

1 Evolution of dispersal can rescue populations from 2 expansion load

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12 **Abstract**

13 Understanding the causes and consequences of range expansions or range shifts has a
 14 long history in evolutionary biology. Recent theoretical, experimental, and empirical
 15 work has identified two particularly interesting phenomena in the context of species
 16 range expansions: (i) gene surfing and the relaxation of natural selection, and (ii) spatial
 17 sorting. The former can lead to an accumulation of deleterious mutations at range
 18 edges, causing an expansion load and slowing down expansion. The latter can create
 19 gradients in dispersal-related traits along the expansion axis and cause an acceleration
 20 of expansion. We present a theoretical framework that treats spatial sorting and gene
 21 surfing as spatial versions of natural selection and genetic drift, respectively. This model
 22 allows us to study analytically how gene surfing and spatial sorting interact, and to
 23 derive the probability of fixation of pleiotropic mutations at the expansion front. We
 24 use our results to predict the co-evolution of mean fitness and dispersal rates, taking
 25 into account the effects of random genetic drift, natural selection and spatial sorting, as
 26 well as correlations between fitness- and dispersal-related traits. We identify a "rescue
 27 effect" of spatial sorting, where the evolution of higher dispersal rates at the leading
 28 edge rescues the population from incurring expansion load.

29 Introduction

30 Understanding the demographic, ecological, and evolutionary forces that determine the
 31 evolution of a species range has been a central area of research since the early days of
 32 evolutionary biology (Darwin, 1859; Sexton et al., 2009). Over the last decade, the fact
 33 that species range expansions impact multiple evolutionary and ecological processes in
 34 peripheral populations has been thrown into the spotlight both theoretically and empiri-
 35 cally (see e.g., Bosshard et al., 2017; Brown et al., 2013; Burton et al., 2010; Fronhofer and
 36 Altermatt, 2015; González-Martínez et al., 2017; Hallatschek and Nelson, 2010; Klop-
 37 stein et al., 2006; Peischl et al., 2013; Shine et al., 2011; Travis et al., 2007; Van Dyken
 38 et al., 2013; Weiss-Lehman et al., 2017). This shift in thinking about the dynamic pro-
 39 cesses forming species ranges has led to the observation that evolutionary and ecological
 40 dynamics at the front of a range expansion can differ considerably from those in the core
 41 of a species range. The set of traits that allow a species to colonize and expand its range
 42 might thus be very different from those that allow a species to successfully persist in
 43 new habitat. In this work, we study the co-evolution of two traits that are highly rel-
 44 evant in the context of species range expansion, namely an individual's fitness and its
 45 dispersal abilities.

46 A first key process in determining evolutionary processes during a range expansion
 47 is genetic drift. In a seminal paper, Edmonds et al. (2004) showed that strong genetic
 48 drift at the front of range expansions can lead to the rapid increase of random neutral
 49 variants along the expansion axis, a process now known as gene surfing (Klopfstein
 50 et al., 2006). Gene surfing also affects selected variants (Travis et al., 2007) and can
 51 lead to an accumulation of deleterious mutations in marginal populations (Hallatschek
 52 and Nelson, 2010). This accumulation of deleterious mutations has been termed
 53 expansion load and has been the subject of several theoretical (Gilbert et al., 2017; Peis-
 54 chl et al., 2013; Peischl and Excoffier, 2015; Peischl et al., 2015), experimental (Bosshard
 55 et al., 2017; Weiss-Lehman et al., 2017), and empirical studies (González-Martínez et al.,
 56 2017; Henn et al., 2016; Peischl et al., 2018; Willi et al., 2018). Expansion load stems
 57 from the repeated founder events at expanding wave fronts that reduce the efficiency of
 58 selection which would otherwise purge most incoming deleterious mutations. In this
 59 sense, the evolutionary dynamics at the front of expanding populations are similar to
 60 that of mutation accumulation experiments (Bosshard et al., 2017). Several factors con-
 61 tribute to the dynamics and severity of expansion load. Theoretical work has identified
 62 that fast-growing species with low dispersal rates are most likely to accumulate harm-

ful mutations (Peischl et al., 2013). The distribution of fitness effects and the degree of dominance of mutations also have a strong impact on the evolution of expansion load (Gilbert et al., 2018; Peischl et al., 2013; Peischl and Excoffier, 2015).

A second important process that can arise during range expansions is the evolution of dispersal-related traits (see, e.g., Bouin and Calvez, 2014; Deforet et al., 2017; Phillips and Perkins, 2017; Phillips et al., 2006; Simmons and Thomas, 2004; Travis and Dytham, 2002), which has been termed spatial sorting (Shine et al., 2011). When a population possesses heritable variation in dispersal abilities, colonists at the range front result disproportionately from individuals with greater dispersal propensity. Individuals are thus sorted over space according to their dispersal abilities, with more dispersive individuals at the range edge, similar to the increase in frequency of beneficial mutations over time due to natural selection (Phillips and Perkins, 2017; Shine et al., 2011). Spatial sorting thus increases dispersal propensity at the front as these individuals mate assortatively, potentially accelerating the speed of a range expansion (Burton et al., 2010; Cwynar and MacDonald, 1987; Hughes et al., 2003; Phillips et al., 2008; Travis et al., 2007). Spatial sorting has most notably been described in the invasive expansion of cane toads (*Rhinella marina*) across Australia (Phillips et al., 2006), but has been observed in several other systems (Fronhofer and Altermatt, 2015; Simmons and Thomas, 2004; Van Ditmarsch et al., 2013; Weiss-Lehman et al., 2017).

A few theoretical studies have focused on the co-evolution of fitness- and dispersal-related traits during range expansions. Using individual-based simulations, Burton et al. (2010) studied the evolution of resource allocation for three life-history traits during range expansions: dispersal, reproduction, and competitive ability. They found that dispersal and reproductive abilities generally increase on the expansion front, whereas competitive abilities decrease as compared to the core. Using a deterministic serial founder effect model with discrete demes, Phillips and Perkins (2017) showed that a mutation that alters both fitness and dispersal abilities will be positively selected on an expansion front if the product of migration rate and fitness is greater than that of an individual with the wild-type allele. Deforet et al. (2017) study the evolution of expansion speed using a deterministic reaction-diffusion type model in continuous space, finding that a mutation can invade the expansion front if it leads to an increase in expansion speed. The expansion speed in their model is proportional to the square root of the product of migration rate and growth rate, and hence any mutation that increases the product of migration rate and growth rate will be positively selected at the expansion

front. Despite modelling differences, the conclusions of Phillips and Perkins (2017) and Deforet et al. (2017) are strikingly similar, in the sense that the product of fitness (or a fitness-related trait such as growth rate) and dispersal rates is what determines whether a mutation is adaptive for expansion or not. The reason for their similar conclusions is that both studies focus on a deterministic model with two key aspects: the ability of reaching the front (determined by dispersal rates) and the chance of surviving on the front (determined by fitness or growth rates). It remains unclear, however, how genetic drift, mutation rates, correlations between traits, and the relationship between fitness, growth rates and expansion speed may influence evolutionary dynamics at expansion fronts.

