

# Ecological genetic conflict: Genetic architecture can shift the balance between local adaptation and plasticity

Olof Leimar<sup>1,\*</sup>, Sasha R. X. Dall<sup>2</sup>, John M. McNamara<sup>3</sup>, Bram Kuijper<sup>4</sup>, and Peter Hammerstein<sup>5</sup>

<sup>1</sup>Department of Zoology, Stockholm University, SE-106 91 Stockholm, Sweden <sup>2</sup>Centre for Ecology and Conservation, University of Exeter, Penryn TR10 9EZ, UK <sup>3</sup>School of Mathematics, University of Bristol, University Walk, Bristol BS8 1TW, UK <sup>4</sup> Environment and Sustainability Institute, University of Exeter, Penryn Campus, Penryn TR10 9EZ, UK <sup>5</sup>Institute for Theoretical Biology, Humboldt University Berlin, D-10115 Berlin, Germany

\*Corresponding author: e-mail: olof.leimar@zoologi.su.se

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#### Abstract

Genetic polymorphism can contribute to local adaptation in heterogeneous habitats, for instance as a single locus with alleles adapted to different habitats. Phenotypic plasticity can also contribute to trait variation across habitats, through developmental responses to habitat-specific cues. We show that the genetic architecture of genetically polymorphic and plasticity loci may influence the balance between local adaptation and phenotypic plasticity. These effects of genetic architecture are instances of ecological genetic conflict. A reduced effective migration rate for genes tightly linked to a genetic polymorphism provides an explanation for the effects, and they can occur both for a single trait and for a syndrome of co-adapted traits. Using individualbased simulations and numerical analysis, we investigate how among-habitat genetic polymorphism and phenotypic plasticity depend on genetic architecture. We also study the evolution of genetic architecture itself, in the form of rates of recombination between genetically polymorphic loci and plasticity loci. Our main result is that for plasticity genes that are unlinked to loci with between-habitat genetic polymorphism, the slope of a reaction norm is steeper in comparison with the slope favored by plasticity genes that are tightly linked to genes for local adaptation.

Keywords: Genetic conflict, local adaptation, phenotypic plasticity, ecotypes, genetic architecture, linkage

### <sup>1</sup> Introduction

Genetic local adaptation and phenotypic plasticity can both contribute to trait vari-2 ation in heterogeneous habitats. Our aim here is to investigate how their relative 3 contribution depends on genetic architecture. In particular, we examine the influ-4 ence on trait variation of the rate of recombination between genetically polymorphic 5 loci, responsible for local adaptation, and plasticity loci influencing the intercept 6 and slope of a reaction norm (fig. 1). As we explain, this issue is less straightfor-7 ward than is generally understood, even though it involves phenomena that have a 8 long history of study in evolutionary ecology (e.g., van Tienderen 1991, 1997; West-9 Eberhard 2003; DeWitt and Langerhans 2004; Kawecki and Ebert 2004; Richards 10 et al. 2006; Griffith and Sultan 2012). The reason is that there can be ecological 11 genetic conflict between plasticity genes that are tightly vs. loosely linked to a ge-12 netically polymorphic locus. Ecological genetic conflict has been studied previously 13 (Leimar et al. 2006; Dall et al. 2015; Leimar et al. 2016), but its importance for local 14 adaptation and plasticity is not widely recognized. To shed light on the issue, we 15 demonstrate the influence of genetic conflict on between-habitat genetic polymor-16 phism and phenotypic plasticity of a trait, in addition to the already well-studied 17 effects of factors like gene flow between habitats, the strength of selection, and en-18 vironmental cue accuracy (Tufto 2000; Sultan and Spencer 2002). As we will show, 19 ecological genetic conflict depends on genetic architecture. 20

In general, genetic conflict occurs when genetic elements, such as alleles at differ-21 ent loci influencing the same phenotypic trait, differ in their selective circumstances, 22 and thus in their evolutionary interests. This possibility has been given much at-23 tention (Hurst et al. 1996; Werren and Beukeboom 1998; Burt and Trivers 2006; 24 Gardner and Ubeda 2017), resulting in the insight that genetic conflict can be im-25 portant for evolutionary change and innovation, as well as influence phenomena like 26 sex determination (Werren 2011). Most work has focused on genetic conflict with a 27 basis in the properties of genetic transmission systems. For instance, the different 28 pathways of transmission for nuclear and mitochondrial genes has been put forward 29 as a source of genetic conflict (e.g., Frank and Hurst 1996; Perlman et al. 2015). 30

It is known that the balance between polymorphism and plasticity is influenced by the level of between-habitat migration (e.g., Tufto 2000; Sultan and Spencer 2002; Leimar et al. 2006), with less migration favoring between-habitat genetic polymorphism. It is also known that the 'effective migration rate' (Bengtsson 1985; Barton and Bengtsson 1986), estimated for a neutral locus linked to a selected, genetically polymorphic locus, is lower for tighter linkage to the polymorphism, and is also <sup>37</sup> lower for stronger selection maintaining the polymorphism. Putting these observa-<sup>38</sup> tions together suggests that the selective circumstances for plasticity loci depend <sup>39</sup> on their linkage to a polymorphic locus, with weaker selection for plasticity when <sup>40</sup> linkage is tight and the effective migration rate thus is low. This difference in how <sup>41</sup> selection operates on plasticity genes, depending on their position in the genome, <sup>42</sup> can be interpreted as an ecological genetic conflict.

In our evolutionary modeling, we examine between-habitat variation of one trait, 43 as well as of two different traits, for which the optimum differs between habitats. 44 The kind of genetic architecture we are concerned with is the degree of linkage be-45 tween genetically polymorphic loci, epistatic modifiers of the effects at these loci, 46 and genes influencing a reaction norm slope. We emphasize the distinction between 47 the case where all loci are tightly linked together and that where modifier and slope 48 loci are unlinked to genetically polymorphic loci (fig. 1B), but we also investigate 49 intermediate cases, for instance a polymorphic locus with a tightly linked modifier 50 and an unlinked locus determining the slope of a reaction norm. We also investi-51 gate effects of previously well-studied factors such as the rate of between habitat 52 migration, the strength of selection, and the accuracy of environmental cues. Fi-53 nally, we study the question of the evolution of the rate of recombination between 54 polymorphic loci, modifiers, and plasticity loci. Our aim is to determine if either 55 tight or loose linkage of plasticity loci to a genetically polymorphic locus is favored 56 by selection. 57

For the analysis, we use individual-based evolutionary simulations of diploid pop-58 ulations, with several local populations in each habitat, as well as numerical analysis 59 of evolutionary equilibria for a model with a very large population in each habitat. 60 For simplicity, we let the sex of an individual be randomly determined (Perrin 2016). 61 A complication we need to deal with is that, because of the presence of genetic con-62 flict, natural selection need not in general lead to a unique outcome for plasticity 63 loci with different linkage to a polymorphism. The alleles in the polymorphism can 64 also evolve in ways that counteract evolutionary change at plasticity loci. These 65 kinds of 'arms races', where the outcome is influenced by such things as supply of 66 mutations, position in the genome, and limits to gene expression are common for 67 genetic conflicts (Hurst et al. 1996; Werren 2011). Our main way of dealing with this 68 is to examine situations where there is selectively maintained genetic polymorphism 69 at one or more loci, but where we do not focus on the possible evolution of the 70 corresponding alleles. Instead, we examine the evolution of genes that modify the 71 phenotypic effects of the polymorphism, for instance modify intercepts and slopes of 72 a reaction norm (fig. 1). As an alternative, we also examine a situation where alleles 73

<sup>74</sup> at a polymorphic locus can mutate and evolve, but there are limits to the sizes of<sup>75</sup> the allelic effects.

