

# Robustness vs productivity during evolutionary community assembly: short-term synergies and long-term trade-offs

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## 1 Abstract

2 The realization that evolutionary feedbacks need to be considered to fully grasp  
3 ecological dynamics has sparked recent interest in the effect of evolution on com-  
4 munity properties like coexistence and productivity. However, modern coexistence  
5 theory being limited to pairwise interactions, little is known about coevolution and  
6 diversification in rich communities. We leverage the recent multidimensional co-  
7 existence theory metrics, together with a structural community robustness metric,  
8 to study such properties in a general trait-based model of competition on a niche  
9 axis. We show that the effects of coevolution on coexistence are two-fold. In the  
10 short-term our results show synergies emerging between increasing productivity and  
11 reinforcing coexistence, while in the long-term, diversification and niche-packing  
12 destabilize communities, thus inducing a long-term trade-off between productivity  
13 and coexistence. In light of classical and recent work, our findings help advance  
14 understanding of evolutionary effects in high-dimensional systems. We illustrate  
15 how our theoretical predictions echo in observed empirical patterns. Finally, we  
16 discuss their implications and provide testable hypotheses.

17 *Keywords:* Eco-evolutionary dynamics, evolutionary stable communities, diver-  
18 sification, coexistence theory, structural stability, niche difference, fitness difference,  
19 productivity, Pareto optimality

20

## 1 Introduction

21 While ecology and evolutionary biology have long developed as separate disciplines (Lewon-  
22 tin, 2003), recent years have seen an effort to better grasp the feedbacks that link vari-  
23 ations in the biotic environment (inter- and intraspecific interactions) and evolutionary  
24 trajectories (McPeek, 2017). The realization that evolutionary changes can happen over  
25 relatively short time-scales (Lankau and Strauss, 2007; Fussmann et al., 2007; Yoshida  
26 et al., 2003; Hairston Jr et al., 2005) fostered a renewed interest in the interface be-  
27 tween ecology and evolution, an area of research now named eco-evolutionary dynamics  
28 (Hendry, 2009). The feedback loop between ecological interactions and evolutionary  
29 change in characters can give rise to complex dynamics beyond the simple optimization  
30 of growth rates, such as disruptive selection and evolutionary branching (Geritz et al.,  
31 1998).

32 Understanding the effects of evolution on the maintenance of species diversity remains  
33 an open question. At the single species level, evolutionary adaptation may save species  
34 from extinction under specific conditions, a phenomenon named evolutionary rescue (Bell  
35 and Gonzalez, 2009). But accounting for density- or frequency-dependent selection can  
36 open up scenarios where a species evolve towards self-extinction (Matsuda and Abrams,  
37 1994), a case dubbed evolutionary suicide in the adaptive dynamics literature (Ferriere  
38 and Legendre, 2013). In a two species scenarios, there can be both evolution towards  
39 stronger niche differentiation (Lankau, 2009) or one species can push the other to extinc-  
40 tion (Dercole et al., 2006). Recently, Pastore et al. (2021) studied the effect of co-evolution  
41 of niche positions in a model of two competing species in the framework of Chesson’s co-  
42 existence theory (Chesson, 2000; Barabás et al., 2018a). Their work shows that evolution  
43 tends to have a negative effect on coexistence by increasing competitive imbalance, an  
44 outcome matching the experimental results of Hart et al. (2019).

45 In communities, we have no reason to expect that evolution, which is driven by dif-  
46 ferences in individual fitness, will lead to optimization of emergent properties at a larger  
47 organizational scale (here, communities) (Metz et al., 2008). Brännström et al. (2012)

48 highlighted the dual nature of evolution in rich competitive communities, which can lead  
49 to the increase of diversity at different levels through generation of polymorphism and  
50 speciation but also to competitive exclusion and evolutionary murder. Classical ecological  
51 theory also predicts that stability in arbitrary large systems may be difficult: in randomly  
52 interacting communities local dynamical stability decreased with richness (May, 1972),  
53 but adding non-random interactions such as adaptive foraging or eco-evo dynamics can  
54 counter this effect (Kondoh, 2003). Loeuille (2010) showed that evolution stabilizes mod-  
55 erately rich communities but destabilizes more species-rich systems.

56 Importantly, dynamical stability - the return to equilibrium after a state perturbation  
57 - is only half the picture (Rohr et al., 2014), the other half being feasibility - the exis-  
58 tence of an equilibrium state where all species have positive abundances. The structural  
59 framework developed by Saavedra et al. (2017) allows us to quantify the tolerance to per-  
60 turbations in the parameter space (such as changes in growth rates due to environmental  
61 factors). The study of feasibility of competing species on a niche axis is tightly linked to  
62 what classical theory calls species packing, a concept which dates back to Hutchinson's  
63 work on species niches (Hutchinson, 1957, 1979) and was mathematically formalized by  
64 MacArthur (1969). In their paper, they studied the limit in similarity between species,  
65 i.e., how close species can be on the niche axis while still coexisting. Case (1981) extended  
66 this approach by including coevolutionary forces, and studied limiting similarity under  
67 the constraint imposed by evolution. To our knowledge, a study of the evolutionary effect  
68 on structural stability has never been proposed to this day.

69 If species interactions are important drivers of evolutionary change we expect that  
70 communities arising from co-evolution, for which Edwards et al. (2018) suggested the term  
71 Evolutionary Stable Communities (ESCs), should comprise a highly non-random subset  
72 of all possible combinations of species. This has been advocated already by Rummel and  
73 Roughgarden (1983), which analyzed island communities modeled through coevolution  
74 processes versus random colonization. More recently, Aubree et al. (2020) showed that  
75 coevolved communities were generally more productive, more stable, more resistant to

76 invasion than collections which were randomly assembled from the species pool.

77 Here, we aim to study how evolution and the emergence of polymorphism interplay  
78 with coexistence and productivity constraints. Using the structural stability approach to  
79 coexistence theory (Saavedra et al., 2017), we expand on previous approaches based on  
80 species pairs and dynamical stability, to focus on multispecies co-evolutionary commu-  
81 nities. Our aim is to go beyond previous works by 1) explicitly linking eco-evolutionary  
82 dynamics and structural coexistence metrics; 2) discussing how these links vary in con-  
83 trasting ways depending on the time scale that is considered; 3) Uncovering and explaining  
84 the emergence of positive or negative correlations among various community properties,  
85 particularly diversity, productivity and coexistence, along evolutionary trajectories.

86 To do so, we follow the structural indicators of niche and fitness difference. For a  
87 given niche difference, small fitness differences increase robustness through equalizing  
88 effects; similarly, for a given fitness difference, niche differences increase robustness due  
89 to stabilizing effects. However, as is widely recognized, these two metrics do not exist  
90 independently (Barabás et al., 2018b; Song et al., 2019), and we find they are insufficient  
91 to paint a complete picture of extinction risks. Therefore, we make use of a structural  
92 metric which quantifies the community robustness in the face of perturbations that would  
93 cause loss of one or more of its constituent species (Medeiros et al., 2021). We track  
94 all three metrics along evolutionary trajectories. Finally, we consider changes in total  
95 community productivity, and contrast it with measures of coexistence.

96 We expect evolution to cause character displacement along the resource axis (Macarthur  
97 and Levins, 1967; Dieckmann and Doebeli, 1999; Grant and Grant, 2006), which should  
98 first result in an increase of niche differentiation. At the same time, divergence of niche  
99 positions ultimately results in a decrease in the growth rate of phenotypes situated further  
100 from the resource optimum, and in a larger imbalance of growth rates in the community  
101 and greater fitness differences. We therefore expect that effects of evolution on coex-  
102 istence and productivity may vary depending on the scale (environmental width) and  
103 timeframe considered.

104

## 2 Materials and methods

105

### 2.1 Ecological dynamics

106 We study a niche-based model of competition based on generalized Lotka-Volterra dy-  
107 namics (Volterra, 1931), following the  $\alpha$ - $r$  parametrization (Mallet, 2012). Each phe-  
108 notype is defined by its position  $\mu_i$  on a niche axis representing resources. Position on  
109 the niche axis affects both the intrinsic growth rates  $r(\mu_i)$  and the strength  $\alpha(\mu_i, \mu_j)$  of  
110 density-dependent competition interactions between types. This kind of model has been  
111 commonly used to model competition on a niche axis (MacArthur, 1972), and has been  
112 shown to readily lead to the evolutionary emergence of polymorphism (Dieckmann and  
113 Doebeli, 1999). We use this behavior to explore the conditions of coexistence throughout  
114 the diversification process. Population dynamics of a type  $i$  then follow:

$$\frac{dN_i}{dt} = N_i \cdot \left( r(\mu_i) - \sum_{j=1}^S \alpha(\mu_i, \mu_j) \cdot N_j \right) \quad i = 1, \dots, S. \quad (1)$$

115 We assume a Gaussian function for  $r(\mu_i)$ , with fecundity decreasing with distance from  
116 a resource optimum  $\mu_R$ , while including a small density-independent intrinsic mortality  
117  $m$  to prevent phenotypes with unreasonably large or small trait values from persisting at  
118 virtually nil abundances:

$$r(\mu_i) = f_{\max} e^{-\frac{(\mu_i - \mu_R)^2}{2\sigma_R^2}} - m. \quad (2)$$

119 The parameter  $f_{\max}$  represents the maximum fecundity rate at the resource optimum,  
120 while  $\sigma_R$  depicts the width of resources availability on the niche axis.