There is striking evidence for both spatial sorting and expansion load from experimental evolution studies. Using *Escherichia coli*, Bosshard et al. (2017) has shown that fitness decreased during expansion on agar plates due to a random accumulation of new incoming mutations. Intriguingly, there are signals for an increase in expansion speed during early phases of the experiment, potentially due to loss of function in genes related to flagella production, which might allow bacteria to reach the expansion front more easily (Bosshard et al., 2018). However, in the long term, expansion speed was found to decrease over time due to reduced growth rates and competitive abilities, corroborating theoretical results (Peischl et al., 2015). Van Ditmarsch et al. (2013) performed similar experiments with *Pseudomonas aeruginosa* where they found strong signals of convergent evolution of a "hyperswarming" phenotype with increased numbers of flagella per individual. Even though growth rates in the evolved strains were lower as compared to the wild-type, the expanded populations out-competed ancestral populations, seemingly due to their increased dispersal abilities (Deforet et al., 2014). In addition to using different species, another key difference between these two experimental studies is the viscosity of the agar environment, and hence the mechanisms of dispersal in the bacteria. While Bosshard et al. (2017) used solid agar (at a concentration of 1.5% (w/v)) where bacteria are "pushed" to the front, Van Ditmarsch et al. (2013) used soft agar (at a concentration of 0.3% (w/v)) that allowed for active dispersal of bacteria via swarming. The extent to which these differences have contributed to the different outcomes of the two experiments remains unclear. These examples of disparate outcomes for evolution of dispersal and fitness emphasize the need to fully understand the theoretical underpinnings of expansion load and spatial sorting and to identify when they may complement or disrupt each other.

In this study, we derive theoretical expectations for when and how interactions between genetic drift, natural selection, and spatial sorting may unfold. Our framework allows a detailed analytic treatment and can be used to predict the co-evolutionary dynamics at expansion fronts. A key analytic result is the derivation of the fixation probability of a pleiotropic mutation affecting both fitness and dispersal-related traits.

Model and Results

We model the evolutionary dynamics of allele frequencies at the front of a one-dimensional range expansion, combining the approaches of [Peischl et al. \(2013, 2015\)](#); [Phillips and Perkins \(2017\)](#); [Slatkin and Excoffier \(2012\)](#). Consider an infinite stepping-stone model of demes, labelled $d = 1, 2, 3, \dots, n$. The carrying capacity of each deme is denoted K . Initially, only a subset of demes is colonized, and all other demes are empty. $d_f(t)$ will denote the most recently colonized deme at time t , which we call the expansion front. Individuals are haploid, and we consider a single locus with two alleles denoted a and A . These alleles can affect either fitness or dispersal rates, or both. Let p denote the frequency of the mutant allele A at the expansion front, that is, in deme d_f . Note that the dependence on t is omitted for the sake of simplicity. The fitness of wild-type and mutant alleles are denoted w_a and w_A , respectively, and the selection coefficient s of the mutant allele A is given by $s = \frac{w_A}{w_a} - 1$. During the dispersal phase, wild-type individuals migrate to neighboring demes with probability m_a and mutants with probability m_A . Analogous to the selection coefficient s , we define the effect on dispersal rate from a mutant allele as $s_m = \frac{m_A}{m_a} - 1$.

A key simplifying assumption in our model is that we model the colonization of new demes as discrete founder events occurring every T generations (see e.g., [Peischl et al., 2013, 2015](#)). When a deme is at carrying capacity, a propagule of size F is placed into the next empty deme $d_f(t) + 1$. The population then grows exponentially for T generations until the new deme's carrying capacity is reached. The size of the propagule is determined by the dispersal abilities of individuals at the expansion front. Let $\bar{m}_f = pm_A + (1 - p)m_a$ denote the average migration rate in the population. The size of the propagule is then $F = K\bar{m}_f/2$. The factor $1/2$ is due to the fact that individuals migrate to each of the two neighboring demes with the same probability. During the growth phase, migration is ignored. Assuming exponential growth at rate $r = \log(R)$, this yields $T = \log(2/\bar{m}_f)/r$ ([Peischl et al., 2013](#)). This model is a good approximation to range expansions with continuous gene flow when growth rates are larger than migration

164 rates (Peischl et al., 2013). We also consider the limiting case where r is so large that
 165 a deme grows to carrying capacity within a single generation $T = 1$, independently
 166 of the number of founders F . Figure 1 shows a sketch of the model that illustrates
 167 how mutations can be positively selected on expanding wave fronts based on either an
 168 increase in migration rates (Figure 1A) or an increase in relative fitness (Figure 1B).

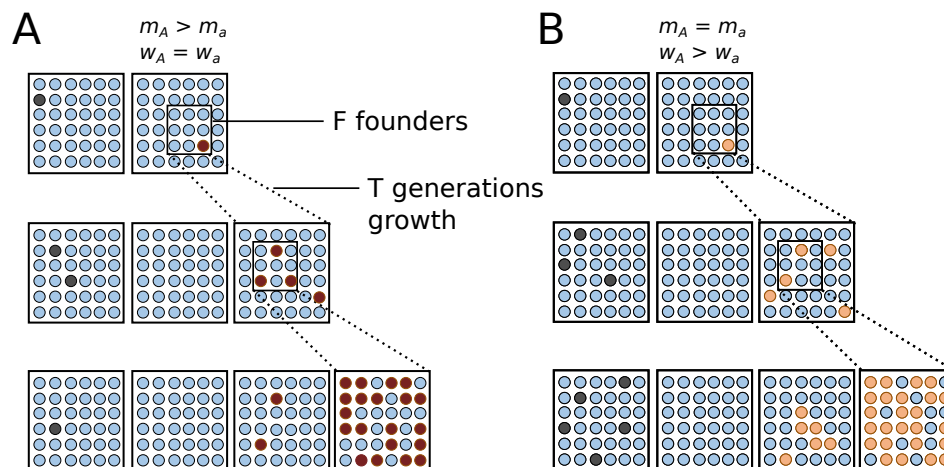


Figure 1: Sketch of the model. A: a mutation with higher migration rate (red) but same fitness as wild-type individuals (blue). The mutation can increase in frequency at the expansion front because it is more likely to be among the F founders as compared to wild-type mutations. B: a mutation with higher fitness than the wild type, but with same migration rate. The mutation (orange) has the same probability to be among the F founders as the wild type (blue), but it can spread at the expansion front due to higher reproductive success during the T generations of growth during which natural selection acts. In both panels, dark gray circles show the evolution of an equivalent mutation in the core of the species range for comparison.

169 Fixation of new mutations

170 We show in Appendix A that the probability of fixation of a mutation with initial fre-
 171 quency p_0 at the expansion front is given by

$$\frac{e^{-4N_e s_e p_0} - 1}{e^{-4N_e s_e} - 1}, \quad (1)$$

172 where we define an effective selection coefficient $s_e = sT + s_m$ and an effective popu-
 173 lation size $N_e = F$. Equation (1) shows that mutations can be under positive selection
 174 at the expansion front for two reasons: (i) increasing an individual's fitness ($s > 0$) or
 175 (ii) increasing the migration rate ($s_m > 0$). If $s_m = 0$, we recover the fixation probability
 176 on the expansion front derived in Peischl et al. (2013). Then, natural selection is most
 177 efficient when R is small (Figure 2B) and \bar{m}_f is large (Peischl et al., 2013, Figure S4B).
 178 If $s = 0$, the fixation probability of a mutation modifying the dispersal probability by

a factor of $1 + s_m$ is equivalent to that of a mutation with selective advantage s_m in a stationary population of size F (Kimura, 1962). This shows that spatial sorting can indeed be viewed as an analog to natural selection across space as proposed by simulation studies (Shine et al., 2011) and deterministic models (Phillips and Perkins, 2017). Note that our model can be seen as a stochastic version of the model presented in Phillips and Perkins (2017) if $T = 1$, i.e., if a new deme is colonized each generation.