Our main finding is that for plasticity genes that are unlinked to a genetic poly-76 morphism, the slope of a reaction norm will be steeper in comparison with the slope 77 favored by plasticity genes that are tightly linked to genes for local adaptation. 78 This holds in particular for intermediate rates of between-habitat migration. Fur-79 thermore, we do not find selection favoring either tight or loose linkage of plasticity 80 genes to a genetic polymorphism, suggesting that this genetic architecture is set 81 by other influences on genome organization. We discuss our results in relation to 82 empirical work on the genomics of ecotypic variation and on the relative importance 83 of local adaptation and plasticity for trait variation. 84

### $_{85}$ Methods

We first present our two-habitat metapopulation model for a single trait u, then extend it to two traits  $u_1$  and  $u_2$ , followed by an explanation of our individual-based simulations. We have also performed a numerical analysis of a model with a very large population in each habitat, which is described in appendix A, with results reported in Table A1 and figure A1.

#### <sup>91</sup> Single trait

The population is divided into  $N_p$  patches, each containing a local population with on average K diploid individuals with non-overlapping generations, and with survival selection operating in each patch. An individual's sex is randomly determined, with equal chances for female and male, and each offspring is formed by randomly selecting a mother and a father from the local population. There is a genotype-cuephenotype mapping, determining an individual's phenotype u as a weighted sum of a 'genetic effect' z and an environmental cue  $x_{juv}$  observed by juveniles, such that

$$u = \alpha z + \beta x_{juv},\tag{1}$$

where z and the weights  $\alpha$  and  $\beta$  are each determined by a diploid locus. This means that there is epistasis between the locus for the genetic effect z and the locus coding for  $\alpha$ .

<sup>102</sup> A patch is in either of two environmental states, corresponding to two types <sup>103</sup> of habitat, which could, for instance, be low and high resource availability, risk of predation, or salinity. The two habitats are denoted by i = 1, 2, with juvenile-toadult survival for phenotype u in habitat i given by

$$s_i(u) = s_0 + (1 - s_0) \exp\left(-\frac{(u - \theta_i)^2}{2\sigma^2}\right),$$
 (2)

where  $s_0$  is a basic survival rate,  $\theta_i$  is the optimal phenotype in habitat *i* and  $\sigma$  is the width of the Gaussian survival function. An individual can get information about which habitat it is in through the juvenile cue, given by

$$x_{\rm juv} = \theta_i + \epsilon_{\rm juv},\tag{3}$$

where  $\theta_i$  is the mean cue in habitat *i*, for simplicity assumed to be the same as the optimal phenotype, and  $\epsilon_{juv}$  is a normally distributed random error with mean 0 and standard deviation  $\sigma_{juv}$ .

There is a probability m of juvenile dispersal to a patch randomly selected in the entire metapopulation, including the patch of origin. The local populations are regulated such that a patch produces K juveniles, each of which has a probability mto disperse. There are equal numbers of patches for the two habitats, which means that the probability for a dispersing individual to change habitat is 1/2.

The life cycle of individuals is as follows: (i) selection, with survival in habitat is as a function of phenotype u as in equation (2); (ii) within-patch random mating, forming K offspring in each local population, after which the adults die; (iii) each juvenile (independently) observes an environmental cue, as given in equation (3), and has its phenotype determined based on its genotype and the environmental cue; (iv) each juvenile has a probability m of migrating to a randomly chosen patch; and the cycle then returns to (i).

At the locus for z there are alleles  $\zeta_k$ , which we represent as real values limited to an interval. We are interested in situations where there is adaptively maintained genetic polymorphism at this locus. In principle the alleles  $\zeta_k$  can mutate, be selected, and evolve, but in order to aid the interpretation of our results, we first make the simplification that there are only two alleles with fixed values  $\zeta_1$  and  $\zeta_2$ . However, in appendix A we show results for a case where alleles can mutate and evolve within set limits (fig. A2).

We think of the effects of the alleles at the locus for z as 'genetic cues', in the sense that they can provide statistical information to an individual about which habitat it is in (Leimar et al. 2006; Leimar and McNamara 2015; Dall et al. 2015). The nature of the information is that the allele frequencies at the locus for z can

differ between habitats, giving rise to a correlation between habitat and z. The 135 value z then acts as a statistical predictor of the habitat in a similar manner as the 136 environmental cue  $x_{iuv}$ . The idea helps understanding how selection acts on the loci 137 for  $\alpha$  and  $\beta$ , because these are weights on z and  $x_{iuv}$  for determining the phenotype 138 (equation 1). A better match between habitat and phenotype gives higher survival 139 (equation 2), and selection at the loci for  $\alpha$  and  $\beta$  acts towards achieving high 140 survival. As mentioned in the Introduction, this selection depends on the effective 141 migration rate between habitats that a locus is exposed to. 142

The locus for the weight  $\alpha$  in equation (1) can be seen as a 'modifier' locus, with 143 alleles  $\alpha_k$ , that influence gene expression at the cue locus (note also that evolutionary 144 changes of a modifier that is fully linked to z has similar effects as evolutionary 145 changes of the genes at the locus for z). We represent the alleles  $\alpha_k$  as real values 146 in an interval. The phenotype in equation (1) is also influenced by the juvenile cue, 147 mediated by the locus for the weight  $\beta$ , with alleles  $\beta_k$ . In terms of plasticity,  $\beta$  is 148 the slope of a reaction norm, and the alleles at the locus can be regarded as plasticity 149 genes. We assume the loci are positioned in the order  $z, \alpha, \beta$  along a chromosome, 150 with  $\rho_{z\alpha}$  the recombination rate between the cue locus and the modifier locus  $\alpha$ , 151 and  $\rho_{\alpha\beta}$  the rate between the modifier locus and the plasticity locus  $\beta$ . 152

The alleles at a locus are additive, producing diploid values as the sum of maternal and paternal allelic values. For instance, at the genetic effect locus we have  $z = \zeta_{\text{mat}} + \zeta_{\text{pat}}$ . The value z is referred to as a genetic effect or 'genetic cue', which can be polymorphic across habitats. For the loci giving the weights in equation (1), we are interested in cases where the modifier and slope effects,  $\alpha = \alpha_{\text{mat}} + \alpha_{\text{pat}}$  and  $\beta = \beta_{\text{mat}} + \beta_{\text{pat}}$ , display fairly little genetic variation over the metapopulation, but still evolve over the longer term.

#### 160 Two traits

We extend the situation above to two traits,  $u_1$  and  $u_2$ , determined as

$$u_1 = \alpha_1 z_1 + \beta_1 x_{juv}$$
(4)  
$$u_2 = \alpha_2 z_2 + \beta_2 x_{juv}.$$

The genetic effects  $z_1$  and  $z_2$  are each determined by a locus with additive alleles, as in the case for a single trait above, and the juvenile environmental cue is given by equation (3). The modifiers  $\alpha_1$ ,  $\alpha_2$  and slopes  $\beta_1$ ,  $\beta_2$  are determined genetically by separate loci. The juvenile-to-adult survival in habitat i is given by

$$s_i(u_1, u_2) = s_0 + (1 - s_0) \exp\left(-\frac{(u_1 - \theta_{1i})^2 + (u_2 - \theta_{2i})^2}{2\sigma^2}\right).$$
 (5)

The loci are positioned in the order  $z_1$ ,  $z_2$ ,  $\alpha_1$ ,  $\alpha_2$ ,  $\beta_1$ ,  $\beta_1$  along a chromosome. In the same way as for the single trait, we make the simplification that there are only two alleles at each of the loci for  $z_1$  and  $z_2$ , with fixed values  $\zeta_{11}$  and  $\zeta_{12}$  for  $z_1$ and  $\zeta_{21}$  and  $\zeta_{22}$  for  $z_2$ . Concerning recombination rates,  $\rho_{zz}$  is the recombination rate between the loci for the genetic effects  $z_1$  and  $z_2$ ,  $\rho_{z\alpha}$  is the recombination rate between the locus for  $z_2$  and the locus for  $\alpha_1$ , and  $\rho_{\alpha\beta}$  is the recombination rate between neighboring loci for  $\alpha_1$ ,  $\alpha_2$ ,  $\beta_1$  and  $\beta_2$ .

