

Evolutionary diversification of ecological specialists under informed resource choice

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Abstract

Behavior can be an important modulator of eco-evolutionary dynamics and genetic evolution that is not always taken into account in models of evolutionary diversification. On the one hand, classical models for the evolution of ecological specialization have been extended to account for flexible behavioral aspects such as diet choice or matching habitat choice, but only in a subset of all spatial settings relevant for adaptive speciation. On the other hand, models of adaptive speciation have been synthesized to encompass different types of spatial settings — either promoting diversification through within-habitat frequency-dependent forces arising from competition, or between-habitat local adaptation independent of competition — albeit with non-flexible behavior. Here, we study individual-based simulations of the emergence of multiple resource specialist strategies, when individuals are allowed to choose their resource based on profitability in a two-resource, two-habitat environment, inspired from previous models of adaptive diversification. We join other authors in finding that active resource choice favors resource conservatism and the maintenance of a single specialist when individuals are already somewhat specialized on one resource. When individuals start off as generalists, however, active resource choice maintains the convergence stability of the generalist branching point — meaning that selection leads to the diversification of two specialists — in the face of strong ecological trade-offs, where evolution towards a single specialist would have otherwise been favored without resource choice. Active resource choice also turns frequency-independent processes of diversification (relying on local adaptation) into frequency-dependent ones (relying on competition) when resources are spatially restricted. Consequently, the chances of ecological divergence in spatially heterogeneous scenarios are maximized at intermediate levels of resource choice accuracy, where frequency-dependent selection balances out selection for resource conservatism. Finally, we find that active resource choice promotes divergence when resources are compartmentalized within microhabitats, by avoiding the waste of energy that foraging in poor microhabitats represents.

Keywords — ecological specialization, optimal foraging, frequency-dependent selection, speciation, resource compartmentalization, matching habitat choice

36 Introduction

37 The adaptive diversification of ecologically specialized species is critical for the emergence and
38 maintenance of biodiversity on Earth (Hutchinson, 1959; Tilman, 1982; Schluter, 2000). Many
39 of the species found in the world’s most diverse ecosystems are ecological specialists, i.e., species
40 having evolved adaptations to efficiently utilize specific ecological niches in their environment
41 (as opposed to generalists, which can use a broader panel of niches but exploit any single one
42 somewhat less efficiently, Rosenzweig, 1981). Examples of particularly specialized species in-
43 clude hummingbirds feeding on the nectar of flowers (Tinoco et al., 2017), phytophagous insects
44 feeding on specific host plants (Del Campo et al., 2003; Lewinsohn & Roslin, 2008), or galling
45 crabs living on particular species of corals in shallow reefs (Bravo et al., 2024). The prevalence
46 of ecological specialization in nature makes unraveling the mechanisms underlying the diversifi-
47 cation of species into ecological specialists a key step for a thorough understanding the origins
48 of biodiversity (Schluter, 2000).

49
50 Under adaptive speciation theory, the divergence of species or morphs specializing on distinct
51 ecological niches is predicted by factors such as habitat heterogeneity in the type of available
52 resources (Nosil, 2012; Rettelbach et al., 2013), and/or frequency-dependent selection arising
53 from interspecific competition for limited resources (Dieckmann & Doebeli, 1999; Dieckmann
54 et al., 2004; Rueffler et al., 2006). In divergent selection, the diversification of specialist ecotypes
55 is driven by distinct habitats having different resources present, thus favoring local adaptation
56 to the most abundant resource in the environment (Nosil, 2012; Rettelbach et al., 2013). In
57 frequency-dependent selection, intraspecific competition for resources creates an advantage to
58 adapt towards under-utilized resources, which may result in a split of multiple specialized eco-
59 types within the population (Dieckmann & Doebeli, 1999; Kondrashov & Kondrashov, 1999;
60 Rueffler et al., 2006; Weissing et al., 2011). Key to these selective forces leading to the diver-
61 sification of specialist ecotypes is the existence of physiological or utilization trade-offs among
62 resources or niches, such that no biologically attainable combination of traits can fully adapt a
63 species to all niches — adaptation to one niche must come at a cost in terms of adaptation to
64 another niche (Levins, 1962; Egas et al., 2004; Rueffler et al., 2006).

65
66 As an evolved property of organisms, behavior can be instrumental in achieving a high fit-
67 ness through the active choice of resources (Stephens & Krebs, 1986), whether those resources
68 are food (Parent et al., 2014), habitats (Morris, 2003; Ronce, 2007; Edelaar et al., 2017), mates
69 (Andersson, 1994; Reynolds & Fitzpatrick, 2007) or breeding sites (Turelli et al., 1984). The
70 ability to modulate foraging efforts or habitat exploitation through behavior has been docu-
71 mented not just in animals, but also in bacteria (Moreno-Gómez, 2022), fungi (Fukasawa &
72 Ishii, 2023) and even plants (Kelly, 1992), and has been recognized as an important factor
73 affecting the dynamics of selection, possibly directing genetic evolution (Rueffler et al., 2007;
74 Edelaar et al., 2008; Edelaar & Bolnick, 2012; Bolnick & Otto, 2013). For example, while eco-
75 logical generalists are predicted to evolve under weak physiological trade-offs when behavior is
76 random (i.e. organisms cannot modulate how they experience the environment, Levins, 1962;
77 Rueffler et al., 2006), this is no longer true when optimal foraging or matching habitat choice

78 is considered. Instead, adaptive choice behavior tends to favor ecological specialists and not
79 generalists whenever individuals can direct their efforts towards the resource or niche they are
80 good at exploiting, which in turn modifies the selection pressures to promote specialization even
81 more (Rueffler et al., 2007; Ravigné et al., 2009).

82

83 Most of previous work on eco-evolutionary dynamics in the face of active choice behavior
84 has focused on exploring rules for the coexistence of specialists and generalists in the context of
85 multiple resources in a single habitat (building on optimal foraging theory, e.g. Stenseth, 1984;
86 Matsuda and Namba, 1989; Abrams, 1999; Svanbäck and Bolnick, 2005; Rueffler et al., 2007),
87 or in the context of multiple habitats containing one resource each (habitat selection models,
88 e.g. Brown, 1990, 1996; Ravigné et al., 2004, 2009). The few models combining both contexts
89 focus on establishing the conditions for the ecological coexistence of specialist species once they
90 already exist as separate ecotypes, but do not focus on the mechanisms of adaptive diversifica-
91 tion resulting in the emergence of such ecotypes (Vincent et al., 1996; Abrams, 2006). Some
92 models of adaptive speciation do combine both within-habitat frequency-dependent selection
93 and between-habitat divergent adaptation in driving the emergence of distinct ecotypes (e.g.
94 Rettelbach et al., 2013), but they have not yet been extended to account for flexible behavior.

95

96 In this study, we use individual-based simulations to ask how the propensity of resource
97 choice affects the likelihood of evolutionary divergence of ecological specialists. We developed a
98 model with two resources and two habitats, allowing us to explore the continuum from within-
99 habitat, competition-driven diversification to between-habitat diversification driven by adapta-
100 tion to local resources (similar to Rettelbach et al., 2013). Our model imposes a certain level
101 of adaptive decision making upon resource encounter, mediated by the accuracy with which
102 the profitability of each resource is assessed by consumer organisms. In turn, the profitability
103 of a resource depends on some underlying ecological trait of the individual, and is subject to
104 a utilization trade-off between resources (similar to Rueffler et al., 2007). We investigate the
105 interaction between modes of diversification and resource choice ability by varying choice accu-
106 racy and resource partitioning among habitats. We study the generality of our findings across
107 strengths of the utilization trade-off between resources and rates of resource discoverability
108 (which is akin to search time during foraging).

109 Methods

110 The model

111 We consider a population of individuals living across two separate habitat patches connected
112 by dispersal (Fig. 1A). In each habitat, two resources are available for individuals to feed on.
113 The concentration R_{ij} of resource i in habitat j is given by the matrix

$$\mathbf{R} = \{R_{ij}\} = \begin{pmatrix} 1 & h \\ h & 1 \end{pmatrix} \quad (1)$$

114 where h is the habitat symmetry parameter ($0 \leq h \leq 1$). When $h = 1$, both resources are
115 equally distributed between habitats. As h decreases, resources become increasingly restricted
116 to a single habitat, until, when $h = 0$, resource 1 is only found in habitat 1 and resource 2 in
117 habitat 2 (Fig. 1A).

118

119 A population of consumers dwells in this landscape. Each individual has a certain quan-
120 titative trait x , which indirectly determines its affinity for each resource through a utilization
121 trade-off, such that a value of x increasing the affinity for one resource reduces the affinity for
122 the other, and vice versa. For any given resource i , this affinity, or consumption rate, is given
123 by

$$C_i(x) = \exp\left(-s(x - x_i^{\text{opt}})^2\right) \quad (2)$$

124 where x_i^{opt} is the trait value needed for maximum consumption of resource i ($x_i^{\text{opt}} = -1$ for
125 resource 1 and $+1$ for resource 2), and s is the trade-off strength parameter ($s \geq 0$), controlling
126 the widths of the two Gaussian curves (Fig. 1B). The higher the value of s , the narrower the
127 curves and the stronger the trade-off. Because they promote ecological specialization, utilization
128 trade-offs between resources or ecological niches are critical components of adaptive speciation
129 models (Levins, 1962; Rueffler et al., 2006).

130

131 The population consists of N individuals. The size of the population is fixed, and gener-
132 ations are discrete and non-overlapping, meaning that at each generation, all adults die and
133 are replaced by N newborns, which will be the adults of the next generation. Reproduction
134 is asexual, and the parents of the newborns are sampled from a weighted lottery where, for
135 each offspring, the probability of an individual to be sampled as the parent of that offspring is
136 proportional to the amount of resource accumulated by that potential parent during its lifetime
137 (which we thus assume to be a measure of its reproductive success).

138

139 Resource accumulation occurs during a feeding season, within which individuals must choose
140 one resource to feed on, and can only utilize that resource. To make this choice, individuals
141 evaluate their expected gains from choosing one resource over the other. They then choose the
142 resource that is most profitable to them with a certain probability (see below). The perceived
143 gains of an individual on a particular resource depend on the affinity of that individual for

144 that resource (Eq. 2) and how much of that resource is available in the local habitat. This,
 145 in turn, depends on how many individuals have already chosen that same resource, and on the
 146 consumption rates of these individuals. The perceived profitability of a resource is the following
 147 share of the available resource, proportional to the focal individual’s consumption rate relative
 148 to others,

$$C_i(x) / \sum_{k \in S_{ij}} C_i(x_k), \quad (3)$$

149 where the sum is over all individuals k from the set S_{ij} , containing all individuals living in
 150 habitat j and having chosen resource i thus far.

151

152 The amount of resource shared among individuals is not necessarily all of the resource that is
 153 present in the habitat. If this were the case, it could result in cases where a few migrants arriving
 154 in a new, previously unoccupied habitat, would automatically enjoy a large fitness benefit from
 155 feeding on some locally abundant resource, simply because of a lack of local competition and
 156 despite possibly having low consumption rates for that resource. Such situations are reminiscent
 157 of models of *soft selection*, in which the spread of alleles depends more on local, relative fitness
 158 than on global, absolute fitness (Levene, 1953; Ravigné et al., 2004; Reznick, 2016). This may of
 159 course occur in nature, but in this study we wanted to be able to modulate the degree to which
 160 individuals are also impacted by their absolute fitness — a hallmark of *hard selection* (Dempster,
 161 1955; Ravigné et al., 2004) — irrespective of any local relief from competition. Notably, models
 162 of hard selection are much less prone to the coexistence of multiple genotypes than models of
 163 soft selection, and this distinction was shown to be important for the evolution and coexistence
 164 of ecological specialists and generalists (Ravigné et al., 2004, 2009). To fine-tune the behavior
 165 of our model along this continuum, we introduce a new parameter, the resource discovery rate
 166 δ , which controls how much of the total resource in a habitat is available to the population of
 167 consumers. This amount of resource discovered is given, for resource i in habitat j , by

$$R_{ij}^D = R_{ij} \left(1 - \exp \left(- \delta \sum_{k \in S_{ij}} C(x_k) \right) \right), \quad (4)$$

168 which is a saturating function of the cumulative consumption rate $\sum_{k \in S_{ij}} C(x_k)$ of all consumers
 169 having chosen that resource, asymptotically reaching R_{ij} (Fig. 1C). The resource discovery rate
 170 δ is therefore the initial slope of this saturating curve ($\delta > 0$), and could be interpreted as the
 171 approximate amount of resource i that would be found by a single hypothetical individual, alone
 172 in the habitat and with consumption rate $C_i(x) = 1$ (the maximum achievable consumption
 173 rate).

174

175 During a feeding season, individuals are taken in random order, and make their decision
 176 based on their perceived share (Eq. 3) of the resources discovered so far (Eq. 4) by all individ-
 177 uals earlier in the queue. To limit the influence of the order in which individuals are choosing,
 178 each generation consists of n_R feeding rounds, or seasons, where a new random order is sampled
 179 every time.

180

181 Once the perceived profitability has been calculated for each resource for a given individual
 182 in the feeding queue, the individual chooses one of the two resources to exploit. At this point,
 183 the resource with the higher perceived profitability may be considered a better option for the
 184 individual (but see below). The individual then picks that better resource with probability

$$P_{\text{best}} = (1 - \beta) P_0^{\text{best}} + \beta \quad (5)$$

185 where β is the resource choice accuracy, determining the weight of the perceived profitability in
 186 the decision-making process of the individual ($0 \leq \beta \leq 1$). If $\beta = 1$, individuals always choose
 187 the most advantageous resource. If $\beta = 0$, profitability is irrelevant, and resource choice then
 188 depends on the probability of encountering the better resource, P_0^{best} . This probability is in
 189 turn given by

$$P_0^{\text{best}} = 1/2 (1 - \alpha) + \alpha R_{\text{best}}/R_{\text{tot}} \quad (6)$$

190 where R_{best} is the total concentration of the better resource in the local habitat (R_{1j} or R_{2j}),
 191 R_{tot} is the sum of both total resource concentrations ($R_{1j} + R_{2j}$), and α is the resource abun-
 192 dance weight parameter, determining how much the probability of encounter of a resource is
 193 determined by its abundance ($0 \leq \alpha \leq 1$). When $\alpha = 1$, the probability of encounter of a
 194 resource is equal to its abundance relative to the other resource, $R_{\text{best}}/R_{\text{tot}}$ (Fig. 1D). When
 195 $\alpha = 0$, this probability is $1/2$, irrespective of resource abundances. Note that the probability of
 196 encounter of the best resource is also $1/2$ whenever $h = 1$, regardless of α , because in that case
 197 the resource concentrations in each habitat are equal. Hence, parameter α really only makes a
 198 difference when $h < 1$ (i.e. $R_{1j} \neq R_{2j}$). Scenarios where $\alpha \neq 1$ and $R_{1j} \neq R_{2j}$ could correspond,
 199 for example, to situations where resources can be found in two particular species of trees, both
 200 occupying equal surfaces in the habitat, but with different yields of usable resource per unit
 201 area. Note that although parameter β controls the probability of accurately choosing the better
 202 of the two local resources given their known profitability, for all intents and purposes this is
 203 equivalent to the accuracy with which individuals perceive this profitability (i.e. in this study
 204 choice accuracy is indistinguishable from assessment accuracy).

205

206 Actual resource gains are not distributed until all individuals have chosen which resource to
 207 utilize. This means that the actual gains individuals receive may not be the same as the gains
 208 they expected upon choosing a resource based on perceived profitability. Once all individuals
 209 have chosen (i.e. at the end of a feeding season), the final amounts of resource discovered R_{ij}^D
 210 are computed (as per Eq. 4) and individuals receive their share, proportional to their con-
 211 sumption rate relative to others (as per Eq. 3). The next feeding season then starts, where
 212 individual choice occurs in a new random order. The gains from all the feeding seasons are
 213 accumulated to constitute a total amount of resource obtained across seasons. This amount of
 214 accumulated resources then serves as the weight in the weighted lottery of asexual reproduction.

215

216 During the reproduction step, each newborn individual is an almost identical clone of its

217 parent, albeit with a potential mutation applied to it. Mutations occur at rate μ (the per capita
 218 mutation rate) and affect the value of trait x . When a mutation occurs, a deviation is sampled
 219 from a normal distribution with standard deviation σ_x (the mutational standard deviation) and
 220 applied to the trait value of the newborn. Newborns at first inherit their parent's habitat, but
 221 after mutation has occurred, can disperse to the alternative habitat with probability m (the per
 222 capita migration rate).

223 Analysis

224 We tracked the mean trait value in the population through time across simulations, as well as
 225 measures of ecological isolation and spatial isolation between nascent ecotypes (i.e. clusters of
 226 individuals above versus below the mean trait value of the whole population at a given time).
 227 We refer to the clear separation and divergence of both ecotypes along the ecological trait axis
 228 x as *evolutionary branching*, which is diagnosed based on the value of the ecological isolation
 229 statistic.

230

231 The degree of ecological isolation is measured by

$$\text{EI}_0 = 1 - \frac{N_1 V_1 + N_2 V_2}{N V} \quad (7)$$

232 where N_i is the number of individuals in ecotype i , V_i is the variance in trait value x within
 233 ecotype i , N is the total population size and V is the total variance in x across the entire
 234 population. This statistic ranges between 0 (full overlap) and 1 (full clustering). However, due
 235 to the impossibility of both ecotypes to overlap in trait value (the mean trait value being a
 236 hard cut-off between the two), EI_0 rarely goes below $2/\pi \approx 0.64$ (which can be shown to be
 237 the value the statistic should take in a perfectly normally distributed population split in two
 238 halves), which renders interpretation difficult. To correct for that, we rescaled EI_0 into

$$\text{EI} = (\text{EI}_0 - 2/\pi)/(1 - 2/\pi), \quad (8)$$

239 which technically ranges between $1/(1 - \pi/2) \approx -1.75$ and 1, but will typically be around 0 for
 240 a unimodal, normally distributed sample of trait values.

241

242 The degree of spatial isolation between the two ecotypes is computed as

$$\text{SI} = \left| \frac{n_{11} n_{22} - n_{12} n_{21}}{\sqrt{n_{1.} n_{2.} n_{.1} n_{.2}}} \right| \quad (9)$$

243 where n_{ij} is the number of individuals from ecotype i in habitat j , $n_{i.}$ is the total number of
 244 individuals in ecotype i (across both habitats) and $n_{.j}$ is the total number of individuals in
 245 habitat j (across both ecotypes). This statistic quantifies the restriction of each ecotype to
 246 one habitat, and ranges between 0 (for complete habitat overlap of the ecotypes) and 1 (for
 247 complete spatial separation).

