groups of species differing by 214 their ecological requirements: forested areas or open semi-natural habitats. Each species 215 group was represented by 1000 randomly selected points in their respective habitats, defined 216 according to the land cover map (categories with resistance = 1 in Table 1). 217

The identification of climate analogues was based on four climate change scenarios (two 218 219 RCPs and two future time periods) and different ecological hypotheses. First, we considered that species do not change their ecological requirements, such that climate analogues had to 220 221 be in the same habitat type as the initial points. Because of this, all the analogues of the forest points were searched in forests and all the analogues of the points located in semi-natural 222 223 open environments were searched in semi-natural open environments. We then considered the application or not of a distance limit between climate analogues. To select a realistic 224 dispersal, we referred to data from Littlefield et al. (2017) who used a similar method to 225 model climate connectivity, and who reported that most species had an average dispersal rate 226 of 0.5 to 5 km/yr. Here we retained an intermediate value such that maximum dispersal 227

distance was rounded to 100 km in 35 years. Therefore, distance limit was set assuming a 228 maximum dispersal distance of 2857.15 m yr-1, meaning that species would have travelled at 229 most 100 km to reach their climatic analogue in 2050 (100 000 m / 35 yrs = 2857.15 m yr-1). 230 Maximum movement distance in this hypothesis would thus be 157 km in 2070. In the case 231 where no distance limit was defined, and providing that other biological and environmental 232 conditions are fulfilled, species could migrate at infinite speed to reach their best possible 233 climate analogues. This allows us to have an overview of the potential climate connectivity of 234 species with a virtually infinite dispersal capacity (realistic for highly dispersing species such 235 as some birds or large mammals that could easily travel hundreds of kms in this time frame) 236 and, in the case where we impose a dispersal constraint, to determine the more realistic 237 pathways for species with limited dispersal capacity. In total, eight scenarios per habitat type 238 were used to identify the climate analogues of the 1000 random starting points. Each starting 239 240 point (in blue on Figure 1) thus had 8 different climate analogues (in red or green on Figure 1). In addition, we selected for each starting point a random point that did not correspond to a 241 climate analogue. These locations were defined according to the same set of rules (in the same 242 habitat type – forest or open – and with or without dispersal limit) except that it did not 243 correspond to a location with similar climate in future conditions. This was done to compare 244 connectivity models built with the aim of simulating movements of range-shifting species to 245 246 models representing a non-climatic ecological connectivity.

Analogue points were obtained by considering all 19 bioclimatic variables with the same weight. We calculated the climatic similarity of all grid cells to each 1000 random starting point using functions in the "analogues" R package. Each grid cell was given a value of climatic similarity and we defined the climate analogue as the one with the highest climate analogy, i.e. the value closest to 1 (100% similarity). In order to identify climate analogues within a distance limit, we defined a radius around the point and selected the most climatically similar point located within this radius.

In order to test whether the choice of 1000 points was sufficient to represent climate connectivity across Europe, we generated connectivity maps (produced with the method outlined below) consisting of the sum of the connectivity across random combinations of 100 to 1000 (by increments of 100) pairs of starting points and their analogues. We calculated the correlation between each subset and the connectivity map produced with the whole set of 100 points, and concluded that 1000 points were sufficient (Appendix A, Figure A2).

#### 260 2.3. Connectivity modelling

In order to model connectivity between the initial random points and their respective climate 261 analogues, it was first necessary to define the resistance of the environment to species 262 movement. Resistance was defined according to land cover categories, using a geometric 263 sequence of common ratio 4, where the lowest resistance value was set to 1, i.e. the easiest to 264 cross. The highest resistance value corresponded to highly anthropized land cover types that 265 we assumed to be highly unsuitable for species, while impassable areas (all aquatic 266 267 environments) were set to NA, i.e. infinite resistance (Table 1). The use of a geometric sequence has been chosen because there is evidence that landscape resistance increases non-268 linearly with decreasing environmental suitability (Keeley et al., 2016). Here, the choice of a 269 ratio of 4 allowed for strong differences between low resistance and high resistance land 270 cover types. The land cover map was reclassified to obtain two resistance maps: one 271 corresponding to the resistance of the landscape to the movement of forest species and 272

another corresponding to the resistance of the landscape to the movement of species living inopen semi-natural habitats.