121 In line with previous works (MacArthur and Levins, 1967), we suppose that com-  
122 petition strength is defined by the similarity among types. It then follows a Gaussian  
123 function centered in  $\mu_i$ , so that  $\alpha(\mu_i, \mu_j) = \alpha(\mu_j, \mu_i)$ , and interaction strength reaches its  
124 maximum  $\alpha_{\max}$  for  $\mu_i = \mu_j$ :

$$\alpha(\mu_i, \mu_j) = \alpha_{\max} e^{-\frac{(\mu_i - \mu_j)^2}{2\sigma_{\alpha}^2}}. \quad (3)$$

125 The parameter  $\sigma_{\alpha}$  controls the width of the niche. This model is conceptually a size  $S$

126 extension of the model used by Taper and Case (1992); Dieckmann and Doebeli (1999);  
 127 Doebeli and Dieckmann (2000). The form of the competition function presents the agree-  
 128 able property of being dissipative *sensu* Volterra (Volterra, 1931; Logofet, 1994). This  
 129 implies that ecological dynamics possess one and only one globally stable equilibrium  
 130 point. Moreover, if there exists a feasible equilibrium, i.e.,  $N_i^* > 0$  for all  $i$ , all ecolog-  
 131 ical dynamics will converge to it regardless of initial abundances. Note that a feasible  
 132 equilibrium must fulfill  $r(\mu_i) = \sum_{j=1}^S \alpha(\mu_i, \mu_j) \cdot N_j^*$  for each  $i$ .

133 *2.2 Evolutionary dynamics*

134 We study evolution within the adaptive dynamics framework (Metz et al., 1995; Bränström  
 135 et al., 2013), thereby accounting for both frequency- and density-dependent selection. The  
 136 adaptive dynamics framework assumes clonal reproduction where mutations are infinites-  
 137 imally small and rare, and a separation of timescales where ecology is assumed faster than  
 138 evolution. This implies that advantageous mutations always go to fixation. The evolution  
 139 of a quantitative trait is determined by the invasion fitness function, defined as the *per*  
 140 *capita* growth rate of a rare mutant of traits  $\mu'_i$  in a resident population of  $S$  phenotypes  
 141 with traits  $\mu_1, \dots, \mu_S$  at their ecological equilibrium (Metz et al., 1992):

$$\omega_i(\mu'_i | \mu_1, \dots, \mu_S) = r(\mu'_i) - \sum_{j=1}^S \alpha(\mu'_i, \mu_j) \cdot N_j^*. \quad (4)$$

142 Note, that each phenotype  $i$  has its invasion fitness function  $\omega_i$ .

143 The evolution over time of niche positions is determined by the canonical equation of  
 144 adaptive dynamics (Dieckmann and Law, 1996):

$$\frac{d\mu_i}{dt} = c \cdot N_i^* \cdot \frac{\partial \omega_i(\mu'_i | \mu_1, \dots, \mu_S)}{\partial \mu'_i} \bigg|_{\mu'_i = \mu_i} \quad i = 1, \dots, S. \quad (5)$$

145 The partial derivatives

$$\frac{\partial \omega_i(\mu'_i | \mu_1, \dots, \mu_S)}{\partial \mu'_i} \bigg|_{\mu'_i = \mu_i} \quad i = 1, \dots, S \quad (6)$$

146 correspond to selection gradients, and they determine the direction of co-evolution of the

147  $S$  niche positions. The coefficient  $c$  incorporates the mutation rate and the mutational  
148 variance, which we assume to be equal for all phenotypes. Without loss of generality, we  
149 can set  $c = 1$  by rescaling the time axis. We integrate the canonical equation (Equ. 5)  
150 numerically until an evolutionary singular strategy is reached, i.e, a set of traits  $\mu_1^*, \dots, \mu_S^*$ ,  
151 at which all selection gradients vanish

$$\frac{\partial \omega_i(\mu'_i | \mu_1, \dots, \mu_S)}{\partial \mu'_i} \bigg|_{\mu'_i = \mu_i = \mu_i^*} = 0 \quad i = 1, \dots, S. \quad (7)$$

152 We start with a single monomorphic population and once we reach an evolutionary  
153 singular strategy, we evaluate the evolutionary stability condition (invasibility of the  
154 strategy) for each phenotype (Brännström et al., 2013; Metz et al., 1995). Invasibility  
155 happens when the second derivative of the invasion fitness function (Equ. 4) with the  
156 mutant trait is positive at the singular strategy (Dieckmann and Law, 1996; Geritz et al.,  
157 1998). Evolutionary branching ensues, and we introduce a new phenotype at a small  
158 distance of the singularity, thereby augmenting the dimensionality of the system. We  
159 continue the simulation until all strategies are evolutionary stable, thereby forming an  
160 Evolutionary Stable Community (ESC) *sensu* Edwards et al. (2018). The algorithm is  
161 detailed in the Supplementary Materials S1.

### 162 2.3 Classical coexistence metrics

163 Along co-evolutionary trajectories where  $S > 2$ , we assess coexistence metrics. Within  
164 the structural approach, coexistence is quantified by structural niche differences  $\Omega$  and  
165 structural fitness differences  $\theta$ . These two metrics are extensions to species-rich systems of  
166 the stabilizing niche difference and the fitness ratio, defined within the modern coexistence  
167 theory framework (Chesson, 2000), and were introduced in their multispecies form by  
168 Saavedra et al. (2017). The niche difference  $\Omega$  quantifies the possibility of coexistence and  
169 is mathematically defined as the solid angle of the domain of intrinsic growth rates leading  
170 to coexistence given the interaction strength  $\alpha(\mu_i, \mu_j)$ , defining the feasibility domain  
171 (Figure 1). In turn, the fitness difference  $\theta$  quantifies the deviation from neutrality, i.e.,

172 to what extent a phenotype dominates the system. The fitness difference  $\theta$  depends on  
 173 both the intrinsic growth rates vector  $\mathbf{r}(\mu_i)$  and the competition strength  $\alpha(\mu_i, \mu_j)$ . To  
 174 compare values of niche difference across communities of different sizes  $S$ , we consider  
 175 the standardized niche difference  $\hat{\Omega} = \sqrt[s]{\Omega}$  (Song et al., 2018).

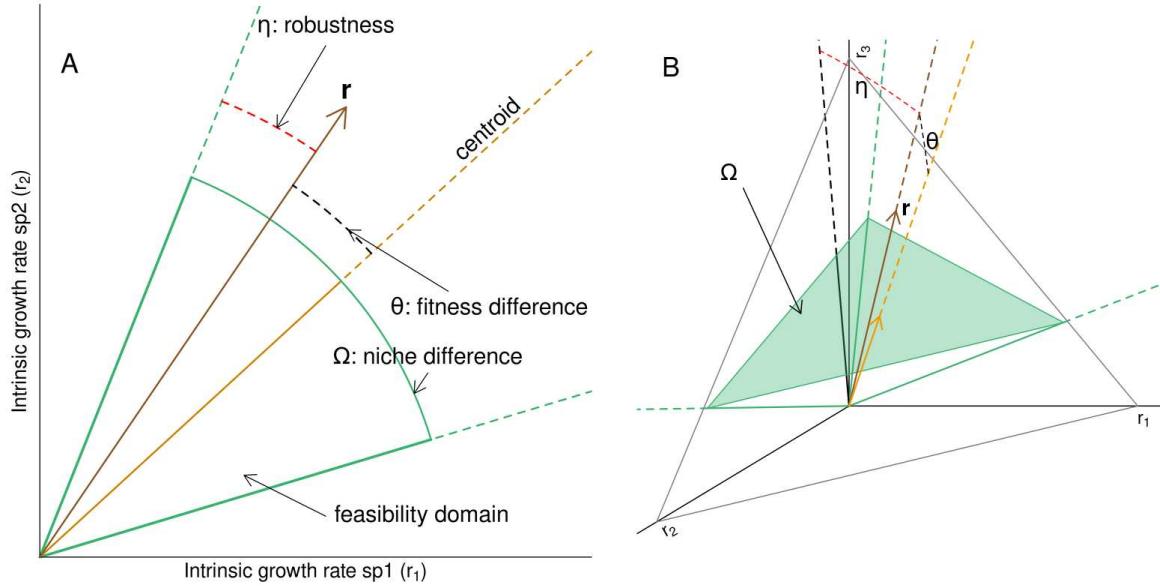


Figure 1: Geometric representation of the structural coexistence metrics for a 2-species system (panel A) and a 3-species system (panel B). On both panels, the green cone, called domain of feasibility, determines the set of intrinsic growth rates leading to a positive equilibrium for each species, i.e., to coexistence. The niche difference  $\Omega$  is given by the amplitude of this cone, while the fitness difference  $\theta$  measures the deviation of the vector of intrinsic growth rates ( $\mathbf{r}$ ) from the centroid of the cone. Finally,  $\eta$  – structural measure of robustness – is given by the smallest angle between the vector of intrinsic growth rates and the border of the feasibility domain. It gives an indication of the fragility of the community with respect to perturbations on  $\mathbf{r}$ . Note that we illustrated the structural metrics for  $S = 2$  and  $S = 3$ , but they are effectively defined and computed for  $S$ -rich systems (though they become hard to represent).