248 **Adaptive dynamics** In parallel, interpretation of the simulations was aided with numer-
249 ical analyses based on adaptive dynamics theory (Metz et al., 1992; Geritz et al., 1998), a
250 body of mathematical tools used to predict the effect of selection on eco-evolutionary dynamics
251 (see Appendix). We used these to verify our simulations when compared to dynamics derived
252 independently from the simplest case of no resource choice ($\beta = 0$).

253 **Simulations**

254 Simulations were run across various combinations of parameters for a total of T_{\max} generations,
255 with each simulation starting with N identical individuals with initial trait value x_0 . Unless
256 specified otherwise in the figures, parameters and their values are as per Table 1.

257 **Specifications**

258 The simulation code in this study was written in the programming language C++20 using
259 standard libraries. Analyses were performed both within the simulations as well as externally,
260 in the R computing environment, version 4.3.3 (R Core Team, 2025). See accompanying code
261 for details.

Results

Symmetric habitats under random choice

We first focus on the case of high habitat symmetry ($h = 1$) where both resources are in equal concentrations in each habitat. Until further notice, we also focus on $\alpha = 1$, in which resources are encountered proportionately to their local abundance (we relax this assumption at the end of this section). Figure 2 shows an example outcome of one simulation in such a case, with intermediate resource choice accuracy β . Analysis of the adaptive dynamics of a deterministic approximation of the model shows general agreement with our stochastic individual-based simulations in the simple case of absence of active resource choice ($\beta = 0$, Fig. S2).

When resource choice is fully random ($\beta = 0$), the strength s of the utilization trade-off between resources determines whether evolutionary branching takes place (Fig. 3, S3). If the trade-off is very weak, a population of generalists evolves, capable of utilizing both resources equally well (i.e. with trait values close to $x = 0$, Fig. 3, S3A). This happens when $s < 0.5$, as in that case the average consumption rate over both resources is maximized at a generalist strategy when each resource is encountered half of the time (Fig. S3C). If the trade-off is very strong (and how strong is too strong depends on the resource discovery rate δ , see Fig. 3), the penalty for deviating from the original phenotype is so high that the population remains specialized on the one resource it is already well adapted to (here, resource 1, when the population starts at $x_0 = -0.9$, Fig. 3, S3A). Branching occurs when the trade-off is strong enough that the total consumption of a specialist (e.g. $x \approx \pm 1$) exceeds that of a generalist ($s > 0.5$), but weak enough that generalists first enjoy a frequency-dependent advantage from utilizing the under-exploited alternative resource (e.g. resource 2), in a population mostly consisting of specialists of one resource (e.g. resource 1). A generalist phenotype then replaces the initial specialist, but loses its advantage once it becomes common, as specialists are now favored that could exploit each resource more fully. Since a specialist of a single resource would restore the asymmetric depletion of the resources that favored generalists in the first place, the only stable outcome is the emergence of two alternative specialists, each depleting its own respective resource — evolutionary branching has occurred ($s = 1$ in Fig. S3A). Because both resources are present in equal quantities in both habitats, the newly split divergent ecotypes diversify in sympatry and spatially overlap (Fig. 3C, S3B). These results are in agreement with the predicted adaptive dynamics of the model in the absence of active resource choice ($\beta = 0$, Fig. S2A).

Resource choice in symmetric habitats

The first notable effect of a nonzero resource choice accuracy β is to prevent the evolution of a generalist phenotype under weak trade-offs ($s < 0.5$), instead promoting branching into two specialists (Fig. 3, S4). As individuals become better at picking the most advantageous resource, the same frequency-dependent advantage as described previously pushes individuals slightly less well adapted to the initially preferred resource 1 to actively choose the alternative resource 2 once the former becomes depleted (Fig. S4). Because the trade-off is relatively weak,

304 individuals choosing the alternative resource for which they are rather maladapted enjoy a simi-
305 lar fitness as those choosing the depleted resource (Fig. S4B). In turn, this splits the population
306 into two groups, with individuals from one end of the phenotypic spectrum consistently exposed
307 to their originally preferred resource, and individuals from the other end consistently driven to
308 the alternative resource by competition. Biased exposure to different resources at both ends of
309 the phenotypic spectrum leads to the adaptive divergence of two alternative specialist strategies,
310 this time without a generalist phase (Fig. S4A); branching occurs.

311

312 The second clear effect of resource choice is a reduction in the maximum trade-off strength
313 s still allowing branching (Fig. 3, S5). That is, under high trade-offs which would have nor-
314 mally favored frequency-dependent branching under random choice ($\beta = 0$), increasing choice
315 accuracy β prevents branching, and increases the probability that the population will remain
316 as a single resource specialist instead. When the trade-off is strong, the most advantageous
317 resource will often be the resource the initial population of specialists is already adapted to
318 (here, resource 1), because deviating from that specialized phenotype (and adapting to resource
319 2) comes at a heavy cost (Fig. S5). For branching to occur, this cost must be offset by the
320 frequency-dependent advantage gained from avoiding competition for the depleted resource, and
321 this advantage must be strong enough to compensate for the fact that higher choice accuracy
322 β reduces exposure to the under-exploited resource (i.e. the trade-off must be weaker). The
323 same mechanism of active, behavioral bias in exposure operates that favors two specialists over
324 one generalist under weak trade-offs (as presented in the previous section), except that under
325 strong trade-offs, the frequency-dependent advantage to utilize the under-exploited resource is
326 no longer sufficient for adaptation to that resource to occur.

327

328 *Asymmetric habitats under random choice*

329

330 The maximum trade-off strength still allowing branching slightly goes down as habitat asym-
331 metry increases ($h < 1$) when resource choice is random ($\beta = 0$, Fig. 3). In asymmetric habitats
332 (low h), one resource is rare and the other is common in any given habitat (e.g. with $h = 0.1$
333 resource 1 is encountered $10/(10 + 100) \approx 9\%$ of the time in habitat 2 and the same is true
334 for resource 2 in habitat 1). Therefore, specialist individuals migrating into a yet unoccupied
335 habitat find themselves with their preferred resource (the one they are already well adapted
336 to) being the rare one. If the preferred resource is encountered too rarely, and the alternative,
337 maladaptive abundant resource is utilized most of the time, the fitness of foraging individuals
338 is no longer sufficient to spread and establish a viable population in the new habitat (Fig. S6).
339 Instead, the population remains mostly restricted to its ancestral habitat, with few migrants
340 arriving at each generation in the alternative habitat (here at rate $m = 0.01$, see Methods and
341 Table 1), but rarely leaving offspring for the next generation (owing to a low fitness after all
342 feeding rounds, Fig. S6D). Branching no longer occurs where more symmetric habitats (higher
343 h) would have promoted it, and the strength s of the trade-off must be lowered to increase
344 the fitness of specialists of one resource utilizing the other resource, and allow branching again.
345 Once the trade-off is appropriately reduced, branching can occur in asymmetric habitats, and

346 involves both a high degree of spatial isolation between diverging ecotypes, as well as rapid eco-
347 logical divergence, without an intermediate generalist phase (Fig. S7). Then, the rare encounter
348 of one resource in each habitat selects against generalists and for specialists. We note that the
349 frequency-dependent advantage of utilizing the under-exploited resource disappears in highly
350 asymmetric habitats, as adaptation to a new resource occurs when its utilization becomes more
351 advantageous than the alternative, much-too-rare ancestral resource, regardless of its state of
352 depletion. In asymmetric habitats, branching is driven more by frequency-independent local
353 adaptation than by frequency-dependent disruptive selection arising from competition.

354

355 *Resource choice in moderately asymmetric habitats*

356

357 In moderately asymmetric habitats where the rare resource is still discoverable by a non-
358 negligible minority of individuals (i.e. here, when $h = 0.1$ and $\delta = 0.04$), increasing the resource
359 choice accuracy β restores the frequency-dependent advantage of generalists, leading to branch-
360 ing under stronger trade-offs, but only up to a point ($\beta \approx 0.25$ in Fig. 3), beyond which
361 the maximum trade-off allowing branching decreases again (Fig. 3, S8). This is because as
362 β increases, individuals pick the preferred but rare resource more often after migrating to a
363 new habitat (when the rare resource is rare but not too rare, picking that resource pays off,
364 but non-choosy individuals, i.e. $\beta = 0$, would typically not encounter it). That rare resource,
365 in turn, is sufficiently rare that a few accurate choosers are enough to deplete it, now giving
366 a frequency-dependent fitness advantage to individuals choosing the alternative resource and
367 favoring adaptation to it — branching occurs. When choice accuracy β becomes high (beyond
368 the apparent threshold of $\beta \approx 0.25$), however, the same phenomenon happens as in symmetric
369 habitats ($h = 1$), where individuals choose the preferred resource (the one they are already
370 well adapted to) so often, and exposure to the alternative resource is so reduced, that a higher
371 consumption efficiency on the latter is no longer beneficial (Fig. S8). As a result, the breadth
372 of trade-off values allowing branching peaks at intermediate levels of resource choice accuracy β .

373

374 This effect of resource choice on frequency-dependent disruptive selection in asymmetric
375 habitats is further exemplified when the migration rate m is reduced. As the rate of migration
376 decreases, the pattern described above becomes more pronounced, and branching occurs under
377 stronger trade-offs at intermediate choice accuracies β (except this time the peak is located
378 at $\beta \approx 0.3$, Fig. S9, S10). Under reduced migration, two things happen (Fig. S10). First,
379 competition intensifies in the ancestral, highly populated habitat, as fewer individuals leave
380 every generation, thus generally reducing fitness and lowering the representation of this habitat
381 in the reproductive pool compared to the new habitat (which experiences less competition),
382 and therefore amplifying the frequency-dependent advantage of adaptation to the alternative
383 resource in the new habitat. This intensification of competition due to less emigration is neg-
384 ligible in the new habitat, which is too poorly populated. In contrast, and second, reduced
385 (im)migration decreases competition for the rare resource in the new habitat, thereby increas-
386 ing the fitness of individuals utilizing it. With increased fitness from choosing the ancestral
387 resource, the depletion of that resource triggers the frequency-dependent advantage of utilizing

388 the alternative, more common resource. Through these two processes, reduced migration ampli-
389 fies the strength of frequency-dependent selection driving branching, when selection takes place
390 in asymmetric resource availability conditions (i.e. when habitat asymmetry is moderate and
391 the rarity of the rare resource is not prohibitively high). Importantly, the migration rate does
392 not affect other model outcomes across parameter space (Fig. S9), strengthening the idea that
393 the observed effect of resource choice accuracy on branching in asymmetric habitats is indeed
394 mediated through its role as modulator of frequency-dependent selection.

395

396 The phenomenon of branching being most favored at intermediate resource choice accuracy
397 also relies on the rare resource not being completely unavailable. If that resource is rare, but
398 can still be discovered a non-negligible proportion of the time (e.g. $h = 0.1$ and $\delta = 0.4$), then
399 the preferred resource is depleted even at low choice accuracy, and branching occurs under a
400 similar range of trade-offs as in symmetric habitats (Fig. 3). If the rare resource is too rare (e.g.
401 $h = 0.01$), or moderately rare but too difficult to discover (e.g. $h = 0.1$ and $\delta = 0.004$), it is
402 never advantageous enough to be depleted, in which case there is no rise in maximum trade-off
403 allowing branching, because frequency-dependent selection never happens (Fig. 3, see below).

404

405 *Highly asymmetric habitats*

406

407 All aforementioned effects of resource choice accuracy disappear in highly asymmetric habi-
408 tats (e.g. $h = 0.01$), or at least asymmetric habitats where the rare resource is particularly
409 difficult to find (i.e. a low resource discovery rate δ is also needed, e.g. $h = 0.1$ but $\delta = 0.004$, or
410 $h = 0.01$ as long as $\delta < 0.4$). In those conditions, the maximum trade-off still allowing branching
411 is more-or-less constant with respect to resource choice (Fig. 3, S11). This is because the rare
412 resource has become so rare or unavailable, that a higher choice accuracy no longer contributes
413 to depleting it more (the previously mentioned phenomenon by which individuals choosing the
414 alternative abundant resource become more advantaged). As a consequence, whether evolu-
415 tionary branching occurs almost exclusively depends on the strength of the trade-off s in highly
416 asymmetric habitats. If the trade-off is very strong, the new habitat is a demographic sink for
417 all choice strategies: non-choosy individuals (low β) utilize the abundant resource but are too
418 maladapted to it, and choosy individuals (high β) find it more advantageous to pick the rare
419 resource, but it still yields so little, that the establishment of a viable local population (and
420 therefore adaptation) is not possible either way. Once the trade-off becomes sufficiently weak,
421 the abundant resource becomes advantageous over the rare one, and it ends up being picked
422 by choosy individuals (high β , by active choice) as well as by non-choosy individuals (low β ,
423 by passive exposure). Whether this results in local adaptation and (frequency-independent)
424 branching then depends on whether that abundant resource yields a high-enough fitness to es-
425 tablish a viable local population, and this, again, depends on the strength of the trade-off which
426 affects all resource choice strategies β in the same way (Fig. S11).

427

428 That said, there seems to be a very slight increase in the maximum trade-off allowing branch-
429 ing at high choice accuracy when resource discovery is very low and habitat symmetry is low

430 but not too low ($h = 0.1$ and $\delta = 0.004$, Fig. 3, S12, S13). This is because as the trade-
431 off becomes weak enough that it is advantageous for choosier individuals (higher β) to pick
432 the more abundant resource, therefore utilizing resources in similar proportions to non-choosy
433 individuals (which predominantly choose the abundant resource due to passive exposure), still
434 some non-choosy individuals pick the rare resource a higher proportion of the time than choosier
435 individuals. That proportion is dictated by the habitat symmetry parameter h (e.g. passive
436 choosers still choose the rare resource 9% of the time when $h = 0.1$, 0.9% when $h = 0.01$, etc.).
437 In contrast, in such conditions, the exposure of choosy individuals to the abundant resource
438 tends to be slightly higher (as it is motivated by active choice), thus increasing the advan-
439 tage to adapt to that resource, and allowing this adaptation to happen up to slightly stronger
440 trade-offs. Because the increase in exposure from active choice is even more slight when habitat
441 symmetry is lower (e.g. it can only be from 99.1% to 100% when $h = 0.01$), this weak pattern
442 is more visible in more moderately asymmetric habitats (e.g. $h = 0.1$ instead of $h = 0.01$, Fig.
443 S13). This phenomenon only applies when resource discovery is low (here, $\delta = 0.004$) because
444 as resource discovery increases, the rare resource can now be depleted enough that frequency
445 dependence kicks in and overrides this weak pattern (Fig. 3).

446

447 Finally, highly asymmetric habitats strongly select against generalists. Again, because of
448 biased exposure to a locally abundant resource, the frequency-dependent advantage of general-
449 ists disappears, and divergent specialists evolve locally adapted to the more abundant resource
450 in each habitat, even at very weak trade-offs as long as they are nonzero (when $s = 0$ genetic
451 drift takes over and no divergence occurs between the habitats due to substantial migration,
452 $m = 0.01$, Fig. 3, S14). The same phenomenon is separately predicted by the adaptive dynam-
453 ics of the model without active choice ($\beta = 0$, Fig. S2B). Note that generalists may still be
454 favored in asymmetric habitats if individuals could disperse between habitats multiple times in
455 their lifetime, but we did not consider this option here. Resource choice accuracy β does not
456 affect this pattern, as choosier individuals (high β) get exposed to the locally abundant resource
457 just the same, albeit by choice (Fig. S14).

458

459 *Effect of the resource discovery rate*

460

461 Other than through interactions with the parameters explored above, the main effect of the
462 resource discovery rate δ is to increase the range of trade-offs suitable for branching by allowing
463 branching at stronger trade-offs (Fig. 3). When resource discovery increases, individuals can
464 access more of the resource contained in a habitat, which tends to (1) increase the fitness of indi-
465 viduals specializing on a rare resource in asymmetric habitats (Fig. S15), and (2) contribute to
466 resource depletion, and therefore frequency-dependent disruptive selection, in more symmetric
467 habitats (thereby also making moderately asymmetric habitats more equivalent to symmetric
468 habitats with respect to branching dynamics, and making resource choice accuracy β matter
469 more in those cases, e.g. $h = 0.1$ in Fig. 3). Notably, the resource discovery rate has no effect
470 on the minimum trade-off strength s needed for branching, as this depends entirely on the par-
471 titioned exposure to the two resources brought about by habitat asymmetry (low h) or active

472 choice (high β), in conditions where generalists would otherwise be favored if resources were
473 encountered equally often (under weak trade-offs, the discovery rate is not limiting the amount
474 of resource utilized). The expansion of the range of trade-offs suitable for branching with higher
475 resource discovery rate is separately predicted by the deterministic adaptive dynamics of the
476 model under random choice ($\beta = 0$, Fig. S2).

477

478 *Effect of resource compartmentalization*

479

480 We studied a version of our model where resource encounter is independent of resource abun-
481 dance, by setting the resource abundance weight parameter α to zero (see Methods). When
482 $\alpha = 0$, non-choosy individuals ($\beta = 0$) encounter each resource half of the time, even if one is
483 rare (i.e. regardless of h). This could correspond to a situation where resources are nested in
484 compartments within the habitat (i.e. microhabitats) — e.g. fruits within host plants, preferred
485 heights within trees, or host species of corals within coral reefs — with the constraint that the
486 two microhabitats occupy the same area in the habitat. This may not be an accurate descrip-
487 tion of all ecosystems, but this extension of the model at least allows to explore the roles of
488 stratification and compartmentalization of resources in the environment. The implementation
489 is such that choosy individuals (higher β) are able to direct their foraging efforts more towards
490 the microhabitat that yields the highest payoff (see Methods for details).

491

492 When resource encounter and abundance are fully decoupled, the microhabitat that hosts
493 the rare resource in asymmetric habitats (low h) has a particularly low yield, but is still ex-
494 plored by non-choosy individuals. Exposure to the abundant resource in a habitat is therefore
495 reduced, increasing the profitability that this resource must have in order to be advantageous to
496 specialize on (i.e. to compensate for the cost of deviating from a phenotype specialized on the
497 rare resource). Hence, compartmentalization of the resources through the decoupling between
498 encounter and abundance decreases the maximum trade-off strength still allowing branching
499 under random resource choice ($\beta = 0$, Fig. S16, S17, also visible in analyses of the adaptive
500 dynamics of the model under random choice, Fig. S2C). Increasing the resource choice accuracy
501 β makes the resource abundance weight α irrelevant, as highly choosy individuals (e.g. $\beta = 1$)
502 direct their efforts purely based on expected payoff and target the most advantageous resource
503 regardless of the rate of passive encounter. Therefore, the foraging and branching dynamics
504 of $\beta = 1$ when $\alpha = 0$ are identical to when $\alpha = 1$ (all of the aforementioned Results). This
505 produces patterns similar to the case without compartmentalization in the more symmetric
506 habitats (high h): $\alpha = 0$ yields the same foraging dynamics as $\alpha = 1$ in fully symmetric habi-
507 tats ($h = 1$), because then no resource is more rare than the other, irrespective of microhabitats.
508 In asymmetric habitats (low h), however, this means that branching will occur under stronger
509 trade-offs s at higher choice accuracies β , owing to choosier individuals wasting less energy for-
510 aging in microhabitats where the yield is low, and generally increasing fitness and the potential
511 for local adaptation (Fig. S16).