We applied an electric circuit approach to compute connectivity between the initial points 275 and their climate analogues (Figure 1). Circuit theory makes an analogy between the 276 resistance to species movement in the environment and the resistance to current flow in 277 electrical circuits (McRae et al. 2016). Contrary to other methods such as the identification of 278 279 least-cost paths, circuit-based methods assign a value to each pixel of the resistance map, allowing to interpret the output in a probabilistic way. To model movement corridors from the 280 initial random points to their future climate analogues across the resistance map, we used the 281 Circuitscape software (Version 1.5.3, Anantharaman et al. 2019). To obtain each connectivity 282 map, we provided Circuitscape with the coordinates of 1000 starting points, the coordinates of 283 their analogues and a resistance map. We set the "connect four neighbors only" parameter to 284 285 false to allow each grid cell to be connected to eight contiguous grid cell, thus identifying all possible paths. Each pair of points was subjected to a virtual electric current representing the 286 potential movement of the species (pairwise analysis). This approach results in a cumulative 287 288 map that sums the connectivity between all 1000 pairs of points, highlighting the areas of potential passage. In total, 16 connectivity maps based on 1000 pairs of points were obtained, 289 corresponding to each scenario and hypothesis (forest vs. open, RCP 2.6 vs. RCP 8.5, 2050 290 291 vs. 2070, unlimited vs. limited dispersal). The entire procedure was also carried out for nonanalogue points. 292

#### 293 2.4. Analysis of connectivity maps and analogues

294 First, we tested some prior hypotheses about the location of analogue points in response to the different scenarios. This was also a way to evaluate how those scenarios influenced the pairs 295 of analogues used for modelling connectivity before analysing connectivity models 296 297 themselves. Our hypotheses were that, in the furthest and most pessimistic scenario (RCP 8.5 in 2070), future climate analogues would be at greater distances, higher elevations, higher 298 latitudes and have lower climatic similarity than their counterparts in current climate than in 299 the closest and most optimistic scenario (RCP 2.6 in 2050). To do so, we extracted the 300 elevation of each point from the elevation raster available on the Worldclim website 301 (https://worldclim.org/), derived from the SRTM project. We calculated the geographical, 302 latitudinal and altitudinal distance between pairs of analogues, and extracted their climate 303 304 similarity as obtained at the stage of identification of climate analogues. We then determined which of the scenario, time interval, and dispersal hypothesis most influenced these variables 305 by performing multi-factor analyses of variance (ANOVA) with two-way interactions. 306

Besides visual inspection of connectivity maps, we first examined the variability of
connectivity depending on the climate scenario by calculating Pearson's correlation between
pairs of maps of the same habitat type, with and without dispersal limitations. Correlation was
also calculated between each climate connectivity map and the corresponding connectivity
model built from non-analogue points.

We then determined for each connectivity map the contribution of each country to the total, European-wide, connectivity, i.e. the amount of virtual current passing through each country. The purpose of this analysis was to determine which countries contributed the most to species' potential movement under various hypotheses. To do this, we extracted the sum of connectivity within each 44 European countries represented in our models. We also extracted their total area, such that we could plot the relationship between connectivity and area. Countries above the regression line were thus countries that contributed disproportionally to European climate connectivity compared to their area; reciprocally, countries below theconnectivity / area regression line contributed less than expected to the total connectivity.

Additional analyses were conducted to assess the overlap between modelled climate 321 connectivity and protected areas. All connectivity maps were combined into a single map to 322 facilitate calculations, but we also report results for forest and open habitats separately. We 323 conducted analyses for two different groups of protected areas: first considering all protected 324 325 areas to determine the amount of connectivity through these areas, then considering only strict nature reserves (IUCN category Ia). As described above, we plotted total connectivity within 326 protected areas per country against the amount of protected areas of each country to identify 327 countries where protected areas overlapped particularly well or not with climate connectivity. 328

We synthesized results into a comprehensive bivariate map showing climate connectivity 329 and its variation across hypotheses and scenarios. For this, we calculated the coefficient of 330 variation of connectivity for three groups of connectivity maps: all scenarios, forest habitats 331 only, and open semi-natural habitats only. In the three resulting bivariate maps, we partitioned 332 connectivity and its coefficient of variation into three quantiles, therefore producing 9 333 categories. The proportion of each category, as a whole and within protected areas only, was 334 extracted to identify the amount and stability of connectivity as a function of habitat type and 335 environmental protection. 336

