

1      Evolutionary diversification of ecological specialists under  
2      informed resource choice

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7      **Abstract**

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

34      **Keywords** — ecological specialization, optimal foraging, frequency-dependent selection,  
35      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; Schlüter, 2000; Tilman, 1982). 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 (Schlüter, 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 (Egas et al., 2004; Levins, 1962; 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 (Edelaar et al., 2017; Morris, 2003; Ronce, 2007), 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-Gámez, 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 (Bolnick & Otto, 2013;  
74 Edelaar & Bolnick, 2012; Edelaar et al., 2008; Rueffler et al., 2007). 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 (Ravigné et al., 2009; Rueffler et al., 2007).

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. Abrams, 1999;  
86 Matsuda and Namba, 1989; Rueffler et al., 2007; Stenseth, 1984; Svanbäck and Bolnick, 2005),  
87 or in the context of multiple habitats containing one resource each (habitat selection models,  
88 e.g. Brown, 1990, 1996; Ravigné et al., 2009, 2004). 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 diversification  
91 resulting in the emergence of such ecotypes (Abrams, 2006; Vincent et al., 1996). 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  $x_i^{\text{opt}}$  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., 2009, 2004). To fine-tune the behavior  
 165 of our model along this continuum, we introduce a new parameter, the resource discovery rate  
 166  $\delta$ , 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  $\delta$  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  $\beta$  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,  $P_0^{\text{best}}$ . 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  $R_{\text{best}}$  is the total concentration of the better resource in the local habitat ( $R_{1j}$  or  $R_{2j}$ ),  
 191  $R_{\text{tot}}$  is the sum of both total resource concentrations ( $R_{1j} + R_{2j}$ ), and  $\alpha$  is the resource abundance weight parameter, determining how much the probability of encounter of a resource is  
 192 determined by its abundance ( $0 \leq \alpha \leq 1$ ). When  $\alpha = 1$ , the probability of encounter of a  
 193 resource is equal to its abundance relative to the other resource,  $R_{\text{best}} / R_{\text{tot}}$  (Fig. 1D). When  
 194  $\alpha = 0$ , this probability is  $1/2$ , irrespective of resource abundances. Note that the probability of  
 195 encounter of the best resource is also  $1/2$  whenever  $h = 1$ , regardless of  $\alpha$ , because in that case  
 196 the resource concentrations in each habitat are equal. Hence, parameter  $\alpha$  really only makes a  
 197 difference when  $h < 1$  (i.e.  $R_{1j} \neq R_{2j}$ ). Scenarios where  $\alpha \neq 1$  and  $R_{1j} \neq R_{2j}$  could correspond,  
 198 for example, to situations where resources can be found in two particular species of trees, both  
 199 occupying equal surfaces in the habitat, but with different yields of usable resource per unit  
 200 area. Note that although parameter  $\beta$  controls the probability of accurately choosing the better  
 201 of the two local resources given their known profitability, for all intents and purposes this is  
 202 equivalent to the accuracy with which individuals perceive this profitability (i.e. in this study  
 203 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  $\mu$  (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  $\sigma_x$  (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

$$EI_0 = 1 - \frac{N_1 V_1 + N_2 V_2}{N V} \quad (7)$$

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

$$EI = (EI_0 - 2/\pi) / (1 - 2/\pi), \quad (8)$$

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

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

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 numerical  
249 analyses based on adaptive dynamics theory (Geritz et al., 1998; Metz et al., 1992), 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.

262 **Results**

263 *Symmetric habitats under random choice*

264

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

272

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

295

296 *Resource choice in symmetric habitats*

297

298 The first notable effect of a nonzero resource choice accuracy  $\beta$  is to prevent the evolution  
299 of a generalist phenotype under weak trade-offs ( $s < 0.5$ ), instead promoting branching into  
300 two specialists (Fig. 3, S4). As individuals become better at picking the most advantageous  
301 resource, the same frequency-dependent advantage as described previously pushes individuals  
302 slightly less well adapted to the initially preferred resource 1 to actively choose the alternative  
303 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  $\beta$  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  $\beta$  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  $\beta$  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  $\beta$  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  $\beta$  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  $\beta$ .

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  $\beta$  (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 amplifies  
389 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 habitats  
408 (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  $\delta$  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  $\beta$ ) utilize the abundant resource but are too  
418 maladapted to it, and choosy individuals (high  $\beta$ ) 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  $\beta$ , by active choice) as well as by non-choosy individuals (low  $\beta$ ,  
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  $\beta$  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  $\beta$ ) 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  $\beta$  does not  
456 affect this pattern, as choosier individuals (high  $\beta$ ) 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  $\delta$  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  $\beta$  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  $\beta$ ), 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 abundance,  
481 by setting the resource abundance weight parameter  $\alpha$  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 description  
487 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  $\beta$ ) 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 explored by non-choosy individuals. Exposure to the abundant resource in a habitat is therefore  
494 reduced, increasing the profitability that this resource must have in order to be advantageous to  
495 specialize on (i.e. to compensate for the cost of deviating from a phenotype specialized on the  
496 rare resource). Hence, compartmentalization of the resources through the decoupling between  
497 encounter and abundance decreases the maximum trade-off strength still allowing branching  
498 under random resource choice ( $\beta = 0$ , Fig. S16, S17, also visible in analyses of the adaptive  
499 dynamics of the model under random choice, Fig. S2C). Increasing the resource choice accuracy  
500  $\beta$  makes the resource abundance weight  $\alpha$  irrelevant, as highly choosy individuals (e.g.  $\beta = 1$ )  
501 direct their efforts purely based on expected payoff and target the most advantageous resource  
502 regardless of the rate of passive encounter. Therefore, the foraging and branching dynamics  
503 of  $\beta = 1$  when  $\alpha = 0$  are identical to when  $\alpha = 1$  (all of the aforementioned Results). This  
504 produces patterns similar to the case without compartmentalization in the more symmetric  
505 habitats (high  $h$ ):  $\alpha = 0$  yields the same foraging dynamics as  $\alpha = 1$  in fully symmetric habitats  
506 ( $h = 1$ ), because then no resource is more rare than the other, irrespective of microhabitats.  
507 In asymmetric habitats (low  $h$ ), however, this means that branching will occur under stronger  
508 trade-offs  $s$  at higher choice accuracies  $\beta$ , owing to choosier individuals wasting less energy for  
509 aging in microhabitats where the yield is low, and generally increasing fitness and the potential  
510 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  $\beta$  (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  $\beta$  (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  $\beta$ .  
 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 repellor, 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  $\delta$  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 (Ravigné et al., 2009;  
589 Rueffler et al., 2007). Ravigné 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 (Egas et al., 2004; Kisdi, 2002), selection for dispersal (Edelaar et al., 2017; Kisdi,  
592 2002; Ronce, 2007) 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 utilizing 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 may  
615 actually facilitate adaptive diversification when the initial population is an ecological generalist.

616 We note, however, that this relies on the resources being in similar abundance and availability  
617 in the environment: if one resource becomes utilized in sufficiently greater proportion, active  
618 choice will promote specialization for that resource. Hence, this phenomenon may be more read-  
619 ily observed (if at all) in natural systems where resources are relatively abundant and easy to  
620 find. On the one hand, this could be the case for *pursuer* species of foragers (sensu MacArthur  
621 and Pianka, 1966), which are not limited by resource scarcity and instead invest in optimizing  
622 handling time (MacArthur & Pianka, 1966; Norberg, 2021; Norberg, 1977; Stephens & Krebs,  
623 1986). Hummingbirds and honeyeaters are usually cited as belonging to that category (Norberg,  
624 2021). On the other hand, *searcher* species (limited by long search times) have been proposed  
625 to be more often generalists than pursuer species (Brown, 1990; MacArthur & Pianka, 1966).  
626 Alternatively, this phenomenon could also apply to species whose resource encounter is high  
627 enough that they can afford “laziness” most of the time, such as short-tailed shrews *Blarina*  
628 *brevicauda* (Herbers, 1981; Martinsen, 1969), or sit-and-wait predators such as *Anolis* lizards  
629 (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 (Doebeli & Dieckmann, 2003; Endler, 1977;  
638 Nosil, 2012), parapatric biogeography (Gavrilets et al., 2000; Mallet, 2005) and hybrid zones  
639 (Abbott, 2017; Barton & Hewitt, 1989; Endler, 1977). 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; Rettel-  
650 bach et al., 2013). In their model, assortative mating and sexual reproduction are key to this  
651 third mode of speciation. Although our results do not necessarily qualify as a new mode of  
652 speciation, here we show that frequency-dependent and spatially divergent selection also inter-  
653 act in an asexual model, independent of assortative mating, when resource choice behavior is  
654 taken into account. Evidence for both frequency-dependent disruptive selection and divergent  
655 local adaptation acting in concert has been documented in relation to speciation in three-spine  
656 sticklebacks *Gasterosteus aculeatus* (Bolnick, 2004; Bolnick & Stutz, 2017; Hendry et al., 2009;  
657 Schlüter, 2003).

658

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. Brown, 1990; Stenseth, 1984; 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 (Bernays, 1998; Pyke, 1984;  
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 (Kisdi, 2001;  
698 Levene, 1953; 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 Ravigné et al., 2009; Rueffler et al., 2007).  
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  $\delta$ . With high  $\delta$ , 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  $\delta$ , 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, or  
730 any specific kind of functional response of the consumer (e.g. Holling's disc equation as in Brown,  
731 1990, 1996; Rueffler et al., 2007; Stenseth, 1984; Vincent et al., 1996). This is because our model  
732 is not based on a demographic model. However, the resources can still be depleted and give  
733 rise to frequency-dependent selection, and the resource discovery rate allows us to approximate  
734 different types of density regulation. Different patterns may emerge under different kinds of  
735 population dynamical aspects such as predator-prey fluctuations (Abrams, 1999, 2006; Abrams  
736 & Matsuda, 2003, 2004; Matsuda & Namba, 1989) and/or more realistic sensory and movement  
737 strategies (e.g. Netz et al., 2022). Second, we did not consider separate trade-offs for various  
738 additional aspects of the foraging process such as search time or handling time (unlike models  
739 based on optimal foraging theory, e.g. Rueffler et al., 2007; Vincent et al., 1996). However,  
740 Rueffler et al. (2007) found that a suitable trade-off in one of these dimensions usually leads  
741 to diversification even if selection does not promote diversification in others, and so this may

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

758 **Acknowledgments**

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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.

770 **Tables**

Table 1: Overview of model parameters and default values.

