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Spatial sorting and range shifts: consequences for evolutionary potential and genetic signature of a dispersal trait

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Abstract

Species are shifting their ranges under climate change, with genetic and evolutionary consequences. As a result, the spatial distribution of genetic diversity in a species’ range can show a signature of range expansion. This genetic signature takes time to decay after the range stops expanding and it is important to take that lag time into account when interpreting contemporary spatial patterns of genetic diversity. In addition, the return to spatial equilibrium on an ecologically relevant timescale will depend on migration of genetic diversity across the species’ range. However, during a range shift alleles may go extinct at the retracting range margin due to spatial sorting. Here we studied the spatial pattern of genotypes that differ in dispersal rate across the species range before, during and after a range shift, assessed the effect of range retraction on this pattern, and quantified the duration of the ephemeral genetic signature of range expansion for this trait. We performed simulation experiments with an individual-based metapopulation model under several contemporary climate change scenarios. The results show an increase of the number of individuals with high dispersal rate. If the temperature increased long enough the allele coding for low dispersal rate would go extinct. The duration of the genetic signature of range expansion after stabilisation of the species’ distribution lasted up to 1200 generations after a temperature increase for 60 years at the contemporary rate. This depended on the total displacement of the climate optimum, as the product of the rate of temperature increase and its duration. So genetic data collected in the field do not necessarily reflect current selection pressures but can be affected by historic changes in species distribution, long after the establishment of the current species’ range. Return to equilibrium patterns may be hampered by loss of evolutionary potential during range shift.

Keywords: Metapopulation; individual-based model; climate change;
**Introduction**

Species are shifting their ranges under climate change (Chen et al., 2011; Parmesan and Yohe, 2003), which has genetic and evolutionary consequences (Excoffier et al., 2009; Kubisch et al., 2014; Parmesan, 2006). The genetic diversity present at the expanding range margin is smeared across the landscape on the expansion wave (Excoffier and Ray, 2008). This is the case for neutral as well as adaptive genetic variation (Cobben et al., 2012a; Cobben et al., 2011; Edmonds et al., 2004; Hewitt, 1996; Ibrahim et al., 1996; Klopstein et al., 2006; Travis et al., 2010; Travis et al., 2007). Under these conditions, the genetic configuration of the newly colonised populations is largely the result of the demographic process under range expansion, and not of selection (Travis et al., 2007), although some traits are selected for under range expansion, particularly the ability to disperse and traits related to population growth rate (Hill et al., 2011; Moreau et al., 2011; Phillips et al., 2010).

With regard to the ability to disperse, both theoretical and empirical studies report an increased dispersal capacity as a result of spatial sorting under range expansions (e.g. Burton et al., 2010; Phillips et al., 2006; Thomas et al., 2001; Travis and Dytham, 2002; Travis et al., 2013). Good dispersers gather at the range margin and from there they colonise new territory, while the poor dispersers lag behind. This eventually leads to a spatial gradient in dispersal capacity across the species’ range, which can be regarded as a genetic signature of range expansion (Phillips et al., 2010). After the range stops expanding, it takes time for such a genetic signature of range expansion to decay, especially when the variation for these traits needs to migrate from the centre of the range or to develop through *de novo* mutations (Dytham et al., 2014; Phillips et al., 2010). During that time period populations are in spatial disequilibrium as a result of the past range expansion. In a time when many investigators are gathering genetic data from natural populations to study selection pressures and micro-evolution, accounting for different explanations of genetic make-up is crucial (Currat et al.,...
Investigating the genetic signature of range expansion is therefore of importance for the interpretation of contemporary spatial patterns in genetic diversity.