512

513 *Generalist starting point*

514

515 When the population starts as a generalist ($x_0 = 0$), branching occurs up to much stronger
 516 trade-offs than when starting as a specialist (Fig. S18). This occurs for both symmetric
 517 and asymmetric habitats, albeit because of different reasons. In asymmetric habitats (low
 518 h), branching occurs up to stronger trade-offs s for a generalist, because a generalist population
 519 is exactly at the right starting point to readily respond to the two opposite regimes of direc-
 520 tional selection operating within each habitat (i.e. selection towards specializing to the most
 521 locally abundant resource) — it does not need to de-specialize, unlike a specialist population
 522 which must overcome the cost of deviating from its well-adapted initial phenotype in order to
 523 successfully colonize the alternative habitat (Fig. S19). This is independent of resource choice
 524 strategy, and therefore, the maximum trade-off still suitable for branching remains relatively
 525 constant with respect to resource choice accuracy β (Fig. S18). The upper limit in trade-off
 526 strength still allowing branching is the point where directional evolution in the starting habi-
 527 tat occurs so fast relative to the colonization of the alternative habitat (owing to differences
 528 in habitat-specific population size) that the population only specializes on the resource most
 529 abundant in its environment (this is visible when the resource discovery rate is sufficiently low,
 530 e.g. $\delta = 0.004$ in Fig. S18, which reinforces the penalty suffered by migrants once the popu-
 531 lation starts to specialize). We note that under strong trade-offs, it is likely that a generalist
 532 population will go extinct before any adaptation can occur, as the population is not sufficiently
 533 adapted to either resource to start with (being a generalist), but this cannot happen in our
 534 model where the population size is fixed (here $N = 1000$).

535

536 In symmetric habitats (high h), the upper limit to the strength of the trade-off still allow-
 537 ing branching increases with resource choice accuracy β (Fig. S18). This is because active
 538 resource choice from a generalist reinforces the frequency-dependent processes responsible for
 539 branching in sympatry. In symmetric habitats, evolutionary branching under random choice
 540 ($\beta = 0$) is typically preceded by the evolution of a transient generalist phenotype (e.g. Fig.
 541 S3), because that phenotype is a branching point: an attractor of the adaptive dynamics which
 542 becomes unstable once reached (this can be visualized by performing an invasion analysis, see
 543 Fig. S2) — starting at the branching point therefore maximizes the chances of branching. As
 544 the strength s of the trade-off increases, specialists become increasingly fitter than generalists,
 545 and may tend to remain specialists, but this depends on the exact starting trait value. Indeed,
 546 phenotypes that are sufficiently close to the branching point ($x = 0$) may still fall within its
 547 basin of attraction, instead of evolving towards one of the two specialist equilibrium phenotypes
 548 ($x = \pm 1$). By exploring evolutionary endpoints of simulations across starting phenotypic values
 549 x_0 , we find that the basin of attraction of the generalist branching point shrinks faster with
 550 increasing trade-off strength s under active resource choice than under random choice (Fig. 5),
 551 but that the branching point itself remains a branching point up to higher trade-off strengths,
 552 thus explaining the more widespread branching observed at higher resource choice accuracy β .
 553 Upon closer inspection, we find that this robustness of the branching point to stronger trade-
 554 offs stems from active resource choice keeping the profitability of the two resources in check
 555 with respect to one another — every time a resource is used slightly more, the alternative re-

556 source automatically becomes more advantageous to generalists, and the next individual in the
557 queue chooses it (Fig. S20), thus creating frequency dependence that maintains the generalist
558 phenotype as an evolutionary attractor. This breaks down when s reaches very high values,
559 at which point the basin of attraction of the generalist equilibrium shrinks to zero and this
560 equilibrium becomes a repeller, i.e. the junction between the basins of attractions of the two
561 remaining stable equilibria (the specialist phenotypes $x = \pm 1$). Then, an initial generalist typ-
562 ically falls on one side or the other (depending on the stochasticity of drift and mutations) and
563 undergoes purely directional selection towards specialization on a single resource (e.g. Fig. S20).

564

565 We note that the lower end of trade-off values s suitable for branching are not affected by
566 the starting phenotype (Fig. S18). In asymmetric habitats, weak nonzero trade-offs always
567 favor ecological divergence regardless of resource choice (see previous sections). In symmetric
568 habitats, weak trade-offs select for a single generalist as stable endpoint of evolution (Fig. S4),
569 and so starting there does not affect these dynamics. Besides, and for the same reasons as
570 mentioned before, the resource discovery rate δ expands the range of trade-off values suitable
571 for branching in generalist starting populations as well (Fig. S18), by reducing the penalty
572 suffered by less-well-adapted individuals when the population starts to deviate from the central
573 phenotype $x = 0$.

574 Discussion

575 In this study we explored the eco-evolutionary dynamics of an individual-based model of adap-
576 tive diversification under various degrees of resource choice behavior. We find the following. (1)
577 Choosiness promotes ecological specialization and resource conservatism (i.e. prevents adapta-
578 tion to other resources present in the environment), selecting against generalists as long as the
579 population does not start off as a generalist. (2) This resource conservatism reduces the scope
580 for within-habitat frequency-dependent diversification (in the form of evolutionary branching)
581 when the population starts as a specialist. (3) While higher choosiness favors conservatism in
582 specialists, the generalist strategy remains a branching point for stronger trade-offs for those
583 strategies starting sufficiently close to it. (4) The effect of resource choice is negligible when
584 the resources are highly spatially partitioned. (5) The scope for branching is maximized at
585 intermediate choosiness when resources are moderately partitioned spatially.

586
587 Previous models of diet and habitat choice have identified the phenomenon whereby in-
588 creased choosiness tends to select for specialists and against generalists (Rueffler et al., 2007;
589 Ravnigné et al., 2009). Ravnigné et al. (2009) highlighted that this “raises the bar” for evolution-
590 ary explanations of ecological generalism, which must then include factors such as fluctuating
591 environments (Kisdi, 2002; Egas et al., 2004), selection for dispersal (Kisdi, 2002; Ronce, 2007;
592 Edelaar et al., 2017) or suboptimal decision making due to incomplete information (Egas et al.,
593 2004; Rueffler et al., 2007). This stands in stark contrast with models lacking diet or habitat
594 choice, which predict that generalists are favored when the shape of the utilization trade-off
595 curve between resources is convex (equivalent to our trade-off strength parameter s being below
596 the threshold value of $s = 0.5$; Levins, 1962; Rueffler et al., 2006). By biasing exposure to
597 any preferred resource, choosiness increases how profitable the alternative resource must be for
598 there to be a frequency-dependent advantage to utilize it. Selection favors adaptation to the
599 resources that are already disproportionately used (as already noted in Rueffler et al., 2007).
600 In turn, this directs evolution towards niche conservatism, and hinders adaptive diversification
601 of specialists in sympatry, for a species that is already somewhat specialized on a given resource.

602
603 For an initial generalist, however, frequency-dependent diversification within a habitat can
604 occur up to stronger trade-offs under active (i.e. more accurate) resource choice than under
605 random choice. With active choice, a population of generalists will keep the profitability of two
606 initially equally abundant resources in check with respect to each other, by systematically al-
607 ternating foraging efforts towards the slightly less depleted resource. This makes the generalist
608 phenotype a convergent attractor of the evolutionary dynamics despite strong ecological trade-
609 offs, maintaining the population under disruptive selection where random choice would have
610 pushed the system towards specialization on a single resource as soon as a stochastic utilization
611 bias emerges. Rueffler et al. (2007) found that diet choice reduces the chances to approach
612 a branching point when not starting as a generalist, but did not comment on the robustness
613 of the convergence stability of the generalist branching point under strong trade-offs. To our
614 knowledge, this phenomenon has not been described before, and suggests that active choice
615 may actually facilitate adaptive diversification when the initial population is an ecological gen-

616 eralist. We note, however, that this relies on the resources being in similar abundance and
617 availability in the environment: if one resource becomes utilized in sufficiently greater propor-
618 tion, active choice will promote specialization for that resource. Hence, this phenomenon may
619 be more readily observed (if at all) in natural systems where resources are relatively abundant
620 and easy to find. On the one hand, this could be the case for *pursuer* species of foragers (sensu
621 MacArthur and Pianka, 1966), which are not limited by resource scarcity and instead invest
622 in optimizing handling time (MacArthur & Pianka, 1966; Norberg, 1977; Stephens & Krebs,
623 1986; Norberg, 2021). Hummingbirds and honeyeaters are usually cited as belonging to that
624 category (Norberg, 2021). On the other hand, *searcher* species (limited by long search times)
625 have been proposed to be more often generalists than pursuer species (MacArthur & Pianka,
626 1966; Brown, 1990). Alternatively, this phenomenon could also apply to species whose resource
627 encounter is high enough that they can afford “laziness” most of the time, such as short-tailed
628 shrews *Blarina brevicauda* (Martinsen, 1969; Herbers, 1981), or sit-and-wait predators such as
629 *Anolis* lizards (Andrews, 1971).

630

631 The effect of resource choice on diversification dynamics disappears as resources become
632 highly spatially isolated. Hence, resource choice is predicted to play little role in colonization-
633 driven, frequency-independent speciation. This, however, is not true for moderately spatially
634 partitioned resources. Rather, we show that active resource choice brings frequency-dependent
635 selection back into spatially heterogeneous habitats where it would have been negligible under
636 random choice. Intermediate resource partitioning may be a hallmark of common settings for
637 ecological speciation, such as ecological gradients (Endler, 1977; Doebeli & Dieckmann, 2003;
638 Nosil, 2012), parapatric biogeography (Gavrilets et al., 2000; Mallet, 2005) and hybrid zones
639 (Endler, 1977; Barton & Hewitt, 1989; Abbott, 2017). In fact, Doebeli and Dieckmann (2003)
640 found that a spatial ecological gradient in resource availability must be steep, but not too steep,
641 to maximize disruptive frequency-dependent selection and speciation in a parapatric setting.
642 Our results suggest that active resource choice may be another way to reinforce this frequency-
643 dependent process of competition in spatially heterogeneous landscapes.

644

645 Rettelbach et al. (2013) showed that disruptive frequency-dependent selection driven by com-
646 petition within habitats could interact with between-habitat, frequency-independent divergent
647 selection to produce a third type of scenario beyond the more *competitive* (within-habitat and
648 through frequency-dependent selection) and the more *ecological* (between-habitat and through
649 local adaptation) modes of adaptive speciation (Dieckmann et al., 2004; Nosil, 2012; Rettelbach
650 et al., 2013). In their model, assortative mating and sexual reproduction are key to this third
651 mode of speciation. Although our results do not necessarily qualify as a new mode of specia-
652 tion, here we show that frequency-dependent and spatially divergent selection also interact in
653 an asexual model, independent of assortative mating, when resource choice behavior is taken
654 into account. Evidence for both frequency-dependent disruptive selection and divergent local
655 adaptation acting in concert has been documented in relation to speciation in three-spine stick-
656 lebacks *Gasterosteus aculeatus* (Schluter, 2003; Bolnick, 2004; Hendry et al., 2009; Bolnick &
657 Stutz, 2017).

659 Under moderate spatial partitioning of the resources, the scope for evolutionary branching
660 of an initial specialist is maximized at intermediate resource choice accuracy (i.e. choosiness)
661 — enough to promote resource depletion, but not so high as to promote resource conservatism.
662 We note that in our model the capacity for resource choice does not evolve, i.e. the assessment
663 accuracy is fixed (similar to matching habitat choice in Ravigné et al., 2004). Models of resource
664 choice have traditionally analyzed the evolution or coexistence of ecological morphs once opti-
665 mal behavioral preference has been achieved, that is, assuming that behavior has evolved, and
666 done so rapidly (e.g. Stenseth, 1984; Brown, 1990; Vincent et al., 1996; borrowing from optimal
667 foraging theory, Stephens and Krebs, 1986). In contrast, it may be argued that neural mecha-
668 nisms that are critical for optimal decision making during foraging evolve slowly, compared to
669 ecological and/or morphological adaptations related to niche use (Pyke, 1984; Bernays, 1998;
670 Railsback, 2022). Suboptimal foraging and incomplete information about the environment may
671 be the rule more than the exception, due to the various biological functions trading-off with
672 energetically demanding cognitive abilities (Mayhew, 1997; Morris, 2003). Hence, we consider
673 it important to not only look at purely adaptive behavior, but to explore the continuum of pos-
674 sibilities between random choice and fully accurate decision making (e.g. as in Rueffler et al.,
675 2007). Doing so allowed us, for example, to find that the scope for evolutionary branching is
676 maximized at intermediate resource choice accuracy in moderately asymmetric habitats, which
677 may be highly relevant in systems where resource choice behavior is not optimal.

678

679 We also find that resource choice may matter more for diversification when resources are
680 spatially heterogeneous but compartmentalized in microhabitats within habitats (e.g. fruits
681 within trees within a locality, if this is the scale at which resource exploitation is relevant,
682 Morris, 1987). We consider the case where microhabitats are in similar propensity but of dif-
683 ferent profitability in different localities. When this type of resource structure is considered,
684 more adaptive resource choice allows to direct efforts towards the most profitable microhabi-
685 tat, which facilitates between-habitat divergence through local adaptation. This highlights the
686 negative effect that resource compartmentalization has on adaptive divergence across habitats
687 in the absence of active choice, as it causes non-choosy organisms to waste energy by foraging
688 in unprofitable microhabitats, a phenomenon connected to incomplete information about the
689 environment and documented, for example, in phytophagous insects (Mayhew, 1997).

690

691 Previous studies have highlighted the role of the scale of competition and density regulation
692 in habitat selection models (reviewed in Ravigné et al., 2009). Under soft selection, density is
693 regulated within habitats and the number of individuals produced in each habitat, i.e. habitat
694 output, is constant (Levene, 1953). Under hard selection, density regulation via competition is
695 global, and habitat output is variable (Dempster, 1955). Protected polymorphism (i.e. diversity
696 of coexisting morphs) is typically only allowed in population genetics models of soft selection,
697 as in hard selection models the fittest genotype overtakes the entire population (Levene, 1953;
698 Kisdi, 2001; Ravigné et al., 2004). In a hybrid model with local density regulation but variable
699 habitat output (which would normally behave like a hard selection model in the absence of habi-

700 tat choice), Ravigné et al. (2004) showed that matching habitat choice makes selection softer
701 by expanding the impact of local density regulation on the maintenance of a polymorphism.
702 Considering habitat choice analogous to resource choice (as in many classical models of match-
703 ing habitat choice, the choice affects which habitat to migrate into based on the profitability of
704 the local resources, reviewed in Ravigné et al., 2009), we find the opposite pattern (and so do
705 Rueffler et al., 2007). This is because the model by Ravigné et al. (2004) is a population genet-
706 ics model with haploid genetics, where extreme phenotypes are within a single mutational step
707 from each other. Instead, a continuously varying (possibly polygenic) specialization trait leads
708 to a bistable system where both extremes are stable and may well coexist (Rueffler et al., 2007),
709 but directional selection will typically direct within-habitat trait evolution one way or the other
710 in the absence of frequency-dependent selection (as in Rueffler et al., 2007; Ravigné et al., 2009).

711

712 While our model has a fixed consumer population size and technically does not follow the
713 usual classification of soft versus hard selection models, habitat output is variable and our model
714 is technically closer to a hard selection one. Nevertheless, we can tune the dynamics between
715 softer-selection outcomes and harder-selection outcomes with our resource discovery parame-
716 ter δ . With high δ , foragers share all of the resource in a habitat (proportionately to their
717 utilization efficiency), allowing poor consumers of a local resource to nevertheless enjoy a high
718 fitness in the absence of competition; migrants can easily establish and evolve to specialize on
719 the local resource (this is essentially soft selection favoring protected polymorphism). With low
720 δ , the amount of resource obtained is constrained by utilization efficiency, and a poor consumer
721 of a local resource will typically not establish, regardless of the absence of competition (this is
722 hard selection). Consistent with that, higher resource discovery generally broaden the range
723 of trade-off strengths suitable for evolutionary branching when resources are asymmetrically
724 distributed, and we propose that such an approach could be a useful implementation linking
725 the two types of density regulation through a continuum (as proposed e.g. by Débarre and
726 Gandon, 2011), rather than having to consider only the two ends of the spectrum.

727

728 Like any model, ours makes several simplifying assumptions which may have affected our
729 results. First, we did not consider continuous resource and consumer population dynamics,
730 or any specific kind of functional response of the consumer (e.g. Holling’s disc equation as in
731 Stenseth, 1984; Brown, 1990, 1996; Vincent et al., 1996; Rueffler et al., 2007). This is because
732 our model is not based on a demographic model. However, the resources can still be depleted
733 and give rise to frequency-dependent selection, and the resource discovery rate allows us to ap-
734 proximate different types of density regulation. Different patterns may emerge under different
735 kinds of population dynamical aspects such as predator-prey fluctuations (Matsuda & Namba,
736 1989; Abrams, 1999; Abrams & Matsuda, 2003, 2004; Abrams, 2006) and/or more realistic
737 sensory and movement strategies (e.g. Netz et al., 2022). Second, we did not consider separate
738 trade-offs for various additional aspects of the foraging process such as search time or handling
739 time (unlike models based on optimal foraging theory, e.g. Vincent et al., 1996; Rueffler et al.,
740 2007). However, Rueffler et al. (2007) found that a suitable trade-off in one of these dimensions
741 usually leads to diversification even if selection does not promote diversification in others, and

742 so this may have little influence on the validity of our conclusions. Third, we focused on the
743 diversification rather than on the coexistence of different morphs, the latter sometimes being
744 permitted under broader conditions. For example, while diversification only produced distinct
745 specialists with exclusive diets in the model by Rueffler et al. (2007), evolutionarily stable co-
746 existence was allowed between specialists and generalists, and between specialists and some
747 intermediate types (neither fully specialists nor generalists). Broader conditions for coexistence
748 than diversification are also supported by Ravigné et al. (2009) and models with more morphs
749 (Brown, 1996; Egas et al., 2004). Such coexistence without in situ diversification may arise,
750 for example, from migration or large mutational steps (but see Rueffler et al., 2007 for poly-
751 morphisms arising without branching points and yet through small mutational steps). Finally,
752 we studied an asexual model without the possibility for assortative mating to evolve, although
753 assortative mating is a key ingredient for speciation (Weissing et al., 2011). Extensions of our
754 study that allow sexual reproduction should shed light on the reproductive aspect of adaptive
755 diversification under behavioral resource choice. We expect resource choice and mate choice
756 to reinforce each other at least in some conditions if, for example, resource choice biases the
757 encounter of potential mates assortatively (Servedio et al., 2011).