#### 173 Simulation models

For our individual-based simulations in figures 2 and 5, we started with a dimorphism 174 at the locus for z, and kept the values of the two alleles fixed, while  $\alpha$  and  $\beta$  evolved. 175 For parameter values for which  $\alpha$  became close to 0, the dimorphism at the locus 176 for z was sometimes not maintained, because one of the alleles was lost from the 177 population through genetic drift. In such a case, for figure 2 we ran that replicate 178 simulation one more time, but for figure 5 we used the replicates where one of the 179 alleles was lost. In either case, this did not influence the evolutionary outcome for 180 plasticity loci. As mentioned, we used intervals for the allowed range of the values 181 of alleles. For the simulations in Figs. 2 and 3 we used  $\zeta_1 = -0.4$ ,  $\zeta_2 = 0.4$  and 182 the range [0.0, 0.5] for alleles at the loci for  $\alpha$  and  $\beta$ . Mutational increments had a 183 Laplace (reflected exponential) distribution with a standard deviation of 0.04, but 184 allelic values were constrained to stay within the interval. The simulations were run 185 for  $50\,000$  generations with a mutation rate of 0.0050, to generate enough genetic 186 variation for adaptation to proceed, followed by 50 000 generations with a mutation 187 rate of 0.0001, to remove excess genetic variation. The simulations in Figs. 5 and 6 188 were performed in a similar way. 189

To investigate the evolution of rates of recombination between the loci for  $z_1$ and  $z_2$ , and between these and the plasticity loci in simulations similar to those in figure 5, we introduced additional loci controlling recombination. There were 9 loci along a chromosome, coding for  $z_1$ ,  $z_2$ ,  $\rho_{zz}$ ,  $\rho_{z\alpha}$ ,  $\rho_{\alpha\beta}$ ,  $\alpha_1$ ,  $\alpha_2$ ,  $\beta_1$ , and  $\beta_2$ . The loci for the recombination rates were tightly linked to the locus for  $z_2$ , in order to maximize the chances of the evolution of tighter linkage of the plasticity loci to the polymorphic complex of  $z_1$  and  $z_2$ . The effects of alleles at the recombination loci were constrained to lie in the interval [0.0, 0.25], resulting in diploid recombination values in [0.0, 0.5]. The C++ source code for the different computer programs used in this work is available as an online electronic enhancement.

### $_{200}$ Results

The effect of genetic architecture on local adaptation and phenotypic plasticity is 201 illustrated in figures 2 and 3, with data from individual-based simulations. There 202 is a single trait u, with optimal survival at trait value  $\theta_1$  and  $\theta_2$  in habitat 1 and 203 2 (equation 2). The determination of the phenotype is given by  $u = \alpha z + \beta x_{juv}$ , 204 where z is a genetic effect,  $\alpha$  is an epistatic modifier of z,  $x_{iuv}$  is an environmental 205 cue (equation 3), and  $\beta$  is a plasticity effect, giving the slope of a reaction norm 206 (equation 1). Each of z,  $\alpha$ , and  $\beta$  is determined by a single diploid locus with additive 207 allelic effects, and we are comparing the case where the loci are tightly linked with 208 that where they are all unlinked (figs. 2, 3). As seen in figure 2, for intermediate 209 rates of migration between habitats the genetic architecture strongly influences the 210 values of  $\alpha$  and  $\beta$ , and thus the contributions of genetic polymorphism and plasticity 211 to variation in u. For tightly linked loci, the genetic contribution to the variation 212 is larger than for unlinked loci, and the reverse is true for the contribution from 213 plasticity. For high enough rate of migration, the modifier  $\alpha$  evolves to be close 214 to zero (fig. 2A), so that z in equation (1) has little influence on the phenotype u. 215 There is then little section at the locus for z, and the allele frequencies undergo 216 genetic drift, with an average frequency of 0.5 for the 'locally fit' allele (fig. 2B). 217 A similar situation, but where the allelic values at the locus for z can mutate and 218 evolve is shown in figure A2. 219

The influence of genetic architecture is further exemplified by the reaction norms 220 for migration rate m = 0.10 between local populations (corresponding to a migration 221 rate of 0.05 between habitats), which are shown in figure 3, together with the dis-222 tributions of the environmental cue that adults in the different habitats observed as 223 juveniles. For the linked case, there are reaction norms with shallower slopes, with 224 different mean intercepts for individuals in habitats 1 and 2 with different genotypes 225 (red and blue lines in fig. 3 represent habitat 1 and 2 specialists, cf. fig. 1A). There 226 is genetic variation in z in each habitat: there are two alleles, each better adapted to 227 one of the habitats, giving rise to alternative homozygotes and heterozygotes, with 228 different frequencies in the habitats (fig. 2B; in principle, these alleles can evolve; 229 see fig. A2). For the unlinked case, there is a single reaction norm with a steeper 230

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slope (dashed line in fig. 3), corresponding to a phenotypically plastic generalist.
Note that the only difference in model parameters between the linked and unlinked
cases is the genetic architecture, demonstrating that ecological genetic conflict can
have a pronounced influence on phenotype determination.

This effect of genetic architecture hinges on whether genes tightly linked to one 235 of the alleles at the polymorphic locus for z, adapted to one of the habitats, has 236 an appreciable chance of recombining to become associated with an allele locally 237 adapted to the other habitat, as well as migrating to that habitat. The way this can 238 happen is if a modifier or slope allele occurs in a heterozygote between alleles at the 239 locus for z, each adapted to different habitats. The strength of selection against such 240 a heterozygote influences the chance for the modifier or slope allele to recombine 241 to the other locally adapted allele. For the linked case shown in figure 3, this 242 chance is small, illustrating that genes for specialism have their evolutionary future 243 mainly in their own habitat. While studying between-habitat genetic polymorphism, 244 Bengtsson (1985) and Barton and Bengtsson (1986) introduced the concept of an 245 effective migration rate for a neutral locus that is linked to a selected, genetically 246 polymorphic locus. For instance, using equation (4) in Yeaman and Whitlock (2011), 247 and ignoring the effects of plasticity, we find an effective migration rate of 0.0002 248 for a linkage of  $\rho = 0.001$  to z (fig. 3), so for such genes the two habitats are fairly 249 isolated from each other. 250

An alternative and potentially more informative way of showing how the selective 251 circumstances vary with the degree of linkage to a between-habitat polymorphism 252 is to examine how the reproductive value for a modifier or slope allele of being 253 associated (linked) with an allele adapted to one or the other habitat depends on 254 the rate of recombination. The reason this is informative is that the reproductive 255 value measures the long-term representation in the population of a plasticity allele 256 in a given position (i.e., linked to the locally more fit or the locally less fit allele 257 at the locus for z). We have performed a numerical analysis of a model with a 258 very large population in each habitat (see appendix A for model description), but 259 otherwise similar to the simulation model with results in figures 2 and 3. The 260 results of the numerical analysis, which takes into account plasticity, are given in 261 Table A1 and figure A1. The outcome of the analysis using reproductive values 262 is less extreme but qualitatively similar to the consideration of effective migration 263 rates. As seen in Table A1, for m around 0.1 ( $m_{12}$  around 0.05) and with modifier 264 and slope loci tightly linked to z, the reproductive value of being associated with 265 the locally adapted allele at the genetic effect locus is around four times higher than 266 that of being associated with the other allele, whereas these values are nearly equal 267

<sup>268</sup> for loosely linked modifiers.