Symbol	Name	Default
$h$	Habitat symmetry	1
$s$	Trade-off strength	1
$x_i^{\text{opt}}$	Optimal trait value for resource $i$	$\pm 1$
$\delta$	Resource discovery rate	0.04
$\beta$	Resource assessment accuracy	0
$\alpha$	Resource abundance weight	1
$m$	Migration rate	0.01
$\mu$	Mutation rate	0.01
$\sigma_x$	Mutational standard deviation	0.02
$N$	Population size	1 000
$n_R$	Number of rounds per feeding season	10
$T_{\text{max}}$	Simulation time	100 000
$x_0$	Starting trait value	-0.9

771 **Figures**

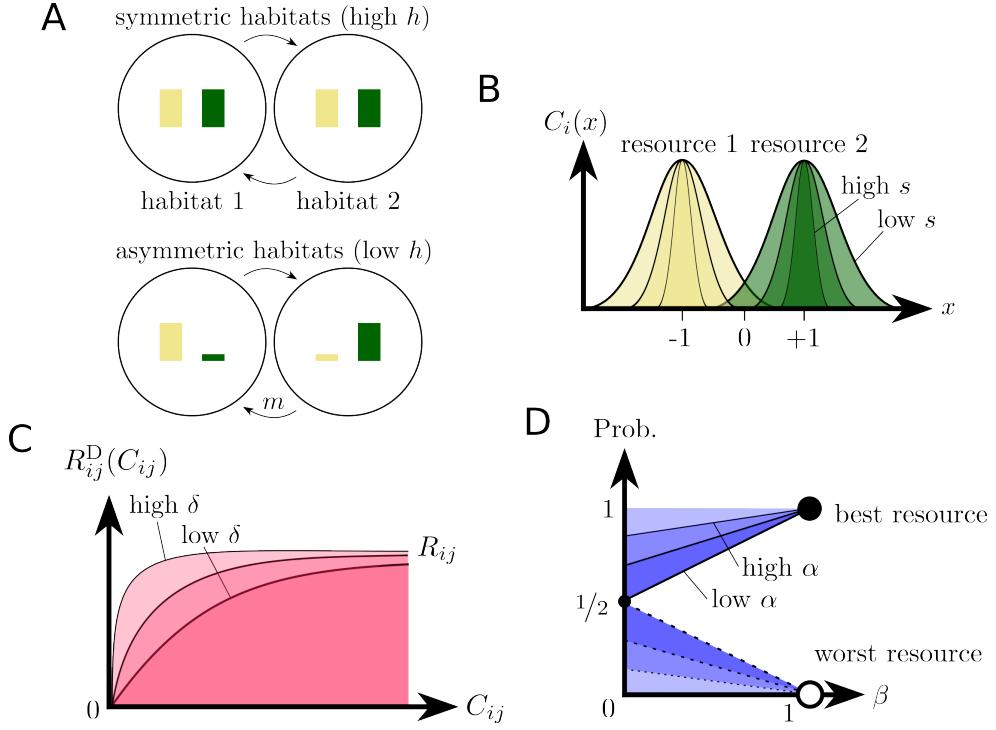


Figure 1: Model overview. (A) Symmetric versus asymmetric partitioning of the resources between habitats depending on the habitat symmetry parameter  $h$ . Migration occurs between habitats at per capita rate  $m$  per generation. (B) Consumption rates on either resource follow two Gaussian curves whose 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  $\delta$ , 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  $\beta$ . The resource abundance weight  $\alpha$  controls the baseline probability of choosing either resource in the absence of choice. This baseline probability approaches the relative proportion of each resource as  $\alpha$  goes up (in this example the local concentration of the best resource is 1, while that of the worst resource is 0).

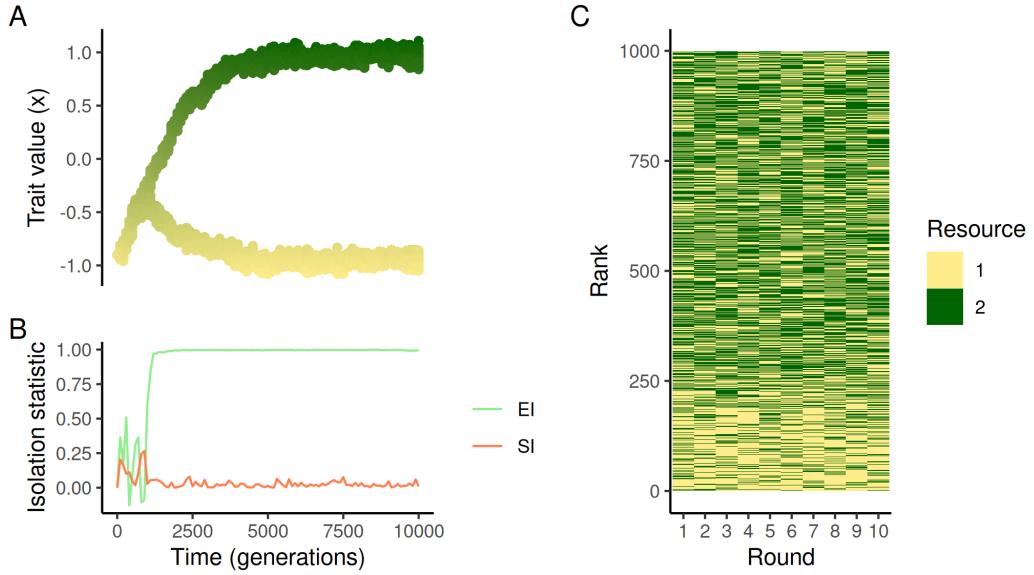


Figure 2: Example simulation in symmetric habitats ( $h = 1$ ) with intermediate assessment accuracy ( $\beta = 0.5$ ). (A) Individual trait values through time. (B) Isolation statistics through time. EI, ecological isolation; SI, spatial isolation (see Methods). Here, evolutionary branching occurs early 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 1 000, 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.

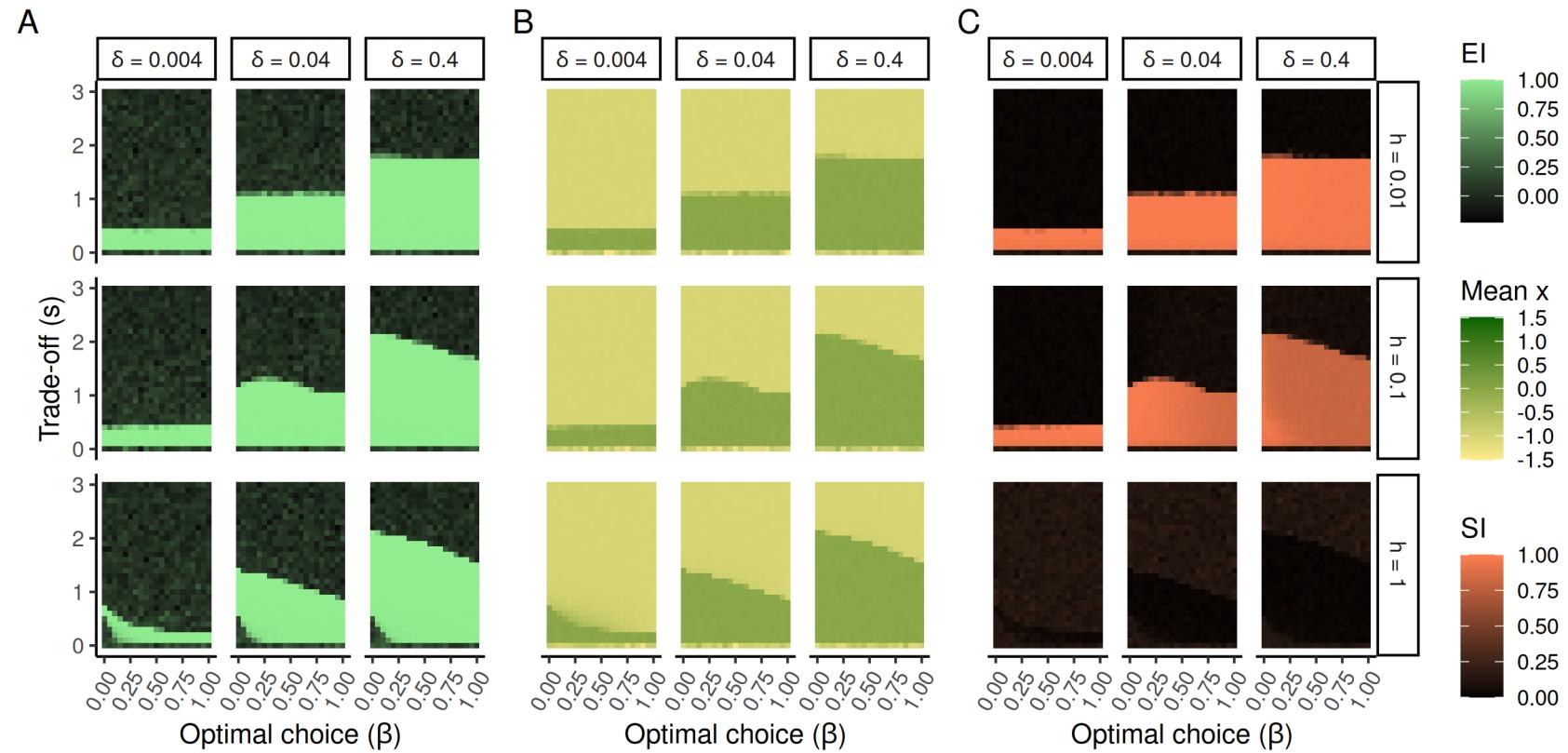


Figure 3: Outcomes of simulations across values of resource assessment accuracy  $\beta$ , trade-off strength  $s$ , resource discovery rate  $\delta$  and habitat symmetry  $h$ . Each cell shows the mean final value after 100 000 generations of (A) the ecological isolation statistic EI, (B) the average 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).