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763 **Author Contributions**

764 RS conceived and designed the study, wrote final simulation code, produced and analyzed
765 simulation data, and wrote the manuscript draft. JD wrote initial simulation code, gener-
766 ated preliminary data and results used for exploratory purposes, and provided feedback on the
767 manuscript. GSvD developed initial simulation code and conceptualized the model, provided
768 supervision and critical feedback on analyses and the manuscript. RSE provided supervision
769 and critical feedback on study design, model implementation, analyses and on the manuscript.

Table 1: Overview of model parameters and default values.

Symbol	Name	Default
h	Habitat symmetry	1
s	Trade-off strength	1
x_i^{opt}	Optimal trait value for resource i	± 1
δ	Resource discovery rate	0.04
β	Resource assessment accuracy	0
α	Resource abundance weight	1
m	Migration rate	0.01
μ	Mutation rate	0.01
σ_x	Mutational standard deviation	0.02
N	Population size	1 000
n_{R}	Number of rounds per feeding season	10
T_{max}	Simulation time	100 000
x_0	Starting trait value	-0.9

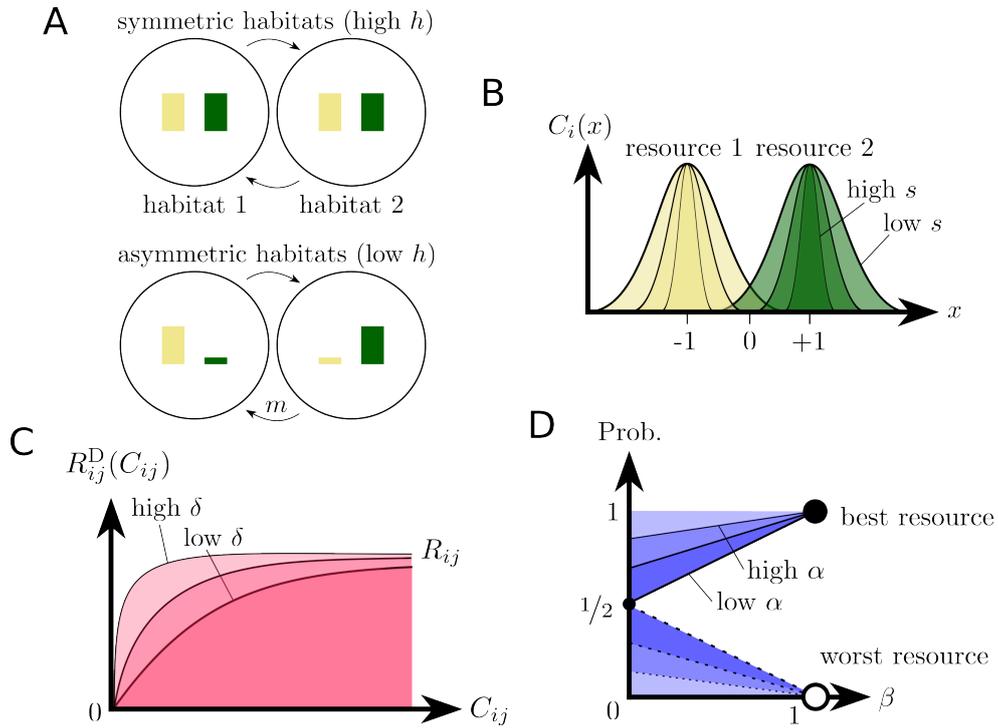


Figure 1: Model overview. (A) Symmetric versus asymmetric partitioning of the resources between habitats depending on the habitat symmetry parameter h . Migration occurs between habitats at per capita rate m per generation. (B) Consumption rates on either resource follow two Gaussian curves whose widths are controlled by the trade-off parameter s . (C) The amount of resource discovered by a population of feeders having chosen resource i in habitat j is a saturating function of their combined consumption rates, with initial slope parameter δ , the rate of resource discovery. (D) After payoffs have been estimated, the probability of choosing the best (and conversely, the worst) resource depends on the resource choice parameter β . The resource abundance weight α controls the baseline probability of choosing either resource in the absence of choice. This baseline probability approaches the relative proportion of each resource as α goes up (in this example the local concentration of the best resource is 1, while that of the worst resource is 0).

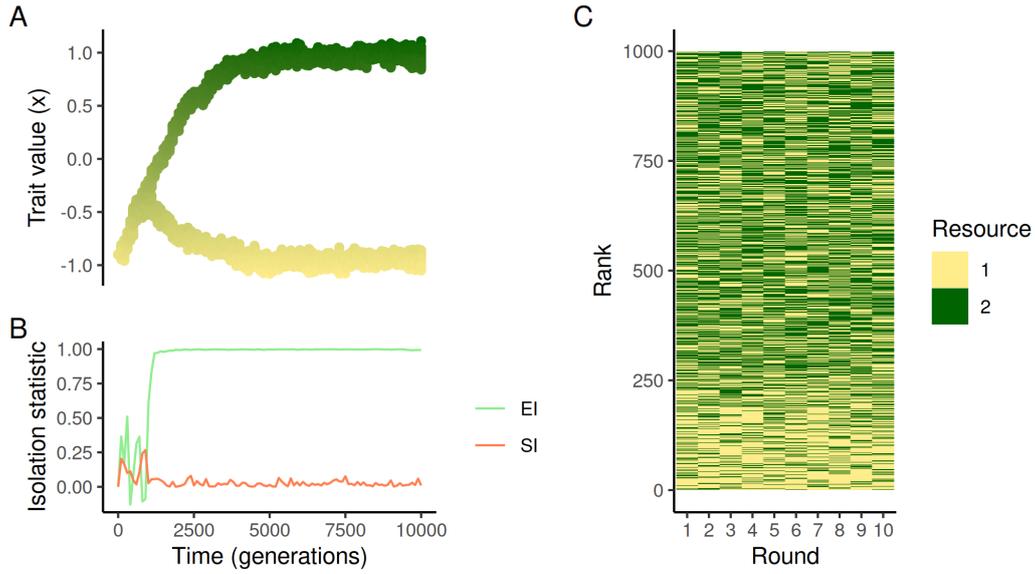


Figure 2: Example simulation in symmetric habitats ($h = 1$) with intermediate assessment accuracy ($\beta = 0.5$). (A) Individual trait values through time. (B) Isolation statistics through time. EI, ecological isolation; SI, spatial isolation (see Methods). Here, evolutionary branching occurs early and in relative spatial isolation due to the level of habitat asymmetry. (C) Overview of the resources chosen by each individual in each feeding round in generation 1000, when the population is still mostly adapted to resource 1. Due to active resource choice, resource 1 tends to be chosen early on in the feeding queue, until it is no longer more advantageous and resource 2 ends up being chosen just as often. Parameters: $h = 1$, $\beta = 0.5$, $s = 0.8$, $\delta = 0.04$, $m = 0.01$. Other parameters as per Table 1.

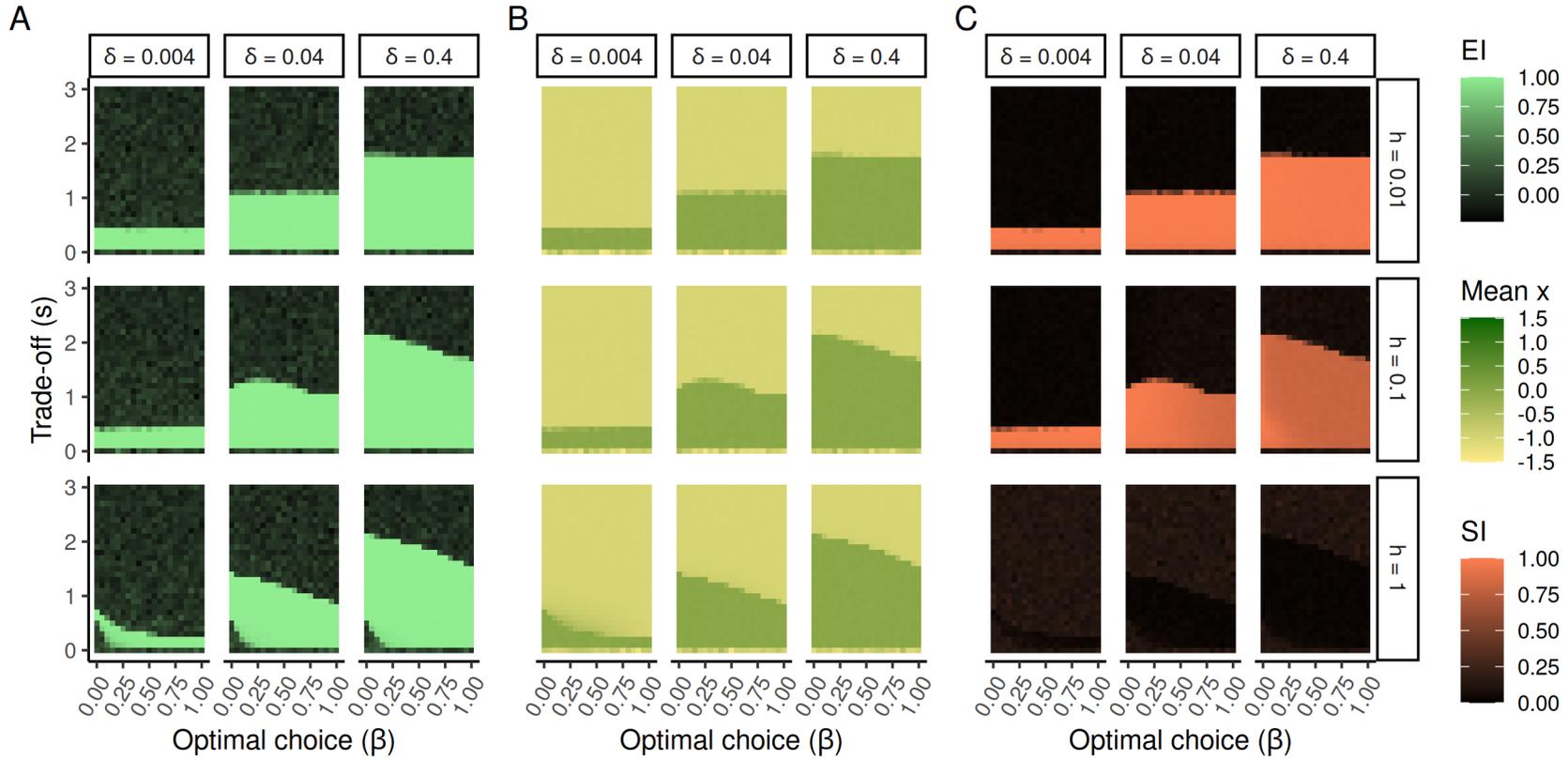


Figure 3: Outcomes of simulations across values of resource assessment accuracy β , trade-off strength s , resource discovery rate δ and habitat symmetry h . Each cell shows the mean final value after 100 000 generations of (A) the ecological isolation statistic EI, (B) the average trait value in the population (“mean x ”), or (C) the spatial isolation statistic SI (see Methods), across 10 replicate simulations per parameter combination. Other parameters as per the default values in Table 1. Regions of parameter space where mean $x \approx -1$ correspond to a single specialist outcome (of resource 1), regions where mean $x \approx 0$ and EI is low correspond a single generalist, and regions where mean $x \approx 0$ and EI is high correspond to two specialists (one of each resource, see Fig. 4 for examples).

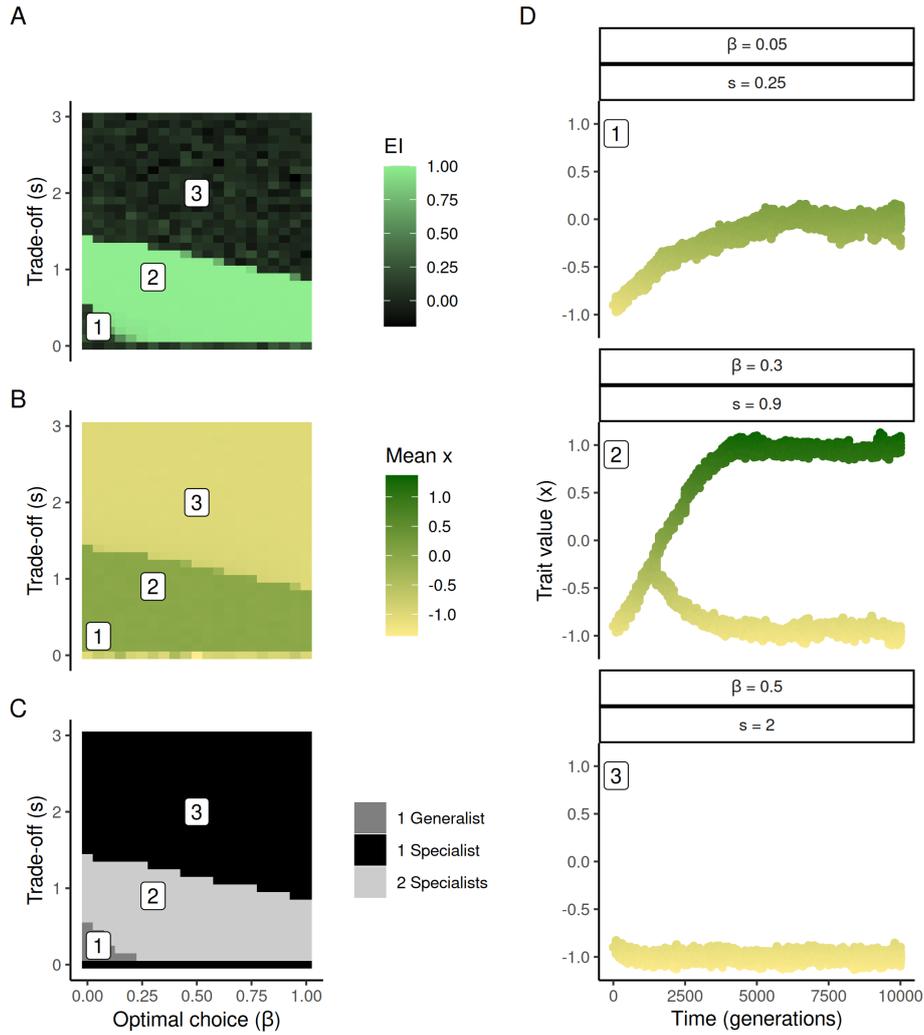


Figure 4: Types of outcomes. (A) Mean final ecological isolation (EI) across parameter space (restricted to the panel with $h = 1$ and $\delta = 0.04$ in Fig. 3A). (B) Same for the mean trait value at the end of the simulations (Fig. 3B). (C) Aggregate of A and B summarizing where the outcome is a single generalist (mean $x \approx 0$, low EI), a single specialist (mean $x \approx -1$, low EI) or two specialists (mean $x \approx 0$, high EI). (D) Example simulations illustrating these three outcomes. Numbered labels in A through C show where these simulations are located in parameter space.

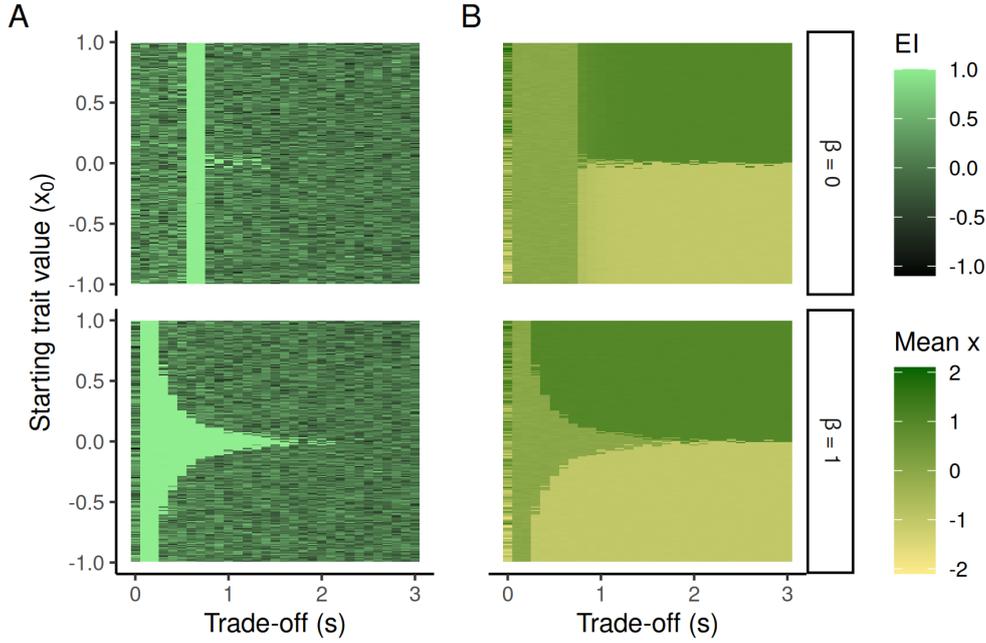


Figure 5: Basin of attraction of the generalist branching point ($x = 0$) across a transect of trade-off strengths s , at two extreme levels of resource assessment accuracy ($\beta = 0$ and $\beta = 1$). One simulation was run for each combination of parameter values. For each combination, we show (A) the value of the ecological isolation statistic EI and (B) the mean ecological trait value in the population, at the last generation of the simulation. The simulations resulting in evolutionary divergence (high EI) are the ones where the population is attracted to the branching point at the center of ecological trait space ($x = 0$). With increasing trade-off strength s the basin of attraction of this branching point becomes more narrow (and conversely those of the two specialist equilibrium phenotypes at $x = \pm 1$ expand). Note that with high choice accuracy ($\beta = 1$), the basin of attraction of the branching point starts to decay at lower trade-off strength s compared to random choice ($\beta = 0$), but the equilibrium remains a branching point up to higher values of s .