In any case, for a migration rate above a critical value, phenotype determination for the linked case (as well as for the unlinked case) is dominated by plasticity, because the modifier  $\alpha$  in equation (1) approaches zero. For instance, in figure 2 the critical migration rate is m = 0.14. The critical migration rate for a wider range of parameters is shown in figure 4. In general, stronger selection between habitats and less accurate juvenile environmental cues favor genetic polymorphism in z, and thus a higher value of the critical migration rate (fig. 4).

For two traits,  $u_1$  and  $u_2$ , each with different optima in the habitats, as given 276 by equation (5), we again find a pronounced influence of genetic architecture on the 277 relative importance of genetic polymorphism and plasticity (figs. 5, 6). For each 278 trait,  $u_1$  and  $u_2$ , there is a separate genetic effect,  $z_1$  and  $z_2$ , coded by one locus, 279 with modifier  $\alpha_1$  and  $\alpha_2$  and reaction norm slope  $\beta_1$  and  $\beta_2$ , but the same juvenile 280 environmental cue  $x_{juv}$  for both reaction norms, as given in equation (4). Three 281 cases are illustrated in figure 5, one where all loci are linked, another where the 282 two genetic effect loci are linked and the loci for  $\alpha_1$ ,  $\alpha_2$ ,  $\beta_1$  and  $\beta_2$  are unlinked 283 from each other and from the genetic effect loci, and a third case where all loci 284 are unlinked. From this figure, and the example in figure 6, it appears that the 285 influence of genetic architecture is qualitatively similar but even stronger for a two-286 trait syndrome compared to a single trait. The reason for this difference between 287 figures 2 and 5 is most likely that the total selection between habitats is stronger 288 for the two-trait syndrome. 289

For the two-trait syndrome, we explored the evolution of genetic architecture 290 using individual-based simulations. Instead of specifying the recombination rates 291  $\rho_{zz}$ ,  $\rho_{z\alpha}$  and  $\rho_{\alpha\beta}$ , we let these be coded by three loci. We found that tight linkage 292 between the two polymorphic effect loci  $z_1$  and  $z_2$  promptly evolved (i.e.,  $\rho_{zz}$  became 293 close to zero; Table 1), so these loci emerge as a polymorphic complex. However, for 294  $\alpha_1, \alpha_2, \beta_1$  and  $\beta_2$  we did not find notable selection for either tighter or looser linkage 295 to the  $z_1$  and  $z_2$  complex. Considerable genetic variation for the recombination rates 296  $\rho_{z\alpha}$  and  $\rho_{\alpha\beta}$  persisted in the population, perhaps as a result of mutation-drift balance 297 (see Table 1 and figures 7 and A3 for illustrations of these simulations). Overall, 298 the outcome for the modifiers  $\alpha_1$ ,  $\alpha_2$  and plasticity slopes  $\beta_1$ ,  $\beta_2$ , shown in Table 1, 299 was similar to the middle (gray) case in figure 5, with tightly linked  $z_1$  and  $z_2$  and 300 unlinked loci for modifiers and slopes. 301

### 302 Discussion

Both local adaptation and plasticity are well-studied phenomena (van Tienderen 303 1991, 1997; West-Eberhard 2003; DeWitt and Langerhans 2004; Richards et al. 304 2006; Griffith and Sultan 2012). Even so, by examining ecological genetic conflict, 305 we have identified phenomena that were not studied before. Compared to previous 306 work, the major new aspect here is that we show the influence of genetic architec-307 ture on the balance between genetic polymorphism and phenotypic plasticity, and 308 that we interpret our results in terms of genetic conflict. We find that the rate of 309 recombination between genetic effect, modifier and plasticity loci influences the evo-310 lutionary outcome, with more plasticity and less genetic polymorphism for unlinked 311 loci. The effect of genetic architecture is strongest for intermediate migration rates 312 (figs. 2 and 5). 313

As we have mentioned, one way of understanding the effect of genetic architecture 314 is as a low effective migration rate for loci tightly linked to a genetic polymorphism 315 (Barton and Bengtsson 1986; Charlesworth et al. 1997; Bürger and Akerman 2011; 316 Yeaman and Whitlock 2011; Aeschbacher and Bürger 2014; Aeschbacher et al. 2017). 317 Another approach is to compute reproductive values of modifier and slope alleles, 318 as we have done (Table A1). Based on our analysis we find that a plasticity allele 319 tightly linked to a polymorphic genetic effect locus can leave copies of itself to 320 future generations also when linked to a genetic effect allele that is not adapted to 321 the current habitat, because migration can transport it to the other habitat. Thus, 322 migration makes the distinction between linked and unlinked genetic architectures 323 a matter of degree rather than kind. 324

In fact, the general pattern of variation of the modifier  $\alpha$  and plasticity slope  $\beta$ 325 with the migration rate m is qualitatively similar for different genetic architectures, 326 with a shift from mainly genetic polymorphism to mainly phenotypic plasticity as 327 m increases (figs. 2, 5, A1, A2). One way of explaining this shift is in terms of the 328 statistical information about the habitat that is contained in the 'genetic cue' z in 329 comparison with the environmental cue  $x_{iuv}$  (see Methods section and Leimar et al. 330 2006; Leimar and McNamara 2015; Dall et al. 2015). Tufto (2000) provides a discus-331 sion of earlier papers dealing with this topic. For higher values of m, allele frequency 332 differences between habitats are smaller (figs. 2B, 5B and Table A1), thus being less 333 statistically informative about the habitat compared to the environmental cue  $x_{iuv}$ . 334 An optimal phenotype determination strategy will therefore put less emphasis on 335 the genetic and more on the environmental cue for higher values of m. For high 336 enough rates of migration, and provided that environmental cues are sufficiently 337

accurate, phenotypic plasticity dominates completely, as illustrated by figure 4. For
a much simpler model with binary cues, inspired by the work of Sultan and Spencer
(2002), an analytical solution is possible, leading to qualitatively similar results (see
equation 4 and figure 5 in Leimar et al. 2006).

Our main result that reaction norm slopes can depend on the genetic architecture 342 (Figs. 2, 3, 5, 6, A1, A2) is new, and there appear to be no empirical data directly 343 examining this question. It is known that ecotypic traits differ in how they are 344 determined, with the variation in some traits being mainly genetic and in other traits 345 mainly plastic (Lucek et al. 2014), but the possible influence of genetic architecture 346 is unknown. There are observations showing that plasticity can decrease during the 347 formation of an ecotype (Hasan et al. 2017), but the genomic basis of the reduction 348 in plasticity is not known. Also, a study of so called expression quantitative trait 349 loci (eQTLs) shows that 'distant', trans-regulatory changes on average had different 350 effects than 'local', *cis*-regulatory changes, and were also more responsive to the 351 environment (Ishikawa et al. 2017), which is at least suggestive of an influence of 352 genetic architecture on trait expression. 353