Particularly for a dispersal trait, it can be expected that the decay of this genetic signature of range expansion may take a long time: after the range expansion stops the net effect of selection will be for lower dispersal capacity, yet low dispersal genotypes are by definition slow dispersers. Kubisch et al. (2010) showed that the establishment of genotypes with low dispersal rates after range expansion is the result of the migration of variation, and does not involve the establishment of new, beneficial mutations. In their study we see a slow return to equilibrium dispersal rate values at the expanding range margin after range expansion, but this result is not specifically quantified or discussed. In contrast, Henry et al. (2014) claimed a fast return to equilibrium values after range expansion, resulting from selection for beneficial mutations. In a mechanistically more realistic model, Dytham et al. (2014) showed that the state of spatial disequilibrium can last for a substantial time period after the range expansion. In addition, they observed that the speed of range expansion depends on selection for existing variation rather than for new variation due to mutations, even under a high mutation rate (Dytham et al., 2014). This is in line with theory and empirical data on micro-evolution, in which evolutionary changes that are relevant at ecological timescales depend on changes in allele frequencies rather than on new mutations. However, none of the above studies have explicitly discussed or tried to quantify the time period during which populations are in spatial disequilibrium.

Under a climate change scenario with continued temperature increase, many species’ ranges are expected to retract at the margin with the deteriorating thermal conditions, resulting in a range shift rather than a range expansion (Gillings et al., 2015; Thomas et al., 2006). This means that slow dispersers might be lost as a result of spatial sorting and
subsequent extinction of populations at the retracting margin (Cobben et al., 2012a; Cobben et al., 2011), while return to equilibrium values after range expansion likely depends on such existing genetic diversity as argued above (Dytham et al., 2014; Kubisch et al., 2010). Many studies have reported increased dispersal under range expansion (e.g. Burton et al., 2010; Phillips et al., 2006; Thomas et al., 2001; Travis and Dytham, 2002; Travis et al., 2013), but little attention has been given to the consequences of a retracting range margin for dispersal traits (but see Henry et al., 2014).

In this paper we investigated how range shifts affect the distribution of different genotypes coding for low, medium and high rates of dispersal across the species range. In addition, we studied the duration of the ephemeral genetic signature of range expansion. For both, we used a mechanistically realistic model of a range shifting species under climate change. We simulated a period of temperature increase in several scenarios, during and after which we registered the distribution of the different dispersal rate genotypes in the landscape up until 5000 years after model initialisation. To warrant sufficient variation in the existing genetic diversity under stationary conditions, we employed diploid inheritance, fragmented habitat and temperature variability. The genetic architecture was designed to get a clear signature of spatial sorting as well as distinct differences between phenotypes.

Methods

We used a spatially explicit, individual-based simulation model of a sexually reproducing species with overlapping generations. The model is called METAPHOR (Verboom et al., 2001; Vos et al., 2001) and has previously been extended to allow for stochastic temperature increase (Schippers et al., 2011) and with a genetic module for neutral (Cobben et al., 2012b; Cobben et al., 2011) and adaptive traits (Cobben et al., 2012a). It was parameterised based on empirical data for the middle spotted woodpecker (Dendrocopus medius), a woodland bird
Table 1 gives an overview of the parameters used. For the current study we allowed the dispersal rate to be adaptive and investigated the spatial distribution of the allele frequencies at the dispersal gene during and after range shift.

**Landscape**

The landscape was 15 km wide (horizontal axis) by 2000 km long (vertical axis). In horizontal direction the borders were merged, creating a (vertical) tube. The landscape contained 3000 randomly distributed habitat patches, equalling a total of 5% habitat area. In vertical direction, a Gaussian temperature curve was shifting across the landscape at various speeds, and with a standard deviation of 140 km (see Appendix A). This temperature was used to determine the local habitat quality and each row in the landscape was as such characterised by a time-specific habitat quality $q$ [0..1], being 1 at the peak of the distribution and 0 outside the curve area. We used three temperature increase scenarios, based on work by the IPCC (2013), of a respective temperature increase of 1°C, 2°C and 4°C by the year 2100, resulting in equivalent isocline shift rates of 2, 4 and 8 km per year, respectively (see Appendix A).

**Population demography**

The patches in the landscape were inhabited by individuals that were characterised by their sex, and their dispersal alleles (see below under Genetics). The yearly life cycle consisted of recruitment, dispersal, and survival, in this order. Per patch, in each year, the number of offspring per unique pair of individuals of opposite sex was dependent on local habitat quality $q$ and population density $d$, following:
\[ R = R_{d0q1} \left( 1 - \left( 1 - \frac{R_{d0q0}}{R_{d0q1}} \right) * (1 - q) \right) \left( 1 - \left( 1 - \frac{R_{d1q1}}{R_{d0q1}} \right) * d \right) \]

with \( R_{d0q1} \) is the number of offspring at \( d = 0 \) and \( q = 1 \), and similar for \( R_{d0q0} \) and \( R_{d1q1} \).