In the following we denote mutations with $s_e > 0$ as adaptive for expansion, since they can spread at the front because of the joint actions of natural selection and spatial sorting. We refer to mutations with $s_e < 0$ as maladaptive for expansion since they can only establish via genetic drift. Equation (1) shows that natural selection is most efficient if \bar{m}_f is large and R is small (see also Peischl et al. 2013). Likewise, spatial sorting is most efficient if \bar{m}_f is large because drift during founder events decreases with increasing migration rates (Figure S2). The growth rate R has no impact on the fixation probability if $s = 0$ (Figure S1), since it only affects the length of the growth phase during which natural selection acts but not the number of founders, F , or the probabilities of individuals to migrate to a new deme.

Pleiotropic mutations

We next consider mutations that affect both the fitness as well as the dispersal ability of a carrier. As expected, mutations that increase both fitness and migration rates ($s, s_m > 0$) are always positively selected (solid lines in Figure 2C) and mutations with $s, s_m < 0$ are always negatively selected at the expansion front (dashed lines in Figure 2A). In both cases, the efficacy of selection for expansion decreases with increasing growth rate R (Figure S2) because the time T during which natural selection can act becomes shorter.

If there is a trade-off between fitness- and dispersal-related traits such that $s_m < 0 < s$ or $s < 0 < s_m$, the growth rate of the population, R , affects the strength as well as the direction of selection for a given mutation (Figure 2). In general, if growth rates are low, natural selection is more effective than spatial sorting because of the longer periods, T , between consecutive founder events during which selection can act (Figure 2), whereas spatial sorting is only acting during the sampling of new founders (Figure 1). Thus, for low R , fixation probabilities are close to that of mutations with effect s in stationary populations of size F . On the other hand, if R is large such that T is close to 1, both spatial sorting and natural selection contribute equally to the fixation probability (Figure 2), which is then similar to a mutation with effect $s + s_m$ in a stationary population of size

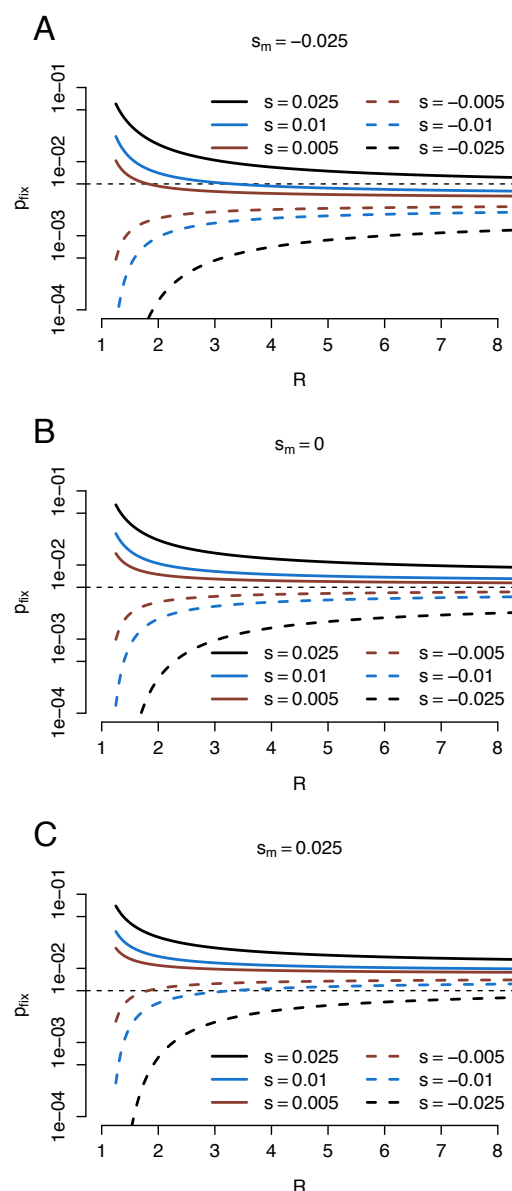


Figure 2: Fixation probability of pleiotropic mutations as a function of population growth rate. Dashed lines indicate deleterious mutations (negative selection coefficient, s) while solid lines indicate beneficial mutations.

212 F . We find that a mutation with $s > 0 > s_m$ has a higher fixation probability as compared
 213 to a neutral mutation ($s = s_m = 0$) if $R < m^{s/s_m}$ (Figure 2A), and a mutation with
 214 $s < 0 < s_m$ has a higher fixation probability if $R > m^{s/s_m}$. Taken together, this means
 215 that spatial sorting and expansion load should more readily impact populations with
 216 high growth rates, especially if increasing dispersal rates is costly in terms of fitness.

217 Figure 3 illustrates the effect of the average migration rate at the expansion front
 218 on the fixation probability of mutations. For very small values of \bar{m}_f , the number of
 219 founders is close to $F = 1$ and selection for expansion is therefore virtually absent (as in
 220 mutation accumulation experiments). The fixation probability is then close to that of a

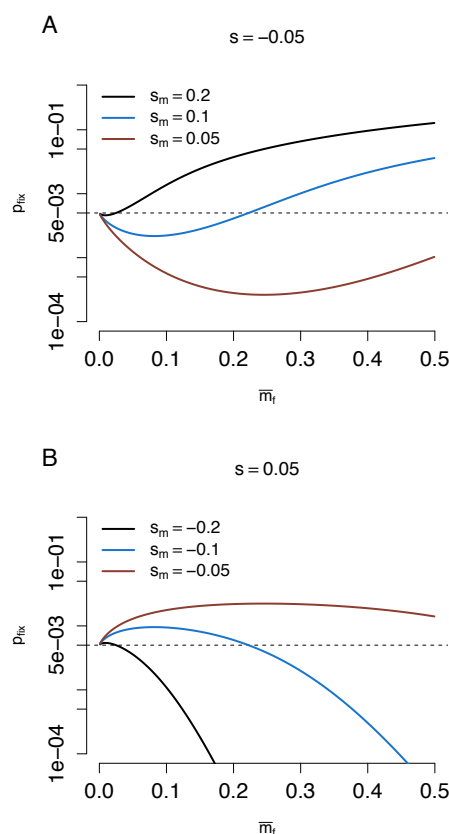


Figure 3: Fixation probability of pleiotropic mutations as a function of mean migration rate at the front. s is the selection coefficient for fitness-impacting mutations while s_m is the selection coefficient for dispersal-impacting mutations.

221 neutral mutation ($\bar{m}_f = 0$ in Figure 3). As \bar{m}_f increases (to ≈ 0.1 in Figure 3) the fixation
 222 probability of a pleiotropic mutation is driven more by the action of natural selection
 223 rather than the action of spatial sorting. This is when T is sufficiently large to allow
 224 the contribution of natural selection to outweigh that of spatial sorting in the effective
 225 selection coefficient s_e . For even larger values of \bar{m}_f , the time between founder events
 226 will decrease as propagule size increases and eventually approach $T = 1$ such that s and
 227 s_m will contribute equally to the fixation probability. Thus, if $s + s_m > 0$ and $s < 0 < s_m$
 228 (Figure 3A) or if $s + s_m < 0$ and $s_m < 0 < s$ (Figure 3B), the direction of selection for
 229 pleiotropic mutations may change with increasing \bar{m}_f . A mutation with $s < 0 < s_m$
 230 has a higher fixation probability as compared to a neutral mutation ($s = s_m = 0$) if
 231 $\bar{m}_f > R^{s_m/s}$. Conversely, pleiotropic mutations with $s > 0$ and $s_m < 0$ will have a higher
 232 fixation probability as compared to neutral mutations if $\bar{m}_f < R^{s_m/s}$.