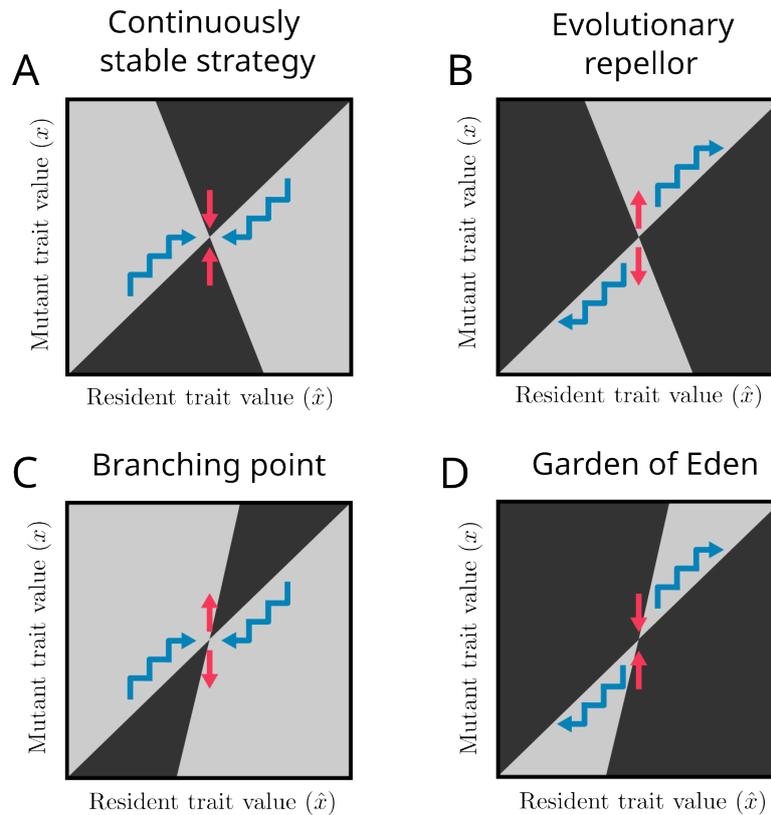


Figure S1: A pairwise invasibility plot (PIP) is a phase plot showing, for each possible value of a trait fixed in a theoretical (monomorphic) *resident* population (here \hat{x}), the range of other values of the same trait that a rare *mutant* arising in a population of residents could have, and what the relative fitness of said mutant (here x) would be, compared to the resident. This relative *invasion fitness* determines whether a mutant can invade, and replace, a given resident. A PIP shows, in two different colors, all pairs of mutant and resident strategies where the mutant can invade (light gray here), and all pairs where the mutant cannot (dark gray). The graphical depiction predicts the dynamics of evolution through successive invasions (of mutants becoming the new residents, and so on, blue arrows). Eventually, a so-called *equilibrium* (or *singular*) strategy may be reached, where the direction of evolution changes (i.e. where the *isoclines* delimiting the invasion boundaries cross). Singularities that evolution by selection leads to (blue arrows) are *convergence stable*, but need not be endpoints of the evolutionary dynamics, as once reached they may be *evolutionarily stable* or not (red arrows). (A) Equilibrium strategies that are both convergence and evolutionarily stable are called continuously stable strategies (CSS) — they are stable endpoints of evolution. (B) Repellors are equilibria which are both convergence and evolutionarily unstable — selection leads away from them. (C) Branching points are convergence-stable attractors that are evolutionarily unstable once reached — they promote diversification into two morphs, each with their own trait value. (D) Gardens of Eden are repellors that would be evolutionarily stable if reached but in practice never are. For more information, see Geritz et al. (1998) and Otto and Day (2007).

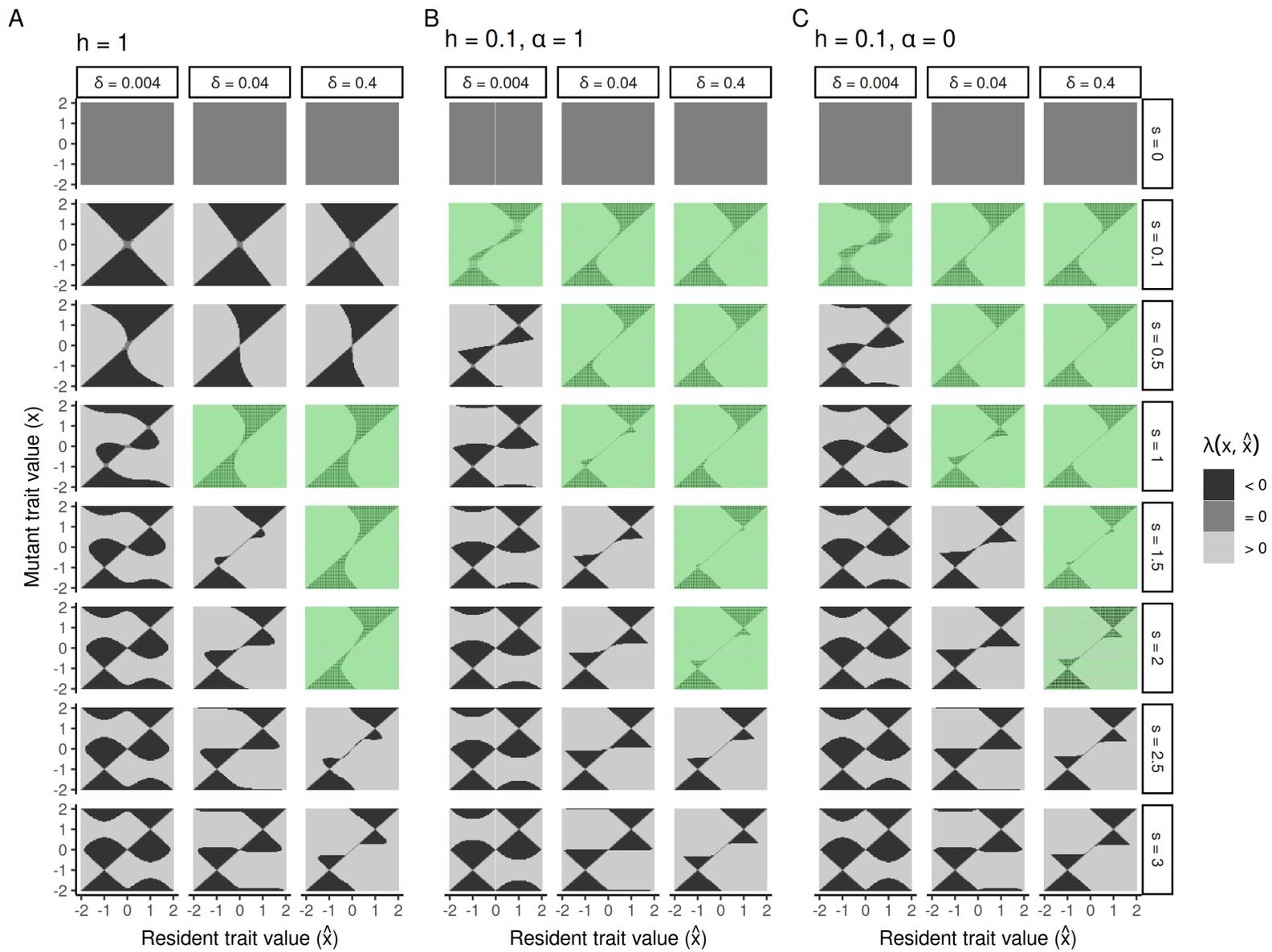


Figure S2: (See next page.)

Figure S2 (continued): Pairwise invasibility plots (PIP) across various parameter combinations explored in our simulations (see Fig. 3 for $h = 1$ and $\alpha = 1$, and Fig. S16 for $\alpha = 0$) when resource choice is random ($\beta = 0$). Note that α is irrelevant when $h = 1$ (see Methods). $\lambda(x, \hat{x})$, invasion fitness of a mutant with trait value x in a resident population with trait value \hat{x} — the mutant invades if $\lambda(x, \hat{x}) > 0$. Green overlays symbolize parameter combinations for which individual-based simulations resulted in successful evolutionary branching of ecological strategies (reaching at least $EI = 0.9$ in ecological isolation at the end of a simulation, on average across replicates in Fig. 3 and S16). These combinations largely match the parameter sets where the generalist strategy ($x = 0$) can be graphically identified as a branching point in the PIP. See Appendix for details and Figure S1 for how to interpret PIPs.

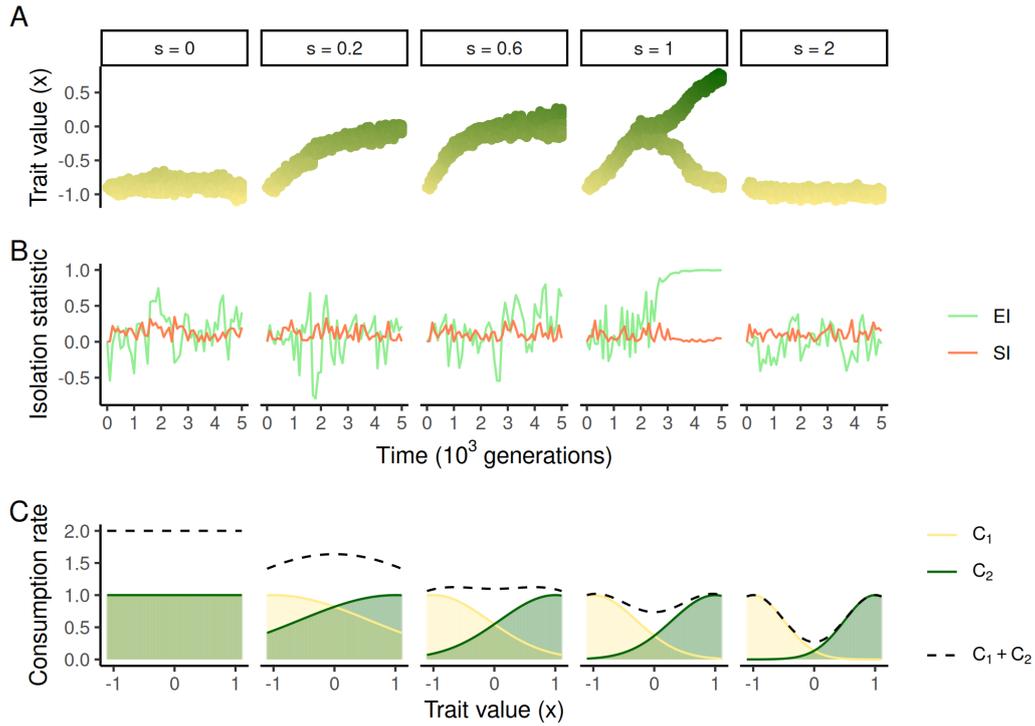


Figure S3: Simulations across trade-off strengths s in symmetric habitats ($h = 1$) and under random choice ($\beta = 0$). Other parameters as per Table 1. (A) Individual traits through time. Evolutionary branching occurs within a specific range of trade-off strengths — a generalist ($x \approx 0$) evolves if s is too low, or the population remains as a single specialist ($x \approx -1$) if s is too high (when $s = 0$ the trait drifts randomly). When branching occurs, it is driven by frequency-dependent selection (i.e. advantage to utilize the less depleted resource) and first goes through a generalist phase. (B) Isolation statistics through time, showing that divergence (whenever EI is high) occurs in sympatry (SI remains low) under these parameter values. EI, ecological isolation; SI, spatial isolation. (C) Consumption curves for both resources (full lines) and sum of both consumption rates (dashed lines). Generalists are favored when $s < 0.5$ as then the cumulative consumption curve peaks at $x = 0$.

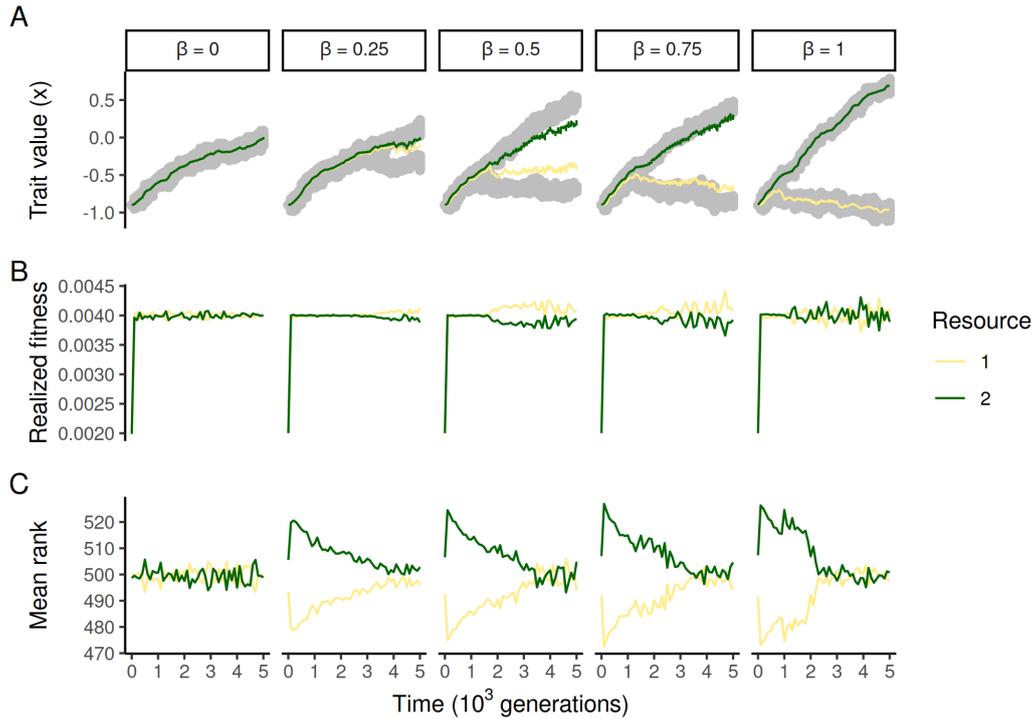


Figure S4: Simulations across resource assessment accuracies β in symmetric habitats ($h = 1$) when the trade-off is weak ($s = 0.2$). Other parameters as per Table 1. (A) Individual trait values through time. Colored lines represent the mean trait value of individuals choosing each resource at each generation. As choosiness β increases, individuals choosing different resources become more phenotypically segregated, and this can be seen already before branching. (B) Mean fitness gain from feeding on each resource at every generation. (Differences in fitness between the resources are negligible and due to imperfect choice in diverging specialists.) (C) Average rank in the queue of individuals choosing each resource, at each generation. At high β , and before branching, resource 2 tends to be chosen later in the queue, when resource 1 has been depleted — a pattern indicating frequency dependence.

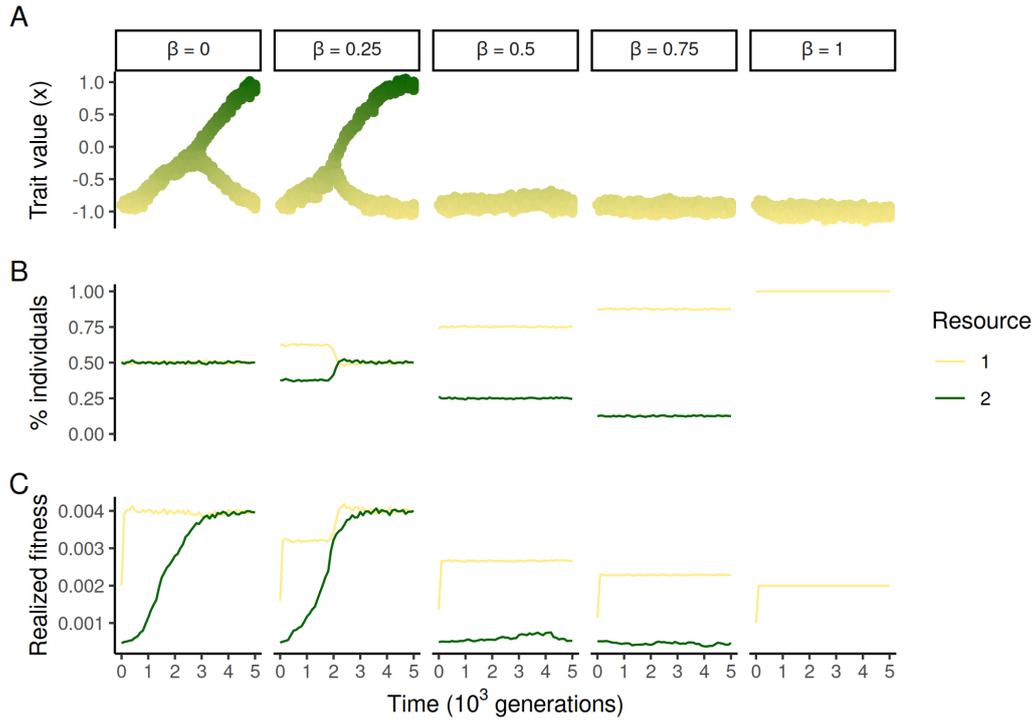


Figure S5: Simulations across resource assessment accuracies β in symmetric habitats ($h = 1$) under a strong trade-off ($s = 1.2$). Other parameters as per Table 1. (A) Individual trait values through time, showing that branching is lost as choosiness increases. (B) Proportion of individuals choosing each resource at each generation. With increased choosiness, more individuals choose resource 1 over resource 2. (C) Mean fitness gain from utilizing each resource at each generation, showing that under a strong trade-off, resource 1 is more advantageous than resource 2 despite being depleted. By biasing choice towards resource 1, this reduces exposure to resource 2 and prevents branching.