176

## 2.4 Structural robustness metric

177 While the classical metrics allow for partitioning coexistence effects, they do not give  
 178 a clear indication of the absolute strength of coexistence. For example, if both  $\Omega$  and  
 179  $\theta$  increase, we do not know whether coexistence is favored or weakened. Hence, we  
 180 utilize a complementary measure which we call  $\eta$ , the smallest angle between the vector  
 181 of intrinsic growth rates  $\mathbf{r}$  and the border of the feasibility domain. This angle, which

182 we call “robustness”, gives an indication of the fragility of the community with respect  
183 to perturbations on  $\mathbf{r}$ , i.e., how close the community is to losing of one or more of its  
184 components. Figure 1 provides a representation of  $\Omega$ ,  $\theta$ , and  $\eta$  in a system of  $S = 2$ . It  
185 is worth noting that a version of this metric was used in Medeiros et al. (2021), where  
186 it is called “full resistance”. The mathematics of how  $\eta$  is computed are detailed in the  
187 Supplementary Materials S2.

188 *2.5 Productivity metrics*

189 Besides coexistence metrics, we also measure the evolution of productivity using the  
190 proxy of total community biomass (or abundance)  $N_{\text{tot}} = \sum_i N_i^*$  is (Tilman et al., 1997).  
191 Biomass diversity effects (overyielding of polymorphic communities compared to mono-  
192 cultures) can be due to complementarity in resource use, as well as selection for high-  
193 yield phenotypes (high carrying capacity  $K_i = r_i/\alpha_{ii}$ ). Both of these mechanisms can be  
194 easily captured under our model, and are expected to vary along coevolutionary time.  
195 Furthermore, we follow relative yield  $Y_{\text{tot}} = (\sum_i N_i^*/K_i)$ , another common currency of  
196 biodiversity-productivity studies Vandermeer (1989); Loreau and Hector (2001).

197 *2.6 Randomizations*

198 We follow the coexistence metrics and productivity along co-evolutionary trajectories and  
199 branching points. However, this does not tell us whether those properties are maximized  
200 by evolution. To this end, we generate between 4'000 and 128'000 (depending on  $S$ )  
201 communities for each singular strategy (branching point or stable strategy) of each sim-  
202 ulation, by sampling uniformly sets of niche positions  $\mu_i$ , conditioned on the resulting  
203 community being feasible ( $N_i^* > 0$  for all  $i = 1, \dots, S$ ). We then compare the metrics  
204 (niche and fitness difference and total biomass) of the sampled communities with the eco-  
205 evolutionary trajectories. We use two different rules to define the range of niche positions  
206 from which we sample. For the first rule, we sample communities within the range of  
207 niche positions observed at the ESC (restricted range), which tells us if the evolutionary  
208 community is unique within this bracket of niche positions. For the second rule (full

209 range), the range of sampled  $\mu_i$  covers all niche positions leading to positive intrinsic  
210 growth rate  $r(\mu_i) > 0$ . Such bounds are given by  $\mu_R \pm \sqrt{2\sigma_R^2(\log f_{\max} - \log m)}$  and allow  
211 us to compare strategies with respect to the whole trait space.

212 *2.7 Choice of parameter values*

213 To reduce the number of free parameters, we first transform the dynamical system into a  
214 nondimensional form. That is, we freely chose the time unit, the abundance unit, and the  
215 scale of the niche axis. Regarding the abundance and time unit, we can set without loss  
216 of generality  $\alpha_{\max} = 1$ ,  $f_{\max} = 1$ . The latter defines the ecological timescale, but adaptive  
217 dynamics assumes that ecological equilibrium is reached before the next mutation occurs.  
218 The combination of both determines the scale of species abundance and can arbitrarily  
219 be rescaled. The additional mortality rate  $m$  remains, and thus cannot be freely chosen.  
220 Regarding the niche axis, we can rescale, without loss of generality, its origin to  $\mu_R = 0$ .  
221 Then we can also choose its scale by setting arbitrarily the resource width  $\sigma_R = 1$  and  
222 explore the effect of the niche width  $\sigma_{\alpha}$ , or alternatively, we can also set arbitrarily the  
223 niche width  $\sigma_{\alpha} = 1$  and explore the effect of the resource width  $\sigma_R$ .

224 For simplicity, in the main text we show the results for one specific parametrization,  
225 set to  $m = 0.01$  and  $\sigma_{\alpha} = 1$ , and  $\sigma_R = 1.7$ . In the Supplementary Material S6 and S7,  
226 we show that our findings are robust relative to different choices of environmental niche  
227 width  $\sigma_R$  and mortality  $m$ .

228

### 3 Results

#### 229 3.1 Co-evolutionary trajectories and branching events

230 Figure (2.A) shows a co-evolutionary trajectory, which undergoes diversification events  
231 until it reaches a stable and convergent community (ESC). After each branching, we  
232 observe a divergence in the niche positions, which leads to a decrease in the level of in-  
233 terspecific competition. The evolutionary endpoint for the monomorphic situation (i.e.,  
234 before the first branching) can be determined analytically. It converges to an evolutionary  
235 singular strategy that is always located at the resource optimum  $\mu^* = \mu_R$ . Its invasibil-  
236 ity depends on the width of the competition function ( $\sigma_\alpha$ ), the width of the resources  
237 ( $\sigma_R$ ), the maximum fecundity rate ( $f_{\max}$ ), and the mortality rate ( $m$ ). We show that  
238 branching occurs if and only if  $\sigma_R^2 \cdot (f_{\max} - m) > \sigma_\alpha^2 \cdot f_{\max}$  (Supplementary Materials S3).  
239 Branching will therefore occur if the resource range is wide enough (as defined by  $\sigma_R$ )  
240 and/or limiting similarity strong enough (small  $\sigma_\alpha$ ). This formula is alike the one derived  
241 by Dieckmann and Doebeli (1999); Doebeli and Dieckmann (2000) but generalized to an  
242 additional mortality term, which has an evolutionary stabilizing effect through reducing  
243 the resident abundance at equilibrium. As expected by niche packing theory MacArthur  
244 and Levins (1967); MacArthur (1969), the number of subsequent branching, and there-  
245 fore, the number of species at the ESC, increases with the resource availability ( $\sigma_R$ ) and  
246 decreases with the competition width ( $\sigma_\alpha$ ), see Supplementary Material S4.

247 Figure 2B shows the evolution of the niche difference  $\hat{\Omega}$ . Between each branching point,  
248 the niche difference increases. This can be expected from panel A showing that niche  
249 positions diverge, which implies an increase in the niche difference and relaxed competi-  
250 tion among phenotypes. But at the branching points, the addition of a new phenotype  
251 very close to an existing one causes the average niche difference to abruptly decrease.  
252 Hence, for a period, coexistence is driven by more neutral (fitness-equivalent) mecha-  
253 nisms, rather than by strong niche separation. More specifically, branching points are  
254 the only contiguous trait regions where indefinitely close phenotypes can coexist, creat-

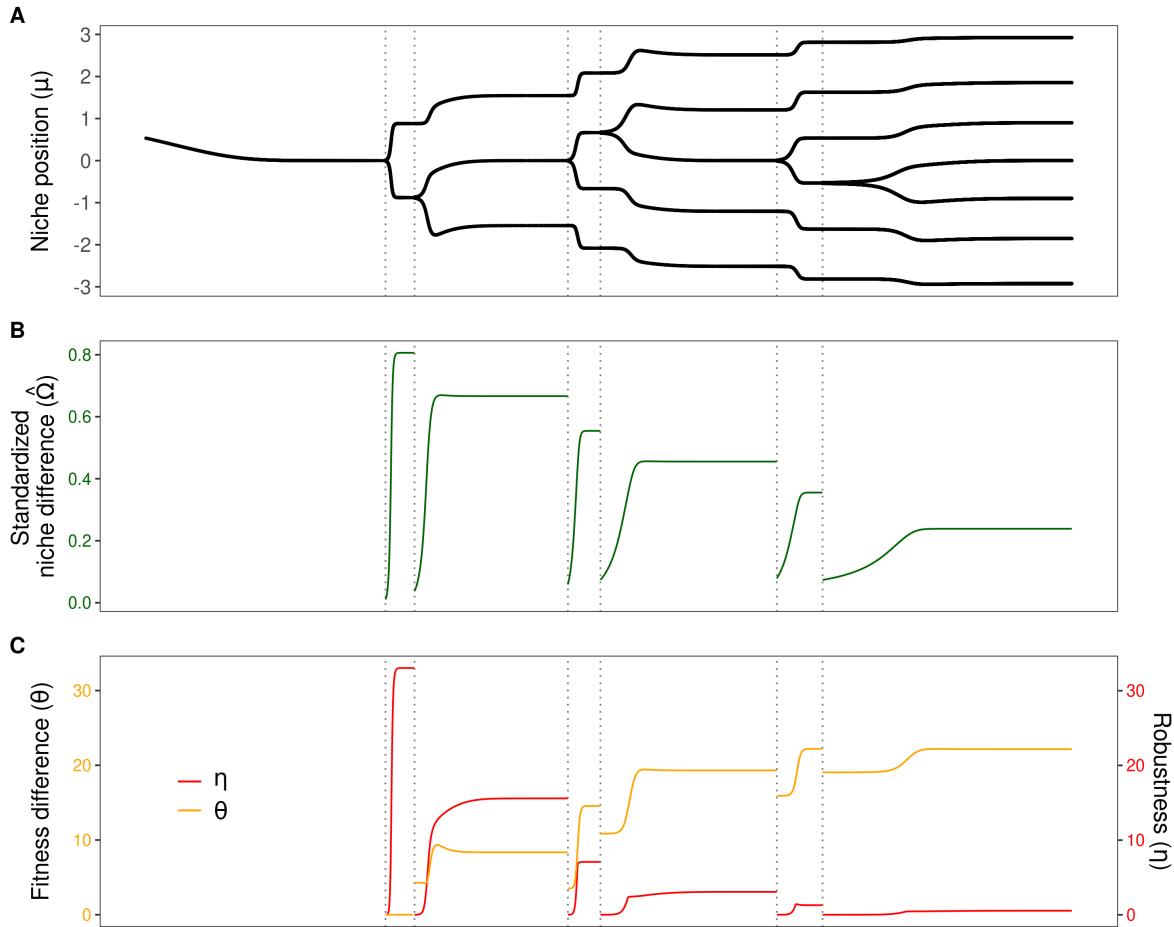


Figure 2: **Evolutionary trajectory.** Panel **A** shows the evolutionary trajectory of the niche positions  $\mu_i$ . The vertical dotted lines indicate evolutionary branching. Panel **B** shows the evolution of the niche difference ( $\hat{\Omega}$ ). Panel **C** shows the evolution of fitness difference ( $\theta$ ) and the robustness ( $\eta$ ). For this figure, the resource width has been set to  $\sigma_R = 1.7$ , the niche width to  $\sigma_\alpha = 1$ , and the mortality term to  $m = 0.1$ .