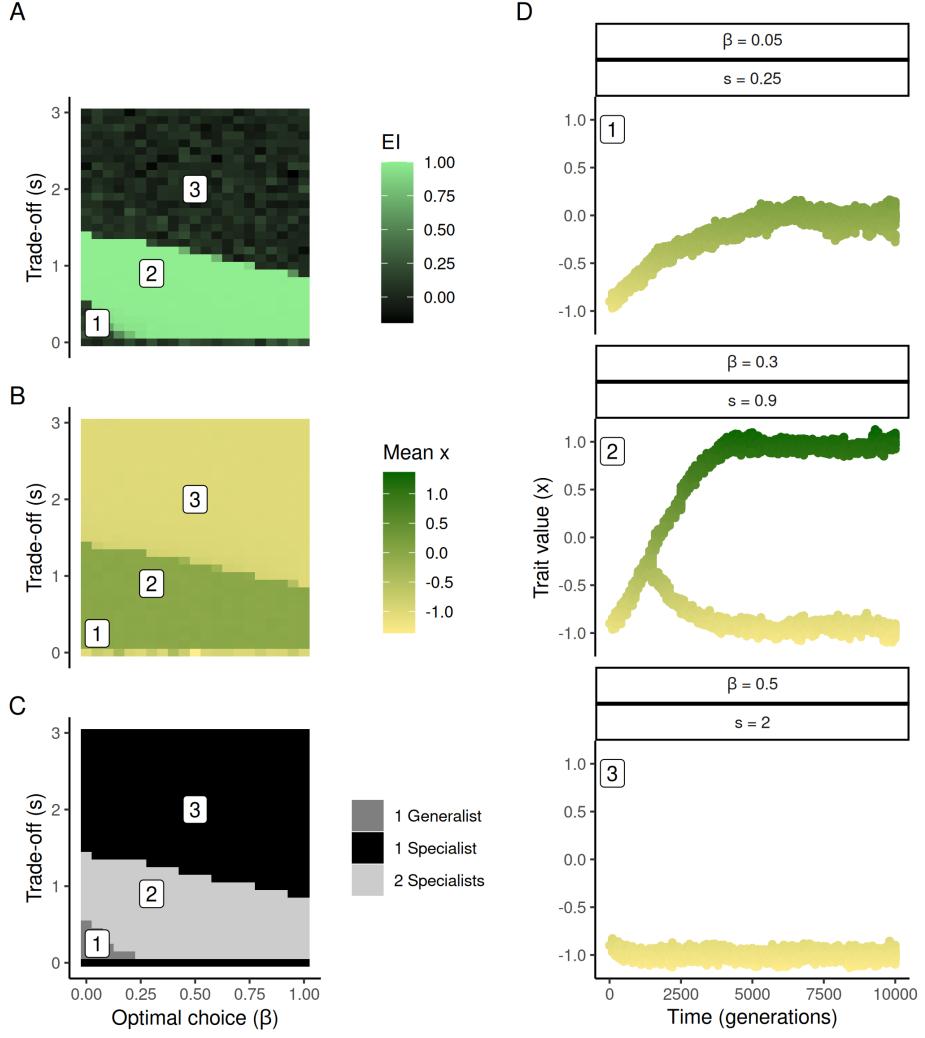


Figure 4: Types of outcomes. (A) Mean final ecological isolation (EI) across parameter space (restricted to the panel with  $h = 1$  and  $\delta = 0.04$  in Fig. 3A). (B) Same for the mean trait value at the end of the simulations (Fig. 3B). (C) Aggregate of A and B summarizing where the outcome is 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.

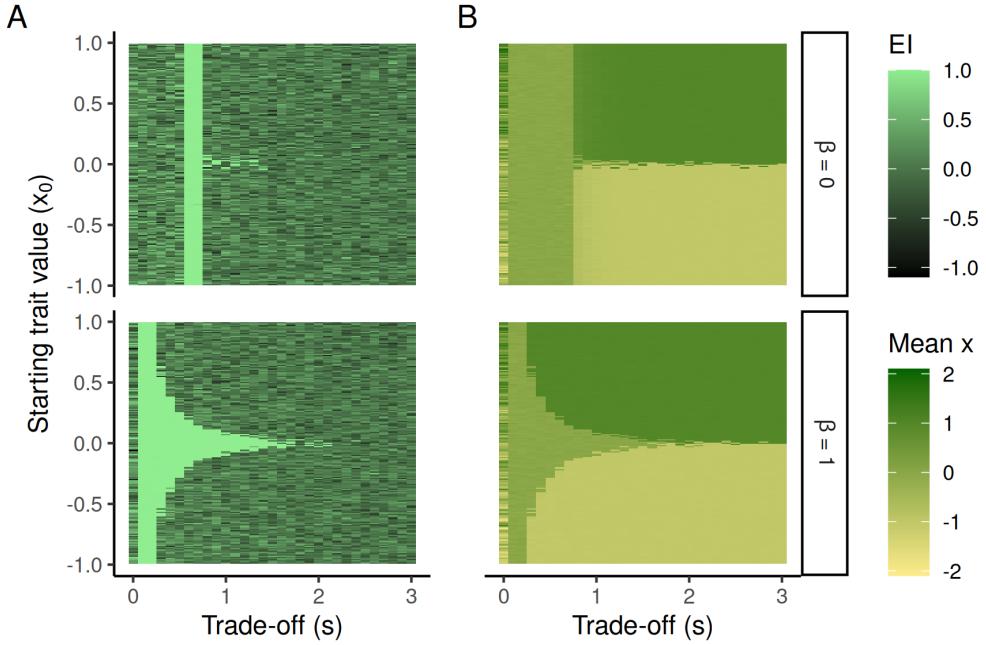


Figure 5: Basin of attraction of the generalist branching point ( $x = 0$ ) across a transect of trade-off strengths  $s$ , at two extreme levels of resource assessment accuracy ( $\beta = 0$  and  $\beta = 1$ ). One simulation was run for each combination of parameter values. For each combination, we 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  $s$ .

772 **Supplementary Figures**

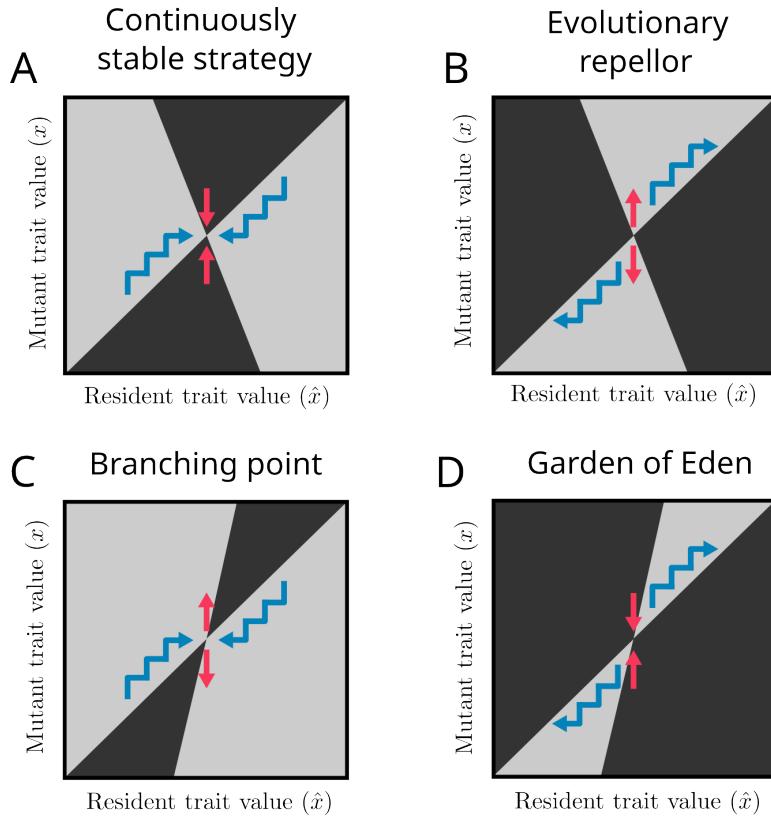


Figure S1: A pairwise invasibility plot (PIP) is a phase plot showing, for each possible value of a trait fixed in a theoretical (monomorphic) *resident* population (here  $\hat{x}$ ), the range of other values of the same trait that a rare *mutant* arising in a population of residents could have, and 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).

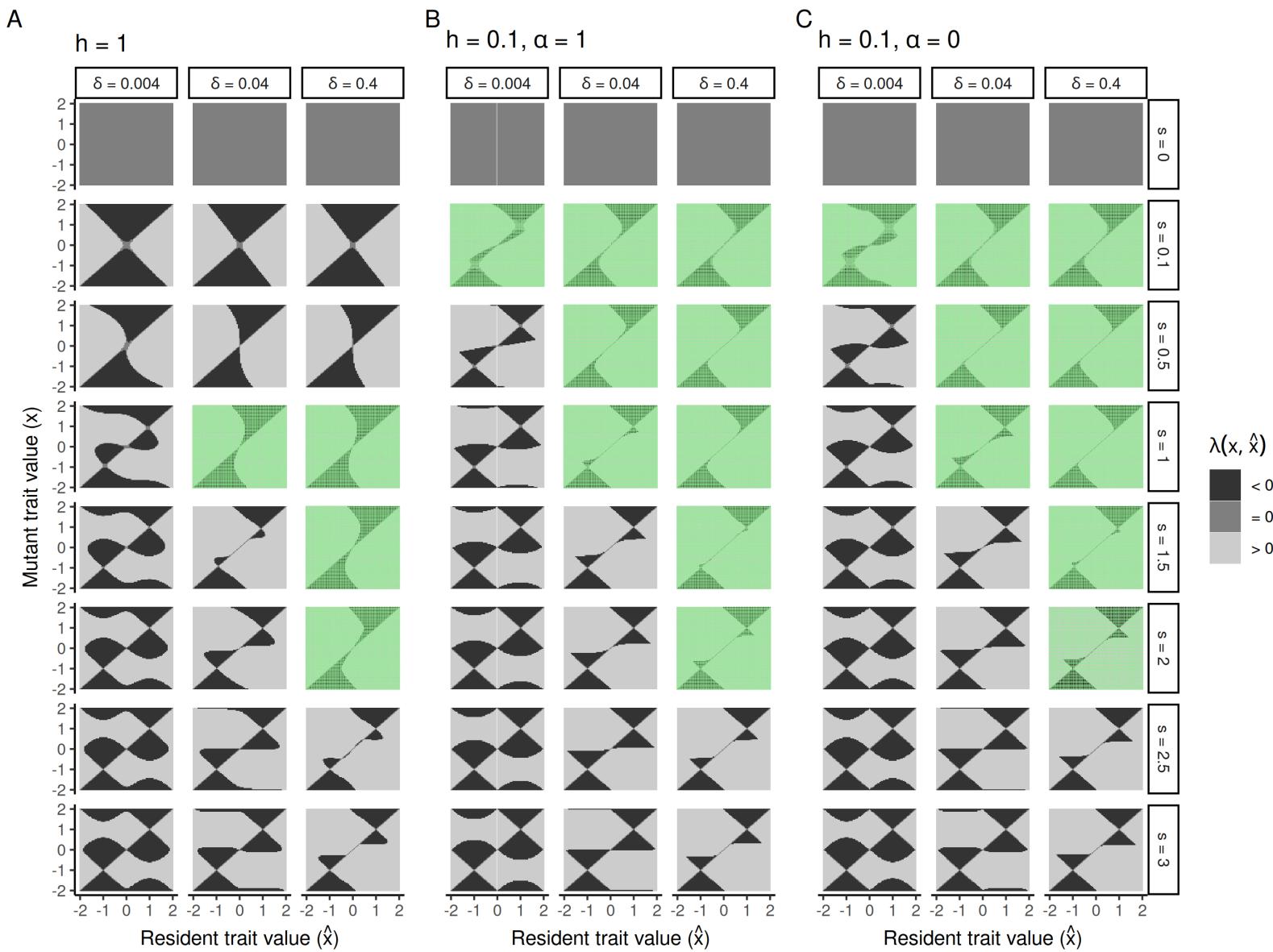


Figure S2: (See next page.)