In our investigation of the evolution of recombination, for a two-trait situation, 354 we found that a low recombination rate between the polymorphic loci for  $z_1$  and  $z_2$ 355 readily evolved (Table 1), and this is in accordance with the traditional understand-356 ing of such situations (Pinho and Hey 2010; Via 2012). On the other hand, we did 357 not detect selection for either tighter or looser linkage between the polymorphic loci 358 and epistatic modifiers and plasticity loci (fig. 7 and Table 1). The question appears 359 not to have been analyzed previously, but perhaps other factors influencing genetic 360 architecture, such as inversions or a tendency towards *cis*-regulatory influences, can 361 play a greater role in determining recombination rates between plasticity loci and 362 polymorphic loci. Overall, we hope that our work can inspire further empirical 363 and theoretical investigation of the genomics of local adaptation and plasticity of 364 ecotypes. 365

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## Tables

Table 1: Evolution of linkage for two-trait simulations similar to fig. 5. The table gives the mean  $\pm$  SD of the population average of individual recombination rates, and the averages of the  $\alpha_n$  and  $\beta_n$ , over 20 replicate simulations. The recombination rate  $\rho_{zz}$  between  $z_1$  and  $z_2$  evolved towards tight linkage, but the other recombination rates reached intermediate average values, with broad distributions over replicate simulations, as illustrated in fig. 7.

m	$ ho_{zz}$	$ ho_{zlpha}$	$ ho_{lphaeta}$	$\alpha_n$	$\beta_n$
0.12	$0.0028 {\pm} 0.0018$	$0.296{\pm}0.091$	$0.261 \pm 0.112$	$0.870 {\pm} 0.011$	$0.067 {\pm} 0.009$
0.18	$0.0017 {\pm} 0.0009$	$0.260{\pm}0.072$	$0.294{\pm}0.097$	$0.807 {\pm} 0.017$	$0.126 {\pm} 0.013$
0.24	$0.0009 \pm 0.0005$	$0.259 {\pm} 0.106$	$0.251{\pm}0.124$	$0.690 {\pm} 0.029$	$0.239 {\pm} 0.027$

### Figures



Figure 1: Sketch of reaction norms and genetic architectures. There are two habitats, with blue and red being used to indicate reaction norms and alleles adapted to each habitat. A. At one extreme, reaction norms could be flat with intercepts adapted to one or the other habitat (blue vs. red lines), corresponding to a pure genetic polymorphism (habitat specialism), and at the other extreme there could be a single reaction norm (dashed gray line), corresponding to pure phenotypic plasticity (habitat generalism). Intercepts and slopes of reaction norms are determined by plasticity genes, either influencing intercepts by epistatically modifying the effects of a genetic polymorphic locus, or by determining the slope of a reaction norm. There is genetic conflict between plasticity genes that are tightly vs. loosely linked to a genetic polymorphism. B. A sketch of the placement of genes along a hypothetical chromosome. A polymorphic locus with two alleles (blue and red) having effects that are suited to each of the habitats. Two cases are illustrated, one with plasticity genes (green) tightly linked to the polymorphic locus, and another with loose linkage.



Figure 2: Phenotype determination for linked and unlinked genetic architectures, as a function of the rate of migration. Panel A shows how the value of the epistatic modifier  $\alpha$  (solid lines) of the genetic effect z and the slope  $\beta$  (dashed lines) of the reaction norm for the environmental cue  $x_{juv}$  depend on the migration rate m and on the genetic architecture. The mean  $\pm$  SD over 10 replicate individual-based simulations is displayed. The left-hand (green) lines correspond to the case where the loci for z,  $\alpha$  and  $\beta$  are all unlinked and the right-hand (orange) lines to the case where the three loci are tightly linked. The lines between these (gray) correspond to an intermediate case where the loci for z and  $\alpha$  are linked but the locus for  $\beta$  is unlinked to these. Panel B shows the average frequency of the 'locally fit' allele, i.e. the average of the frequency of  $\zeta_1$  in habitat 1 and of  $\zeta_2$  in habitat 2. The dashed dark gray line shows the case of pure genetic polymorphism ( $\alpha = 1$ and  $\beta = 0$ ). Survival selection between habitats is given by equation (2) and the phenotype is determined as in equation (1). For the linked case, recombination rates are  $\rho_{z\alpha} = \rho_{\alpha\beta} = 0.001$ , for the unlinked case  $\rho_{z\alpha} = \rho_{\alpha\beta} = 0.5$ , and for the intermediate case  $\rho_{z\alpha} = 0.001, \rho_{\alpha\beta} = 0.5$ . Other parameter values:  $N_p = 200$ ,  $K = 100, s_0 = 0.1, \sigma = 1.0, \theta_1 = -0.75, \theta_2 = 0.75, \zeta_1 = -0.4, \zeta_2 = 0.4, \sigma_{juv} = 0.5.$ 



Figure 3: Example of the effect of genetic architecture (linked or unlinked) on phenotype determination. Mean reaction norms (with slope  $\beta$ ) for habitat 1 specialists: thick and thin blue lines (slightly shifted up and down for clarity) represent individuals in habitat 1 with genotype  $\zeta_1\zeta_1$  and  $\zeta_1\zeta_2$  (with frequencies before migration of 0.76 and 0.22; line widths proportional to frequencies); and habitat 2 specialists: thick and thin red lines represent individuals in habitat 2 with genotype  $\zeta_2\zeta_2$  and  $\zeta_1\zeta_2$  (with frequencies 0.77 and 0.21); and for phenotypically plastic generalists: gray dashed line, slope and intercepts averaged over both habitats). For the generalist, the reaction norm is very similar between habitats (not shown), because  $\alpha$  is small and  $\beta$  does not vary much, but the alleles  $\zeta_1$  and  $\zeta_2$  still segregate at the locus for z. The distributions of the juvenile environmental cue  $x_{juv}$  are shown lightly shaded for adult individuals in habitat 1 (left) and habitat 2 (right; not labeled to avoid visual clutter). The figure corresponds to the cases in figure 2 for migration rate m = 0.10, with tightly linked loci for specialism and unlinked loci for plasticity.



Figure 4: Critical migration rate, above which a genetic polymorphism in z is not selectively maintained, resulting in pure phenotypic plasticity. There is a single trait u and the loci for z,  $\alpha$  and  $\beta$  are tightly linked. The critical rate is defined as the value of m for which the genetic proportion of the trait variance in u is less than 0.01. The critical migration rate is shown as a function of the strength of selection in one habitat against a phenotype locally adapted to the other habitat, defined as  $1 - s_1(\theta_2) = 1 - s_2(\theta_1)$  (see equation 2 for definition of  $s_i$ ). The points correspond to  $s_0 = 0.9, 0.7, 0.5, 0.3, 0.1$ , and the lines are labeled with the juvenile environmental cue error,  $\sigma_{juv}$ . The rightmost point on the line for  $\sigma_{juv} = 0.50$ corresponds to the rightmost point for the linked case in fig. 2A, B. Other parameter values:  $\rho_{z\alpha} = \rho_{\alpha\beta} = 0.001$ ,  $N_p = 200$ , K = 100,  $\sigma = 1.0$ ,  $\theta_1 = -0.75$ ,  $\theta_2 = 0.75$ .