233 Co-evolutionary dynamics

234 We next study the co-evolution of mean fitness and migration rates in expanding popu-
 235 lations taking into account the interactions of mutation rates, the distribution of fitness
 236 effects (DFE) of new mutations, and genetic correlations in fitness and migration-related
 237 traits. In the following we assume that selection is soft such that population mean fit-
 238 ness does not affect growth rates or carrying capacities. Consequently the parameters
 239 T and F are independent of the evolution of mean fitness (cf. [Peischl et al., 2015](#)), and
 240 following equation (1), the evolution of mean fitness does not impact the evolution of
 241 migration modifiers. However, the amount of migration into new empty demes affects
 242 both the parameters F and T ([Peischl et al., 2015](#)), which in turn determine the efficacy
 243 of selection and the strength of drift at the expansion front. We approximate the evolu-
 244 tion of mean fitness and migration rate, analogous to the model in [Peischl et al. \(2015\)](#),
 245 and consider mutations that can affect both migration rates and fitness simultaneously.
 246 Let $u(s, s_m)$ denote the mutation rate of mutations with effect s on fitness and s_m on the
 247 migration rate. We assume that s and s_m are drawn from a bi-variate distribution with
 248 mean \bar{s} and \bar{s}_m , variance V_s and V_m , and correlation ρ . Appendix B shows that we can
 249 approximate the dynamics of mean fitness and migration rate at the front by

$$\frac{d}{dt}\bar{w}_f(t) = w(t)u \left[(F(t) - 1) \left(\rho\sqrt{V_m V_s} + T(t)V_s \right) + \bar{s} \right] \quad (2)$$

250 and

$$\frac{d}{dt}\bar{m}_f(t) = \bar{m}_f(t)u \left[(F(t) - 1) \left(\rho\sqrt{V_m V_s}T(t) + V_m \right) + \bar{s}_m \right], \quad (3)$$

251 where $F(t) = K\bar{m}_f(t)/2$, $T(t) = \log(2/\bar{m}_f(t))/r$ and u is rate at which new mutations
 252 occur per individual and generation.

253 In general, the mean mutational effect of mutations affecting fitness will be negative
 254 ($\bar{s} < 0$) as most new incoming mutations are deleterious ([Eyre-Walker and Keightley,](#)
 255 [2007](#)). Thus, expansion load will generally occur unless one of the following is true:
 256 the variance of the distribution of fitness V_s is sufficiently large (thus increasing the
 257 proportion of beneficial variants in the DFE, see also [Peischl et al. 2013](#)), the covariance
 258 of a mutation's fitness effects with effects on migration related traits is positive and
 259 sufficiently large, or the carrying capacity of demes at the expansion front is sufficiently
 260 large. A negative correlation between fitness-related and migration-related traits can
 261 increase the chance for expansion load to occur.

Spatial sorting is expected to occur if there are sufficiently many new mutations that increase migration rates (\bar{s}_m or V_m large), if mutations that increase migration rates also increase fitness ($\rho > 0$), or if population size is sufficiently large. If fitness- and migration-related traits are independent, that is $\rho = 0$, then $\bar{s}_m \geq 0$ implies that migration rates will always increase at the expansion front. If there is, however, a trade-off between migration- and fitness-related traits such that $\rho < 0$, the average migration rate at the front can decrease despite $\bar{s}_m \geq 0$. Thus, if an increase in migration is costly in terms of fitness, expansion load can constrain spatial sorting. A positive correlation between effects on migration and fitness ($\rho > 0$) will generally increase the chance for spatial sorting to occur, as well as reduce the chance for expansion load to accumulate.

Evolution of dispersal can rescue expanding populations

While a detailed analytic analysis of eqs. (2) and (3) is mathematically challenging and beyond the scope of this paper, we can gain some intuition from the case when growth rate is strong such that newly colonized demes reach carrying capacity within a single generation ($T = 1$). This is usually the case when $r \gg 1$. Here, we find that \bar{w}_f increases over time if

$$\bar{m}_f > m_{crit,w} = 2/K(1 - \bar{s}/(V_s + \rho\sqrt{V_s V_m})) \quad (4)$$

and m_f increases over time if

$$\bar{m}_f > m_{crit,m} = 2/K(1 - \bar{s}_m/(V_m + \rho\sqrt{V_s V_m})). \quad (5)$$

Figure 4 illustrates the dynamics of mean fitness and migration rate at the expansion front as given by eqs. (2) and (3) (arrows in Figure 4) and compares these dynamics with the outcome of individual-based simulations in a serial founder effect model (as depicted in Figure 1). Even though eqs. (2) and (3) can be solved analytically, we proceed by describing the dynamics using a geometric approach that allows us to exhaustively identify all qualitatively different evolutionary regimes. If $\bar{m}_f(0) < m_{crit,w}, m_{crit,m}$, an expanding population will not evolve increased dispersal and will also suffer from expansion load (green lines in Figure 4). On the other hand, if $\bar{m}_f(0) > m_{crit,w}, m_{crit,m}$, both mean fitness and the average migration rate at the expansion front will increase

(red lines in Figure 4). More interestingly, if $m_{crit,m} < \bar{m}_f(0) < m_{crit,w}$, expansion load will accumulate and migration rates will also increase over time. Thus, we eventually observe a "rescue effect" when $\bar{m}_f(t)$ surpasses $m_{crit,w}$, in the sense that founder events become less drastic and selection at the expansion front becomes sufficiently efficient so that mean fitness will start to increase over time (see blue lines in Figure 4).

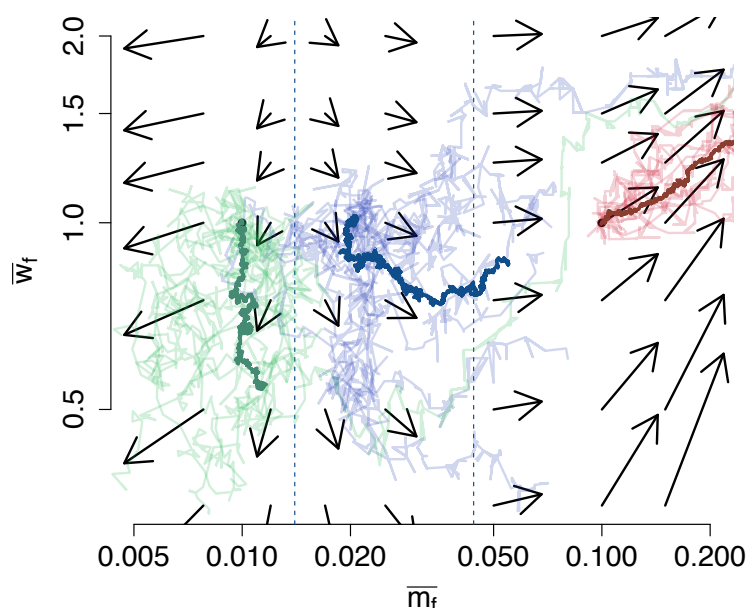


Figure 4: Individual-based simulations of the co-evolution of mean fitness and migration rate in a population undergoing serial founder events, as depicted in Figure 1. The arrows show the vector field generated by the differential equations (2) and (3), and indicate the direction of evolution as predicted by the analytic theory. The thin lines show the outcome of single simulation runs. The thick lines show the average across 10 simulation runs for each initial condition. The different colors correspond to different initial migration rates. Mutations occur at rate $u = 0.01$ per individual per generation and their effects are drawn from a bi-variate Gaussian distribution with parameters $\bar{s}_m = -0.001$, $\bar{s} = -0.005$, $V_m = 0.004$, $V_s = 0.002$ and $\rho = 0$. The remaining parameters are $K = 500$, $\bar{w}_f(0) = 1$, and $\bar{m}_f = 0.01$ (green), 0.02 (blue) and 0.1 (red).