337

## 338 3. Results

339 *3.1. Climate analogues* 

The first analysis revealed a significant effect of scenario, habitat, year, and distance limit on 340 the elevation, climate analogy, distance, and latitude of analogue points (Appendix A, Table 341 A1). On average, future climate analogues were located at higher elevation than in the 342 present, especially in the RCP 8.5 scenario and in 2070 (Figure 2). Climate analogy, distance 343 and latitude difference were influenced mainly by the climate scenario and the distance limit 344 (Figure 2 and Appendix A, Table A1). Climate analogy appeared to be greater in the RCP 2.6 345 scenario and when no distance limit was imposed (Figure 2). Geographical and latitudinal 346 distance between analogues were clearly higher in the RCP 8.5 scenario, although both 347 distances were decreased by the introduction of a dispersal limit (Figure 2). The scenario × 348 349 year factor was generally the most influential interaction, revealing that differences between 2050 and 2070 were exacerbated in the RCP 8.5 scenario (Appendix A, Table A1). 350

351 *3.2. Climate connectivity per country* 

We found that climate connectivity was correlated positively to available area (Figure 3), with 352 lower total climate connectivity in smaller countries and vice-versa. However, there were 353 notable exceptions, such as Malta which had zero connectivity regardless of the area and 354 surface considered. At the same time, the surface of protected areas also showed a clear 355 positive correlation with country area. Still, we observed that some of the largest countries 356 were not always the countries that protected their territory the most. Sweden and Norway 357 were the two countries with the largest areas of Ia IUCN category protected areas, although 358 they are only the 4<sup>th</sup> and 5<sup>th</sup> largest countries in Europe. On the contrary, although Turkey is 359 the largest country covered by our maps (ca. 780,000 km<sup>2</sup> including its Asian part), its surface 360 of protected areas was smaller than that of Denmark, which is only the 28<sup>th</sup> largest European 361 country (ca. 44,000 km<sup>2</sup>, Appendix A, Figure A4). Denmark, however, had lower than 362

363 expected connectivity, regardless of the area considered (in total or within protected areas,

Figure 3). Similarly, it appears that connectivity in Ia IUCN category protected areas was

lower than expected in the United Kingdom. Despite being among the smallest European
 countries, Liechtenstein (137 km<sup>2</sup>) had a connectivity comparable to that of Denmark, while

366 countries, Liechtenstein  $(137 \text{ km}^2)$  had a connectivity comparable to that of Denmark, wh 367 the connectivity of Luxembourg (2,608 km<sup>2</sup>) was close to that of Estonia (45,819 km<sup>2</sup>).

Focusing on forest and open semi-natural habitats, we observed again a strong correlation

between total connectivity and country area (Appendix A, Figure A3). However, Ireland and

- 370 Iceland had no connectivity in forested areas, regardless of the area considered. Another
- exception was Portugal, which had no connectivity in Ia IUCN category protected areas in
- 372 forested areas.

#### 373 *3.3. Stability of climate connectivity*

Overall, when looking at a broad European scale, climate connectivity maps highlighted the 374 same main areas in all hypotheses and scenarios (Appendix A, Figure A5). However, finer 375 examination revealed differences, mostly depending on the type of habitat considered and the 376 existence of a dispersal limit. Correlation analyses showed that, for a given dispersal and 377 habitat type, connectivity maps for the different scenarios and years of future climate change 378 were highly correlated with each other (Pearson's correlation > 0.8), with a slightly stronger 379 correlation without dispersal limits (Appendix A, Figure A5). Looking at the bivariate maps, 380 we observed high connectivity, although with some variations, in a large part of the 381 mountainous areas of Central Europe. The Bosphorus Strait (Turkey) also seemed to be 382 associated with a high climatic connectivity but with an important variation depending of the 383 384 habitat type and dispersal hypothesis. A strong, moderately variable, connectivity extended from northern Turkey to Poland, along the Balkan peninsula and the Alpine massif (Figure 4). 385 Fennoscandian countries, although strongly connected under some hypotheses, also exhibited 386 387 very variable connectivity, especially Sweden and Finland that were located in areas of high connectivity for forest species but not for those using open habitats. On the opposite, Turkey 388 and Ireland showed high connectivity for open semi-natural areas but less for forest habitats. 389 In addition, some areas exhibited high and stable connectivity, such as in the Carpathians in 390 Romania, the Massif Central in France and central Portugal (Figure 4). 391