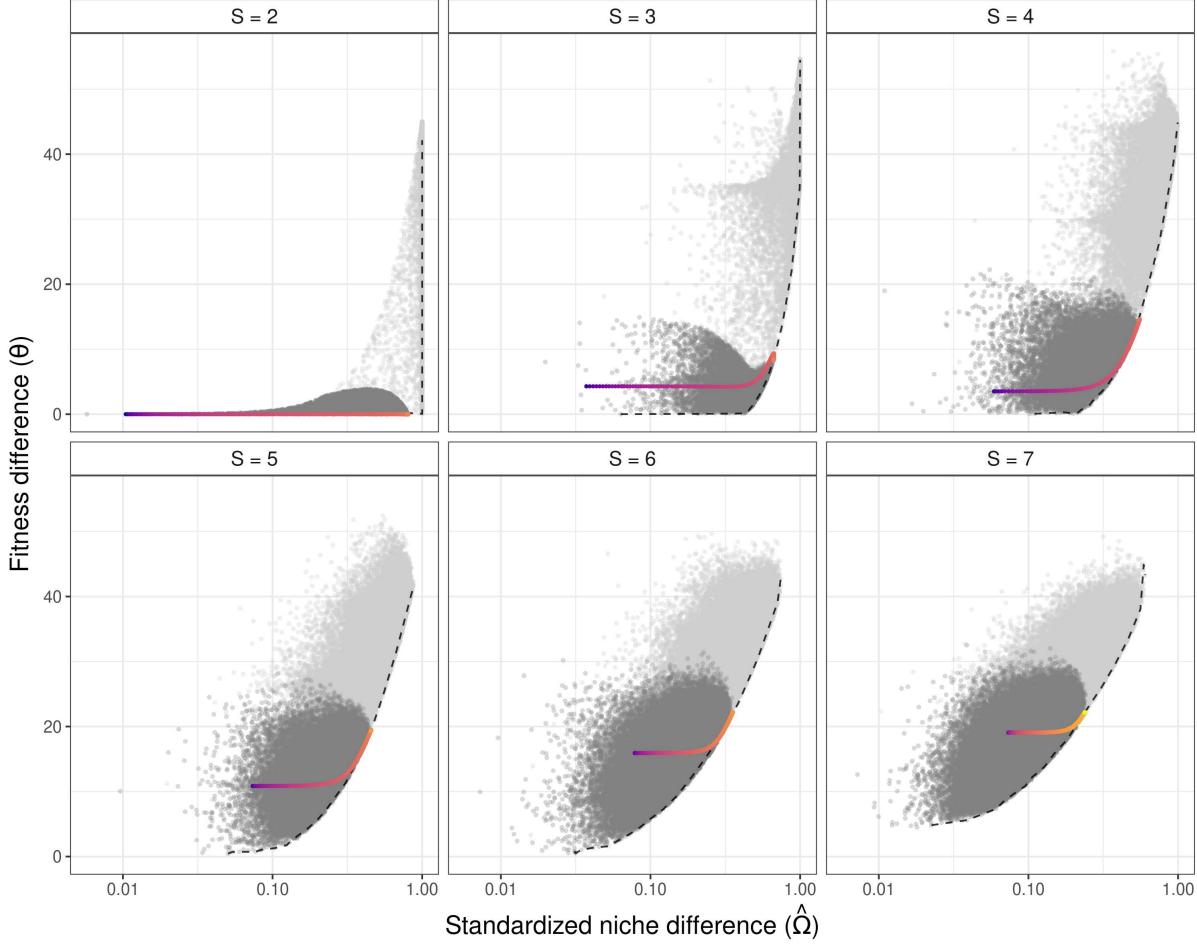
255 ing a niche-neutrality continuum *sensu* Song et al. (2019). Conversely, fitness differences  
256  $\theta$  increase all along evolutionary trajectories, which undermines community robustness  
257 (Fig. 2C). Fitness differences  $\theta$  are initially zero when only two types coexist, as their  
258 trait positions are symmetric around the resource center. They later tend to increase for  
259  $S > 2$  as phenotypes start differing in their intrinsic growth rates, reflecting imbalances  
260 between strong and weak competitors. Note that because evolution here increases both  
261 niche differences (stabilizing effect) and fitness differences (unequalizing effect) between  
262 branching points, its overall effect on coexistence is not obvious. We therefore use the  
263 robustness indicator – the angle  $\eta$  – to assess the evolution of the distance to the border to  
264 the feasibility domain. Figure 2C shows that the robustness  $\eta$  tends to increase between  
265 the branching points, so that the overall effect of a simultaneous increase in both fitness  
266 and niche difference ultimately results in more robust communities. As logically expected,  
267  $\eta$  decreases at each branching point. Moreover, as  $S$  increases and the community be-  
268 comes saturated, the value of  $\eta$  becomes very small. Once maximum phenotype packing  
269 is achieved (at ESC), small perturbations in  $\mathbf{r}$  suffice to lead to non-feasibility, and the  
270 community is structurally more fragile than an undersaturated one (fewer phenotypes  
271 than at the ESC).

272 Supplementary Materials S6 and S7 show that these results are consistent when vary-  
273 ing the resource width  $\sigma_R$  and the mortality rate  $m$ .

### 274 3.2 Projection into coexistence space

275 Figure 3 illustrates how coexistence metrics  $\hat{\Omega}$  and  $\theta$  change along the evolutionary tra-  
276 jectory of figure 2 when  $S \geq 2$ . Remember that large niche differences  $\hat{\Omega}$  and small fitness  
277 differences  $\theta$  enhance persistence. In the background, we show the coexistence metrics of  
278 randomized communities that were produced according to the two rules presented in the  
279 Methods.

280 Considering all randomized communities, it appears that communities that have small  
281 fitness differences most often also have small niche differences. This is due to both



**Figure 3: Variations of coexistence metrics along evolutionary trajectories.** The figure is split into six panels according to the number of phenotypes  $S$ . Each panel therefore represents a slice (in between two branching points) of the trajectory presented in figure 2. The colored line represents the projection of the evolutionary trajectory of figure 2 into the space of coexistence metrics for a given  $S$ , variations of color from purple to orange indicating evolutionary time. The X-axis represents the standardized niche difference ( $\hat{\Omega}$ ), while the Y-axis stands for the fitness difference ( $\theta$ ). Gray points represent randomized feasible communities, with niche position value sampled from the restricted range explored by evolution (1st rule; dark gray) or the full range leading to positive intrinsic growth rates (2nd rule; light gray). Pareto fronts among randomized communities are shown using dashed lines.

282 selecting for feasible communities, but also to ecological constraints under this model,  
283 which generate a global trade-off at the community level between the two coexistence  
284 metrics. Points that optimize one of the two properties relative to the other lie on a  
285 Pareto front. Figure 3 shows that evolution leads to this Pareto-optimality, and more  
286 specifically to the point on this front that also maximizes the standardized niche difference  
287 within the restricted range (first randomization rule). More extreme niche positions would  
288 also allow larger niche differences, but this decrease in competition would come at the  
289 cost of decreasing intrinsic growth rates for the phenotypes further from the resource  
290 optimum, thereby increasing fitness differences  $\theta$ .

291 Regarding community robustness ( $\eta$ ), figure 4 indicates that evolution optimizes it  
292 within the range of niche positions explored by evolution (first randomization rule), but  
293 greater robustness could (rarely) be reached for the second randomization rules. However,  
294 among the niche positions explored by evolution and for a given number of phenotypes,  
295 evolution converges to more robust communities, by optimizing the niche difference  $\Omega$   
296 and the robustness  $\eta$ .

297 Supplementary Materials S6 and S7 show that these results are consistent when vary-  
298 ing the resource width  $\sigma_R$  and the mortality rate  $m$ .