The individual dispersal rate depended on its diploid genotype and can be low (zero), medium (0.1) or high (0.2), indicating its yearly dispersal probability. Upon dispersal, individuals could leave their patch in every direction in a straight line, and were then assigned to the first patch they encountered within their maximum dispersal distance of 15 km. The individual survival probability depended on population density \( d \) and habitat quality \( q \).

\[ s = s_{d0q1} \left( 1 - \left( 1 - \frac{s_{d0q0}}{s_{d0q1}} \right) * q \right) \left( 1 - \left( 1 - \frac{s_{d1q1}}{s_{d0q1}} \right) * (1 - d) \right) \]

with \( s_{d0q1} \) is survival rate at \( d = 0 \) and \( q = 1 \), and similar for \( s_{d0q0} \) and \( s_{d1q1} \).

Genetics

There were two alleles at the dispersal locus. Allele A signified zero dispersal rate and allele B coded for 0.2 dispersal rate (i.e. there was a 20 percent chance that an individual left the habitat patch in which it was born). The individuals were diploid and the alleles codominant. The actual probability of dispersal was calculated as the arithmetic mean of the alleles that they had randomly inherited from each parent: genotype AA is a dispersal rate of 0, AB is 0.1 dispersal rate and BB is a dispersal rate of 0.2. Pilot runs included additional alleles 1. with higher dispersal rates, but these were never maintained beyond model initiation, and 2. with finer-scaled rates, many of which were lost in more or less random patterns, and thus blurring the signal of spatial sorting. Each allele could mutate at a rate of \( 10^{-6} \) (Nachman and Crowell, 2000) to the other allele. Evolution in this genetic model thus depended mostly on standing genetic variation, but had clear differences between phenotypes, while the codominance allowed the spread of the A allele in AB individuals (Bell and Aguirre, 2013; Schluter and Conte, 2009; Van Belleghem et al., 2014).
Simulation experiments

At initialisation of the model, all habitat patches were filled with 10 adult individuals, equalling half the carrying capacity $K$. Each individual was randomly given two dispersal alleles. The temperature optimum $Y_{opt}$ was initialised at 400 km from the southern landscape edge. After initialisation the model was run for 500 generations, equalling 500 years, to establish equilibrium population demography and dispersal rate values. After this burn-in period we applied the stochastic temperature increase for 500 years at three isocline speeds of 2 km yr$^{-1}$, 4 km yr$^{-1}$ and 8 km yr$^{-1}$. In a second experiment we used the same three different isocline shift rates, but combined with three time periods of temperature increase each, and monitored the population establishment and the distribution of genotypes for 5000 years afterwards (see Table 2). We performed 10 replicate simulations for both experiments.

Additional control simulations included different standard deviations of the mean speed of the temperature curve of zero and 280 kilometres, and a different mutation rate of 10$^{-6}$.

Analysis

We summed the average number of genotypes over 10 replicate simulations in landscape blocks of 50 km in vertical direction. The equilibrium genotype distribution was defined as the average number of individuals of each genotype in each 50 km-block across all replicates of all simulations in year 500 (the year in which we start the temperature increase, so after the burn-in phase and prior to any disturbances). To quantify the deviation from the equilibrium range size and location after the stabilisation of the temperature ($Deviation\ I$), we calculated in each year and each 50 km-block the sum of the absolute differences in total number of individuals, corrected for the location of the climate optimum (which is moving during the period of temperature increase). Similarly, to quantify the deviation from the equilibrium
genotype distribution after the stabilisation of the temperature, we calculated in each year and
each 50 km-block the sum of the absolute differences in individual numbers of each of the
different genotypes, corrected for the location of the climate optimum (Deviation II).

\[
\text{Deviation I, } t = \sum_{i=1}^{15} \left( |\sum_{x=1}^{3} (N_{x_i, t=0}) - \sum_{x=1}^{3} (N_{x_i, t})| \right)
\]

\[
\text{Deviation II, } t = \sum_{x=1}^{3} \left( \sum_{i=1}^{15} |N_{x_i, t=0} - N_{x_i, t}| \right)
\]

with number of individuals \( N \) of genotype \( X \) (AA, AB or BB) in range block \( i \) north of the
climatic optimum, with the location of the climate optimum dependent on time and simulation
scenario.