Individual-based range expansion simulations

Using an individual-based model first developed in Peischl et al. (2013), we simulate populations undergoing range expansions with both the evolution of dispersal and fitness. The key difference to the serial founder effect model in Figure 4 is that we simulate the whole species range instead of just the deme at the leading edge, and that gene flow occurs every generation rather than just during colonization events. In particular, we model a linear, 1-dimensional discrete landscape of 1000 demes with a stepping-stone

migration model. Each deme has a carrying capacity of $K = 1000$ and an initial migration rate $m = 0.01$. The 5 left-most demes are initiated at carrying capacity and burned in for 6000 generations, after which free expansion into subsequent empty demes is allowed. Individuals each have the potential to accumulate deleterious load through 1300 bi-allelic, unconditionally deleterious loci, or increase fitness from 700 bi-allelic, unconditionally beneficial loci. Fitness is multiplicative across these 2000 loci with genome-wide mutation rate $U = 0.2$ and an equivalent potential for back-mutations to the wild type. Fitness effects are fixed at ± 0.01 and are additive (heterozygote fitness is perfectly intermediate to homozygotes). Generations are non-overlapping and growth is instantaneous in newly-colonized demes. The dispersal trait is modelled as a quantitative trait such that each individual inherits its migration rate from as the average of both parents' trait value plus a random mutational deviation drawn from a Normal distribution with mean 0 and standard deviation 0.005 or 0.01 for either a low or high rate of dispersal evolution, respectively. Migration rate is constrained between $0 < m < 0.5$.

For these parameter values, equations (4) and (5) indicate that we expect an increase in dispersal rates independently of initial conditions, and that expansion load occurs if $\bar{m}_f \lesssim 0.07$ and ceases to occur if $\bar{m}_f \gtrsim 0.07$. In agreement with these predictions, we find that mutation load does accumulate during expansion as a result of gene surfing of deleterious mutations, but also that as dispersal evolves, spatial sorting leads to the rescue of fitness at the range front (Figure 5). The rescue effect is particularly strong under a higher rate of dispersal evolution (Figure 5C), where migration rate evolves to be close to 0.5. Under both low and high rates of dispersal evolution, fitness loss is reduced and then reversed, an effect opposite to that expected for fast range expansions in the absence of dispersal evolution (Gilbert et al., 2017; Hallatschek and Nelson, 2010; Peischl et al., 2013, 2015). Conversely, in the absence of the evolution of dispersal, fitness is continually lost throughout the course of expansion.

Discussion

The question of what makes an organism successful at colonizing new habitats is highly relevant in evolutionary biology, (Sexton et al., 2009), conservation biology (e.g., for predicting invasiveness of species, Pejchar and Mooney 2009), and evolutionary medicine (e.g., in the context of cancer growth, Waclaw et al. 2015). In this work, we present an analytically tractable theoretical framework for the co-evolution of fitness- and dispersal-related traits that builds upon classical models in population genetics. We show that a

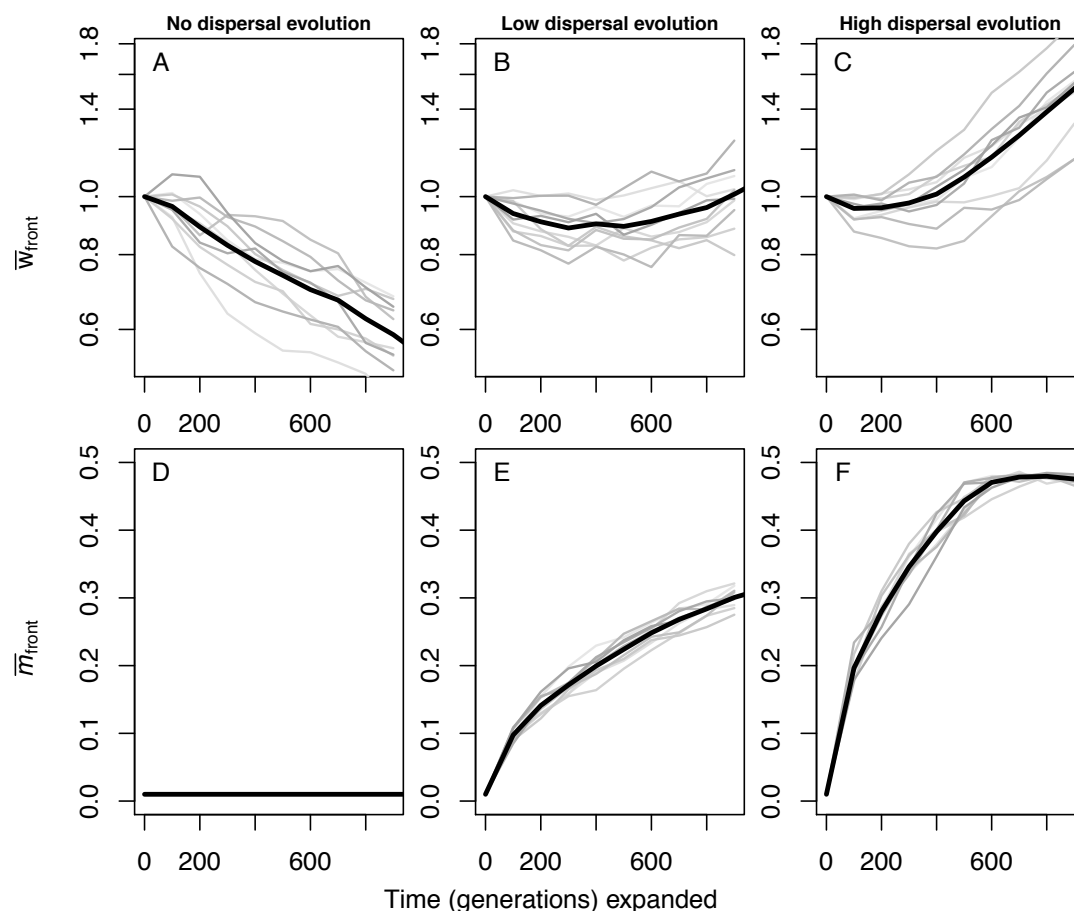


Figure 5: Fitness (A-C) or migration rate (D-F) measured at the front edge of a range expansion either in the absence of the evolution of migration or a low (standard deviation of mutational effect 0.005) or high (standard deviation of mutational effect 0.01) rate of evolution of migration. Individual replicate simulations are shown in gray while the mean is shown by the thick black line. Starting fitness is scaled to 1 for comparison, and all other parameters are as described in text.