Figure 5: Phenotype determination for different genetic architectures, as a function of the rate of migration. Similar to figure 2, but there are two traits,  $u_1$  and  $u_2$ , each with optima that differ between the habitats. There are two genetic effect loci, one for each trait, and modifiers  $\alpha_1$  and  $\alpha_2$  for each of the genetic effects  $z_1$ and  $z_2$ , as well as slopes  $\beta_1$  and  $\beta_2$  for the reaction norms of  $u_1$  and  $u_2$  for the juvenile cue  $x_{juv}$ , following equation (4). Panel A shows how the mean modifier  $(\alpha_1 + \alpha_2)/2$  and mean slope  $(\beta_1 + \beta_2)/2$  depend on the migration rate m and on the genetic architecture. The solid lines show the mean modifier over 10 replicates of individual-based simulations, with the left-hand (green) line giving a case where the loci for the two genetic effects and the modifiers  $\alpha_1, \alpha_2, \beta_1, \beta_2$  are all unlinked. The right-hand (orange) line shows the same thing, except that the six loci are tightly linked. For the middle (gray) line, the two genetic effect loci are tightly linked, but the modifier and plasticity loci are unlinked from these and from each other. The dashed lines show the corresponding reaction norm slopes. The situation is symmetric between the traits, and the results for each trait separately are very similar to those shown here. Panel B shows the average frequency of the 'locally fit' allele, i.e. the average of the frequencies of  $\zeta_{11}$  and  $\zeta_{21}$  in habitat 1 and the frequencies of  $\zeta_{12}$  and  $\zeta_{22}$  in habitat 2 (for every second value of m in panel A). The dashed dark gray line shows the case of pure genetic polymorphism ( $\alpha_1 = \alpha_2 = 1$  and  $\beta_1 = \beta_2 = 0$ ). Survival selection between habitats is given by equation (5). For the linked case, recombination rates are  $\rho_{zz} = \rho_{z\alpha} = \rho_{\alpha\beta} = 0.001$ , and for the unlinked case  $\rho_{zz} = \rho_{z\alpha} = \rho_{\alpha\beta} = 0.5$ . Other parameter values:  $N_p = 200, K = 100, s_0 = 0.1,$  $\sigma = 1.0, \ \theta_{11} = \theta_{21} = -0.75, \ \theta_{12} = \theta_{22} = 0.75, \ \zeta_{11} = \zeta_{21} = -0.4, \ \zeta_{12} = \zeta_{22} = 0.4,$  $\sigma_{\rm juv} = 0.5.$ 



Figure 6: Example of the effect of genetic architecture (linked or unlinked) on phenotype determination. Mean reaction norms (with slope  $(\beta_1 + \beta_2)/2$ ) for habitat 1 specialists: thick and thin blue lines (slightly shifted up and down for clarity) represent individuals in habitat 1 with genotype  $\zeta_{n1}\zeta_{n1}$ ,  $\zeta_{n1}\zeta_{n2}$  and  $\zeta_{n2}\zeta_{n2}$  at each of the two genetic effect loci, n = 1, 2, (with frequencies after migration of 0.74, 0.15) and 0.10; line widths proportional to frequencies); and habitat 2 specialists: thick and thin red lines represent individuals in habitat 2 with genotype  $\zeta_{n2}\zeta_{n2}$ ,  $\zeta_{n1}\zeta_{n2}$  and  $\zeta_{n1}\zeta_{n1}$  at each of the two genetic effect loci (with frequencies 0.73, 0.16 and 0.11); and for phenotypically plastic generalists: gray dashed line, slopes and intercepts averaged over both habitats). For the linked case (specialists), the genotypes at the loci for  $z_1$  and  $z_2$  are highly correlated, both among habitats (correlation of genetic effects: 0.999) and within habitats (0.998). For the unlinked case, the reaction norm is very similar between habitats (not shown), because the  $\alpha_n$  are small and the  $\beta_n$  do not vary much, but the alleles  $\zeta_{21}$  and  $\zeta_{22}$  still segregate at the locus for  $z_2$ , whereas in this example  $z_1$  is fixed for  $\zeta_{12}$ . The distributions of the juvenile environmental cue  $x_{juv}$  are shown lightly shaded for adult individuals in habitat 1 (left) and habitat 2 (right). The figure corresponds to the cases in fig. 5 for migration rate m = 0.24, with tightly linked loci for specialism and unlinked loci for plasticity.



Figure 7: Illustration of the distributions of effect sizes of recombination alleles in the population, for 20 replicate simulations for each studied migration rate. Effects of alleles at the loci for  $\rho_{z\alpha}$  and  $\rho_{\alpha\beta}$  were constrained to lie in the interval [0.0, 0.25]. For each replicate simulation, the allelic effects in the population were put into 5 bins (i.e., on average a proportion of 0.2 per bin), and the mean and SD proportions over the replicates were computed. Panels A and B show effects for  $\rho_{z\alpha}$  and  $\rho_{\alpha\beta}$  for simulations with m = 0.12, B and C show the same for a simulations with m = 0.18, and E and F show the same for a simulations with m = 0.24. Figure A3 shows examples of the population distributions of diploid recombination values from the replicate simulations.

### Appendix A

#### Numerical analysis

Our approach here shows similarity to the numerical analysis by Leimar et al. (2016). The main aim of the analysis is to illustrate the difference in long-term representation in the population for a plasticity allele between being linked to the locally adapted vs. the locally non-adapted allele at a polymorphic genetic effect locus. We achieve this through the use of reproductive values, as illustrated in Table A1. We also show how the modifier  $\alpha$  and the slope  $\beta$  vary as the rate of recombination between these loci and the genetic effect increases from 0 to 0.5 (fig. A1).

Let habitat i, i = 1, 2, support a large population of size  $n_i$  and let  $m_{ij}$  be a rate of migration to habitat i from habitat j, in the sense that, after migration, the respective proportions  $m_{11}$  and  $m_{12}$  of individuals in habitat 1 originate from habitat 1 and 2, and similarly in habitat 2. We are mostly interested in the symmetric case where  $n_1 = n_2$ ,  $m_{11} = m_{22}$  and  $m_{12} = m_{21}$  The life cycle of individuals is a version of that in the main text: (i) within-habitat random mating, forming  $n_i$  offspring in habitat i, conceptualized as random unions from a pool of gametes, drawn from the adults in the habitat (after which the adults die); (ii) each juvenile (independently) observes an environmental cue, as given in equation (3), and has its phenotype determined based on its genotype and the environmental cue; (iii) each juvenile has a probability  $m_{ij}n_i/n_j$  of migrating from its habitat j to habitat i; (iv) selection, with survival in habitat i as a function of phenotype u as in equation (2); and the cycle then returns to (i).

Let us use notation like  $\zeta_k$  to denote alleles at the locus for z. We take (i) as our census point, and let  $p_{ik}$  be the frequency among the gametes (that form the next generation) of allele  $\zeta_k$  in habitat *i*. If we order the gametes as maternal-paternal, the genotype frequencies among the offspring at the census point in habitat *i* are  $p_{ik}p_{il}$ . Concerning environmental cues, note that the mean cue in habitat *i* is  $\theta_i$ , according to equation (3). The survival in habitat *i* of individuals with genotypes with alleles  $\zeta_k$  and  $\zeta_l$  who have observed the juvenile cue in habitat j becomes

$$W_{ijkl} = s_0 + (1 - s_0) \frac{1}{\sqrt{2\pi\sigma_{juv}^2}} \times$$

$$\int \exp\left(-\frac{-(\alpha(\zeta_k + \zeta_l) + \beta(\theta_j + \eta) - \theta_i)^2}{2\sigma^2}\right) \exp\left(-\frac{\eta^2}{2\sigma_{juv}^2}\right) d\eta$$

$$= s_0 + (1 - s_0) \frac{\sigma}{\sqrt{\beta^2 \sigma_{juv}^2 + \sigma^2}} \exp\left(-\frac{1}{2} \frac{(\alpha(\zeta_k + \zeta_l) + \beta\theta_j - \theta_i)^2}{\beta^2 \sigma_{juv}^2 + \sigma^2}\right),$$
(A1)

where the integration variable  $\eta$  represent the environmental cue error. Note that we have the symmetry  $W_{ijkl} = W_{ijlk}$ . Defining an average survival as