When considering both habitat types together, the most frequently observed areas were 392 those of low connectivity and high variation in connectivity (13.6%), while the low-393 394 connectivity associated with low-variation areas were the least frequent (9.0%, Table 2). This 395 was the case also when looking at open semi-natural habitats only, but not for forests where medium connectivity with low variation was the most frequent (Table 2). Protected areas 396 397 appeared to be mostly located in highly connected and very stable areas (21.9%). In total, 46.6% of the surface of all protected areas were located in high-connectivity regions. Looking 398 at each habitat separately, we observed that climate connectivity remained predominantly 399 high in protected areas located in forest and semi-natural open habitats. More than half of 400 strongly protected areas (Ia IUCN category) were highly connected, including 30.8% with 401 high variation. This result remained observable in forested areas but appeared to be different 402 in semi-natural open habitats, where 35% of the strongly protected areas were in low-403 connectivity regions; the distribution of connectivity and its variation was also more 404 homogeneous. 405

#### 406 *3.4. Climate connectivity vs. non-climatic connectivity*

To identify whether connectivity between climate analogues differed from a more general
 definition of ecological connectivity in Europe, we produced connectivity maps between

- 409 random locations. In the case of a forested habitat and without distance limitation, the climate
- and random connectivity maps were correlated at more than 75% and were therefore very
- similar. The maps of semi-natural open spaces showed correlations of more than 85% with the
- 412 connectivity maps obtained from the climate analogue. Looking at the maps produced with
- distance limits, we observed that, in the case of forest as in the case of open semi-natural
- habitats, correlations that did not exceed 60%. Thus, the climate analogue and the non-
- analogue maps appeared in this case distinctly different (Table 3).
- 416

## 417 **4. Discussion**

In a context of climate change, conservation planning must take into account species 418 movements between their current habitat and the habitats they will need to reach in the future 419 to keep pace with temperature warming. Modelling climate connectivity according to 420 different scenarios of climate change and ecological hypotheses allows producing predictions 421 422 of the most likely pathways for species range shifts, and thus aid in the design of biodiversityfriendly landscapes. Several climate connectivity modelling studies have already been 423 conducted to identify species movement and facilitate species range change in increasingly 424 425 fragmented habitats (Littlefield et al., 2019). Such approach has been particularly carried out 426 for species-specific needs, or at the continental scale in America only (e.g. Littlefield et al., 2017; Carroll et al., 2018). Here, for the first time, we provide a mapping of climate 427 428 connectivity across Europe under various scenarios of climate change, landscape resistance and dispersal. We used these results to assess the coverage of climate connectivity by each 429 European country and their protected areas, providing a first step for future enhancements of 430 431 the network of protected areas within Europe.

432 First of all, the process of identifying climate analogues showed that none of the "analogues" we selected for modelling connectivity were 100% climatically similar to the 433 initial locations. This is certainly due to the fact that we computed climate analogy based on a 434 large set of variables (19); however, this also shows that climate change will favour the 435 emergence of novel climates (Williams et al., 2007) that do not currently exist in Europe. 436 True climate analogues may exist further than the extent of Europe that we analysed here; 437 however, since species cannot realistically travel at infinite distances, this suggests that many 438 organisms will face climatic conditions that they did no experience before, even if they are 439 capable of long-distance movements. In reality, species are not adapted to a single climate, 440 441 but possess a more or less wide climatic niche (Quintero and Wiens, 2013). Therefore, the likelihood for species to find suitable climatic conditions in the future will likely depend on 442 their niche properties (Thuiller et al., 2005) and on their ability to reach the areas harbouring 443 these climatic conditions (Estrada et al, 2016). In this regard, our modelling scenario where 444 we set a maximum dispersal distance highlighted further the impact that species' intrinsic 445 movement ability may have on their response to climate change. Here, the diminishing 446 447 climatic analogy shows that species with reduced movement capacity will have to cope with future climatic conditions that can be very different from now. Species with a narrow niche 448 breadth and limited dispersal ability are thus expected to be particularly sensitive to climate 449 change (Thuiller et al. 2005; Ofori et al., 2017). This result, however, is partly dependent on 450 the coarse resolution we used here; microclimate refugia may exist at a smaller scale that 451 could help species persist longer in their current habitats (Suggitt et al., 2011). 452

453 Our study demonstrated that difference in elevation, climate analogy, as well as
454 geographical and latitudinal distances between climate analogues, were influenced by the
455 scenario of future climate change. Specifically, it could be observed that the distance species

will have to travel to find the closest possible climate analogue is maximum in the case of 456 RCP 8.5; in this scenario, climate analogy will also be lower. The most environmentally 457 458 impactful scenarios thus appear to impose greater displacement to more distant and less climatically similar analogues (Littlefield et al. 2017), likely increasing the vulnerability of 459 little mobile species. Similarly, differences in elevation and latitude were also larger in the 460 461 case of RCP 8.5. This was expected as this is the most severe scenario of climate change (IPCC, 2018); our results confirm that species will most likely be impacted more strongly by 462 the most extreme changes in climate, accelerating extinction risk for many species (Urban, 463 2015). 464