### 299 *3.3 Evolution of productivity*

300 In general, co-evolutionary dynamics increase productivity. While strict optimization  
301 can be proven in the monomorphic situation (Supplementary Materials S3), productiv-  
302 ity increase also happens throughout diversification, as illustrated by figure 5. For the  
303 total abundance (panel A), the co-evolutionary trajectory, though not strictly optimal,  
304 is among the most productive communities. This holds for the two randomization rules.  
305 We have seen on figure 3, that within the restricted range of randomization, co-evolution  
306 tends to maximize niche differentiation, and thus to decrease interspecific competition.  
307 Given a fuller range of randomization, however, communities with larger niche differ-  
308 entiation can be sampled. Results for the relative yield metric are presented in the

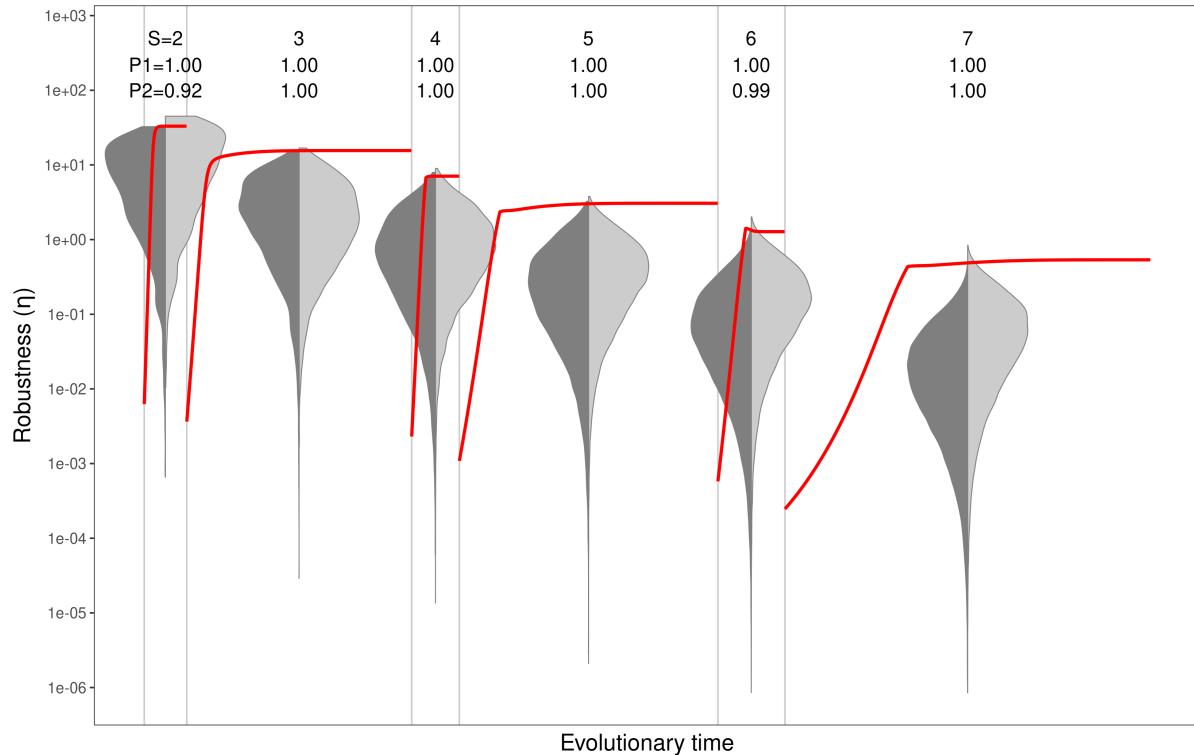


Figure 4: **Evolution of community robustness.** The red line shows the evolution of  $\eta$ . Gray points represent randomized feasible communities, with niche position value sampled from the restricted range explored by evolution (1st rule; dark gray) or the full range leading to positive intrinsic growth rates (2nd rule; light gray). At the singular strategy, the evolved community reaches a value of  $\eta$  greater or equal to than a proportion  $P_1$  and  $P_2$  of randomized communities according to the first and second randomization rules, respectively.

309 Supplementary Materials S5.

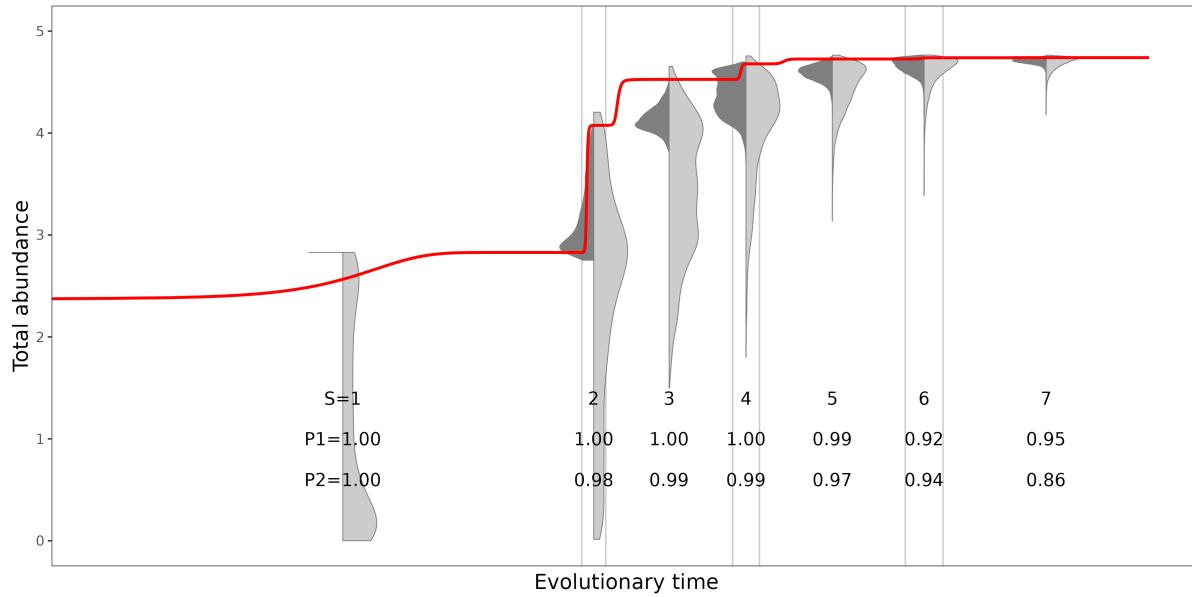


Figure 5: **Productivity along evolutionary trajectories.** Evolution of productivity. Here the numbers  $P1$  and  $P2$  in the second row of text indicate the percentage of randomized communities that are less productive than the ESC. For  $S=1$ , biomass production is optimized at the ESC. For  $S=6$ , the ESC is in the 80th percentile of the most productive communities. Point clouds represent randomization in the narrow (1st rule; dark gray) or the full range (2nd rule; light gray).

310

## 4 Discussion

311 In light of global changes, understanding the way evolution shapes and affects commu-  
312 nity robustness is paramount. Communities are currently reshuffled by the arrival of  
313 non-native species and by range shifts (Alexander et al., 2016). How these novel, evo-  
314 lutionary non-equilibrium communities differ from ESCs is unclear. We also need to  
315 assess under which conditions and how quickly evolution may prevent species from going  
316 extinct following environmental change, considering not only population-level processes  
317 (as in evolutionary rescue) but also larger spatial or organizational scales (Urban, 2016).  
318 Previous studies showed that mechanisms such as adaptive foraging, which can be evo-  
319 lutionary in nature, can stabilize complex systems (Kondoh, 2003; Valdovinos et al.,  
320 2013). Recently a handful of studies, both theoretical and experimental, have explored  
321 this question for pairs of coevolving species within the framework of modern coexistence  
322 theory Pastore et al. (2021); Hart et al. (2019). Here we go further by employing recent  
323 theoretical advances to investigate the question for species-rich systems ( $S > 2$ ) where  
324 diversity arises through subsequent branching events. We show that evolution of coexis-  
325 tence properties follows two distinct trends on two different timescales, while productivity  
326 systematically increases along evolutionary trajectories.

327 The short-term trend, in-between branching events, allows for increased efficiency in  
328 resource partitioning and decreased competition. On this timescale, evolution promotes  
329 niche differentiation ( $\hat{\Omega}$ ), but this effect is counterbalanced by a simultaneous increase  
330 in fitness difference ( $\theta$ ) (Fig. 2). Hence, coexistence is enhanced by favoring niche par-  
331 titioning, rather than by neutral mechanisms. Since we observe both an increase in the  
332 stabilizing mechanisms (increase in niche differences) and a reduction of the equalizing  
333 mechanism (in fitness differences), as in (Pastore et al., 2021), we use the measure of  
334 robustness  $\eta$  introduced earlier to show that between branching communities, evolution  
335 selects less fragile communities with respect to environmental perturbations that impact  
336 growth rates  $\mathbf{r}$  (Saavedra et al., 2017). The fact that coexistence metrics are usually not  
337 independent is already appreciated (Barabás et al., 2018a; Song et al., 2019), but we here

338 show how they are readily coupled throughout evolutionary dynamics. This link is made  
339 obvious in our trait-based model (Fig. 3), where niche position affects both growth rates  
340 and competition strength, and in turn  $\Omega$  and  $\theta$ . This interdependence makes it impossible  
341 to minimize simultaneously niche overlap and fitness differences (Fig. 3). Still, evolution  
342 leads to communities that are not strict optima of  $\Omega$  nor minima of  $\theta$  but rather lie on  
343 a Pareto front, an optimal compromise that is reached between niche increasing differ-  
344 entiation and restricting fitness imbalances. In a rare direct empirical test of this short  
345 term effect, Hart et al. (2019) find an increase in fitness difference after evolution, but  
346 not one in niche difference, which they explain may have been prevented by competition  
347 for a single discrete, non-substitutable resource. Still, evidence of (evolutionary) niche  
348 differentiation measured as character displacement abounds in other natural and experi-  
349 mental settings (Grant and Grant, 2006; Dayan and Simberloff, 2005), hence confirming  
350 that we should also expect evolution to increase both metrics in nature.