333 given mutation with pleiotropic effects can be positively or negatively selected at the
 334 expansion front, depending on the current growth rate and migration rate at the expan-
 335 sion front (see Figures 2 and 3). Furthermore, we show that while migration rates and
 336 growth rates both affect the expansion speed in similar ways, their effect on the strength
 337 and direction of selection at the expansion front can be quite different in finite popula-
 338 tions (see Figures 2 and 3). Finally, we used our results to predict the co-evolutionary
 339 dynamics of fitness and dispersal during range expansions (see equations (2) and (3),
 340 and Figure 4). For the special case of high growth rates and soft selection, our model
 341 allowed us to exhaustively characterize the co-evolutionary dynamics, and to identify
 342 conditions when the evolution of dispersal can rescue a population from expansion load
 343 (see equations (4) and (5), and Figure 4). Individual-based simulations of range expan-
 344 sions confirmed our analytic results (see Figure 5).

Our study generalizes the results of [Phillips and Perkins \(2017\)](#), who studied the co-evolution of dispersal and fitness during range expansions with a constant speed of 1 deme per generation ($T = 1$) using a deterministic model similar to ours. As expected, in the case of infinite population size and $T = 1$, our results are in perfect agreement with those of [Phillips and Perkins \(2017\)](#), meaning that mutations are adaptive for expansion if and only if $s_e = s + s_m > 0$, which, for small s and s_m is equivalent to the conditions derived in [Phillips and Perkins \(2017\)](#). We note that this condition for invasion of new mutations is also equivalent to the condition found by [Deforet et al. \(2017\)](#) in a deterministic continuous space model if one treats growth rate in their model as equivalent to fitness. It would be interesting to see whether and how results from deterministic continuous space models further generalize to finite populations, and to better understand the role of genetic correlations on spatial sorting and expansion load. A direct comparison of our results with those of [Deforet et al. \(2017\)](#) is difficult because their assumptions differ regarding the interplay of fitness, growth rates, and gene flow in the modelling approaches.

The simplicity of our model comes at a cost as we made several simplifying assumptions. Perhaps most importantly, we employ a separation of time scales argument that allows us to model evolution of the leading edge population independently from the core. We have previously shown that this is a good approximation to models with continuous gene flow between demes as long as the growth rate of the population is sufficiently large ([Peischl et al., 2013](#)). We thus expect our results to be valid even if dispersal rates are large, as long as growth rates are even larger (see Figure 5 for results from individual-based simulations). If growth rates are on the order of dispersal rates or lower, we expect our model to underestimate the strength of drift because gene flow will lead to a more gradual decrease in population size towards the expansion front (c.f. [Hallatschek and Nelson, 2010](#)). Therefore, the rescue effect we identified with our model might be less relevant for species with growth rates and migration rates of similar magnitudes. In this case, a more suitable modelling approach would be assuming continuous space, e.g., using reaction-diffusion equations as in [Deforet et al. \(2017\)](#). Including the effects of genetic drift, however, is somewhat harder in continuous-space models (but see e.g. [Barton et al. 2013](#); [Brunet and Derrida 2001](#); [Hallatschek 2011](#)).

We focused on expansion along a one-dimensional habitat. This should be a good approximation for range expansion along a narrow two-dimensional corridor ([Peischl et al., 2013](#)). In wider habitats, the evolutionary dynamics at the expansion front might

be quite different from what is expected in the one-dimensional case (see e.g., Polechová and Barton, 2015). In particular, lateral gene flow perpendicular to the axis of expansion can restore genetic diversity and hence prevent some of the negative consequences of increased drift at the expansion front (Peischl et al., 2013). Previous studies have shown that a wider expansion front can reduce the rate at which expansion load is built up and lead to faster recovery after the expansion (Gilbert et al., 2018). With spatial sorting it remains unclear how a two-dimensional expansion front would affect the outcome. Gene flow might have very different effects on expansion speed and genetic diversity, depending on its direction relative to the expansion axis. One might thus expect that not only the rate or distance of dispersal evolves, but also the direction of dispersal (see e.g., Lindström et al., 2013).

For the sake of simplicity we assumed haploid individuals, but our model can be readily extended to sexually reproducing, diploid organisms (see e.g., Phillips and Perkins 2017). Since the evolutionary dynamics of diploid and haploid individuals are equivalent in the case of co-dominant (multiplicative) mutations (Bürger, 2000), our model can be applied directly to diploid individuals. Additionally, while it would be straightforward to include dominance in our model (Gilbert et al., 2018; Peischl and Excoffier, 2015), adding epistatic interactions across loci would be much more difficult. Furthermore, we ignored the effects of clonal interference in the derivation of equations (1), which could lead to an overestimation of the fixation probability of beneficial mutations. Our results should be good approximations if recombination is strong or if mutations occur infrequently so that multiple segregating mutations rarely interact (that is, if $uK < 1$). However, because mutations are either fixed or lost very quickly at expanding fronts, we expect our results to hold even if mutation rates are fairly high such that $uK > 1$ (see Figure 4).

We assumed that selection is soft, i.e. local carrying capacities and growth rates do not depend on population mean fitness. The co-evolutionary dynamics under hard selection might be somewhat different from those in our model (see e.g., Peischl et al., 2015) since growth rates and carrying capacities affect expansion speed (Skellam, 1951), the amount of genetic drift, and the efficacy of spatial sorting and natural selection at the expansion front (Hallatschek and Nelson, 2010). In particular, while increasing growth rates render natural selection at expansion fronts less efficient due to the reduced time between subsequent colonization events (see Figure 2), the efficacy of natural selection increases with increasing dispersal (see Figure 3). A model with hard selection would

thus lead to additional feedback between evolutionary processes at the expansion front and the efficacy of natural selection and spatial sorting.

We have presented a theoretical study that shows how the evolution of dispersal can serve as a factor to reduce or even eliminate expansion load. To further test our model in experimental or empirical settings, comparing fitness evolution during geographic spread in tandem with dispersal evolution will prove especially illuminating. This is most approachable in experimental evolution studies where understanding these trajectories simultaneously will inform how often dispersal is positively or negatively correlated with changes in fitness. This may also provide insights into understanding the distribution of fitness effects for new mutations and for dispersal-impacting mutations by fitting a model to data from an experimental study (e.g., as in [Bosshard et al. 2017](#)). Understanding both this correlation between fitness and dispersal as well as the distribution of effects for mutations impacting both of these characteristics is a major step forward in evolutionary biology, and could help us explain the different results already found in several experimental studies ([Bosshard et al., 2017](#); [Van Ditmarsch et al., 2013](#)). Additionally, in natural systems this interaction between dispersal and load accumulation may explain important dynamics during colonization events. Given that invasive species present as ideal candidates for accumulation of expansion load in terms of rapid expansion and small founding population sizes, yet seem to exhibit no detrimental fitness effects, this mechanism of rescue and recovery due to increased dispersal could prove as an explanatory factor in their successful invasions as well as the successful spread of other natural range expansions ([Arim et al., 2006](#); [Hanski et al., 2002](#); [Hughes et al., 2007](#); [Lombaert et al., 2014](#); [Monty and Mahy, 2010](#); [Phillips et al., 2006](#); [Simmons and Thomas, 2004](#); [Szűcs et al., 2017](#); [Tayeh et al., 2013](#); [Thomas et al., 2001](#)). The improved understanding of the evolutionary forces and interactions between changes in fitness and dispersal ability will enhance our knowledge of what makes some species particularly successful at colonization, as well as what factors might contribute to formation of species range limits.