The fact that elevation is higher and climate analogy is lower in 2070 than in 2050 shows 465 that climate change will continue to alter species' habitats over time regardless of the scenario 466 and habitat type considered. The movement of species to higher elevations as a response to an 467 468 increase in temperature conditions has already been shown, for example in bird species (Flousek et al., 2015). Species movement to higher elevations and latitudes, however, depends 469 on the species under consideration, with greater shifts in meridional areas (Chen et al., 2011). 470 471 This result has been proven in particular on bumblebee species, which moved to higher altitudes in the southern limits but decreased in distribution in the septentrional limits (Kerr et 472 al., 2015). Plant species, especially herbaceous, also follow a shift to higher elevations in 473 474 response to climate change (Lenoir et al., 2008). However, moving to higher altitudes is not always effective, especially since the topographical configuration of mountains leads to 475 decreasing available area at higher elevations (Elsen and Tingley, 2005). This has resulted, for 476 example, in large losses of butterfly diversity in mountainous areas of Spain (Wilson et al., 477 2007). Poleward range shifts are not necessarily an easier way for species to track their 478 climatic niche. Indeed, latitudinal range shifts often involve long-distance movements in a 479 fragmented landscape and require all ecological requirements to be present in novel habitats 480 (Chen et al., 2011; Devictor et al., 2012). It will therefore be important to preserve existing 481 habitats to slow down the extirpation of cold-adapted species and to promote the migration 482 and settlement of species in new habitats. 483

484 The synthesis of climate connectivity at the country level allowed us to determine which countries contributed the most to the predicted movements across Europe. Since we 485 determined that the amount of climate connectivity within countries is directly related to their 486 size, efforts directed towards assisting species' range shifts must be mainly accomplished in 487 the largest European countries such as France, Spain, Sweden, Norway and Germany, which 488 together host a large part of the predicted climate connectivity. Differences can be observed in 489 490 some countries, though. For example, we observed a remarkable lack of connectivity in Malta that is clearly due to its geographical location and size, Malta being an island at the very 491 southern European limit. The choice of points may also have influenced this result since the 492 probability of obtaining one out of 1000 random points in the smallest countries was low. 493 There was also a complete lack of connectivity in forested areas in Ireland and Iceland, which 494 is easily explained by the fact that very few forest areas are identifiable in these two countries. 495 The probability of getting one initial location in the forest of Ireland or Iceland out of 1000 496 497 points is therefore low; the probability that its future climate analogue is also located in the same country is even lower. To overcome the lack of connectivity in small- to medium-sized 498 islands, a small but non-null connectivity could have been allowed across seas to simulate 499 long-distance dispersers that can cross large water bodies. 500

The effectiveness of protected areas in mitigating the effects of climate change has
already been demonstrated for some species, such as breeding birds (Gaüzère et al., 2016).
Generally, protected areas appear to be a good way of providing protection and climate
refugia for climate-sensitive species at their trailing edge, as well as favouring the

colonisation of expanding species at their leading edge (Thomas et al. 2012; Gillingham et al. 505 2015). Here, we assessed whether European protected areas were placed in locations with 506 507 high climate connectivity, effectively protecting species experiencing range shifts. We first observed that some countries have significantly fewer protected areas than others, despite 508 their large size. For example, Turkey had a much smaller surface of protected areas, compared 509 510 to its total size, than most other countries (Appendix A, Figure A1). Regrettably, the Bosphorus was also the area with the largest sum of connectivity in several scenarios. There, 511 connectivity was very strong in all scenarios without dispersal limit, but also appeared 512 unstable because the assumption of a dispersal limit could prevent passage across the strait. 513 514 Looking closely at land cover in this region, we noticed that predicted movement occurred through heavily anthropized areas. Despite the high resistance values assigned to these land 515 cover types, it appeared to be the only possible way to go from Asia to Europe in search of 516 climate analogues. In fact, other corridors may exist around the Black Sea that could not be 517 modelled here because our land cover data did not encompass this region. Moreover, even if 518 519 we did not consider this possibility in our models, some species could disperse through water, swimming or transported by the current; flying or wind-dispersing species may also be able to 520 cross the short distance separating the European and Asian parts of Turkey (see e.g. Waisel et 521 al., 2008; Martínez-López et al., 2020). However, it is likely that crossing the Bosphorus will 522 523 be necessary for some species tracking their preferred climate, from Anatolia to Balkans for example. Therefore, it seems important that Turkey acknowledges its leading role in this 524 regard, for instance by increasing its network of protected areas to promote the climatic 525 526 connectivity of habitats. This could also be done by prohibiting the construction of human infrastructure in the residual natural areas, by limiting urban expansion along the coast, or by 527 promoting the development of green spaces in the existing urban areas. The construction of 528 529 ecoducts across the strait could also be a solution to remedy the difficulty of passage in this 530 area.