By comparing these two deviations for all years after the stabilisation of the
temperature, we could see the difference between the moment that the full species range is re-
established after climate change stops and the moment that the equilibrium genotype
distribution is re-established. This difference in time was the period in which the observed
genotype distribution in the species range is the consequence of past range expansion, the so-
called ephemeral signature of range expansion, and it was not indicative of current and local
selection pressures.

Results

Loss of the A allele at the trailing edge

At the end of the burn-in phase, so under equilibrium conditions, the metapopulation
consisted of individuals with dispersal rates of 0, 0.1 and 0.2 (genotypes AA, AB and BB
respectively; Figure 1 year 0). All three genotypes were present in the total range but the
individuals with 0 dispersal rate (AA) dominated the centre of the range, while the 0.2
individuals (BB) had the highest frequency in the margins, where habitat quality and
population densities were lower. The 0.1 individuals (AB) took an intermediate position. The
control simulations showed a comparative increase in numbers of dispersive genotypes with
increasing standard deviations of the mean speed of the temperature curve, but overall decreasing population numbers (Figure S1 in Appendix B).

The original distribution of the populations and the genotypes within the metapopulation changed under the range shift that was induced by the temperature increase in the first experiment (Figure 1). The range shifted, but not as fast as the climate optimum moved, and this lag resulted in a decline in the metapopulation size in time. The decline was faster with increasing speed of the climate optimum and an increasing standard deviation of this mean speed across years (Figure S1 in Appendix B). In addition, there was an increase of the proportion of 0.2 dispersal rate individuals and the relative size of the area where they lived, at the expanding edge of the metapopulation. The number of lower dispersal rate individuals decreased and eventually the A allele went extinct as the habitat quality in the southern populations fell below the threshold value (Figure 1 yr 500). This pattern occurred in all simulations in which the metapopulation persisted sufficiently long (e.g. Figure S1 in Appendix B).

The genetic signature of range shift In experiment 2, once the temperature stabilised, the distribution of the genotypes re-established their equilibrium, but only after a certain amount of time (Figure 2). This amount of time depended on the total displacement of the temperature optimum, so the speed of the isotherm shift multiplied by the number of years of temperature increase. This displacement also determined the shape of the deviation curve (deviation II being the measure for how far the metapopulation was from spatial equilibrium with regard to the distribution of the different genotypes), with a typical initial increase in the genotype distribution deviation for large displacements (Figure 3c). The genotype distribution deviation was always larger than the range deviation (Figure 3), but for small displacements the lines ran in parallel (Figure
3a). For intermediate displacements, both deviations converged towards equilibrium values, with the range deviation levelling off first (Figure 3b). This means that when the range reached its equilibrium size, the spatial pattern of the different dispersal genotypes was not yet in the equilibrium situation.

Discussion

We have investigated how range shifts affect the distribution of genotypes of different dispersal rates across the species range. The combination of spatial sorting and range retraction could lead to the extinction of the A allele, coding for low dispersal rate. When the temperature stabilised before the extinction of the A allele, the equilibrium spatial distribution of genotypes was able to recover. This, however, took an extensive amount of time, dependent on the total displacement of the temperature optimum for the investigated range of temperature increase rates. We expected that for low rates of temperature increase the genotype distribution would differ less from the equilibrium distribution under range shift, because then the selection pressures hardly change compared to a stable range. It is thus important to note that for the investigated system any realistic scenario of the rate of temperature increase under climate change had an effect on the spatial distribution of the genotypes. As such, recovery time here did not depend on the rate of temperature increase per se, but on the final total displacement of the temperature optimum. After the range shift was completed and the species’ range had reached its equilibrium size, it took up to 20 times the duration of the temperature increase period before the spatial distribution of genotypes was at equilibrium. Although the genetic signature of range expansion is ephemeral, it can affect the spatial distribution of genotypes in the landscape for a long time period after the range stops expanding. This is important for the interpretation of genetic field data, since signatures of
adaptive evolution in genetic data can mimic signatures of range expansion (Currat et al., 2006; Ray and Excoffier, 2009).