351 Our results suggest that costs in terms of community robustness may happen on a  
352 longer timescale, where evolution leads to diversification along with evolutionary niche  
353 packing. Branching points, by virtue of addition of one more phenotype, whose trait value  
354 and fitness are initially very close to an existing one, have a simultaneously equalizing  
355 and destabilizing effect. As the community size increases, robustness and niche differences  
356 peak at increasingly lower levels. This is a consequence of trying to pack a greater amount  
357 of phenotypes in the same resource width  $\sigma_R$  (Fig. 2). Classical theory shows that the  
358 number of phenotypes that can be packed on a resource axis is a function of resource  
359 width  $\sigma_R$  relative to the niche width of species (MacArthur and Levins, 1967). But when  
360 evolution is allowed in the community, the richness at the ESC is lower than under  
361 strict maximum (non-evolutionary) packing. In fact, ESCs are by definition uninvadable,  
362 because the fitness landscape for any possible invading trait value  $\mu_m$  is zero at the  
363 resident trait values  $\mu_1, \dots, \mu_S$  and negative everywhere else; however, richer, feasible non-  
364 evolutionary configurations do exist for a given environment. This result has been shown  
365 to be consistent across a range of models, for both continuous and discrete resources

366 (e.g., Case (1981); Rummel and Roughgarden (1983); Shores et al. (2008)). It is worth  
367 pointing out that the answer to whether evolution helps or hinders coexistence is context-  
368 dependent: when the community is undersaturated (fewer members than at the ESC),  
369 evolution drives an increase in robustness  $\eta$ . Conversely, by starting with feasible but  
370 supersaturated communities (more diversity than allowed at the ESC),  $\eta$  will repeatedly  
371 drop to zero and we observe a sequence of extinction events until the ESC is reached.  
372 For example, Shores et al. (2008) found that evolution destabilizes communities, which  
373 can be explained by the fact that they started with supersaturated communities. This  
374 also helps explain why (Loeuille, 2010) found evolution to be usually stabilizing for small  
375 communities (likely undersaturated), while its effect is reversed for rich communities  
376 (likely oversaturated). Empirical evidence for the longer term effects of evolution on  
377 community properties is difficult to acquire, but could be found in phylogenetic patterns  
378 of niche conservatism (Mayfield and Levine, 2010). For example, Yguel et al. (2016) report  
379 patterns consistent with increased productivity along time due to niche diversification and  
380 filling.

381 Ultimately, the evolutionary process results in ESCs that are highly non-random with  
382 respect to their structure and properties. Coevolved communities also show more evenly-  
383 spaced trait distributions than non-evolutionary communities in the model of Barabás  
384 and D'Andrea (2016) (but see Bennett et al. (2013)). Our results show that they are  
385 also more structurally robust. This has important implications in the context of global  
386 change: if coevolved communities are more robust to environmental stresses (changes in  
387 temperature, water, nutrients levels), this could have important consequences in terms  
388 of managing global changes that may need to be considered when guiding conservation  
389 efforts. Indeed, studies on the effects of global change often focus on single-species re-  
390 sponses, but many have argued that community-level responses should be given more  
391 attention (Walther, 2010; Alexander et al., 2016; Gilman et al., 2010). Because commu-  
392 nity composition and interactions are modified by climate change and invasive species,  
393 leading to new assemblages that likely depart from coevolved structures (David et al.,

394 2017; Blowes et al., 2019), our results suggest that the robustness of these new commu-  
395 nities may be relatively poor. Further experimental testing of community-level responses  
396 to environmental stress in co-evolved versus randomized assemblages should prove an  
397 exciting and critically needed avenue of research, and the structural approach provides  
398 a useful theoretical framework to tackle the question. Experimental testing could for in-  
399 stance be undertaken using systems that rapidly diversify (e.g., the *Pseudomonas* system  
400 in Rainey and Travisano (1998)).

401 Regarding productivity, rich randomized communities tend to be more productive on  
402 average than poor ones (Fig. 5). By requiring randomized communities to be feasible,  
403 we however introduce a selection bias: in this sense, the positive slope between richness  
404 and productivity is indeed a byproduct of coexistence (Pillai and Gouhier, 2019), since  
405 the conditions that promote coexistence (namely niche differences), are also those that  
406 promote complementarity and greater productivity. Thus, we observe the emergence of  
407 trade-offs between community properties: higher total biomass values are possible among  
408 randomized communities, but these tend to have larger fitness imbalance and/or lower  
409 community robustness. This finding echoes results of Rohr et al. (2016) showing that very  
410 productive communities have low evenness, while those with lower deviation from the fea-  
411 sibility domain center ( $\theta$ ) have intermediate levels of biomass production. In agreement  
412 with Aubree et al. (2020), our results also show that evolutionary communities are often  
413 more productive on average than randomized ones, especially when species richness is low.  
414 Total abundance increases monotonically along evolutionary trajectories, first because of  
415 selection for higher intrinsic growth rates in the monomorphic case (selection effect), then  
416 due to a better use of the resource space by the diversifying phenotypes (complementar-  
417 ity effect) (Fig. 2). This is consistent with the expectation that increased diversification  
418 within and across species should lead to occupation of vacant niche spaces, leading to  
419 increased niche complementarity and utilization, and ultimately increased abundances  
420 at the consumer level (a hypothesis supported by Yguel et al. (2016)). For instance,  
421 some of the largest biodiversity ecosystem-functioning experiments have consistently re-

422 ported an increase in the net effect of biodiversity on biomass production and of niche  
423 complementarity across a decade (Fargione et al. (2007); Marquard et al. (2009); Cardi-  
424 nale et al. (2007)). Similarly, Stefan et al. (2022) showed that plant-plant interactions  
425 shifted towards increased complementarity and yield over just a few generations of coex-  
426 istence, while van Moorsel et al. (2018) report higher productivity in polycultures with  
427 an 8-year co-evolution history compared to identical-composition, but evolutionary naive  
428 plant communities. This evidence strongly supports the idea that the complementarity  
429 effect often observed in biodiversity-ecosystem functioning experiments is expected from  
430 an evolutionary point of view, and likely reinforcing along evolutionary trajectories. A  
431 consequence of this result is that many BEF studies may have underestimated the produc-  
432 tivity gains due to biodiversity, by assembling *de-novo* communities, instead of coevolved  
433 ones. Conversely, relative yield measures might have been overestimated (Supplementary  
434 Material S5).

435 It is, however, possible to find more productive combinations among the random com-  
436 munities. Indeed, strict evolutionary optimization of total biomass at equilibrium only  
437 arises for the monomorphic case in this class of model and is lost in higher-dimensional  
438 systems (Rohr and Loeuille, 2022). Mathematically, this is because for  $S = 1$ , the se-  
439 lection gradient coincides with the gradient of  $r(\mu)$  (Supplementary Materials S3). In  
440 polymorphic systems, maximizing biomass productivity would require further niche dif-  
441 ferentiation, at the expense of increasing fitness imbalance, which is not achieved under  
442 the Pareto-optimality engendered by evolution. This runs contrary to a widespread belief  
443 that evolution begets optimality: while growth rate optimization might be the norm in  
444 simple systems with weak or no interactions, this assumption does not generally hold true  
445 if we introduce realistic eco-evolutionary feedback loops (see also Lion and Metz (2018)  
446 for a treatment in epidemiological models and Metz and Geritz (2016) for a discussion  
447 on optimization principles).

448 Several questions remain open for investigation. We considered only competitive in-  
449 teractions and a single resource axis, so that packing more than two species with the same

450 amount of niche overlap between all of them is impossible Macarthur and Levins (1967).  
451 If phenotypes were arranged in a multidimensional trait space, neutral configurations with  
452 multiple species would be possible. In addition, we considered only evolving niche posi-  
453 tions in our model, but niche width could also evolve, leading to possible diversification  
454 between generalists and specialists species. Nevertheless, the theoretical predictions of  
455 our study need to be further experimentally tested. Although there is empirical support  
456 in biodiversity ecosystem-functioning studies regarding the increase in niche complemen-  
457 tarity and biomass production over time, it remains unclear whether those predictions  
458 would hold in co-evolved communities emerging from diversification. Experimentally,  
459 comparison of properties of ESCs to non-evolutionary communities is complicated by the  
460 fact that ESCs are conceptually useful, but whether they are frequent in nature, and how  
461 to go about identifying them, is unclear (Edwards et al., 2018). Despite these hurdles,  
462 experimental tests of our theoretical results could provide timely evidence that would  
463 contribute to our understanding of evolution of communities properties and their inter-  
464 play in rich systems. Such experiments could be undertaken in microbial systems, where  
465 rapidly evolving communities could be compared to control treatments of non-coevolved  
466 assemblages (Rainey and Travisano, 1998; Altermatt et al., 2015).

## 467 Data and code

468 Numerical simulations of eco-evolutionary trajectories were performed in Julia 1.5.1  
469 (Bezanson, Edelman, Karpinski, and Shah, 2017), while computation of metrics of co-  
470 existence and plotting of results were done in R 4.0.3 (R Core Team, 2020). Code to  
471 reproduce the analyses will be made available on github.

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474

## References

475 Alexander, J. M., Diez, J. M., Hart, S. P., and Levine, J. M. 2016. When Climate  
476 Reshuffles Competitors: A Call for Experimental Macroecology. *Trends in Ecology and Evolution* 21:47:1–11. ISSN 01695347.