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  $\alpha$  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.

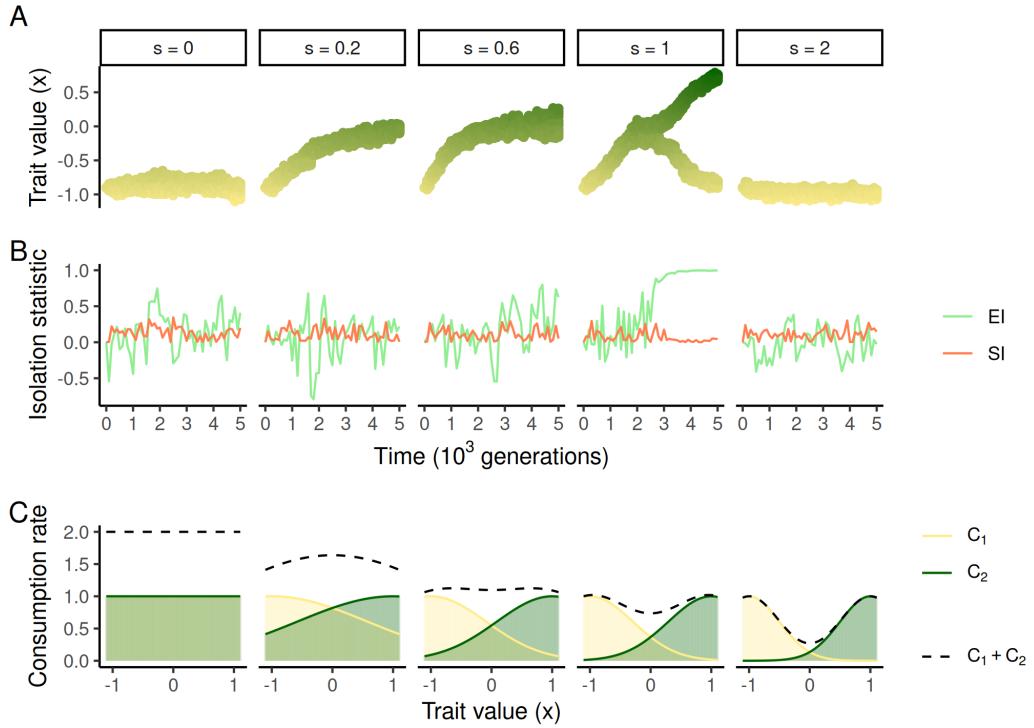


Figure S3: Simulations across trade-off strengths  $s$  in symmetric habitats ( $h = 1$ ) and under random choice ( $\beta = 0$ ). Other parameters as per Table 1. (A) Individual traits through time. Evolutionary branching occurs within a specific range of trade-off strengths — a generalist ( $x \approx 0$ ) 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  $x = 0$ .

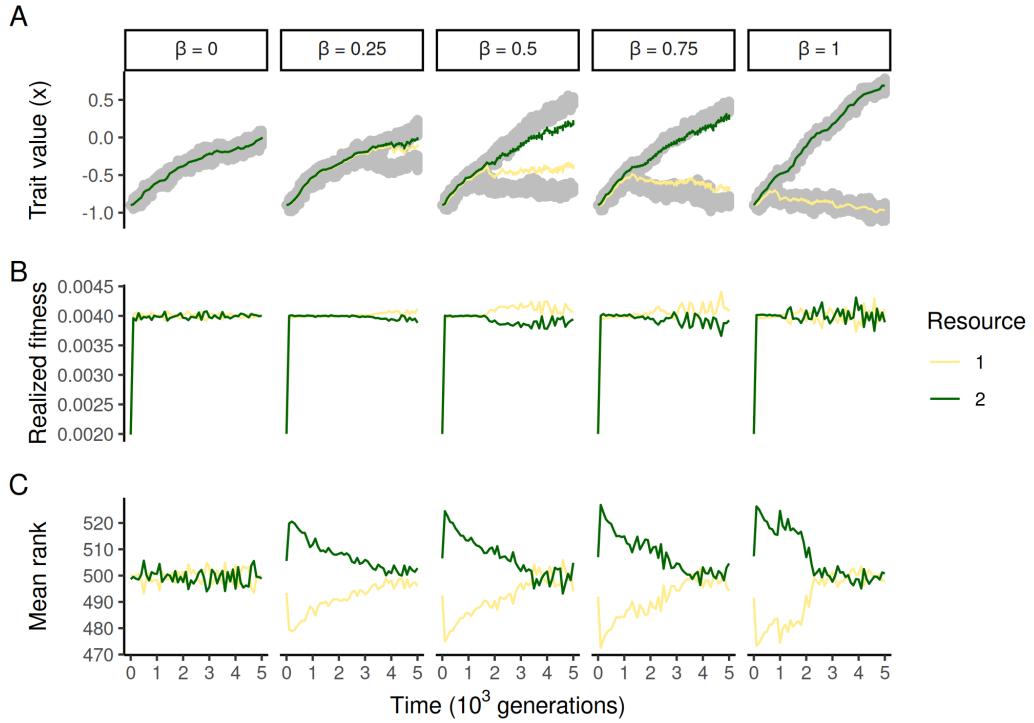


Figure S4: Simulations across resource assessment accuracies  $\beta$  in symmetric habitats ( $h = 1$ ) when the trade-off is weak ( $s = 0.2$ ). Other parameters as per Table 1. (A) Individual trait values through time. Colored lines represent the mean trait value of individuals choosing each resource at each generation. As choosiness  $\beta$  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  $\beta$ , and before branching, resource 2 tends to be chosen later in the queue, when resource 1 has been depleted — a pattern indicating frequency dependence.

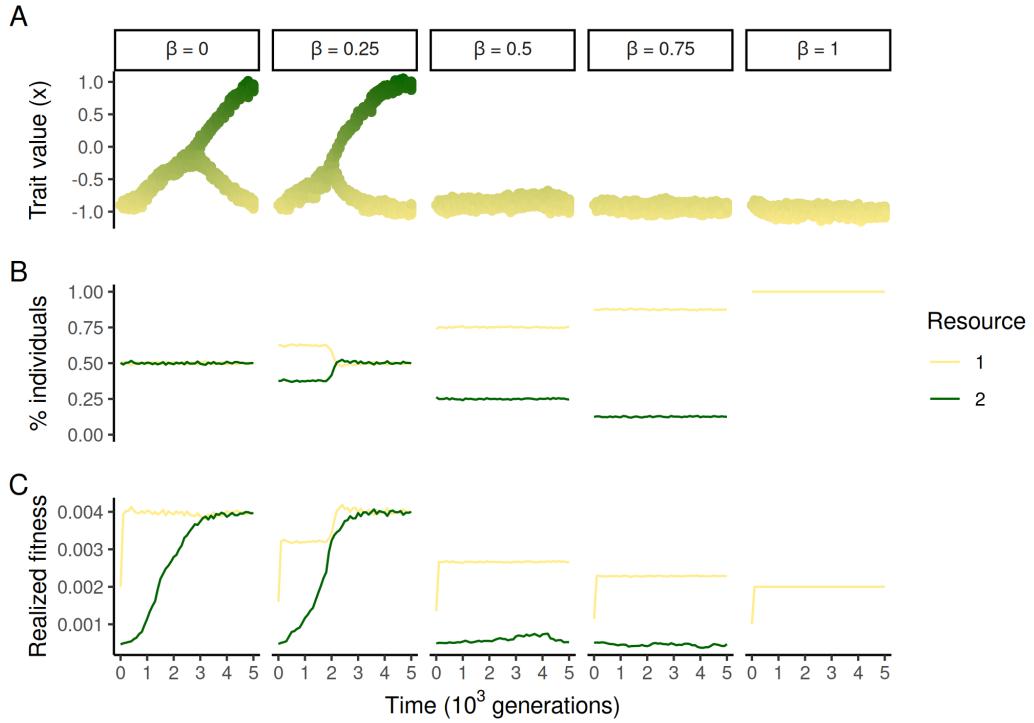


Figure S5: Simulations across resource assessment accuracies  $\beta$  in symmetric habitats ( $h = 1$ ) under a strong trade-off ( $s = 1.2$ ). Other parameters as per Table 1. (A) Individual trait values through time, showing that branching is lost as choosiness increases. (B) Proportion of individuals 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.

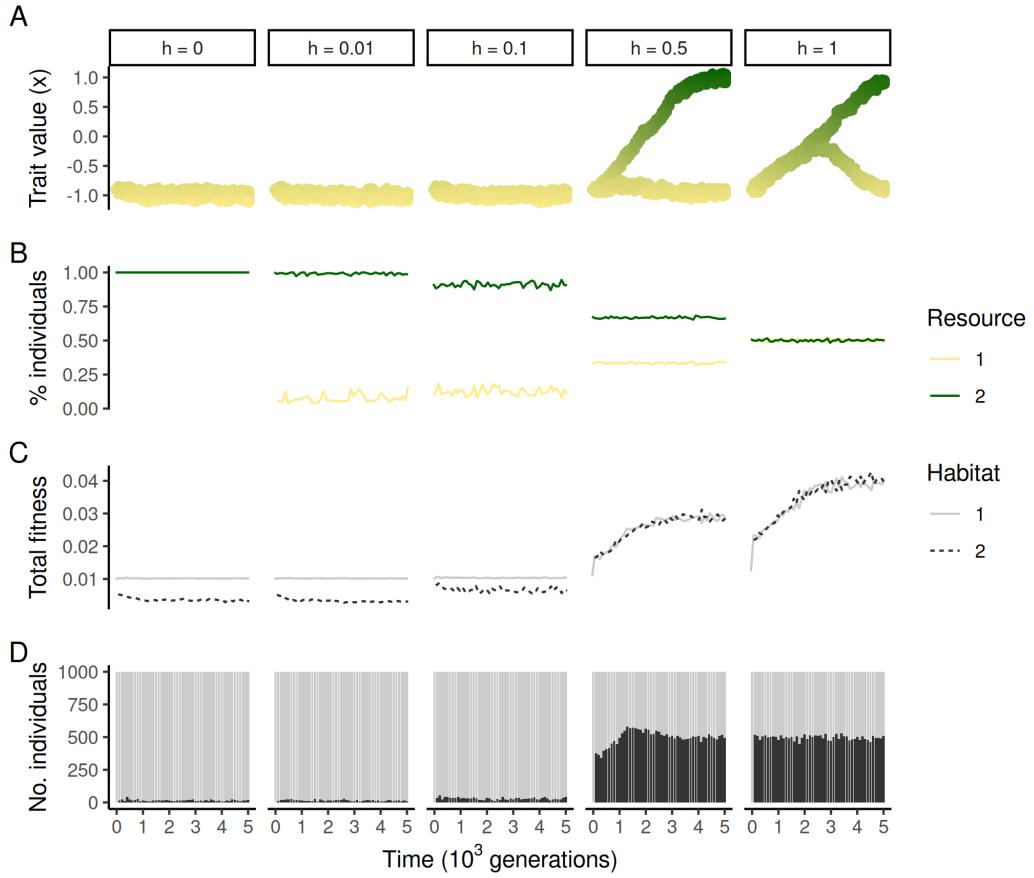


Figure S6: Simulations across habitat symmetry levels  $h$ , under a strong trade-off ( $s = 1.2$ ) and random choice ( $\beta = 0$ ). Other parameters as per Table 1. (A) Trait values through time. Branching is lost as habitat symmetry  $h$  decreases (right to left), because more asymmetric habitats 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.