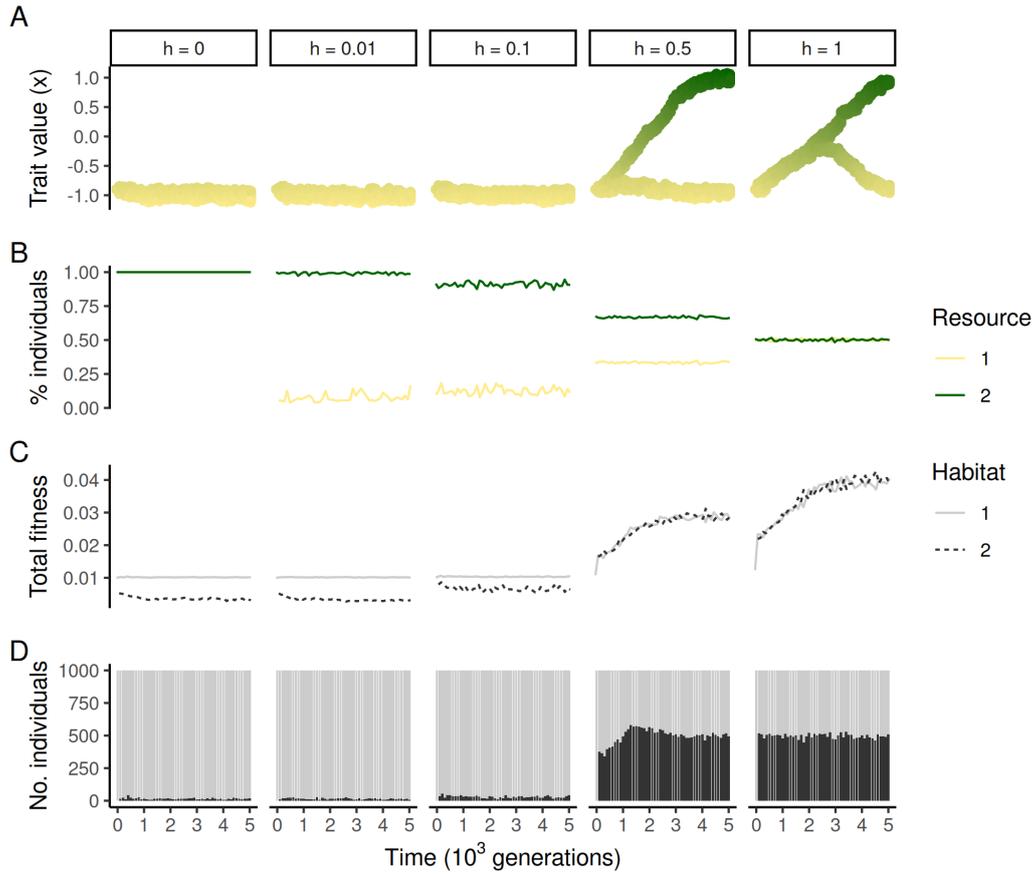


Figure S6: Simulations across habitat symmetry levels h , under a strong trade-off ($s = 1.2$) and random choice ($\beta = 0$). Other parameters as per Table 1. (A) Trait values through time. Branching is lost as habitat symmetry h decreases (right to left), because more asymmetric habitats require a weaker trade-off to be conducive to evolutionary branching. (B) Proportion of individuals choosing each resource at each generation in habitat 2. In more asymmetric habitats, individuals (which, under these parameters, are not choosy) utilize the more abundant resource more, even though they are not well-adapted to it. (C) Mean fitness after all feeding rounds in both habitats, showing that this choice of the maladaptive but abundant resource results in a lower fitness in habitat 2 when habitat asymmetry is high (low h). (D) Population density across habitats and through time. When h is low, individuals from habitat 2 fail at establishing a stable population because their fitness on resource 2 is too low.

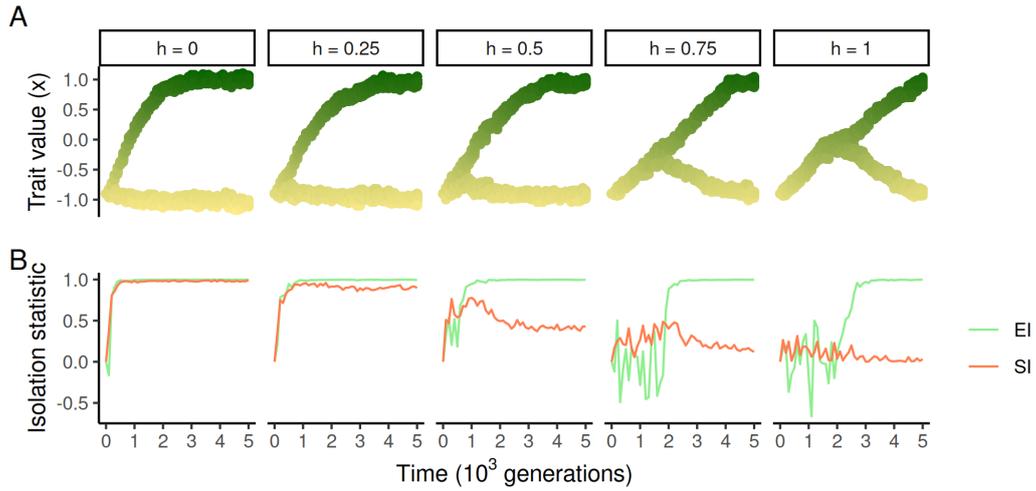


Figure S7: Simulations across habitat symmetry levels h , under moderate trade-off strength ($s = 1$) and random choice ($\beta = 0$). Other parameters as per Table 1. (A) Individual trait values through time. As habitat asymmetry increases (h decreases, right to left), branching becomes more and more driven by local adaptation to the most abundant resource, and less by frequency dependence. (B) Isolation statistics showing that divergence occurs concurrently with spatial isolation when habitat asymmetry is high. EI, ecological isolation; SI, spatial isolation.

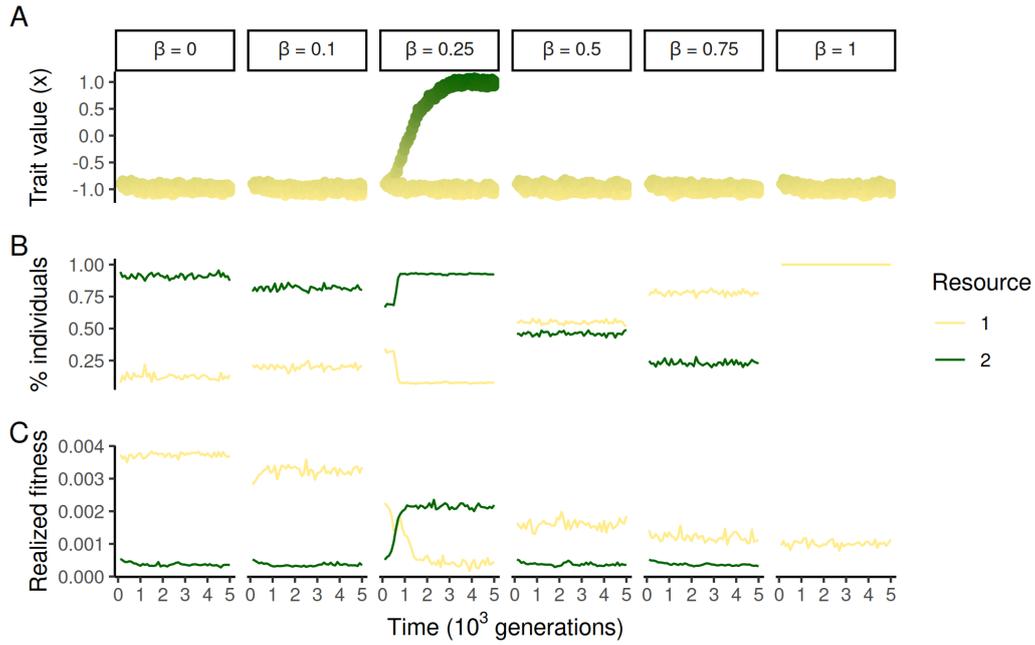


Figure S8: Simulations across resource assessment accuracies β , in asymmetric habitats ($h = 0.1$) and under a strong trade-off ($s = 1.2$). Other parameters as per Table 1. (A) Trait values through time. Under moderate resource asymmetry (i.e. with high-enough resource discovery, $\delta = 0.04$), branching occurs at intermediate choosiness (here, $\beta = 0.25$). (B) Proportion of individuals choosing each resource at each generation in habitat 2. As choosiness increases, the rare resource 1 is chosen more often in habitat 2. (C) Mean fitness gain from choosing each resource through time, in habitat 2, showing that resource 1 is still the most advantageous despite being rare. For branching and adaptation to the maladaptive resource 2 to occur, resource 1 must be chosen often enough that a viable population with sufficient fitness can be established in habitat 2, but not so often that individuals are no longer exposed to resource 2.

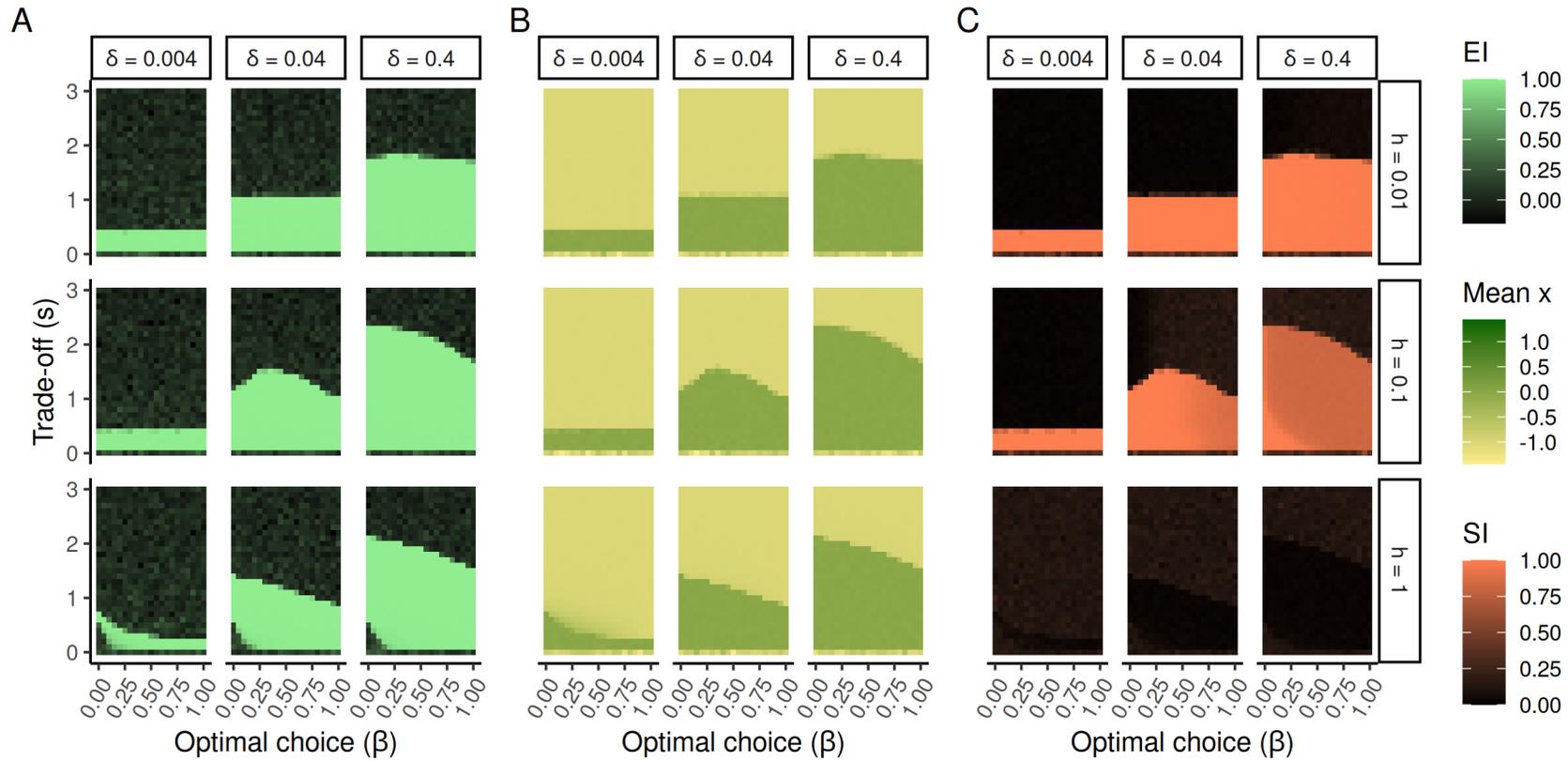


Figure S9: Simulation outcomes across the same parameter space as in Figure 3, but with a lower migration rate ($m = 0.001$). Legend as per Figure 3. All simulations are run under $m = 0.001$ except those under $h = 1$, which are the same as those in Figure 3 (i.e. $m = 0.01$), and are here for comparison, because lowering the migration rate was not expected to cause any difference when habitats are symmetric ($h = 1$) and eco-evolutionary dynamics are driven by within-habitat processes in both habitats simultaneously, rather than by colonization-limited processes (as in asymmetric habitats with low h).

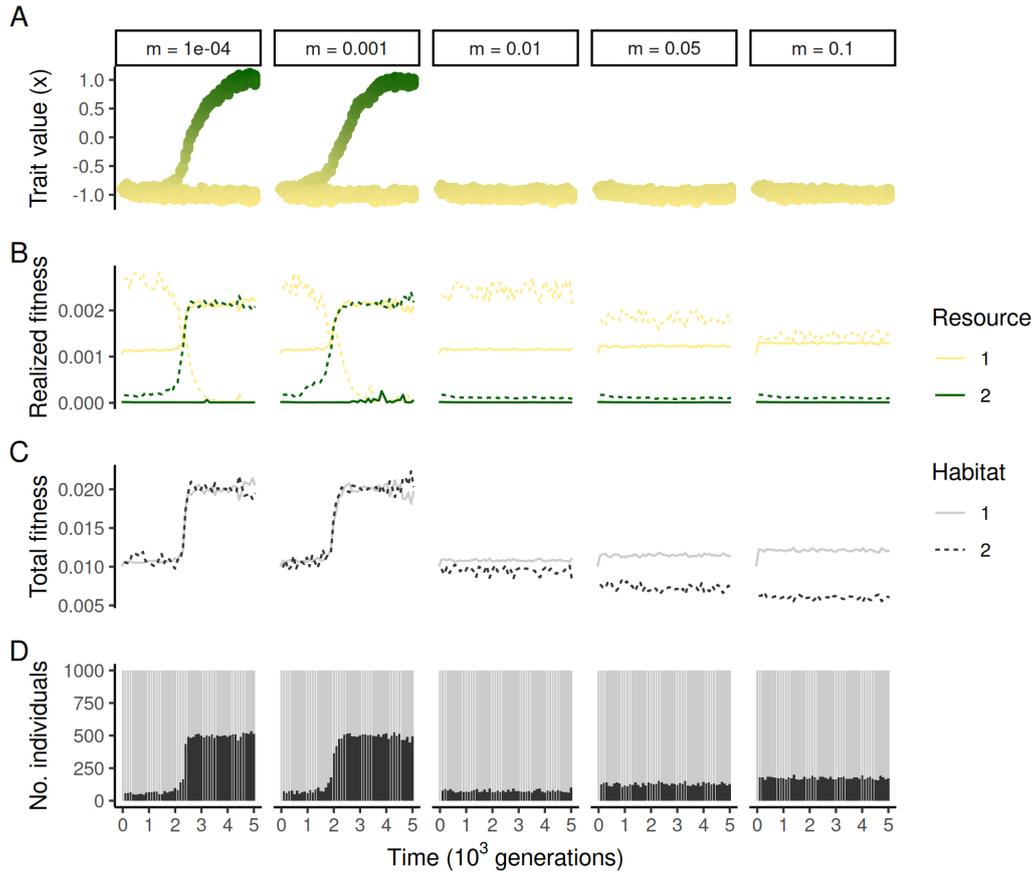


Figure S10: Simulations across migration rates m in asymmetric habitats ($h = 0.1$), under a strong trade-off ($s = 1.5$) and at intermediate resource assessment accuracy ($\beta = 0.3$). Other parameters as per Table 1. (A) Traits through time. Intermediate choosiness maximizes the chances for branching in moderately asymmetric habitats when migration is low — if migration increases, branching is lost. (B) As migration goes down, the rare resource 1 becomes more profitable in habitat 2 (less competition from immigrants), and slightly less so in habitat 1 (more competition with individuals that would have otherwise migrated). (C) Mean fitness of individuals after all feeding rounds in each habitat, where reduced migration increases fitness in habitat 2 relative to habitat 1 — until a viable population can establish and adapt to resource 2 in habitat 2. (D) Population densities across habitats.

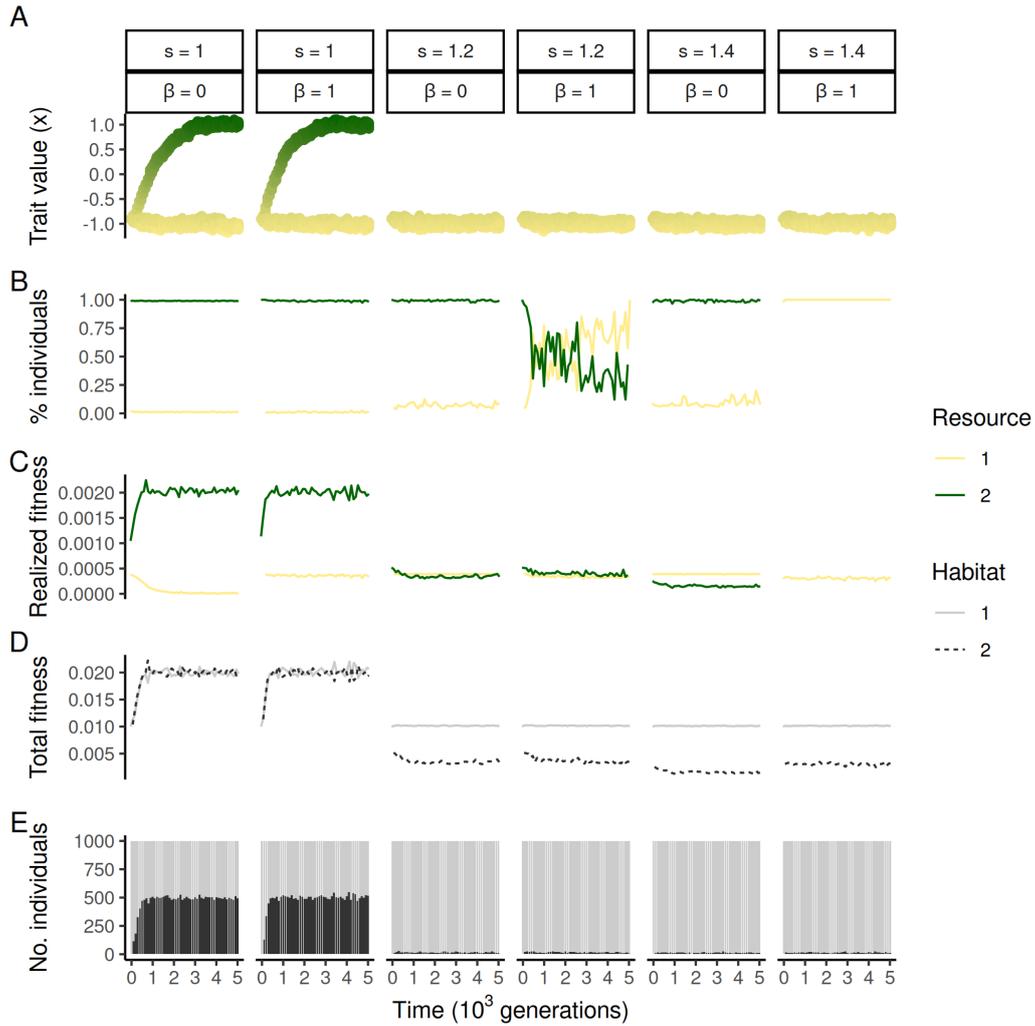


Figure S11: Simulations across trade-off strength s at two levels of choosiness ($\beta = 0$ and $\beta = 1$), in highly asymmetric habitats ($h = 0.01$). Other parameters as per Table 1. (A) Traits through time. Trade-off strength affects both ends of the choosiness spectrum in a similar way — branching occurs if the trade-off is not too strong. (B) Proportion of individuals choosing each resource at each generation in habitat 2. Both choosy ($\beta = 1$) and non-choosy ($\beta = 0$) individuals pick the same resource (the abundant resource 2) when the trade-off is weak enough. (C) Mean fitness gain from choosing each resource in habitat 2, showing that the abundant resource becomes the most profitable as the trade-off weakens. (D) Mean fitness after all feeding rounds in both habitats. A population can establish in habitat 2 once the trade-off is sufficiently weak, at which point choosy and non-choosy individuals are already utilizing the same resource. (E) Population densities across habitats.