477

478 Altermatt, F., Fronhofer, E. A., Garnier, A., Giometto, A., Hammes, F., Klecka, J.,  
479 Legrand, D., Mächler, E., Massie, T. M., Pennekamp, F., Plebani, M., Pontarp, M.,  
480 Schtickzelle, N., Thuillier, V., and Petchey, O. L. 2015. Big answers from small worlds:  
481 A user’s guide for protist microcosms as a model system in ecology and evolution.  
482 *Methods in Ecology and Evolution* 6:218–231. ISSN 2041210X.

483 Aubree, F., David, P., Jarne, P., Loreau, M., Mouquet, N., and Calcagno, V. 2020.  
484 How community adaptation affects biodiversity–ecosystem functioning relationships.  
485 *Ecology Letters* 23:1263–1275. ISSN 14610248.

486 Barabás, G. and D’Andrea, R. 2016. The effect of intraspecific variation and heritability  
487 on community pattern and robustness. *Ecology Letters* 19:977–986. ISSN 14610248.

488 Barabás, G., D’Andrea, R., and Stump, S. M. 2018a. Chesson’s coexistence theory.  
489 *Ecological Monographs* 0:1–27. ISSN 00129615.

490 Barabás, G., D’Andrea, R., and Stump, S. M. 2018b. Chesson’s coexistence theory.  
491 *Ecological Monographs* 88:277–303. ISSN 15577015.

492 Bell, G. and Gonzalez, A. 2009. Evolutionary rescue can prevent extinction following  
493 environmental change. *Ecology Letters* 12:942–948. ISSN 1461-0248.

494 Bennett, J. A., Lamb, E. G., Hall, J. C., Cardinal-McTeague, W. M., and Cahill Jr., J. F.  
495 2013. Increased competition does not lead to increased phylogenetic overdispersion in  
496 a native grassland. *Ecology Letters* 16:1168–1176. ISSN 1461-0248.

497 Bezanson, J., Edelman, A., Karpinski, S., and Shah, V. B. 2017. Ju-  
498 lia: A fresh approach to numerical computing. SIAM review 59:65–98. URL  
499 <https://doi.org/10.1137/141000671>.

500 Blowes, S. A., Supp, S. R., Antão, L. H., Bates, A., Bruelheide, H., Chase, J. M., Moyes,  
501 F., Magurran, A., McGill, B., Myers-Smith, I. H., Winter, M., Bjorkman, A. D.,  
502 Bowler, D. E., Byrnes, J. E. K., Gonzalez, A., Hines, J., Isbell, F., Jones, H. P.,  
503 Navarro, L. M., Thompson, P. L., Vellend, M., Waldock, C., and Dornelas, M. 2019.  
504 The geography of biodiversity change in marine and terrestrial assemblages. Science  
505 366:339–345. ISSN 0036-8075, 1095-9203.

506 Bränström, Å., Johansson, J., Loeuille, N., Kristensen, N., Troost, T. A., Lambers, R.  
507 H. R., and Dieckmann, U. 2012. Modelling the ecology and evolution of communities:  
508 A review of past achievements, current efforts, and future promises. Evolutionary  
509 Ecology Research 14:601–625. ISSN 1522-0613.

510 Bränström, Å., Johansson, J., and von Festenberg, N. 2013. The Hitchhiker’s Guide to  
511 Adaptive Dynamics. Games 4:304–328.

512 Cardinale, B. J., Wright, J. P., Cadotte, M. W., Carroll, I. T., Hector, A., Srivastava,  
513 D. S., Loreau, M., and Weis, J. J. 2007. Impacts of plant diversity on biomass  
514 production increase through time because of species complementarity. Proceedings  
515 of the National Academy of Sciences 104:18123–18128.

516 Case, T. J. 1981. Niche packing and coevolution in competition communities. Proceedings  
517 of the National Academy of Sciences of the United States of America 78:5021–5025.  
518 ISSN 00278424.

519 Chesson, P. 2000. Mechanisms of Maintenance of Species Diversity. Annual Review of  
520 Ecology and Systematics 31:343–366. ISSN 0066-4162.

521 David, P., Thébault, E., Anneville, O., Duyck, P. F., Chapuis, E., and Loeuille, N., 2017.  
522 Chapter One - Impacts of Invasive Species on Food Webs: A Review of Empirical

523 Data. Pages 1–60 in D. A. Bohan, A. J. Dumbrell, and F. Massol, eds. Advances  
524 in Ecological Research, volume 56 of *Networks of Invasion: A Synthesis of Concepts*.  
525 Academic Press.

526 Dayan, T. and Simberloff, D. 2005. Ecological and community-wide character displace-  
527 ment: The next generation. *Ecology Letters* 8:875–894. ISSN 1461-0248.

528 Dercole, F., Ferrière, R., Gragnani, A., and Rinaldi, S. 2006. Coevolution of slow–fast  
529 populations: Evolutionary sliding, evolutionary pseudo-equilibria and complex Red  
530 Queen dynamics. *Proceedings of the Royal Society B: Biological Sciences* 273:983–990.

531 Dieckmann, U. and Doebeli, M. 1999. On the origin of species by sympatric speciation.  
532 *Nature* 400:354–357. ISSN 00280836.

533 Dieckmann, U. and Law, R. 1996. The dynamical theory of coevolution: A derivation  
534 from stochastic ecological processes. *Journal of Mathematical Biology* 34:579–612.  
535 ISSN 14321416.

536 Doebeli, M. and Dieckmann, U. 2000. Evolutionary branching and sympatric speciation  
537 caused by different types of ecological interactions. *American Naturalist* 156. ISSN  
538 00030147.

539 Edwards, K. F., Kremer, C. T., Miller, E. T., Osmond, M. M., Litchman, E., and Klaus-  
540 meier, C. A. 2018. Evolutionarily stable communities: A framework for understanding  
541 the role of trait evolution in the maintenance of diversity. *Ecology Letters* 21:1853–1868.  
542 ISSN 14610248.

543 Fargione, J., Tilman, D., Dybzinski, R., Lambers, J. H. R., Clark, C., Harpole, W. S.,  
544 Knops, J. M., Reich, P. B., and Loreau, M. 2007. From selection to complementarity:  
545 Shifts in the causes of biodiversity–productivity relationships in a long-term biodiver-  
546 sity experiment. *Proceedings of the Royal Society B: Biological Sciences* 274:871–876.

547 Ferriere, R. and Legendre, S. 2013. Eco-evolutionary feedbacks, adaptive dynamics  
548 and evolutionary rescue theory. *Philosophical Transactions of the Royal Society B:*  
549 *Biological Sciences* 368:20120081.

550 Fussmann, G. F., Loreau, M., and Abrams, P. A. 2007. Eco-evolutionary dynamics of  
551 communities and ecosystems. *Functional Ecology* 21:465–477. ISSN 02698463.

552 Geritz, S. A., Kisdi, É., Meszéna, G., and Metz, J. A. 1998. Evolutionarily singular  
553 strategies and the adaptive growth and branching of the evolutionary tree. *Evolutionary  
554 Ecology* ISSN 02697653.

555 Gilman, S. E., Urban, M. C., Tewksbury, J., Gilchrist, G. W., and Holt, R. D. 2010. A  
556 framework for community interactions under climate change. *Trends in Ecology and  
557 Evolution* 25:325–331. ISSN 01695347.

558 Grant, P. R. and Grant, B. R. 2006. Evolution of character displacement in Darwin's  
559 finches. *Science* 313:224–226. ISSN 00368075.

560 Hairston Jr, N. G., Ellner, S. P., Geber, M. A., Yoshida, T., and Fox, J. A. 2005. Rapid  
561 evolution and the convergence of ecological and evolutionary time. *Ecology Letters*  
562 8:1114–1127. ISSN 1461-0248.

563 Hart, S. P., Turcotte, M. M., and Levine, J. M. 2019. Effects of rapid evolution on species  
564 coexistence. *Proceedings of the National Academy of Sciences of the United States of  
565 America* 116:2112–2117. ISSN 10916490.

566 Hendry, A. P. A. P., 2009. Eco-evolutionary dynamics. Princeton University Press,  
567 Princeton. ISBN 9781400883080.

568 Hutchinson, G. E. 1957. Concluding remarks. Cold Spring  
569 Harbor Symposia on Quantitative Biology 22:415–427. URL  
570 <http://symposium.cshlp.org/content/22/415.short>.

571 Hutchinson, G. E., 1979. An introduction to population ecology. Yale University Press,  
572 New Haven, Conn., 1st ed. 1978, 3rd print. 1978 ; 1st ed., 4th print. 1979 edition. ISBN  
573 0300021550.

574 Kondoh, M. 2003. Foraging Adaptation and the Relationship Between Food-Web Com-  
575 plexity and Stability. *Science* 299:1388–1391.

576 Lankau, R. A. 2009. Genetic Variation Promotes Long-Term Coexistence of *Brassica*  
577 *nigra* and Its Competitors. *The American Naturalist* 174:E40–E53. ISSN 0003-0147.

578 Lankau, R. A. and Strauss, S. Y. 2007. Mutual feedbacks maintain both genetic and  
579 species diversity in a plant community. *Science* 317:1561–1563. ISSN 00368075.

580 Lewontin, R., 2003. Building a science of population biology. Pages 7–20 *in* R. S.  
581 Singh and M. K. Uyenoyama, eds. *The Evolution of Population Biology*. Cambridge  
582 University Press, Cambridge.