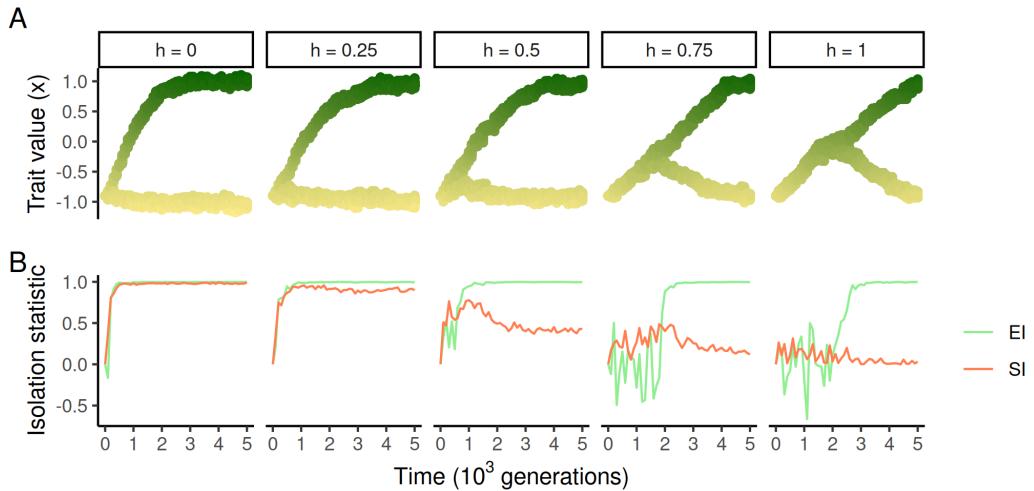


Figure S7: Simulations across habitat symmetry levels  $h$ , under moderate trade-off strength ( $s = 1$ ) and random choice ( $\beta = 0$ ). Other parameters as per Table 1. (A) Individual trait values through time. As habitat asymmetry increases ( $h$  decreases, right to left), branching becomes more 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.

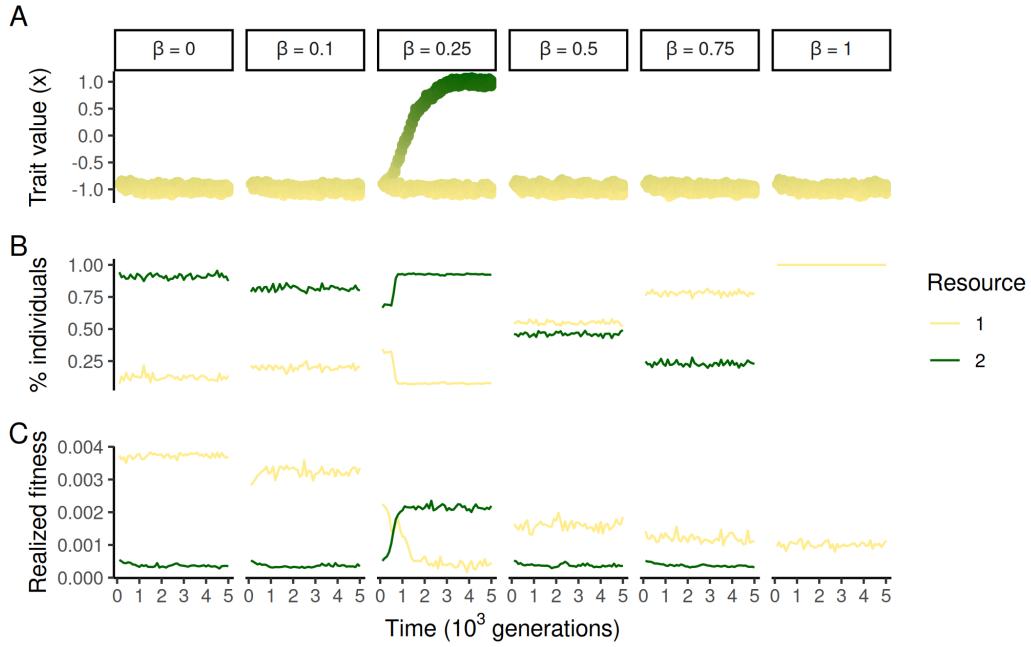


Figure S8: Simulations across resource assessment accuracies  $\beta$ , in asymmetric habitats ( $h = 0.1$ ) and under a strong trade-off ( $s = 1.2$ ). Other parameters as per Table 1. (A) Trait values through time. Under moderate resource asymmetry (i.e. with high-enough resource discovery,  $\delta = 0.04$ ), 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.

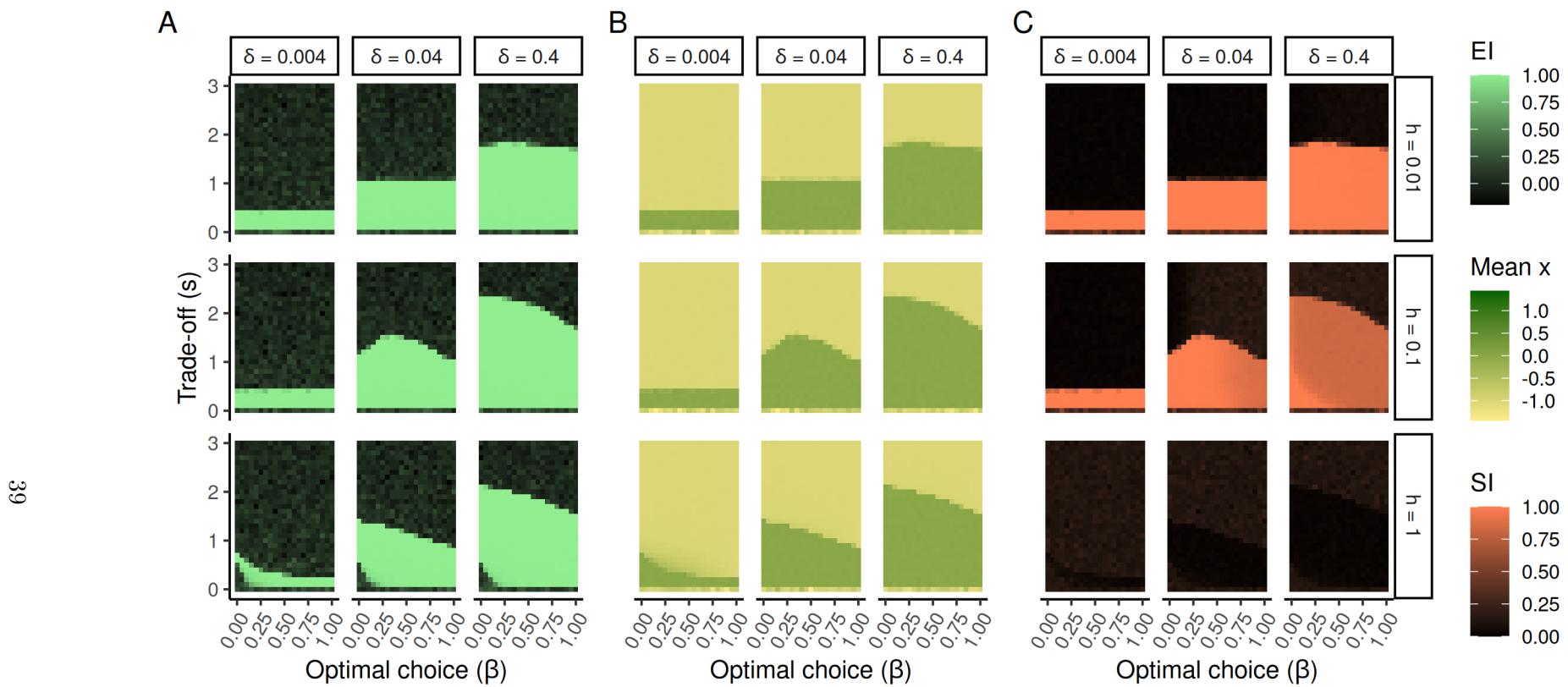


Figure S9: Simulation outcomes across the same parameter space as in Figure 3, but with a lower migration rate ( $m = 0.001$ ). Legend as per Figure 3. All simulations are run under  $m = 0.001$  except those under  $h = 1$ , which are the same as those in Figure 3 (i.e.  $m = 0.01$ ), and are here for 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  $h$ ).

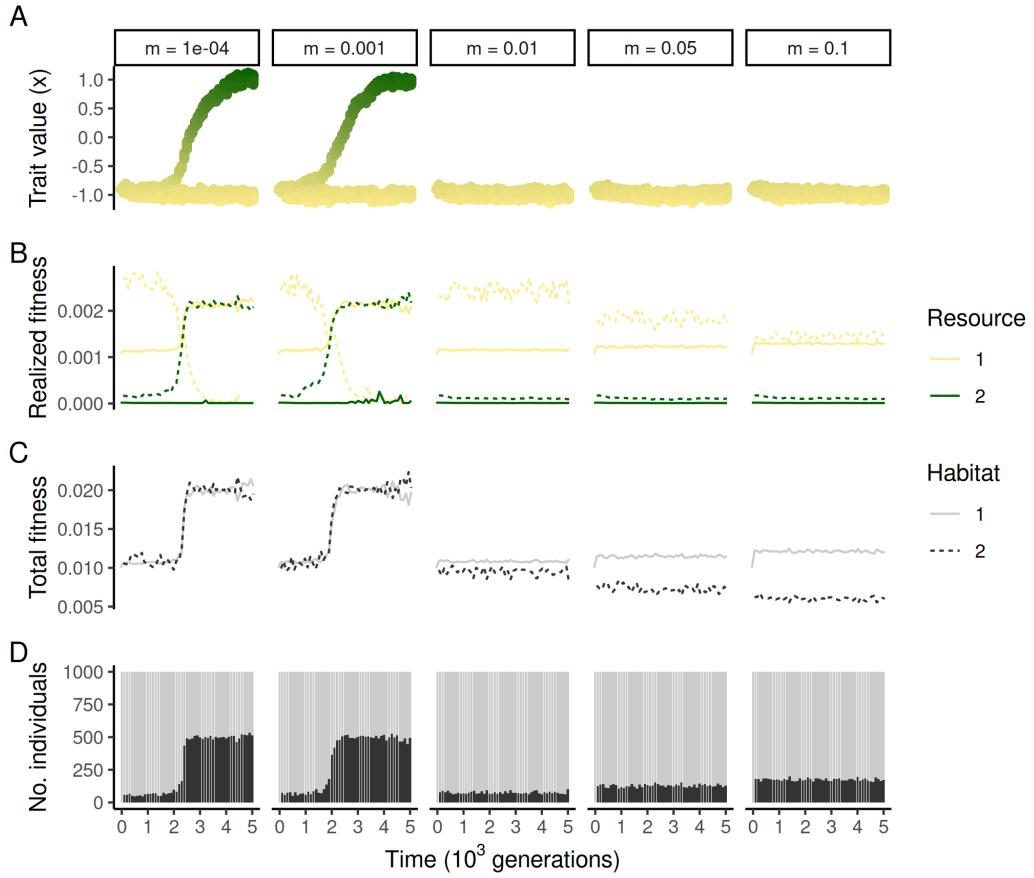


Figure S10: Simulations across migration rates  $m$  in asymmetric habitats ( $h = 0.1$ ), under a strong trade-off ( $s = 1.5$ ) and at intermediate resource assessment accuracy ( $\beta = 0.3$ ). Other parameters as per Table 1. (A) Traits through time. Intermediate choosiness maximizes the chances 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.