$$\bar{W}_{ij} = \sum_{kl} W_{ijkl} p_{jk} p_{jl},$$

we get the genotype frequencies at the end of phase (iv) as

$$P_{111}(p_{..}) = \frac{m_{11}W_{1111}p_{11}p_{11} + m_{12}W_{1211}p_{21}p_{21}}{m_{11}\bar{W}_{11} + m_{12}\bar{W}_{12}}$$
(A2)  

$$P_{112}(p_{..}) = \frac{m_{11}W_{1112}p_{11}p_{12} + m_{12}W_{1212}p_{21}p_{22}}{m_{11}\bar{W}_{11} + m_{12}\bar{W}_{12}}$$
  

$$P_{121}(p_{..}) = \frac{m_{11}W_{1121}p_{12}p_{11} + m_{12}W_{1221}p_{22}p_{21}}{m_{11}\bar{W}_{11} + m_{12}\bar{W}_{12}}$$
  

$$P_{122}(p_{..}) = \frac{m_{11}W_{1122}p_{12}p_{12} + m_{12}W_{1222}p_{22}p_{22}}{m_{11}\bar{W}_{11} + m_{12}\bar{W}_{12}} ,$$

in habitat 1, and

$$P_{211}(p_{..}) = \frac{m_{21}W_{2111}p_{11}p_{11} + m_{22}W_{2211}p_{21}p_{21}}{m_{21}\bar{W}_{21} + m_{22}\bar{W}_{22}}$$
(A3)  

$$P_{212}(p_{..}) = \frac{m_{21}W_{2112}p_{11}p_{12} + m_{22}W_{2212}p_{21}p_{22}}{m_{21}\bar{W}_{21} + m_{22}\bar{W}_{22}}$$
  

$$P_{221}(p_{..}) = \frac{m_{21}W_{2121}p_{12}p_{11} + m_{22}W_{2221}p_{22}p_{21}}{m_{21}\bar{W}_{21} + m_{22}\bar{W}_{22}}$$
  

$$P_{222}(p_{..}) = \frac{m_{21}W_{2122}p_{12}p_{12} + m_{22}W_{2222}p_{22}p_{22}}{m_{21}\bar{W}_{21} + m_{22}\bar{W}_{22}} .$$

in habitat 2. The notation  $P_{ikl}(p_{..})$  means that there is a dependence on the allele frequencies:  $p_{..} = (p_{11}, p_{21}, p_{12}, p_{22})$ . Again, we have the symmetry  $P_{ikl}(p_{..}) = P_{ilk}(p_{..})$ , and the index combination kl means that k is the maternal and l the paternal allele. From one generation to the next, we then have the following iteration for the allele frequencies at the census point:

$$p_{i1}(t+1) = P_{i11}(p_{..}(t)) + P_{i12}(p_{..}(t))$$

$$p_{i2}(t+1) = P_{i21}(p_{..}(t)) + P_{i22}(p_{..}(t)),$$
(A4)

where we have taken into account the symmetry  $P_{i12} = P_{i21}$ . We can note that  $p_{i1}(t+1) + p_{i2}(t+1) = 1$ , as it should, so we only need the equation for  $p_{i1}$ . The iteration (A4) can be used to determine numerically the equilibrium allele frequencies for a given situation, as is done in Table A1. In the following, we let  $\hat{p}_{ik}$  denote such an equilibrium.

#### Mutant invasion

We now consider a rare mutant modifier, that modifies either  $\zeta_1$ ,  $\zeta_2$ ,  $\alpha$  or  $\beta$ , and that has a rate of recombination  $\rho$  with the polymorphic locus for z. To make it simple, we assume that a modifier changes either  $\zeta_1$  to  $\zeta'_1$ , or  $\zeta_2$  to  $\zeta'_2$ , when linked to that allele, or modifies  $\alpha$  to  $\alpha'$  or  $\beta$  to  $\beta'$ . Thus, we examine one particular mutant modifier at a time. Let  $p'_{ik}$  be the frequency in habitat i of a mutant modifier linked to allele k, with  $p'_{ik} \ll \hat{p}_{ik}$ , and let  $W'_{ijkl}$  be the modified survival where the modifier is linked to allele l. Here, we do not distinguish maternal and paternal origin. Similar to equations (A2, A3), we have the first-order terms in mutant frequencies as

$$P_{111}' = \frac{2}{\bar{w}_1} (m_{11} W_{1111}' \hat{p}_{11} p_{11}' + m_{12} W_{1211}' \hat{p}_{21} p_{21}')$$
(A5)  

$$P_{121}' = \frac{2}{\bar{w}_1} (m_{11} W_{1121}' \hat{p}_{12} p_{11}' + m_{12} W_{1221}' \hat{p}_{22} p_{21}')$$
(A5)  

$$P_{112}' = \frac{2}{\bar{w}_1} (m_{11} W_{1112}' \hat{p}_{11} p_{12}' + m_{12} W_{1212}' \hat{p}_{21} p_{22}')$$
  

$$P_{122}' = \frac{2}{\bar{w}_1} (m_{11} W_{1122}' \hat{p}_{12} p_{12}' + m_{12} W_{1222}' \hat{p}_{22} p_{22}'),$$

and

$$P_{211}' = \frac{2}{\bar{w}_2} (m_{21} W_{2111}' \hat{p}_{11} p_{11}' + m_{22} W_{2211}' \hat{p}_{21} p_{21}')$$
(A6)  

$$P_{221}' = \frac{2}{\bar{w}_2} (m_{21} W_{2121}' \hat{p}_{12} p_{11}' + m_{22} W_{2221}' \hat{p}_{22} p_{21}')$$
(A6)  

$$P_{212}' = \frac{2}{\bar{w}_2} (m_{21} W_{2112}' \hat{p}_{11} p_{12}' + m_{22} W_{2212}' \hat{p}_{21} p_{22}')$$
(A6)

where we used the notation  $\bar{w}_1 = m_{11}\bar{W}_{11} + m_{12}\bar{W}_{12}$  and  $\bar{w}_2 = m_{21}\bar{W}_{21} + m_{22}\bar{W}_{22}$ . These represent mutant heterozygote genotypes surviving to the census point, ready to produce gametes for next generation:  $P'_{ikl}$  is the frequency of mutant heterozygotes in habitat *i* where the mutant modifier is linked to the *l* allele. Recombination when forming gametes from  $P'_{i12}$  and  $P'_{i21}$  can transfer the mutant modifier to become linked to the other allele at the locus for *z*. Using this, the iteration from one generation to the next for the  $p'_{ik}$  becomes:

$$p_{11}'(t+1) = \frac{1}{2} \left[ P_{111}'(t) + (1-\rho) P_{121}'(t) + \rho P_{112}'(t) \right]$$

$$p_{21}'(t+1) = \frac{1}{2} \left[ P_{211}'(t) + (1-\rho) P_{221}'(t) + \rho P_{212}'(t) \right]$$

$$p_{12}'(t+1) = \frac{1}{2} \left[ P_{122}'(t) + (1-\rho) P_{112}'(t) + \rho P_{121}'(t) \right]$$

$$p_{22}'(t+1) = \frac{1}{2} \left[ P_{222}'(t) + (1-\rho) P_{212}'(t) + \rho P_{221}'(t) \right] .$$
(A7)

We can write the mutant population projection as

$$p'_{ik}(t+1) = \sum_{jl} A'_{ikjl} p'_{jl}(t), \tag{A8}$$

where  $A'_{ikjl}$  is the population projection matrix. We get

$$A'_{i1j1} = \frac{m_{ij}}{\bar{w}_i} \left( W'_{ij11} \hat{p}_{j1} + (1-\rho) W'_{ij21} \hat{p}_{j2} \right)$$
(A9)  

$$A'_{i1j2} = \frac{m_{ij}}{\bar{w}_i} \rho W' i j 12 \hat{p}_{j1}$$
  

$$A'_{i2j1} = \frac{m_{ij}}{\bar{w}_i} \rho W' i j 21 \hat{p}_{j2}$$
  

$$A'_{i2j2} = \frac{m_{ij}}{\bar{w}_i} \left( W'_{ij22} \hat{p}_{j2} + (1-\rho) W'_{ij12} \hat{p}_{j1} \right).$$

The mutant projection is a  $4 \times 4$  matrix, and each line of equation (A9) represents a partitioning of this matrix into  $2 \times 2$  sub-matrices.