We detected in other areas such as in the vicinity of Zagreb in Croatia, in Liguria (north-531 western Italy) and around the Haut-Languedoc Regional Nature Park (southern France) a high 532 533 variation in the modelled connectivity. The dispersal limit was mostly responsible for this variation because it determines whether some relatively distant natural areas, separated by 534 highly anthropized land cover, could be connected or not. The largest region where variation 535 536 in connectivity was high is in Sweden and Finland. The instability in these areas is also partly due to the establishment of a distance limit, but is mainly caused by variation among habitats. 537 Indeed, Sweden and Finland show almost no connectivity in semi-natural open habitats, in 538 contrast to forest that is the dominant land cover in these countries. There, the maintenance 539 and expansion of efficient climate corridors will thus largely depend on the development of a 540 sustainable management of forested areas, which are currently mostly privately owned for 541 wood production (Chapin et al., 2007). This prominent importance of the dispersal limit in our 542 connectivity models also highlights that, on the opposite, the actual scenario and period of 543 climate change matters a lot less, making it easier to anticipate the effect of climate change on 544 species movements whatever the socio-economical trajectory society will follow. 545

546 It was visible from our connectivity maps that much of Central Europe has strong connectivity and stability, highlighting the high likelihood of climate-driven movements 547 connecting southern to northern Europe. Spain and Portugal are also two countries with strong 548 and very stable connectivity. Many protected areas are present in these regions and, generally, 549 we showed that protection overlapped more often with areas of stable connectivity (with the 550 exception of strict nature reserves, probably because they are mostly located in Fennoscandia, 551 552 where connectivity differs a lot between habitats). However, despite some positive signs (e.g. Gaüzère et al., 2016), the ability of protected areas to truly protect climate-tracking species 553

remains uncertain. For example, there are contrasting empirical evidence of the effectiveness 554 of the Natura 2000 network for biodiversity conservation, with demonstrated positive effect in 555 birds (Princé et al. 2021) but not in butterflies (Rada et al., 2019). In the context of climate 556 change, it is also important to ensure that spatial connectivity between protected areas is 557 effectively capable of providing continuous corridors for species on the move (Caplat et al., 558 559 2016; Santini et al., 2016). It is noticeable that the number of strict nature reserves is overall very low, and even non-existent in some of the most connected areas (e.g. all central Europe). 560 In order to fully protect these areas, it could be interesting to increase the level of protection, 561 including more wilderness areas and national parks that we did not consider separately in our 562 study but that confer a relatively strong protection for biodiversity. An interesting strategy 563 could also be to deliberately implement protected areas in regions of high variation in 564 connectivity in order to protect these areas from future, uncertain, changes. Issues regarding 565 the location of protected areas were also observed in the United States, where most regions of 566 high climate connectivity were not located in protected areas (Carroll et al., 2018). This is 567 therefore a complex challenge to identify the areas that must be protected in a context where 568 climate change drives spatial dynamics of biodiversity, especially when resources are not 569 infinite and must be allocated carefully. 570

Comparing maps of connectivity between climate analogues with non-climatic 571 572 connectivity maps allowed us to determine whether areas of high connectivity were strongly dependent on the exact location of connected points. In our results, climate and non-climatic 573 connectivity maps were very similar when no distance boundary was defined. Indeed, the 574 main areas of predicted movement remain broadly the same, regardless of the habitat 575 considered. This reinforces the idea that the areas we have identified are indeed areas of high 576 probability of passage to be preserved in the future, but also for ensuring ecological continuity 577 regardless of climate change. However, we also noted significant dissimilarities between 578 579 climate and non-climate connectivity maps in the case where we imposed a distance limit. It can be explained by the fact that, while long-distance connectivity always follows the same 580 main European corridors, connectivity within a smaller distance is highly dependent on the 581 582 exact location of the points. This is another confirmation that species' dispersal ability is one of the main factors to consider when predicting movements in a fragmented landscape, which 583 makes complex the implementation of protection strategies for multi-species aims. 584