The metapopulation in equilibrium consisted of individuals with 0, 0.1, and 0.2 dispersal probability. The observed gradient of increasing dispersal rate from the core of the metapopulation to the range margins at equilibrium was as anticipated, as dispersal is essential at the margin due to the variable habitat quality, but only occasionally beneficial in the range core as a result of the high densities there (Clobert et al., 2004; Ronce, 2007). This led to a frequency-dependent selection, with a large proportion of non-dispersers and lower proportions of dispersive individuals in the core of the range. Control simulations initialised with additional alleles coding for higher dispersal rates, showed that the relative costs of dispersal were too high anywhere across the range for their persistence in the metapopulation at equilibrium. During the range shift the relative number of individuals with 0.2 dispersal rate increased as a result of spatial sorting (Shine et al., 2011). Again, in control simulations allowing for mutations with higher dispersal rates, we observed that such mutations could arise, but they never managed to establish themselves in the population during range shift.

While most studies focus on the evolution of dispersal traits under range expansion (Excoffier et al., 2009; Kubisch et al., 2011; Shine et al., 2011; Thomas et al., 2001; Travis and Dytham, 2002), we have here used a range shift scenario. We showed that in the retracting part of the range the genetic variation coding for lower dispersal rates may go extinct (Figure 1). As soon as the lower dispersal individuals were confined to populations near the southern range margin where local selection favours higher dispersal, there was no chance of recovery for the A allele. This happened even though the metapopulation under equilibrium conditions consisted for more than two thirds of individuals carrying an A allele, so with 0 and 0.1 dispersal rates (Figure 1). Such loss of genetic variation at the trailing edge of a shifting metapopulation is a so far non-described consequence of spatial sorting (Shine et
While spatial sorting is usually referred to as the phenomenon that dispersive genotypes gather at the expanding range margin (Phillips et al., 2010; Shine et al., 2011), our results indicate that this concept may be broadened to include the confinement and possible extinction of poor disperser alleles at the retracting range margin under range shift. This may cause a range-wide loss of adaptive genetic variation (Cobben et al., 2012a; Henry et al., 2014). Whether such a phenomenon can be or has been observed in the field we do not know. While the importance of rear edge populations for the conservation of species genetic diversity has been noted (Hampe and Petit, 2005), this concerned populations that are older than the populations in the core and leading range margin and have persisted in an ice age refugium. In addition, loss of genetic diversity in small and isolated populations as a result of genetic drift is a well-established theory and has been observed in empirical studies as well (Pearson et al., 2009). However, observing the loss of genetic diversity as a result of spatial sorting would require a well-advanced range retraction, while the first retractions have only been observed recently (Gillings et al., 2015; Thomas et al., 2006). Historical and ongoing range contractions are plentiful (Channell and Lomolino, 2000; Laliberte and Ripple, 2004; Rodríguez, 2002), but are not necessarily associated with a range shift, and information about the adaptive genetic composition of such species’ ranges is not available at the moment (but for neutral diversity see Koen et al., 2014).

The disequilibrium distribution of genotypes after the range stops expanding, and its long duration may reduce the species’ fitness (Cobben et al., 2012a; Dytham et al., 2014) as it represents a form of maladaptation. Here we showed in an ecologically realistic model that if the variation for a specific trait needs to migrate or develop by de novo mutations, and the trait itself is related to a reduced tendency to disperse, this period of disequilibrium may last for hundreds or even thousands of years. Control simulations with an increased mutation rate of $10^{-4}$ seldom showed the establishment of a low dispersive mutation which then increased
the return rate to equilibrium. Similar studies report a lengthy phase of spatial disequilibrium and a dependency on existing genetic diversity rather than of new mutations (Dytham et al., 2014; Kubisch et al., 2010). In contrast, Henry et al. (2014) mention a fast return to equilibrium distributions of dispersal rates after range expansion due to the establishment of beneficial mutations. They however used a high mutation rate of $10^{-3}$, a larger model population, and haploid inheritance. Under diploid inheritance, the phenotypic effect of a beneficial mutation is diluted in heterozygotes. As long as the allele frequency is low the allele will exist almost completely within heterozygotes. In diploids the return to spatial equilibrium will therefore depend on the migration of existing (standing) genetic variation (Schluter and Conte, 2009; Van Belleghem et al., 2014) rather than on mutations.