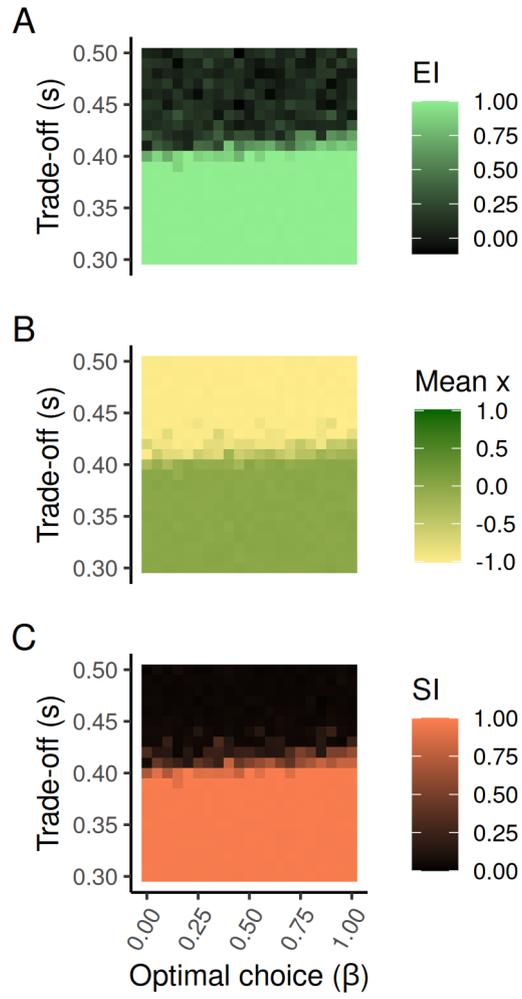


Figure S12: Zoom-in on a particular range of trade-off strengths s in the parameter space explored in Figure 3, when $h = 0.1$ and $\delta = 0.004$. Legend as per Figure 3.

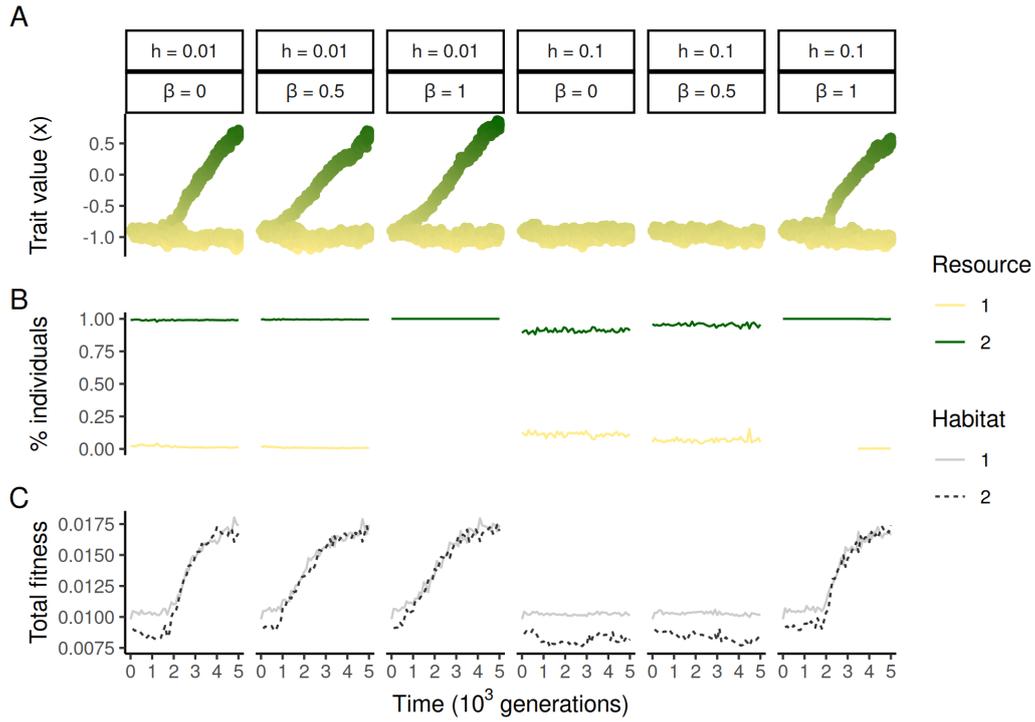


Figure S13: Simulations across levels of habitat symmetry h and resource assessment accuracies β under a weak trade-off ($s = 0.4$) and at low resource discovery rate ($\delta = 0.004$). Other parameters as per Table 1. (A) Traits through time. When resource discovery is low, branching becomes slightly more likely with higher choosiness β when habitats are not too asymmetric ($h = 0.1$) — a pattern that disappears when asymmetry is very high ($h = 0.01$). (B) Proportion of individuals choosing each resource in habitat 2. When habitat symmetry is not too low ($h = 0.1$), noticeably fewer non-choosy individuals (low β) pick resource 2 than choosy individuals (high β), as non-choosy individuals encounter resources purely based on exposure. (C) Mean fitness after all feeding rounds in both habitats. The establishment of a viable population in habitat 2 is rendered more difficult when (non-choosy) individuals pick the rarer resource 1 by passive exposure (a problem that does not present itself when resource 1 is sufficiently rare, $h = 0.01$).

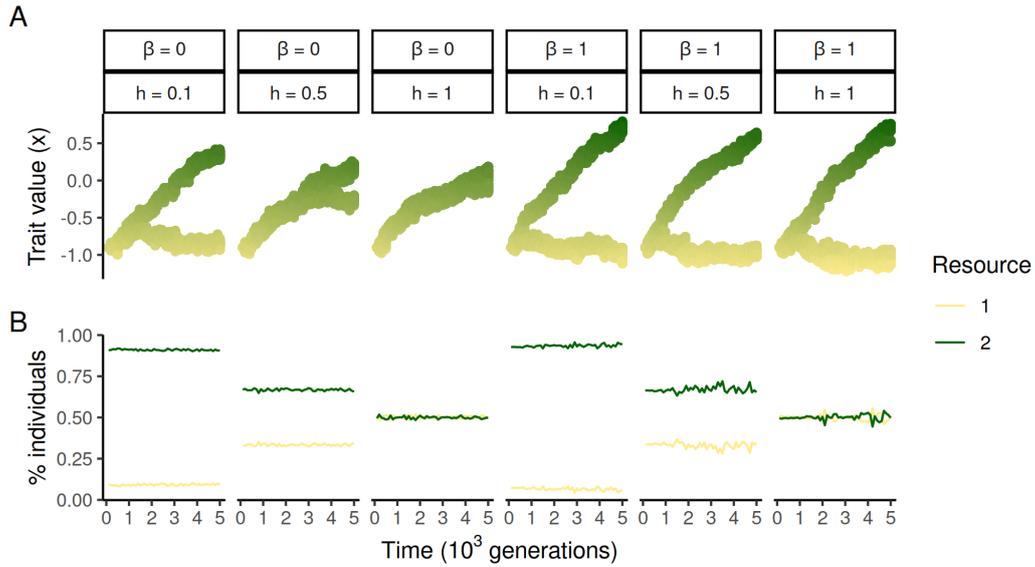


Figure S14: Simulations across levels of habitat symmetry h and resource assessment accuracies β under a weak trade-off ($s = 0.2$). Other parameters as per Table 1. (A) Traits through time. Under random choice ($\beta = 0$) a high habitat asymmetry (low h) is needed for branching when the trade-off would normally favor generalists, while high choosiness promotes branching at all levels of habitat symmetry. (B) Proportion of individuals choosing each resource in habitat 2. Habitat asymmetry introduces a bias in exposure which selects against generalists in non-choosy individuals ($\beta = 0$). This effect is not visible in choosy individuals ($\beta = 1$) because they are already biased in their exposure (even in symmetric habitats) due to active choice.

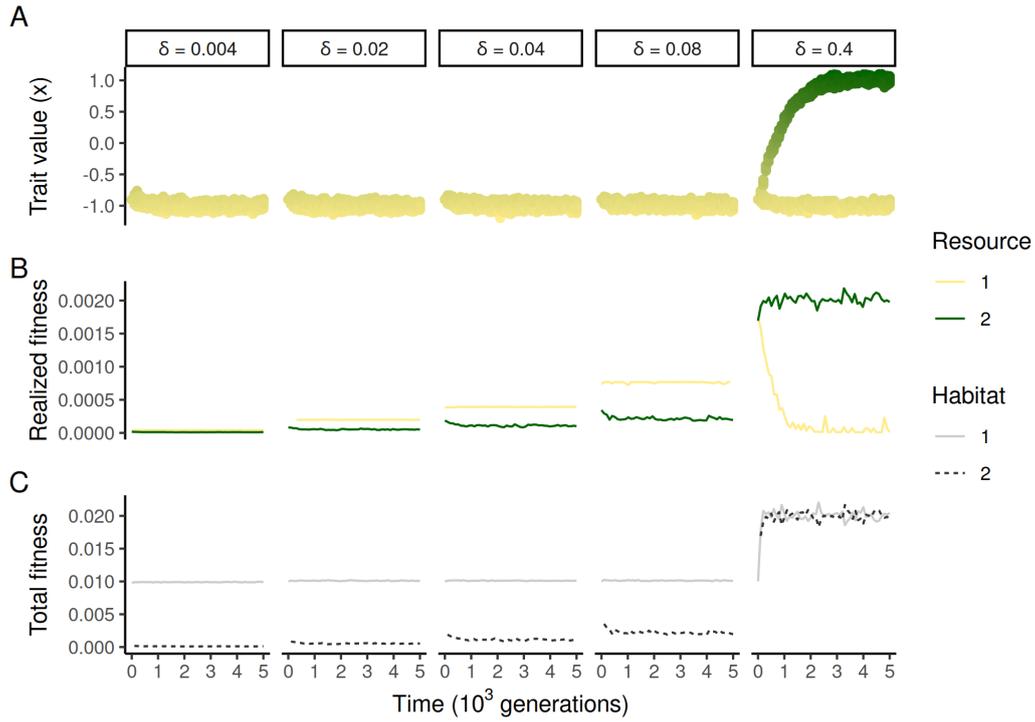


Figure S15: Simulations across resource discovery rates δ in highly asymmetric habitats ($h = 0.01$), under a strong trade-off ($s = 1.5$) and random choice ($\beta = 0$). Other parameters as per Table 1. (A) Traits through time. Generally, resource discovery facilitates branching by alleviating limiting conditions such as strong trade-offs or low resource concentrations. (B) Mean fitness gain from feeding on each resource at each generation in habitat 2. (C) Mean fitness after all feeding rounds in both habitats. By making a rare resource more available, resource discovery is critical in establishing a sufficiently high fitness for a viable colonizing population to be maintained in asymmetric habitats.

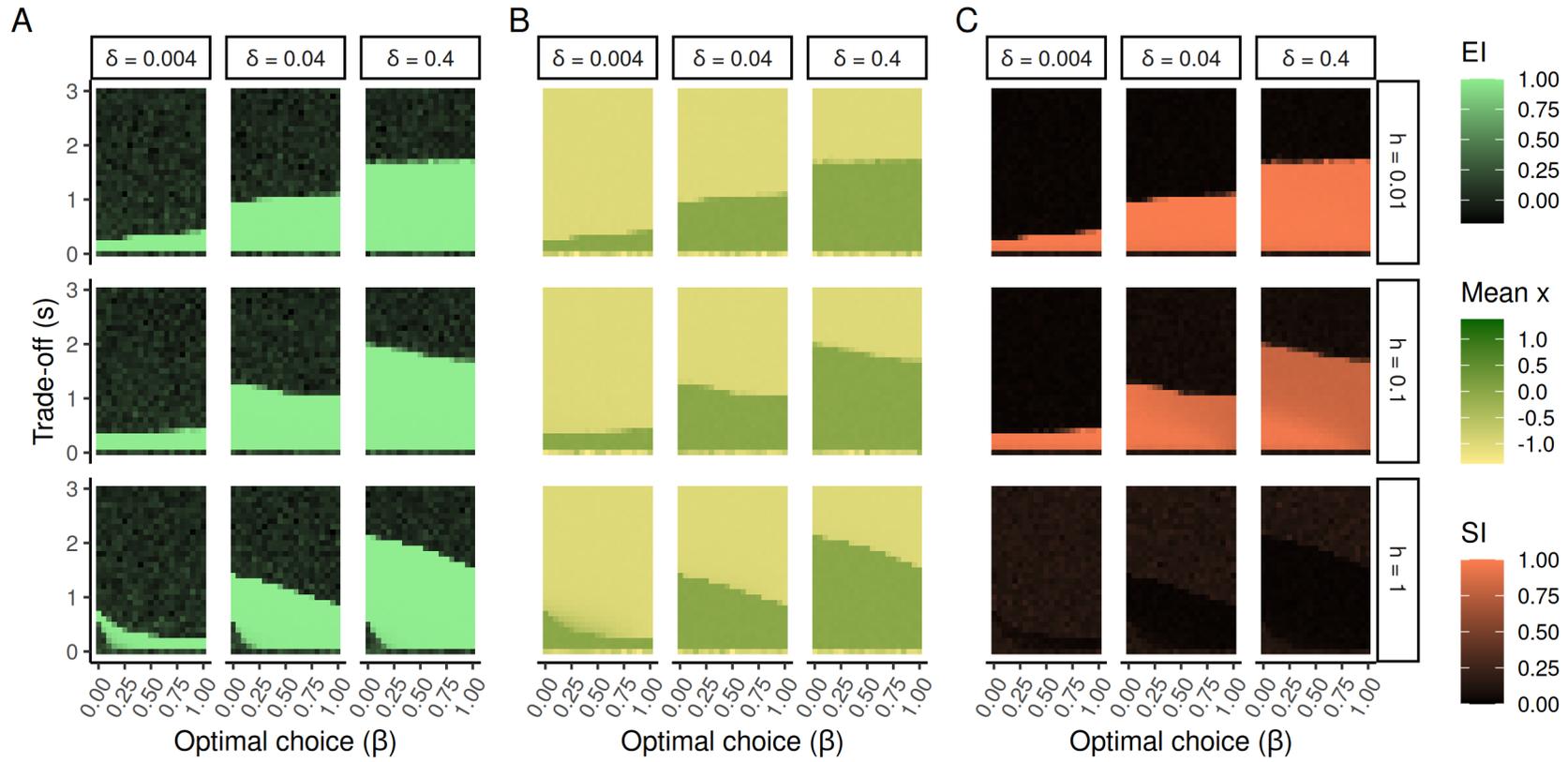


Figure S16: Simulation outcomes across the same parameter space as in Figure 3 but with low resource abundance weight ($\alpha = 0$). Legend as per Figure 3. Note that simulations with $h = 1$ shown here are the same as in Figure 3 (i.e. $\alpha = 1$) because by construction the model under $\alpha = 0$ is equivalent to $\alpha = 1$ when $h = 1$ (resource are equally abundant in both habitats, see Methods).

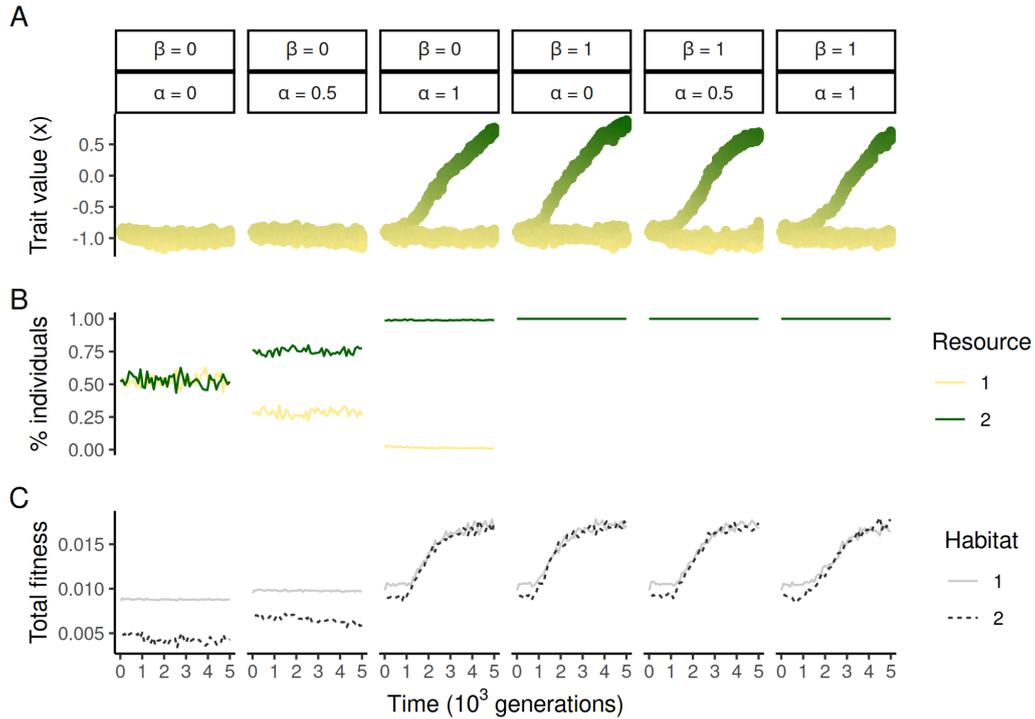


Figure S17: Simulations across resource abundance weights α at two levels of resource assessment accuracy ($\beta = 0$ and $\beta = 1$), in highly asymmetric habitats ($h = 0.01$), at a low resource discovery rate ($\delta = 0.004$) and under a weak trade-off ($s = 0.4$). Other parameters as per Table 1. (A) Traits through time. Decoupling resource abundance and encounter (lower α) reduces the probability of branching in less choosy individuals (low β). (B) Proportion of individuals feeding on each resource at each generation in habitat 2. Lowering the resource abundance weight α forces non-choosy individuals ($\beta = 0$) to pick resource 1 close to half of the time even when it is nearly absent, thus reducing exposure to the more abundant resource 2. (C) Mean fitness after all feeding rounds in both habitats, showing that high exposure to a nearly absent resource is accompanied by a reduction in fitness in habitat 2, hindering the potential of the population to settle and adapt in this habitat. By construction, resource abundance weight has no effect when $\beta = 1$ (see Methods).