583 Lion, S. and Metz, J. A. 2018. Beyond R<sub>0</sub> Maximisation: On Pathogen Evolution  
584 and Environmental Dimensions. *Trends in Ecology and Evolution* 33:458–473. ISSN  
585 01695347.

586 Loeuille, N. 2010. Influence of evolution on the stability of ecological communities.  
587 *Ecology Letters* 13:1536–1545. ISSN 14610248.

588 Logofet, D. O., 1994. Matrices and Graphs: Stability Problems in Mathemat-  
589 ical Ecology, CRC Press, New York (1993)., volume 56. CRC Press. URL  
590 [http://link.springer.com/10.1016/S0092-8240\(05\)80313-1](http://link.springer.com/10.1016/S0092-8240(05)80313-1).

591 Loreau, M. and Hector, a. 2001. Partitioning selection and complementarity in biodiver-  
592 sity experiments. *Nature* 412:72–6. ISSN 0028-0836.

593 MacArthur, R. 1969. Species packing, and what competition minimizes. *Proceedings of*  
594 *the National Academy of Sciences* 64:1369–1371. ISSN 0027-8424.

595 MacArthur, R. and Levins, R. 1967. The Limiting Similarity , Convergence , and Di-  
596 vergence of Coexisting Species Author ( s ): Robert MacArthur and Richard Levins  
597 Source : The American Naturalist , Vol . 101 , No . 921 ( Sep . - Oct ., 1967 ), pp .  
598 377-385 Published by : The University of C. American Naturalist 101:377–385.

599 MacArthur, R. H., 1972. Geographical ecology. Patterns in the distribution of species.  
600 Harper and Row.

601 Mallet, J. 2012. The struggle for existence: How the notion of carrying capacity, K,  
602 obscures the links between demography, Darwinian evolution, and speciation. Evolution-  
603 ary Ecology Research 14:627–665. ISSN 15220613.

604 Marquard, E., Weigelt, A., Temperton, V. M., Roscher, C., Schumacher, J., Buchmann,  
605 N., Fischer, M., Weisser, W. W., and Schmid, B. 2009. Plant species richness and  
606 functional composition drive overyielding in a six-year grassland experiment. Ecology  
607 90:3290–3302. ISSN 1939-9170.

608 Matsuda, H. and Abrams, P. A. 1994. Runaway Evolution to Self-Extinction Under  
609 Asymmetrical Competition. Evolution 48:1764–1772. ISSN 0014-3820.

610 May, R. M. 1972. Will a large complex system be stable? Nature 238:413–414. ISSN  
611 00280836.

612 Mayfield, M. M. and Levine, J. M. 2010. Opposing effects of competitive exclusion on the  
613 phylogenetic structure of communities. Ecology Letters 13:1085–1093. ISSN 1461023X.

614 McPeek, M. A., 2017. Evolutionary Community Ecology, Volume 58. Monographs in  
615 Population Biology ; 77. Princeton University Press,, Princeton, NJ. ISBN 1-4008-  
616 8821-2.

617 Medeiros, L. P., Song, C., and Saavedra, S. 2021. Merging dynamical and structural  
618 indicators to measure resilience in multispecies systems. Journal of Animal Ecology  
619 90:2027–2040. ISSN 1365-2656.

620 Metz, J., Geritz, S., Meszena, G., Jacobs, F., and van Heerwaarden, J., 1995.  
621 Adaptive dynamics: A geometrical study of the consequences of nearly faith-  
622 ful reproduction. Iiasa working paper, IIASA, Laxenburg, Austria. URL  
623 <http://pure.iiasa.ac.at/id/eprint/4497/>.

624 Metz, J., Mylius, S., and Diekmann, O. 2008. When does evolution optimize? *Evol.*  
625 *Evolutionary Ecology Research* 10.

626 Metz, J. A. and Geritz, S. A. 2016. Frequency dependence 3.0: An attempt at codifying  
627 the evolutionary ecology perspective. *Journal of Mathematical Biology* 72:1011–1037.  
628 ISSN 14321416.

629 Metz, J. A. J., Nisbet, R. M., and Geritz, S. A. H. 1992. How should we define ‘fitness’  
630 for general ecological scenarios? *Trends in Ecology & Evolution* 7:198–202. ISSN  
631 0169-5347.

632 Pastore, A. I., Barabás, G., Bimler, M. D., Mayfield, M. M., and Miller, T. E. 2021. The  
633 evolution of niche overlap and competitive differences. *Nature Ecology and Evolution*  
634 Pages 21–23. ISSN 2397334X.

635 Pillai, P. and Gouhier, T. C. 2019. Not even wrong: The spurious measurement of  
636 biodiversity’s effects on ecosystem functioning. *Ecology* 100:1–12. ISSN 00129658.

637 R Core Team, 2020. R: A Language and Environment for Statistical Com-  
638 puting. R Foundation for Statistical Computing, Vienna, Austria. URL  
639 <https://www.R-project.org/>.

640 Rainey, P. B. and Travisano, M. 1998. Adaptive radiation in a heterogeneous environ-  
641 ment. *Nature* 394:69–72. ISSN 1476-4687.

642 Rohr, R. P. and Loeuille, N. 2022. Effects of evolution on niche displacement and  
643 emergent population properties, a discussion on optimality. *Oikos* n/a:e09472. ISSN  
644 1600-0706.

645 Rohr, R. P., Saavedra, S., and Bascompte, J. 2014. On the structural stability of  
646 mutualistic systems. *Science* 345. ISSN 10959203.

647 Rohr, R. P., Saavedra, S., Peralta, G., Frost, C. M., Bersier, L. F., Bascompte, J., and  
648 Tylianakis, J. M. 2016. Persist or produce: A community trade-off tuned by species  
649 evenness. *American Naturalist* 188:411–422. ISSN 00030147.

650 Rummel, J. D. and Roughgarden, J. 1983. Some Differences between Invasion-Structured  
651 and Coevolution-Structured Competitive Communities: A Preliminary Theoretical  
652 Analysis. *Oikos* 41:477. ISSN 00301299.

653 Saavedra, S., Rohr, R. P., Bascompte, J., Godoy, O., Kraft, N. J., and Levine, J. M.  
654 2017. A structural approach for understanding multispecies coexistence. *Ecological  
655 Monographs* 87:470–486. ISSN 15577015.

656 Shoresh, N., Hegreness, M., and Kishony, R. 2008. Evolution exacerbates the paradox of  
657 the plankton. *Proceedings of the National Academy of Sciences of the United States  
658 of America* 105:12365–12369. ISSN 00278424.

659 Song, C., Barabás, G., and Saavedra, S. 2019. On the Consequences of the Interdepen-  
660 dence of Stabilizing and Equalizing Mechanisms. *The American Naturalist* 194:000–  
661 000. ISSN 0003-0147.

662 Song, C., Rohr, R. P., and Saavedra, S. 2018. A guideline to study the feasibility domain  
663 of multi-trophic and changing ecological communities. *Journal of Theoretical Biology*  
664 450:30–36. ISSN 10958541.

665 Stefan, L., Engbersen, N., and Schöb, C. 2022. Rapid transgenerational adaptation in  
666 response to intercropping reduces competition. *eLife* 11:e77577. ISSN 2050-084X.

667 Taper, M. L. and Case, T. J. 1992. Models of character displacement and the theoretical  
668 robustness of taxon cycles. *Evolution* 46:317–333. ISSN 00143820.

669 Tilman, D., Knops, J., Wedin, D., Reich, P., Ritchie, M., and Siemann, E. 1997. The  
670 Influence of Functional Diversity and Composition on Ecosystem Processes. *Science*  
671 277:1300–1302.

672 Urban, M. 2016. Improving the forecast for biodiversity under climate change. *Science*  
673 (New York, N.Y.) 353:1293–1310. ISSN 1095-9203.

674 Valdovinos, F. S., Moisset de Espanés, P., Flores, J. D., and Ramos-Jiliberto, R. 2013.  
675 Adaptive foraging allows the maintenance of biodiversity of pollination networks. *Oikos*  
676 122:907–917. ISSN 1600-0706.

677 van Moorsel, S. J., Hahl, T., Wagg, C., De Deyn, G. B., Flynn, D. F. B., Zuppinger-  
678 Dingley, D., and Schmid, B. 2018. Community evolution increases plant productivity  
679 at low diversity. *Ecology Letters* 21:128–137. ISSN 1461-0248.

680 Vandermeer, J. H., 1989. *The Ecology of Intercropping*. Cambridge University Press.

681 Volterra, V., 1931. *Leçons sur la théorie mathématique de la lutte pour la vie*. Les grands  
682 classiques Gauthier-Villars. J. Gabay, Sceaux. ISBN 2876470667.

683 Walther, G.-r. 2010. Community and ecosystem responses to recent climate change.  
684 *Philosophical Transactions of the Royal Society B: Biological Sciences* 365:2019–24.  
685 ISSN 1471-2970.

686 Yguel, B., Jactel, H., Pearse, I. S., Moen, D., Winter, M., Hortal, J., Helmus, M. R.,  
687 Kühn, I., Pavoine, S., Purschke, O., Weiher, E., Violle, C., Ozinga, W., Brändle, M.,  
688 Bartish, I., and Prinzing, A. 2016. The evolutionary legacy of diversification predicts  
689 ecosystem function. *American Naturalist* 188:398–410. ISSN 00030147.

690 Yoshida, T., Jones, L. E., Ellner, S. P., Fussmann, G. F., and Hairston, N. G. 2003. Rapid  
691 evolution drives ecological dynamics in a predator-prey system. *Nature* 424:303–306.  
692 ISSN 00280836.