Burton et al. (2010) showed in a modelling study that the presence of a competing species in the region to be colonised can greatly reduce the extent to which dispersal traits are selected for in populations at the leading edge. Additionally, in a large metapopulation as the one we modelled, you may expect adaptation to local environmental conditions, although for small subpopulations such adaptation is less common (Leimu and Fischer, 2008). Bourne et al. (2014) showed that the spread of a genotype that is beneficial under climate change only facilitates population rescue when there is no strong local selection against non-specialists. Such local adaptation could therefore be of consequence (Kubisch et al., 2013), especially when there is genetic linkage between the underlying genes or it essentially involves the same traits (Brown et al., 2007).

With continuously changing environmental conditions and land use patterns we can safely assume that many species (or species traits) are permanently in disequilibrium. In Europe, even now, some species are still expanding their ranges after the last glacial maximum. This means that observed phenotypes in the field cannot be assumed to perfectly reflect the local and current selection pressures (Currat et al., 2006; Edmonds et al., 2004;
Klopfstein et al., 2006; Lindström et al., 2013; Ray and Excoffier, 2009). However, field studies investigating changing local population composition in the period after colonisation have shown that less dispersive phenotypes can be found within years (Lindström et al., 2013) to decades (Thomas et al., 2001) after initial colonisation. This could be the result of the genetic architecture and heritability of the dispersal trait, or of trade-offs with other traits. For the cane toads e.g. (in Lindström et al., 2013) the dispersal capacity is the result of a combination of many different traits (Phillips et al., 2010), which are combined during spatial sorting (Shine et al., 2011). The individuals that have replaced the fast dispersers within eight years after initial colonisation had about half of the dispersal distance of the pioneer toads (Lindström et al., 2013), and there was a continuous strong selection against fast dispersal in these toads (Brown et al., 2007). Dispersal traits are very diverse, and it is no exception that they are polygenic, although heritability is mostly greater than 0.3 (Ronce, 2007). Conditional dispersal has been observed (Clobert et al., 2004; Donohue, 1999; Imbert and Ronce, 2001), but often the relative role of genetic changes and plasticity in changes of dispersal phenotypes is unclear (Ronce, 2007). We have modelled dispersal as a single gene trait and completely heritable, in line with particular empirical evidence (Haag et al., 2005; Roff, 1986; Van Belleghem et al., 2014), to quantify the duration of the genetic signature of range expansion for a trait for which the variation either needs to migrate or mutate. Control simulations including more alleles showed qualitatively similar (yet less distinct) patterns and equal recovery times. If the low dispersive A allele was recessive (Roff, 1986), the AB and BB genotypes would have the same phenotype of 0.2 dispersal rate. Additionally, the AB genotype was then expected to be present at low frequencies at the range margin under stable conditions (Wright, 1969). This may be sufficient to prevent the colonisation lag for the A allele and in any case likely shortens the period of spatial disequilibrium. Increasing complexity of genetic architecture to include multiple loci or regulator genes, or including
multiple traits might further influence spatial disequilibrium patterns and evolutionary potential and remains an interesting field of future investigation. The current study shows that co-dominant heritable genetic variation for low dispersal rates on a single gene can cause a long period of spatial disequilibrium and loss of evolutionary potential under range shifts due to spatial sorting.

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References


## Tables

### Table 1. Model parameters and variables used.

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Table 2. Model simulations in both experiments, with rates of isocline shift and the periods for which they were simulated. For experiment II the total displacements of the climate optimum after the period of temperature increase are given.