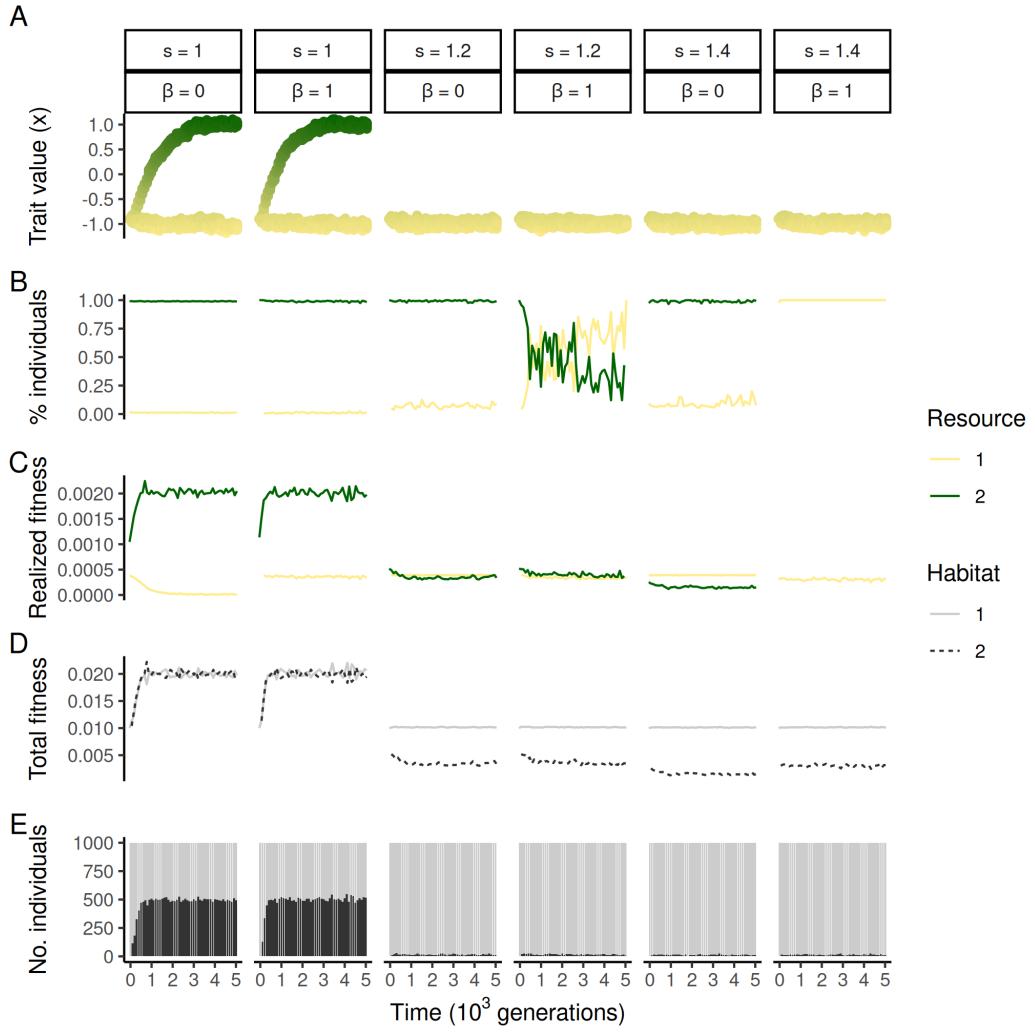


Figure S11: Simulations across trade-off strength  $s$  at two levels of choosiness ( $\beta = 0$  and  $\beta = 1$ ), in highly asymmetric habitats ( $h = 0.01$ ). Other parameters as per Table 1. (A) Traits through time. Trade-off strength affects both ends of the choosiness spectrum in a similar way — 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.

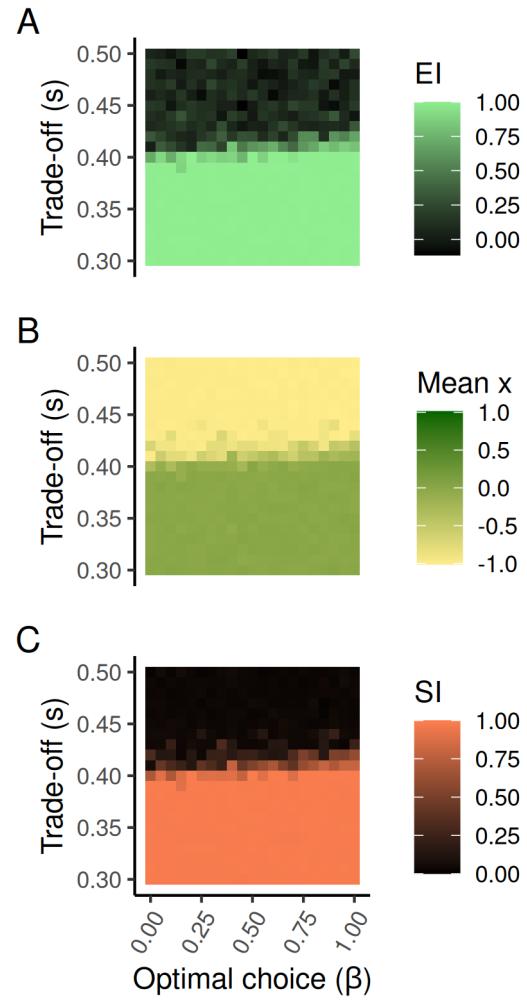


Figure S12: Zoom-in on a particular range of trade-off strengths  $s$  in the parameter space explored in Figure 3, when  $h = 0.1$  and  $\delta = 0.004$ . Legend as per Figure 3.

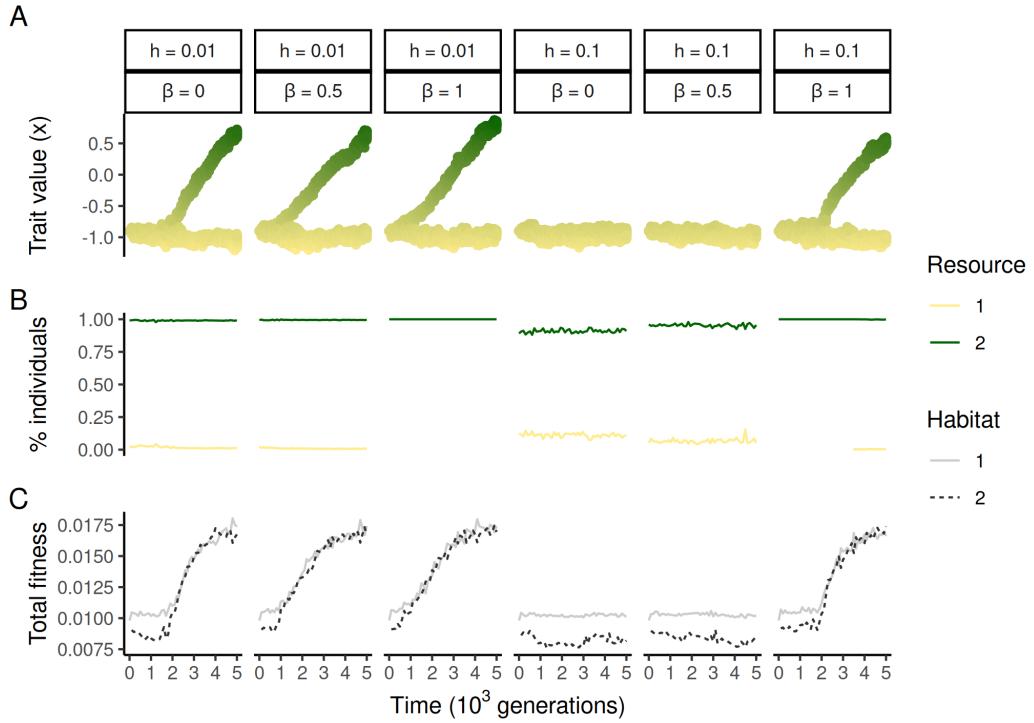


Figure S13: Simulations across levels of habitat symmetry  $h$  and resource assessment accuracies  $\beta$  under a weak trade-off ( $s = 0.4$ ) and at low resource discovery rate ( $\delta = 0.004$ ). Other parameters as per Table 1. (A) Traits through time. When resource discovery is low, branching becomes slightly more likely with higher choosiness  $\beta$  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  $\beta$ ) pick resource 2 than choosy individuals (high  $\beta$ ), 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,  $h = 0.01$ ).