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Online Appendix

A Supplementary Figures

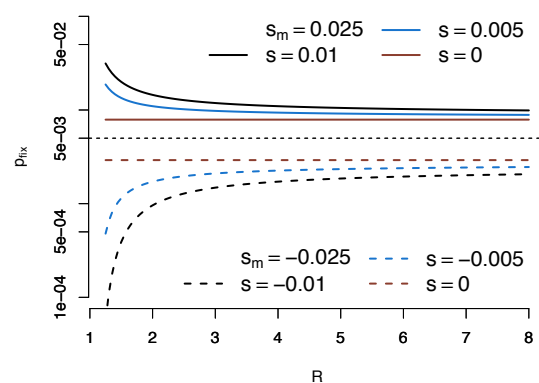


Figure S1: Fixation probability of pleiotropic mutations as a function of the population growth rate.

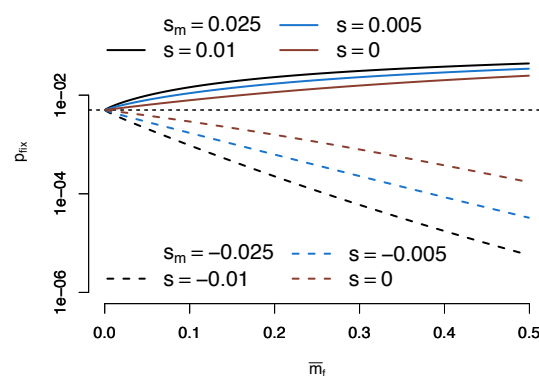


Figure S2: Fixation probability of pleiotropic mutations as a function of the mean migration rate at the expansion front.

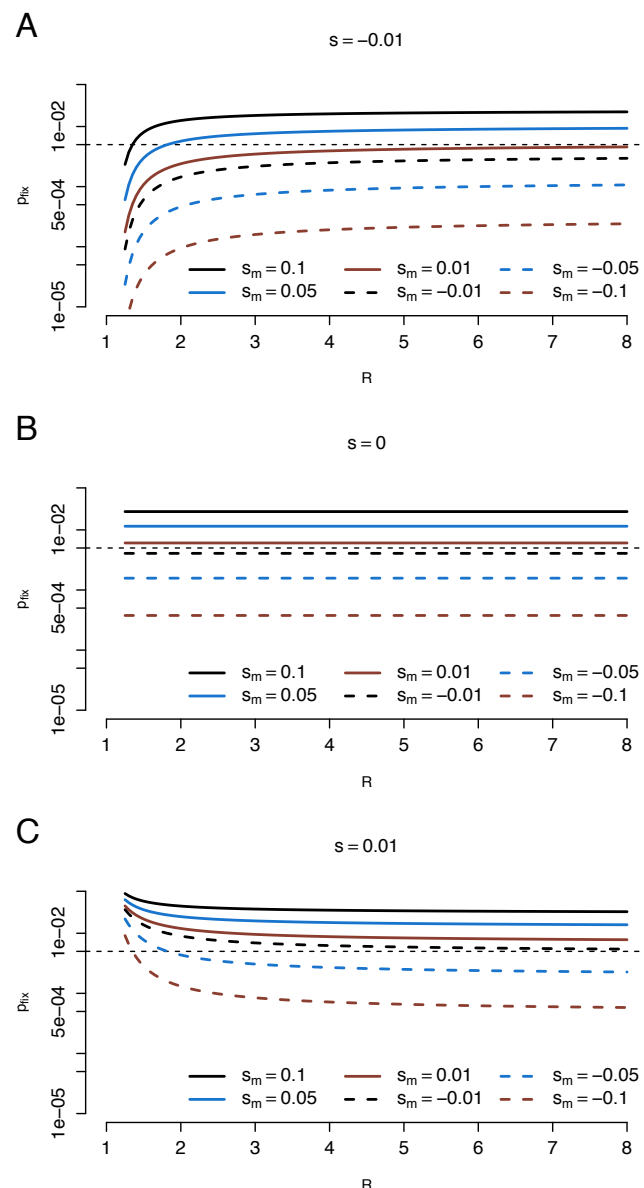


Figure S3: Fixation probability of pleiotropic mutations as a function of the population growth rate under varying scenarios of the mutational effect on fitness, s : deleterious (A), neutral (B), and beneficial (C).

447 B Derivation of fixation probability

We use a diffusion approach to calculate the fixation probability of a mutation that affects fitness and/or migration rates. One "generation" in the diffusion approximation corresponds to the colonization of a single deme and starts just before a new propagule disperses. We consider a mutant that is present at frequency p when the population in deme d_f is at carrying capacity K . The expected frequency of the mutant in the

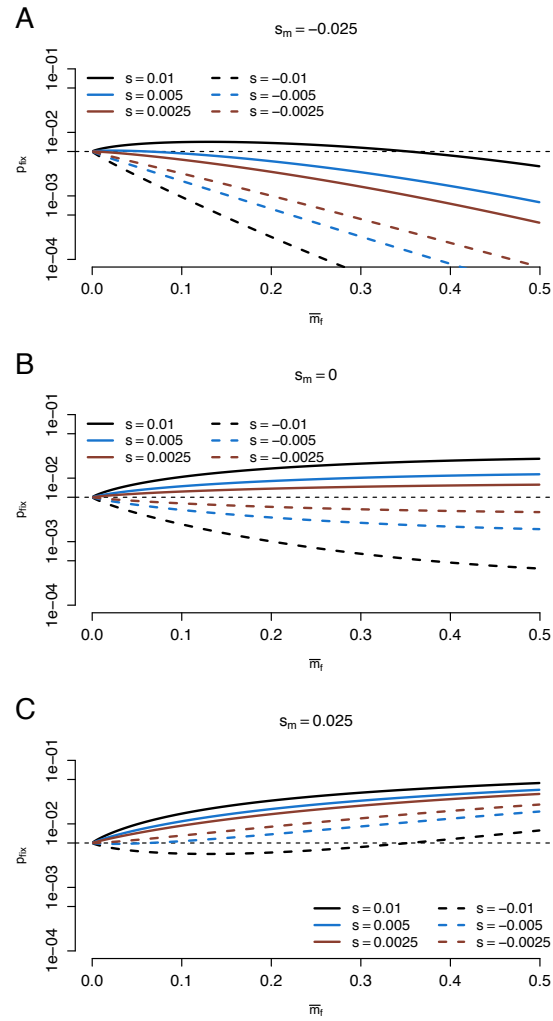


Figure S4: Fixation probability of pleiotropic mutations as a function of the mean migration rate at the expansion front.

propagule is then

$$p' = \frac{pKm_A/2}{pKm_A/2 + (1-p)Km_a/2} = p \frac{m_A}{\bar{m}_f} \left(\approx p \frac{m_A}{m_a} \text{ if } p \ll 1 \right),$$

and the variance due to binomial sampling is

$$V = \frac{1}{F} p' (1 - p').$$

If the mutant's frequency in the propagule is p' , its expected frequency after the growth phase is

$$p'' = \frac{p'w_A^T}{p'w_A^T + (1-p')w_a^T} = \frac{p'(1+s)^T}{p'(1+s)^T + (1-p')}$$

which is equal to

$$p'' = \frac{p \frac{m_A}{m_f} w_A^T}{p \frac{m_A}{m_f} w_A^T + (1-p) \frac{m_a}{m_f} w_a^T} = \frac{p m_A w_A^T}{p m_A w_A^T + (1-p) m_a w_a^T}.$$