585

## 586 **5. Conclusion**

587 We know that global change has a major impact on the movement of species. Limiting the ongoing warming – which undoubtedly involves considerable changes in human activities – is 588 thus crucial to limit the need for species to engage in long-distance movements to higher 589 590 latitudes and elevations. The lower climate analogies between current and future habitats under the more pessimistic scenario reveals the importance of preserving climate refugia. We 591 also demonstrated that species with limited dispersal capacity may face in the near future 592 593 climatic conditions that no longer match the ones they are currently adapted to. In addition, our study showed that species living on islands will be more vulnerable to climate change 594 because of the difficulty of dispersing to the continent. Long-term predictions of climate 595 596 change are also characterised by an increase in the distance between current and future analogue climates. Reducing our impact in the coming decades could therefore help species to 597 reach their climatic analogue more easily. Our approach also identified the most important 598 599 climate corridors, in part to determine which regions should be subject to the establishment of protected areas if they are lacking such. Interestingly, we showed that pan-European climatic 600

connectivity did not differ much between different scenarios of climate change, or between 601 connectivity modelled from now to 2050 or from now to 2070. This implies that modelling 602 603 climate connectivity may be an efficient tool to predict where species could go during range shift whatever the path we take – rapid mitigation of climate change or worst-case scenario 604 with a severe increase in temperature – and the time period considered. However, models 605 606 showed important differences depending on the dispersal hypothesis. Here, we tested only two assumptions of dispersal limit (unlimited or ca. 3 km yr<sup>-1</sup>); in reality each species will 607 disperse at its own pace, hence restructuring communities and complicating the modelling of 608 multi-species connectivity. We also proved that even when considering no climatic 609 connectivity between random locations, the areas of high connectivity remained broadly the 610 same as in climate connectivity models. The protection of climate corridors is therefore an 611 important issue both to help species reach their climate analogue and to limit extinction risk in 612 their initial habitats. It would be possible to go further by comparing our results with 613 historical data of bioclimatic and land use variables to determine if the climate corridors we 614 have identified could have been identified with past data. Our results could also be compared 615 with actual species movement data in Europe, to determine whether the identified climate 616 corridors will actually be used in the future. Finally, the framework we used here has the 617 potential to serve in the construction of land cover change scenarios to identify the best 618

- 619 strategies to improve climate connectivity.
- 620

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- 628

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## **Tables**

Table 1: Resistance values according to land cover categories. The resistance values are
scaled from 1 (low resistance) to 16348 (high resistance). NA values represent infinite
resistance. Land cover categories were classified according to the habitat group considered.
Initial random points and their analogues were selected within habitats with resistance = 1.

Resistance value	Forests	Semi-natural open habitats
1	Agro-forestry areas, Broad-leaved forest, Coniferous forest, Mixed forest	Pastures, Natural grasslands, Moors and heathland, Sclerophyllous vegetation Green urban areas, Sport and leisure facilities, Non-irrigated arable land, Annual crops associated
4	Green urban areas, Fruit trees and berry plantations, Olive groves, Sclerophyllous vegetation	with permanent crops, Complex cultivation patterns, Land principally occupied by agriculture, with significant areas of natural vegetation, Transitional woodland-shrub, Sparsely vegetated areas
16	Complex cultivation patterns, Natural grasslands, Moors and heathland, Transitional woodland-shrub	Permanently irrigated land, Inland marshes
64	Vineyards, Pastures, Annual crops associated with permanent crops, Land principally occupied by agriculture, with significant areas of natural vegetation	Peat bogs, Salt marshes, Salines
256	Sport and leisure facilities, Non-irrigated arable land, Permanently irrigated land, Rice fields, Inland marshes, Peat bogs	Rice fields, Vineyards, Fruit trees and berry plantations
1024	Sparsely vegetated areas	Olive groves, Agro-forestry areas, Broad-leaved forest, Coniferous forest, Mixed forest
4096	Road and rail networks and associated land, Beaches, dunes, sands, Bare rocks, Burnt areas, Glaciers and perpetual snow, Salt marshes, Salines, Intertidal flats	Beaches, dunes, sands, Bare rocks, Burnt areas, Glaciers and perpetual snow, Intertidal flats
16384	Continuous urban fabric, Discontinuous urban fabric, Industrial or commercial units, Port areas, Airports, Mineral extraction sites, Dump sites, Construction sites	Continuous urban fabric, Discontinuous urban fabric, Industrial or commercial units, Road and rail networks and associated land, Port areas, Airports, Mineral extraction sites, Dump sites, Construction sites
NA	Water courses, Water bodies, Coastal lagoons, Estuaries, Sea and ocean, NODATA, UNCLASSIFIED LAND SURFACE, UNCLASSIFIED WATER BODIES, UNCLASSIFIED	Water courses, Water bodies, Coastal lagoons, Estuaries, Sea and ocean, NODATA, UNCLASSIFIED LAND SURFACE, UNCLASSIFIED WATER BODIES, UNCLASSIFIED

Table 2: Summary table of the sum and variation of connectivity in Europe. Each cell in the
table represents the proportion (in %) of grid cells of each quantile of connectivity value and
connectivity variation in Europe, in the whole Europe, in total protected areas and in highly
protected areas.