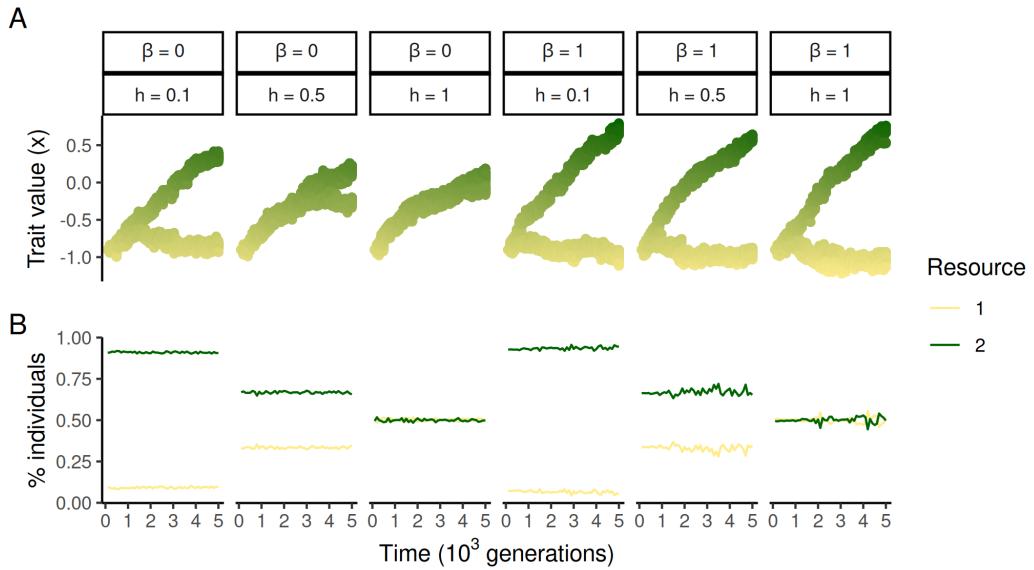


Figure S14: Simulations across levels of habitat symmetry  $h$  and resource assessment accuracies  $\beta$  under a weak trade-off ( $s = 0.2$ ). Other parameters as per Table 1. (A) Traits through time. Under random choice ( $\beta = 0$ ) a high habitat asymmetry (low  $h$ ) is needed for branching 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.

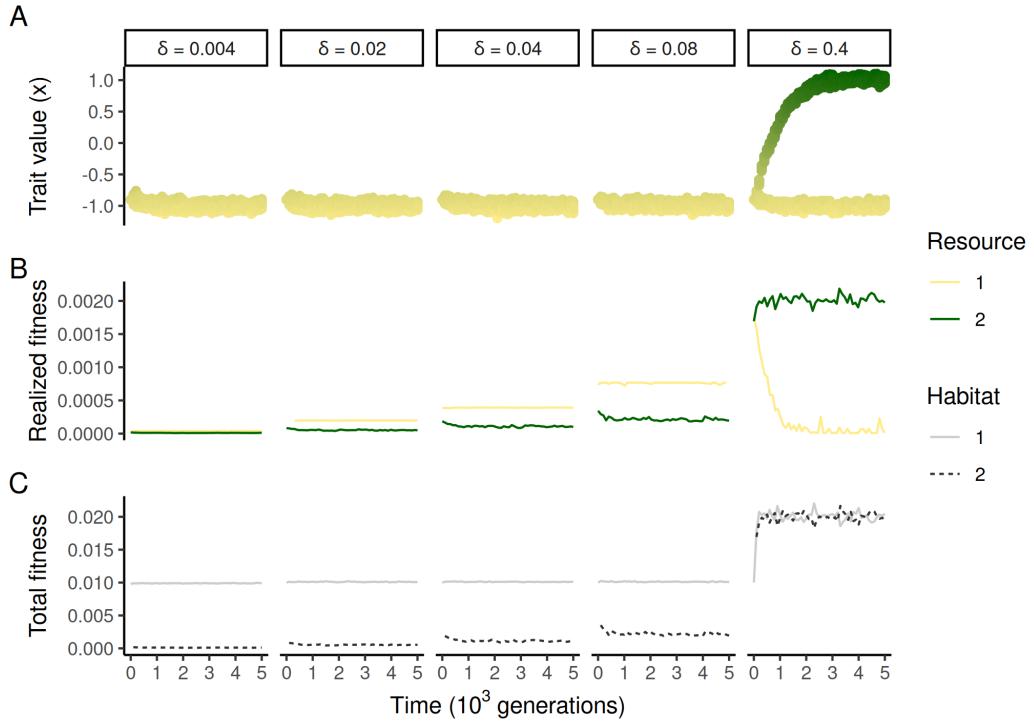


Figure S15: Simulations across resource discovery rates  $\delta$  in highly asymmetric habitats ( $h = 0.01$ ), under a strong trade-off ( $s = 1.5$ ) and random choice ( $\beta = 0$ ). Other parameters as per Table 1. (A) Traits through time. Generally, resource discovery facilitates branching by alleviating 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.

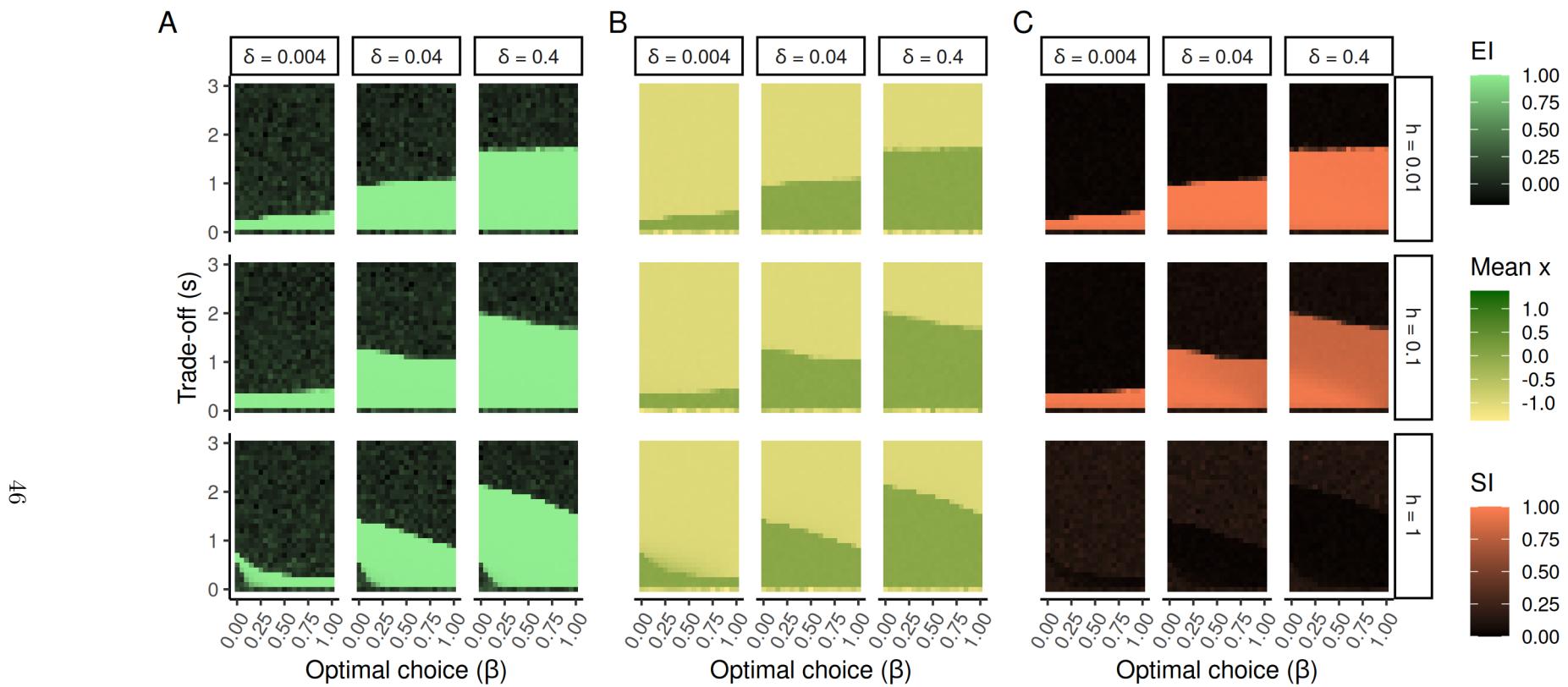


Figure S16: Simulation outcomes across the same parameter space as in Figure 3 but with low resource abundance weight ( $\alpha = 0$ ). Legend as per Figure 3. Note that simulations with  $h = 1$  shown here are the same as in Figure 3 (i.e.  $\alpha = 1$ ) because by construction the model under  $\alpha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

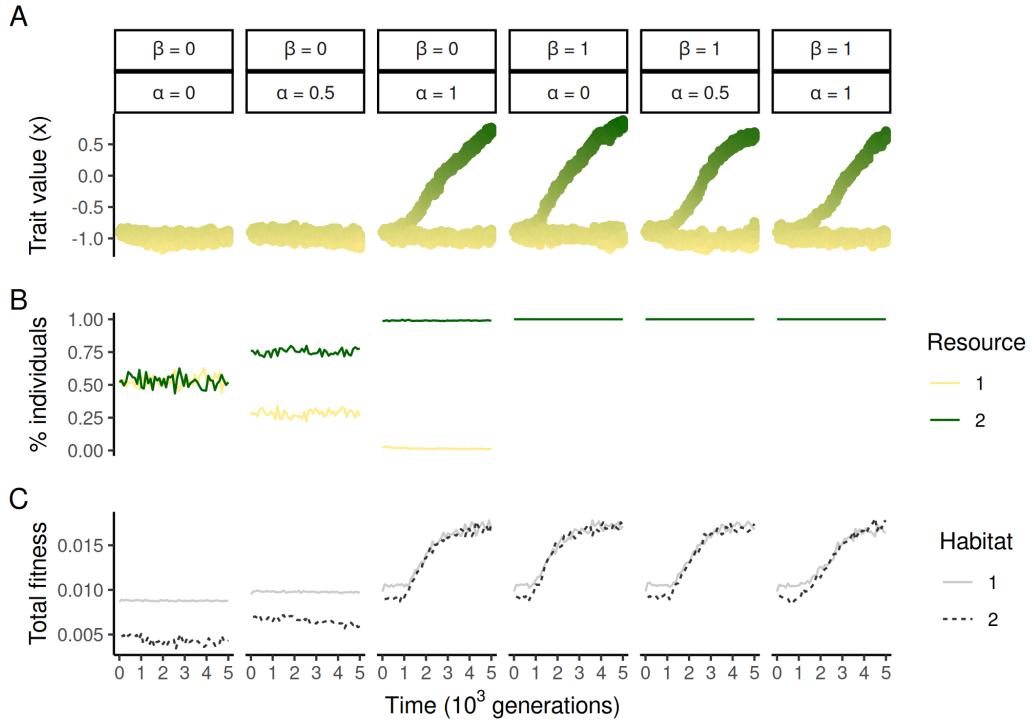


Figure S17: Simulations across resource abundance weights  $\alpha$  at two levels of resource assessment accuracy ( $\beta = 0$  and  $\beta = 1$ ), in highly asymmetric habitats ( $h = 0.01$ ), at a low resource discovery rate ( $\delta = 0.004$ ) and under a weak trade-off ( $s = 0.4$ ). Other parameters as per Table 1. (A) Traits through time. Decoupling resource abundance and encounter (lower  $\alpha$ ) reduces the probability of branching in less choosy individuals (low  $\beta$ ). (B) Proportion of individuals feeding on each resource at each generation in habitat 2. Lowering the resource abundance weight  $\alpha$  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).