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<tbody>
<tr>
<td>2 km year(^{-1})</td>
<td>500 yrs</td>
<td>50, 100, 200 yrs</td>
<td>100, 200, 400 km</td>
</tr>
<tr>
<td>4 km year(^{-1})</td>
<td>500 yrs</td>
<td>30, 60, 120 yrs</td>
<td>120, 240, 480 km</td>
</tr>
<tr>
<td>8 km year(^{-1})</td>
<td>500 yrs</td>
<td>10, 20, 60 yrs</td>
<td>80, 160, 480 km</td>
</tr>
</tbody>
</table>
Figure legends

Figure 1. The distributions of the 0 (light grey), 0.1 (medium grey), and 0.2 (dark grey) dispersal rate individuals in time in the landscape under the range shift caused by the temperature isocline shift rate of 2 km year\(^{-1}\), stopped after 500 years. The bold black bars indicate the spatial locations of the average temperature optimum in the specific years.

Figure 2. The distributions of the 0 (light grey), 0.1 (medium grey), and 0.2 (dark grey) dispersal rate individuals in time in the landscape under the range shift caused by the temperature isocline shift rate of 8 km year\(^{-1}\), stopped after 60 years. The bold black bars indicate the spatial locations of the average temperature optimum in the specific years.

Figure 3. The range size deviation of total numbers of individuals in time (deviation I, -●-) and the genotype distribution deviation (deviation II, -○-) in time since the year of temperature stabilisation for every scenario. The deviations are calculated as the sum of the absolute differences in numbers of individuals between the equilibrium distribution and the distribution in each year, per 50 km-block. a. Scenario of 2 km year\(^{-1}\) isocline shift for 50 years, b. Scenario of 4 km year\(^{-1}\) isocline shift for 60 years, c. Scenario of 8 km year\(^{-1}\) isocline shift for 60 years.
Figure 1

Figure 2
Figure 3a

Figure 3b
Figure 3c
Appendix A

We used three temperature increase scenarios, based on work by the (IPCC, 2013), of a respective temperature increase $I$ of 1°C, 2°C and 4°C by the year 2100. When translating these scenarios to temperature isocline shift rates, we assumed that the simulated metapopulation was situated along the European Atlantic coast. The Hadley Centre (2003) predicts no regional differences in change of temperature within this stretch of coast for the year 2080. We therefore assumed a similar rate of temperature increase across the simulated landscape. Unaffected by mountain ranges, this coast line has a temperature gradient $G$ of 0.0042 °C per km (Schippers et al., 2011). The current average standard deviation of the average temperature $\sigma_t$ in this region is 0.59 °C (Schippers et al., 2011). We used the temperature gradient to convert the explored temperature increase rates and the yearly temperature variability, in °C per year and °C respectively, to geographical distances. This leads to the speed with which temperature isoclines travel north ($T$ km yr$^{-1}$) and the yearly fluctuation of these lines ($\sigma_d$ km), by using:

$$T = \frac{I}{G},$$

$$\sigma_d = \frac{\sigma_t}{G}$$

Following from these, the three temperature increase scenarios resulted in equivalent isocline shift rates $T$ of 2 km yr$^{-1}$, 4 km yr$^{-1}$ and 8 km yr$^{-1}$ respectively, with a standard deviation of the temperature optimum $\sigma_d$ of 140 km. We then calculate the location of the optimal temperature in north-south direction ($Y_{opt}$) in a given year $t$ as:

$$Y_{opt,t} = Y_{opt,0} + T * t + \sigma_d * N_t$$

with $N_t$ is the yearly random number drawn from a standard normal distribution and $Y_{opt,0}$ is the location of the optimal temperature at $t = 0$. 

...
Figure S1. The number of individuals with dispersal probabilities of 0 (0%DP), 0.1 (10%DP), and 0.2 (20%DP) and the other (>0.2 dispersal rate) individuals (rest) and their sum (TOT) in time under the temperature increase scenario of 2 km/year for standard deviations of the mean speed of the shift of the temperature curve of 0 km (a), 140 km (b) and 280 km (c).
c)

![Graph showing population decline over years for different scenarios: 0% DP, 10% DP, 20% DP, rest, TOT. The y-axis represents the number of individuals, and the x-axis represents the year. The graph shows a general decline in population with time.]