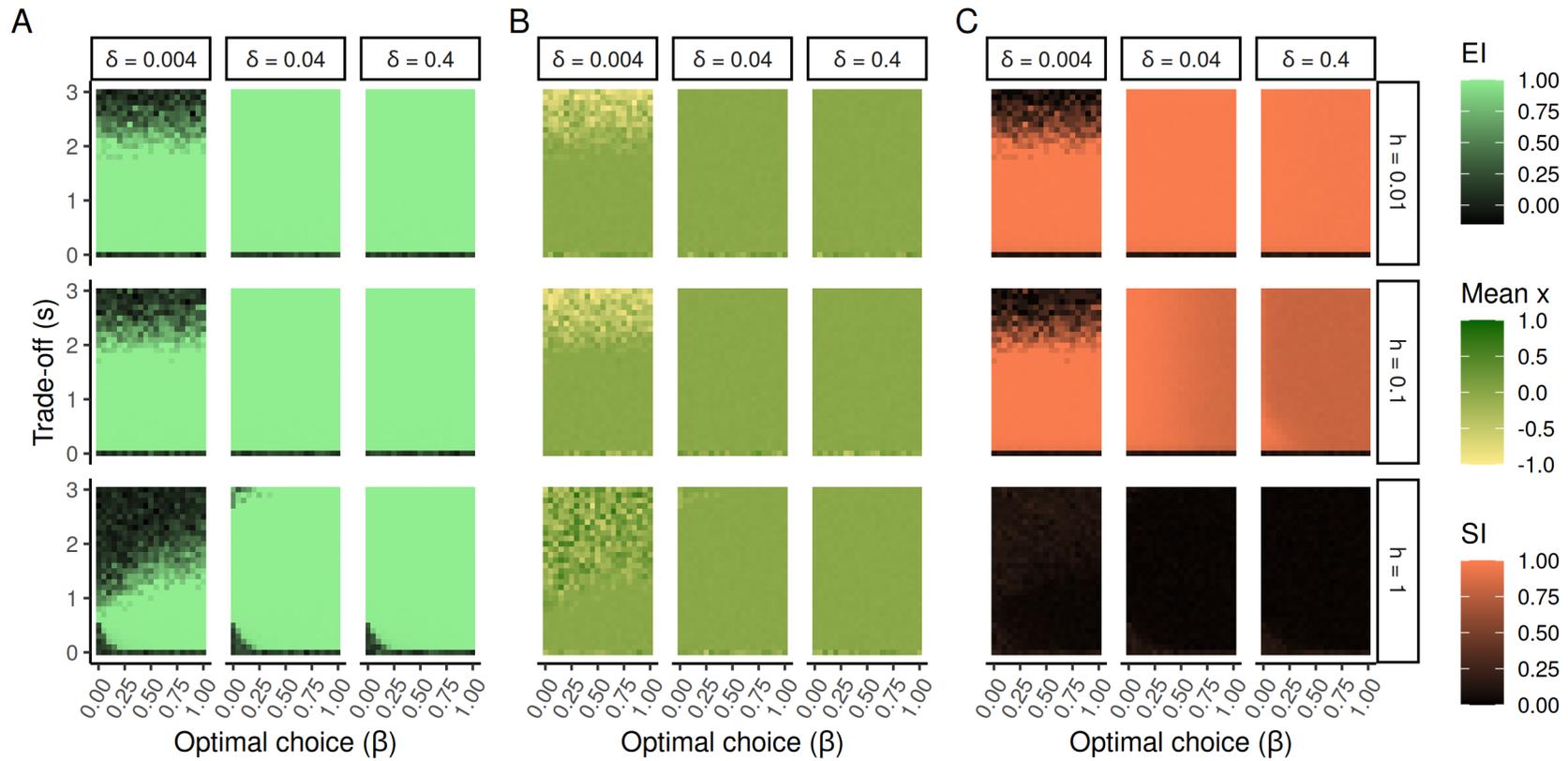


Figure S18: Simulation outcomes across the same parameter space as in Figure 3, but where the population starts as a generalist ($x_0 = 0$). Legend as per Figure 3.

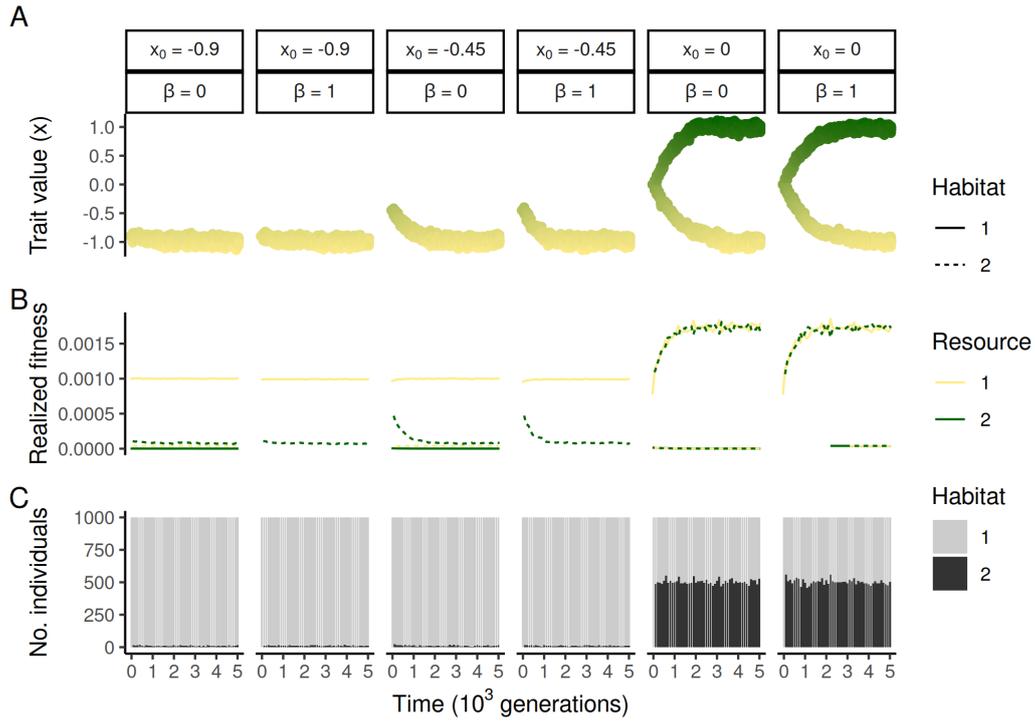


Figure S19: Simulations across starting points x_0 at two resource assessment accuracies ($\beta = 0$ and $\beta = 1$), in highly asymmetric habitats ($h = 0.01$), at a low resource discovery rate ($\delta = 0.004$) and under a moderate trade-off ($s = 1$). Other parameters as per Table 1. (A) Traits through time, showing that branching in asymmetric habitats can still occur under stronger trade-offs if the population starts closer to the generalist strategy $x = 0$. (B) Mean fitness gain from feeding on each resource in both habitats. Adaptation to resource 2 in habitat 2 proceeds once the starting trait value is close enough to zero as to give a sufficiently high fitness to individuals utilizing that resource, which can then respond to directional selection for increased adaptation without having to overcome the cost of deviating from some initial, specialized phenotype. (C) Population densities in both habitats.

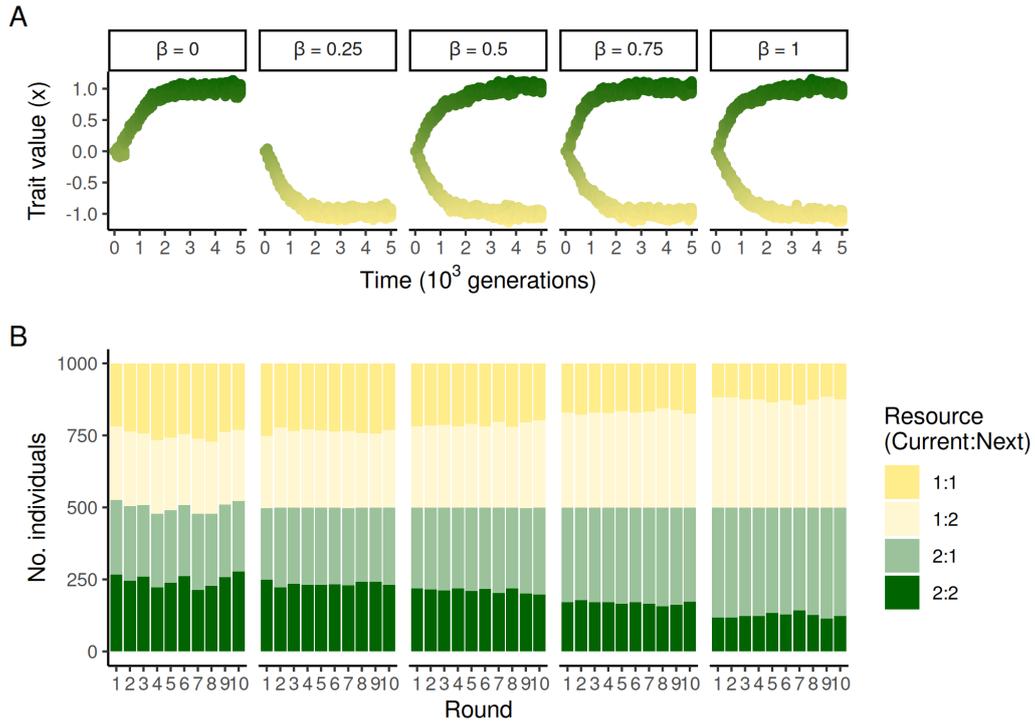


Figure S20: Simulations across resource assessment accuracies β when the population starts off as a generalist ($x_0 = 0$), in symmetric habitats ($h = 1$), at a low resource discovery rate ($\delta = 0.004$) and under a strong trade-off ($s = 1.5$). Other parameters as per Table 1. (A) Traits through time. When starting as a generalist, higher choosiness promotes branching where non-choosy individuals (low β) would have specialized on either resource (at random depending on stochastic fluctuations). (B) Numbers of individuals in each combination of resource chosen by the focal individual (“current”) and resource chosen by the next individual in the queue (in the same habitat, “next”) at generation 0. As choosiness β increases, individuals in the feeding queue are more likely to choose the opposite resource to their predecessor, indicating strong frequency dependence and an advantage to pick the resource that is slightly less depleted (in symmetric habitats where both resources start in equal amount).

773 **Appendix**

774 Here we introduce a deterministic version of our model, which was used for adaptive dynamics
 775 analysis (Metz et al., 1996; Geritz et al., 1998). Adaptive dynamics theory is a body of con-
 776 ceptual tools allowing the analysis of the outcome of evolution by selection using evolutionary
 777 invasion analysis. In this study, we used some of those tools to predict the expected outcome
 778 of our stochastic model in the absence of resource choice based on perceived payoff ($\beta = 0$, see
 779 Methods). We could not derive equations for $\beta > 0$ because the choice then depends on the or-
 780 der in which individuals are taken throughout feeding rounds. The present analysis is therefore
 781 used as benchmark against which to compare and validate the findings from individual-based
 782 simulations in a simplified scenario.

783 **Demographic model** Let there be a mutant with trait value x in a monomorphic resident
 784 population with trait value \hat{x} . The demographic dynamics of a rare mutant are given by

$$\vec{N}_{t+1} = \mathbf{\Lambda}(x, \hat{x}) \vec{N}_t \quad (10)$$

785 where \vec{N}_t is the vector of densities of the mutant across habitat patches at time t , itself given
 786 by

$$\vec{N}_t = \begin{pmatrix} N_1 \\ N_2 \end{pmatrix}_t \quad (11)$$

787 where N_j is the density of mutants in habitat patch j . $\mathbf{\Lambda}$ is the demographic transition matrix
 788 from one generation to the next, given by

$$\mathbf{\Lambda}(x, \hat{x}) = \mathbf{M} \mathbf{Q}(x, \hat{x}) \quad (12)$$

789 where \mathbf{M} is the migration matrix, given by

$$\mathbf{M} = \begin{pmatrix} 1 - m & m \\ m & 1 - m \end{pmatrix}, \quad (13)$$

790 in which m is the migration rate between the two patches, and where \mathbf{Q} is the reproduction
 791 matrix, given by

$$\mathbf{Q}(x, \hat{x}) = \begin{pmatrix} W_1(x, \hat{x}) & 0 \\ 0 & W_2(x, \hat{x}) \end{pmatrix} \quad (14)$$

792 in which $W_j(x, \hat{x})$ is the per capita growth rate of the mutant in habitat patch j .

793

794 The per capita growth rate $W_j(x, \hat{x})$ in a given habitat is equivalent to the probability of an
 795 individual being picked as parent of a new offspring in the weighted lottery of reproduction, and
 796 is directly proportional to the accumulated amount of resources over multiple feeding rounds
 797 (see Methods), which is itself proportional to the average amount of resource obtained in a
 798 round. The resource obtained by an individual in a round depends on (1) the resource chosen,

799 and (2) which other individuals chose the same resource. In the absence of payoff-based choice
800 ($\beta = 0$), these two events are independent. Hence, over many feeding rounds,

$$W_j(x, \hat{x}) \propto \sum_{i=1}^2 P_{ij} \sum_{n=0}^{\hat{N}_j} W_{ij}(x, \hat{x}, n) \mathbb{P}(\hat{N}_{ij} = n) \quad (15)$$

801 where $W_{ij}(x, \hat{x}, n)$ is the amount of resource i accumulated in habitat j by a mutant with trait
802 x , in a resident population with trait \hat{x} where n individuals have also chosen resource i , P_{ij} is
803 the baseline probability of choosing resource i at random, and $\mathbb{P}(\hat{N}_{ij} = n)$ is the probability
804 that the number \hat{N}_{ij} of resident individuals having chosen resource i in habitat j is n .

805

806 The probability P_{ij} of choosing a given resource i in habitat j is given by

$$P_{ij} = 1/2(1 - \alpha) + \alpha R_{ij}/R_j^{\text{tot}} \quad (16)$$

807 where R_{ij} is the (untouched) total amount of resource i in habitat j (either 1 or h , the habi-
808 tat symmetry parameter, see Methods), $R_j^{\text{tot}} = R_{1j} + R_{2j}$ is the total amount of resources in
809 habitat j , and α is the resource abundance weight parameter, tuning how important relative
810 resource abundances are for encounter rates (if $\alpha = 0$ the encounter probability is $1/2$ regardless
811 of resource abundance, see Methods).

812

813 Since every individual chooses randomly ($\beta = 0$), the number \hat{N}_{ij} of (resident) individuals
814 having chosen resource i in habitat j in a given round follows the binomial distribution

$$\hat{N}_{ij} \sim \text{Binom}(\hat{N}_j, P_{ij}) \quad (17)$$

815 where \hat{N}_j is the total number of resident individuals in habitat j . This means that the probability
816 of any given number n of individuals having chosen resource i in habitat j is given by the
817 binomial formula

$$\mathbb{P}(\hat{N}_{ij} = n) = \binom{\hat{N}_j}{n} P_{ij}^n (1 - P_{ij})^{\hat{N}_j - n}. \quad (18)$$

818 In turn, the amount of resource i obtained by a mutant with trait value x in habitat j amidst
819 a resident population with trait \hat{x} , of which n have chosen resource i , is given by

$$W_{ij}(x, \hat{x}, n) = \begin{cases} R_{ij}^{\text{D}}(\hat{x}, n) C_i(x) / C_{ij}(\hat{x}, n) & \text{if } C_{ij}(\hat{x}, n) > 0 \\ 0 & \text{otherwise,} \end{cases} \quad (19)$$

820 where $R_{ij}^{\text{D}}(\hat{x}, n)$ is the amount of resource i discovered in habitat j by a population of n feeders
821 with trait value \hat{x} , $C_i(x)$ is the consumption rate of a mutant with trait value x on resource
822 i , and $C_{ij}(\hat{x}, n)$ is the cumulative consumption rate on resource i of all n residents having also
823 chosen resource i (here we assume the impact of the rare mutant to be negligible on the amount
824 of resource discovered). The amount of resources discovered by those individuals is given by

$$R_{ij}^D(\hat{x}, n) = R_{ij} \left(1 - \exp(-\delta C_{ij}(\hat{x}, n)) \right) \quad (20)$$

825 where δ is the resource discovery rate (see Methods). For a given resource i , and since the
 826 resident population is monomorphic with trait \hat{x} , the cumulative consumption rate on all n
 827 residents having also chosen that resource is given by

$$C_{ij}(\hat{x}, n) = n C_i(\hat{x}) . \quad (21)$$

828 In the numerical computations, we further assumed individuals to be equally distributed
 829 between the two habitats, owing to the fixed total population size and the symmetrical migration
 830 matrix (Eq. 13), that is, $\hat{N}_1 = \hat{N}_2 = N/2$.

831 **Invasion fitness** Following Otto and Day (2007), the population growth rate of a mutant
 832 across both habitat patches is given by the leading eigenvalue of the transition matrix \mathbf{A} (Eq.
 833 12). Here, it is

$$r(x, \hat{x}) = \frac{1}{2} \left((1 - m) (W_1(x, \hat{x}) + W_2(x, \hat{x})) \right. \quad (22)$$

$$\left. + \sqrt{\left((m - 1) (W_1(x, \hat{x}) + W_2(x, \hat{x})) \right)^2 - 4 W_1(x, \hat{x}) W_2(x, \hat{x}) (1 - 2m)} \right) . \quad (23)$$

834 This growth rate can then be compared with that of the resident, $r(\hat{x}, \hat{x})$, to know if any
 835 given mutant can invade and become the new resident, or not, in the form of an *invasion fitness*
 836 function,

$$\lambda(x, \hat{x}) = r(x, \hat{x}) - r(\hat{x}, \hat{x}) . \quad (24)$$

837 In adaptive dynamics analysis, a mutant can invade if $\lambda(x, \hat{x}) > 0$. Otherwise, it goes extinct
 838 and the resident remains. Mapping the value of the invasion fitness across many combinations of
 839 mutants and residents results in a pairwise invasibility plot (PIP), which graphically summarizes
 840 the expected evolutionary dynamics. See Figure S1 for how to interpret PIPs.

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