#### Invasion fitness

The leading eigenvalue  $\lambda$  of the matrix  $\mathbf{A}'$ , with elements  $A'_{ikjl}$ , or rather its logarithm,  $\log \lambda$ , gives the mutant invasion fitness. For the case where the mutant is equal to the resident, we have  $\lambda = 1$ , with  $(\hat{p}_{11}, \hat{p}_{21}, \hat{p}_{12}, \hat{p}_{22})$  as right eigenvector and the reproductive values  $(v_{11}, v_{21}, v_{12}, v_{22})$  as left eigenvector. Furthermore, the mutant can invade if  $\lambda > 1$ . We developed a C++ program that follows a path of small steps through either  $\zeta_1\zeta_2$ -space, or  $\alpha\beta$ -space, each of which increases the invasion fitness, until reaching an accurate approximation of the equilibrium. We first put  $\alpha = 1$  and  $\beta = 0$  and looked for an equilibrium dimorphism  $\zeta_1\zeta_2$ . We then retained this dimorphism and let  $\alpha$  and  $\beta$  evolve to an equilibrium, for different values of the rate of recombination  $\rho$  between the locus for  $\zeta_1\zeta_2$  and the loci for  $\alpha$  and  $\beta$ . In this analysis, we made the assumption that  $\alpha$  and  $\beta$  are tightly linked to each other. The result of the analysis is presented in Table A1. An important point of the analysis appears in the final column, giving the ratio  $v_{11}/v_{12}$  of the reproductive value for a small-effect modifier (in the limit of being neutral) of being associated with the locally favored allele  $\zeta_1$ to being associated with the other allele  $\zeta_2$ . This ratio expresses how much a small increase in survival in one habitat is weighed against a corresponding decrease in survival in the other habitat.

Table A1: Numerical analysis of the alternative model. It is similar to the simulation model explored in the main text, with results in figures 2 and 3. The main difference is that, in the alternative model, each habitat supports a single very large population, instead of several smaller local populations. Phenotype determination follows equation (1) with survival in each habitat given by equation (2) and environmental cues as in equation (3). The rate of migration between habitats is denoted  $m_{12}$  (with  $m_{21} = m_{12}$ ) and corresponds to m/2 in the model in the main text. The table shows the rate of between-habitat migration  $m_{12}$ , the rate of recombination  $\rho$ between the genetic effect locus and the loci for  $\alpha$  and  $\beta$ , the value  $\zeta_1$  of the allele adapted to habitat 1 at the genetic effect locus (with  $\zeta_2 = -\zeta_1$ ), the equilibrium values of the modifier  $\alpha$  and the slope  $\beta$ , the frequencies  $\hat{p}_{11}$  and  $\hat{p}_{12}$  in habitat 1 of the alleles  $\zeta_1$  and  $\zeta_2$  at the time of reproduction, and the reproductive values  $v_{11}$ and  $v_{12}$  of small-effect mutant modifiers, with recombination rate  $\rho$  the the genetic effect locus. The value  $v_{11}$  applies when the mutant modifier is linked to the locally adapted allele  $\zeta_1$  and  $v_{12}$  when linked to the alternative allele  $\zeta_2$ . The final column gives the ratio of the reproductive values, which indicates how strongly modifications that improve performance in habitat 1 are favored. Note that the situation is symmetric, with  $\hat{p}_{21} = \hat{p}_{12}$ ,  $\hat{p}_{22} = \hat{p}_{11}$ ,  $v_{21} = v_{12}$  and  $v_{22} = v_{11}$ . Other parameter values:  $s_0 = 0.1, \sigma = 1.0, \theta_1 = -0.75, \theta_2 = 0.75, \sigma_{juv} = 0.5.$ 

$m_{12}$	ρ	$\zeta_1$	α	β	$\hat{p}_{11}$	$\hat{p}_{12}$	$v_{11}$	$v_{12}$	$v_{11}/v_{12}$
0.005	0.001	-0.377	0.978	0.016	0.991	0.009	1.009	0.022	46.461
0.005	0.10	-0.377	0.973	0.020	0.991	0.009	1.007	0.281	3.588
0.005	0.50	-0.377	0.966	0.026	0.991	0.009	1.003	0.655	1.531
0.01	0.001	-0.380	0.955	0.032	0.981	0.019	1.018	0.041	24.822
0.01	0.10	-0.380	0.945	0.041	0.981	0.019	1.014	0.301	3.362
0.01	0.50	-0.380	0.930	0.054	0.981	0.019	1.006	0.673	1.495
0.03	0.001	-0.388	0.856	0.112	0.937	0.063	1.058	0.132	8.008
0.03	0.10	-0.388	0.817	0.145	0.933	0.067	1.043	0.402	2.596
0.03	0.50	-0.388	0.759	0.195	0.926	0.074	1.019	0.757	1.347
0.05	0.001	-0.395	0.735	0.216	0.880	0.120	1.101	0.255	4.316
0.05	0.10	-0.395	0.642	0.297	0.860	0.140	1.074	0.547	1.961
0.05	0.50	-0.395	0.448	0.459	0.800	0.200	1.028	0.888	1.157
0.055	0.001	-0.396	0.699	0.247	0.863	0.137	1.112	0.293	3.792
0.055	0.10	-0.396	0.579	0.351	0.833	0.167	1.081	0.600	1.802
0.055	0.50	-0.396	0.000	0.677	0.500	0.500	1.000	1.000	1.000



Figure A1: Numerical analysis of the alternative model. The panels show the modifier  $\alpha$  (solid lines) and slope  $\beta$  (dashed lines) for different values of the parameters  $m_{12} = m_{21}$  (orange and green lines),  $\sigma_{juv}$ , and  $s_0$ , as a function of the recombination rate  $\rho$  between the genetic effect locus and the loci for  $\alpha$  and  $\beta$ . Other parameter values:  $\sigma = 1.0, \theta_1 = -0.75, \theta_2 = 0.75$ .



Figure A2: Phenotype determination for linked and unlinked genetic architectures, as a function of the rate of migration, similar to figure 2. The difference from figure 2 is that here the alleles at the locus for z can evolve, such that the value of each allele lies in the interval [-0.5, 0.5]. Except for migration rates for which  $\alpha$  evolved to be close to zero, the allelic vales at the locus for z evolved to be close to the lower and upper limits. As a result, the distributions of values for z are trimorphic, with peaks at -1.0 (both alleles near -0.5), at 0.0 (one allele near -0.5 and the other near 0.5) and at 1.0 (both alleles near 0.5). Other parameter values are the same as in figure 2 and panels A and B show the same quantities as in that figure.



Figure A3: Examples of the distribution of recombination rates over individuals in the population, from three evolutionary simulations. Panels A and B show  $\rho_{z\alpha}$ and  $\rho_{\alpha\beta}$  for a simulation with m = 0.12, B and C show the same for a simulation with m = 0.18, and E and F show the same for a simulation with m = 0.24. The examples are randomly chosen among the simulations for figure 7.