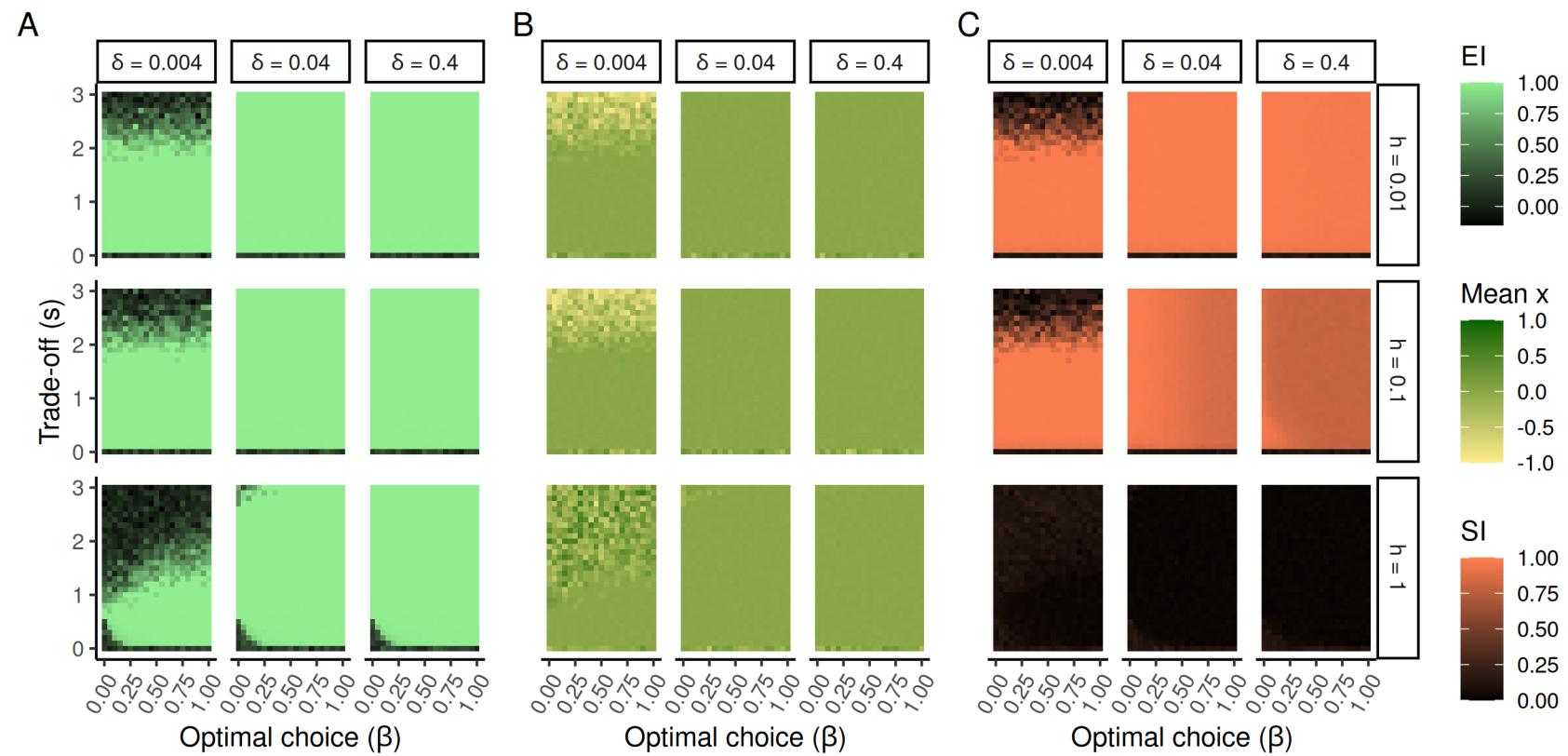


Figure S18: Simulation outcomes across the same parameter space as in Figure 3, but where the population starts as a generalist ( $x_0 = 0$ ). Legend as per Figure 3.

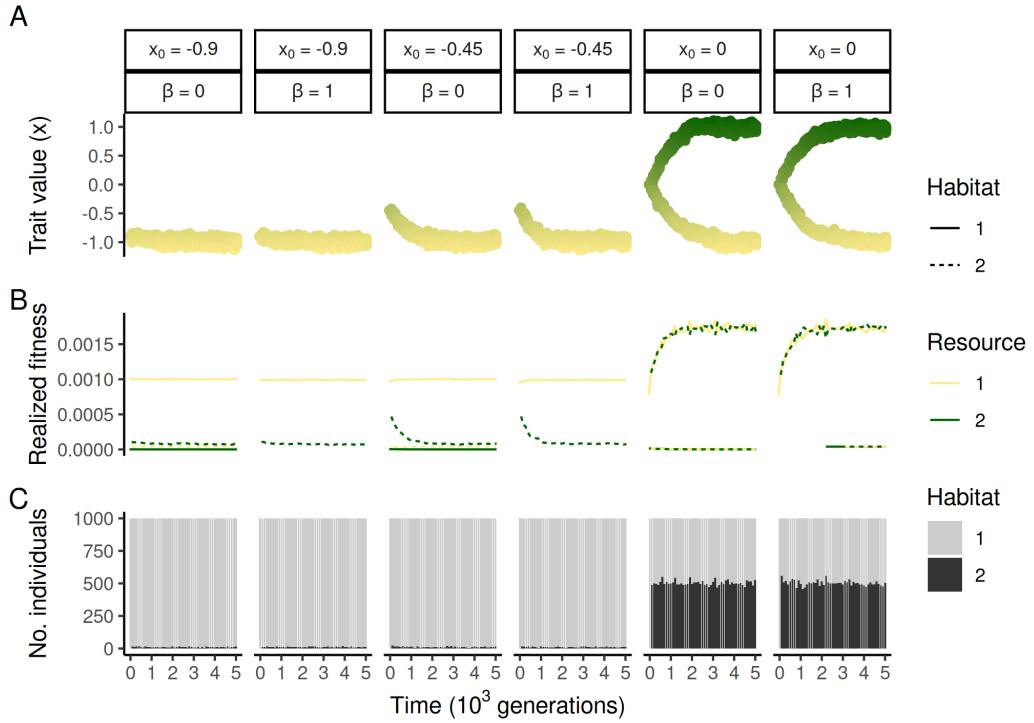


Figure S19: Simulations across starting points  $x_0$  at two resource assessment accuracies ( $\beta = 0$  and  $\beta = 1$ ), in highly asymmetric habitats ( $h = 0.01$ ), at a low resource discovery rate ( $\delta = 0.004$ ) and under a moderate trade-off ( $s = 1$ ). Other parameters as per Table 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.

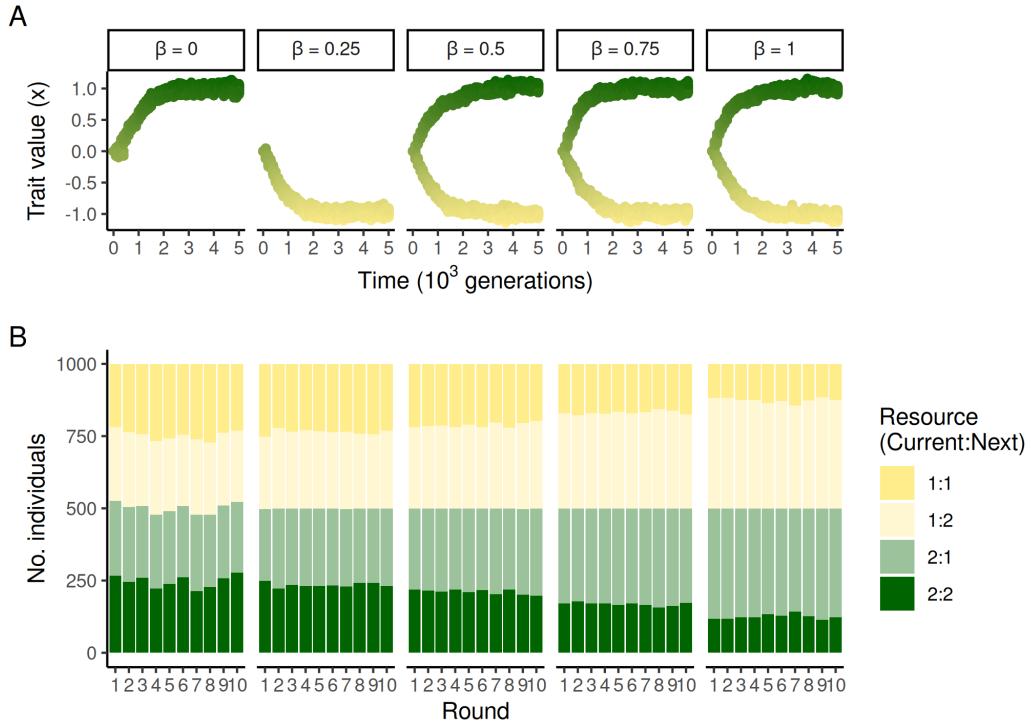


Figure S20: Simulations across resource assessment accuracies  $\beta$  when the population starts off as a generalist ( $x_0 = 0$ ), in symmetric habitats ( $h = 1$ ), at a low resource discovery rate ( $\delta = 0.004$ ) and under a strong trade-off ( $s = 1.5$ ). Other parameters as per Table 1. (A) Traits through time. When starting as a generalist, higher choosiness promotes branching where non-choosy individuals (low  $\beta$ ) 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  $\beta$  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 (Geritz et al., 1998; Metz et al., 1996). 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,

$$801 \quad W_j(x, \hat{x}) \propto \sum_{i=1}^2 P_{ij} \sum_{n=0}^{\hat{N}_j} W_{ij}(x, \hat{x}, n) \mathbb{P}(\hat{N}_{ij} = n) \quad (15)$$

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

$$807 \quad P_{ij} = 1/2(1 - \alpha) + \alpha R_{ij}/R_j^{\text{tot}} \quad (16)$$

807 where  $R_{ij}$  is the (untouched) total amount of resource  $i$  in habitat  $j$  (either 1 or  $h$ , the habi-  
808 tatt symmetry parameter, see Methods),  $R_j^{\text{tot}} = R_{1j} + R_{2j}$  is the total amount of resources in  
809 habitat  $j$ , and  $\alpha$  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

$$815 \quad \hat{N}_{ij} \sim \text{Binom}(\hat{N}_j, P_{ij}) \quad (17)$$

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

$$818 \quad \mathbb{P}(\hat{N}_{ij} = n) = \binom{\hat{N}_j}{n} P_{ij}^n (1 - P_{ij})^{\hat{N}_j - n}. \quad (18)$$

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

$$820 \quad W_{ij}(x, \hat{x}, n) = \begin{cases} R_{ij}^D(\hat{x}, n) C_i(x) / C_{ij}(\hat{x}, n) & \text{if } C_{ij}(\hat{x}, n) > 0 \\ 0 & \text{otherwise,} \end{cases} \quad (19)$$

820 where  $R_{ij}^D(\hat{x}, n)$  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

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

825 where  $\delta$  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

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

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

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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