The expected change in allele frequency is then

$$E[\Delta p] = p'' - p \approx p(1-p) \left(sT + \frac{m_A}{m_a} - 1 \right).$$

We assume that the stochastic sampling effects during colonization of a deme are the main contribution of genetic drift and therefore approximate the variance in allele frequency change by

$$V[\Delta p] = \frac{1}{F} p'(1-p').$$

448 Next, we calculate the fixation probability using standard diffusion methods. The prob-
449 ability of fixation (conditioned on initial frequency p_0) is given by:

$$\frac{\int_0^{p_0} g(x) dx}{\int_0^1 g(x) dx},$$

450 where

$$g(x) = \exp \left[- \int \frac{2E[\Delta p]}{V[\Delta p]} dp \right]$$

After some fundamental algebra we end up with

$$\frac{e^{-4F(sT + \frac{m_A}{m_a} - 1)p_0} - 1}{e^{-4F(sT + \frac{m_A}{m_a} - 1)} - 1},$$

451 for the fixation probability, where F is the propagule size and T is the time until a deme
452 is filled.

Note that this is Kimura's (Kimura, 1962) original equation for the fixation probability

$$\frac{e^{-4N_e s_e p_0} - 1}{e^{-4N_e s_e} - 1},$$

with effective selection coefficient

$$s_e = sT + \frac{m_A}{m_a} - 1$$

453 and effective population size $N_e = F$.

454 C Co-evolutionary dynamics

If mutations affect only fitness but not dispersal probabilities, [Peischl et al. \(2015\)](#) showed that the change in mean relative fitness at the expansion front can be approximated using the following equation

$$\bar{w}_f(t+1) = \bar{w}_f(t) \left(1 + \int_{-1}^{\infty} s u_s(s) K p(s, 0) ds \right)$$

455 where $u_s(s)$ is the mutation rate of mutations with effect s . Here, we also assumed that
 456 mutations evolve independently of each other, that is, we ignore clonal interference. The
 457 parameter F measures genetic drift on the expansion front and T measures the length
 458 of the growth period during which selection occurs. Both F and T can depend on mean
 459 fitness, migration rates, and growth rates in models of hard selection (see [Peischl et al.](#)
 460 [2015](#) or [Bosshard et al. 2017](#) for details). The integral in the equation calculates the
 461 expected long-term effect of each new incoming mutation, that is, whether they will be
 462 fixed or lost from the population at the expansion front, taking into account the effects of
 463 mutation rates, random genetic drift, natural selection, and spatial sorting. This model
 464 has been shown to be a good approximation for the evolution of mean fitness at the
 465 front of range expansions if the growth rate of populations at the front, r , is larger than
 466 the rate of gene flow, m ([Peischl et al., 2013](#)).

We next consider mutations that affect migration traits but have no effect on fitness. Let $\bar{m}_f(t)$ denote the mean migration rate of the population at the edge of the expansion at time t . The evolution of \bar{m}_f can be modelled analogously via

$$\bar{m}_f(t+1) = \bar{m}_f(t) \left(1 + \int_{-1}^{\infty} s_m u_m(s_m) K p(0, s_m) ds_m \right)$$

467 where $u_m(s_m)$ is the mutation rate of migration modifier mutations that increase disper-
 468 sal probabilities by a factor $1 + s_m$.

For pleiotropic mutations that affect both fitness and dispersal, the evolutionary dynamics of both traits follows

$$\begin{aligned} \bar{w}_f(t+1) &= \bar{w}_f(t) \left(1 + \int_{-1}^{\infty} \int_{-1}^{\infty} s u(s, s_m) K p(s, s_m) ds ds_m \right) \\ \bar{m}_f(t+1) &= \bar{m}_f(t) \left(1 + \int_{-1}^{\infty} \int_{-1}^{\infty} s_m u(s, s_m) K p(s, s_m) ds ds_m \right), \end{aligned}$$

469 where $u(s, s_m)$ denotes the mutation rate of mutations with effect s on fitness and s_m on

the dispersal probability. We assume that s and s_m are given by a bi-variate distribution with mean values \bar{s} and \bar{s}_m , variances V_s and V_m , and correlation ρ (e.g., a bi-variate Normal distribution). Switching to continuous time, the dynamics can be approximated by

$$\frac{d}{dt}\bar{w}_f(t) = \bar{w}_f(t) \int_{-1}^{\infty} \int_{-1}^{\infty} su(s, s_m)Kp(s, s_m)dsds_m \quad (\text{A1})$$

$$\frac{d}{dt}\bar{m}_f(t) = \bar{m}_f(t) \int_{-1}^{\infty} \int_{-1}^{\infty} s_mu(s, s_m)Kp(s, s_m)dsds_m. \quad (\text{A2})$$

We define

$$g(s, s_m) := sKp(s, s_m)$$

and approximate it by

$$\tilde{g}(s, s_m) = \sum_{n=0}^2 \frac{1}{n!} \left((s - \bar{s}) \frac{\partial}{\partial s} + (s_m - \bar{s}_m) \frac{\partial}{\partial s_m} \right)^n g(s, s_m) |_{(s, s_m) = (\bar{s}, \bar{s}_m)}.$$

This allows us to approximate the integrals in equations (A1) by

$$\int_{-1}^{\infty} \int_{-1}^{\infty} u(s, s_m) \tilde{g}(s, s_m) ds ds_m.$$

Equation (A2) can be treated analogously. These integrals can then be solved straightforwardly but the solutions are lengthy and uninformative and hence not shown here. Importantly, however, because $\tilde{g}(s, s_m)$ is quadratic in $(s - \bar{s})$ and $(s_m - \bar{s}_m)$, the contribution of the distribution of mutational effects $u(s, s_m)$ can be expressed solely in terms of the mean (\bar{s}, \bar{s}_m) , variances V_s and V_m , and the correlation ρ . To gain a better intuition we proceed by further approximation. We re-scale the quantities s_m , \bar{s} , \bar{s}_m , V_s , V_m and ρ by s , and expand in a Taylor series around $s = \bar{s}$. Ignoring third- and higher-order terms in s (and switching back to the original variables) we can then approximate the dynamics of mean fitness and migration rate at the front by

$$\frac{d}{dt}\bar{w}_f(t) = w(t)u \left[(F(t) - 1) \left(\rho\sqrt{V_m V_s} + T(t)V_s \right) + \bar{s} \right]$$

and

$$\frac{d}{dt}\bar{m}_f(t) = \bar{m}_f(t)u \left[(F(t) - 1) \left(\rho\sqrt{V_m V_s}T(t) + V_m \right) + \bar{s}_m \right],$$

where $F(t) = K\bar{m}_f(t)/2$, $T(t) = \log(2/\bar{m}_f(t))/r$ and $u = \int_{-1}^{\infty} \int_{-1}^{\infty} u(s, s_m)dsds_m$.

If mutations affect either migration rates or fitness, but not both, we obtain

$$\frac{d}{dt}\bar{w}_f(t) = w(t)u_s [(F(t) - 1) T(t)V_s + \bar{s}]$$

and

$$\frac{d}{dt}\bar{m}_f(t) = \bar{m}_f(t)u_m [(F(t) - 1) V_m + \bar{s}_m],$$

475 where u_m and u_s are the mutation rates of migration- and fitness-related mutations,
476 respectively.

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