		All Europe		All protected area		Strong protected area				
					Sum of connectivity					
		Low	Medium	High	Low	Medium	High	Low	Medium	High
Variation of connectivity	Total									
	High	13.6	10.0	9.7	9.7	6.5	7.5	15.3	13.5	30.8
	Medium	9.8	11.3	12.1	6.9	9.4	17.2	6.2	6.8	13.4
	Low	9.0	12.5	12.0	7.9	13.0	21.9	3.0	3.1	7.9
	Forests									
	High	11.4	10.0	11.9	8.5	9.9	13.9	7.0	9.8	28.3
	Medium	9.8	11.7	11.9	5.8	11.0	17.1	3.7	6.9	14.6
	Low	8.3	13.5	11.5	6.0	12.4	15.4	4.9	8.0	16.8
	Open semi-natural habitat									
	High	14.8	9.4	9.1	9.8	7.9	12.9	12.9	8.9	8.2
	Medium	9.4	11.5	12.5	7.2	10.2	19.4	12.9	11.9	13.7
	Low	7.8	13.1	12.4	6.0	9.7	16.9	11.7	7.6	12.2

Table 3: Pearson's correlation coefficients between climate connectivity maps and random
connectivity maps. A value of 1 represents perfect similarity, and a value of 0 represents no
similarity.

		Limited dispersal	Full dispersal			
Forest						
RCD 2 6	2050	0.79	0.58			
NCI 2.0	2070	0.84	0.59			
RCP 8 5	2050	0.81	0.58			
NCI 0.5	2070	0.75	0.60			
Open semi-natural habitats						
RCP 2 6	2050	0.85	0.49			
NCI 2.0	2070	0.85	0.55			
RCD 8 5	2050	0.85	0.49			
1.01 0.5	2070	0.87	0.52			

## 874 Figure captions

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**Figure 1**: Illustration of the methods used for modelling climatic connectivity. A randomly

chosen point (blue) and its different climate analogues (red and green) are shown in A.

878 Connectivity modelled using Circuitscape between the initial point and its analogue in 2050 in

- the RCP 8.5 scenario with full dispersal is shown in B, and with its analogue in 2070 in the
- 880 RCP 8.5 scenario with no dispersal in C. We produced in total 16 European-scale connectivity
- maps by modelling connectivity between 1000 pairs of points for forest and open species,
  with or without dispersal limit, for two climate change scenarios and two future time periods.

883

- **Figure 2**: Difference between the 1000 random points and their future climate analogues.
- 885 Box-and-whisker plots show the altitudinal and latitudinal difference, as well as the
- geographical distance and climate analogy, between the initial points and their analogues, for
- forest and open semi-natural habitats, for both RCP 2.6 and 8.5 scenarios, and with or without
- dispersal limits. Mean values are shown as black crosses within boxplots.

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Figure 3: Relationship between the sum of climate connectivity within each Europeancountry and their total area (left), within protected areas (middle), and within strictly protected

- country and their total area (left), within protected areas (middle), and within strictly protected
   areas only (IUCN category Ia) (right). Countries are represented by three letters from the ISO
   atom dord. Both connectivity and area were log transformed
- standard. Both connectivity and area were log-transformed.

894

**Figure 4**: Bivariate maps representing the sum of connectivity and its variation (coefficient of

- variation), calculated for all connectivity maps merged together or for connectivity maps in
- forests or in open semi-natural habitats only. Areas represented in grey have no connectivity
- 898 value.

А

Analogue RCP 8.5 - 2070 (full dispersal)

Analogue RCP 2.6 - 2070 (full dispersal)

Analogue RCP 8.5 - 2050 (full dispersal)

Analogue RCP 8.5 - 2070 (limited dispersal)

**Random initial point** 

Analogue RCP 2.6 - 2050 (full dispersal)

Analogue RCP 2.6 - 2070 (limited dispersal)





Analogue RCP 8.5 - 2070 (limited dispersal)

Random initial point









