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3 **Adaptive phenotypic diversification along a temperature-depth gradient**

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19 **Online appendix A**: Model variables, functions, and parameters

20 **Online appendix B**: Model functions and their estimation from empirical data

21 **Online appendix C**: Sensitivity analysis

22 **ABSTRACT**

23 Theoretical models suggest that sympatric speciation along environmental gradients might be
24 common in nature. Here we present the first data-based model of evolutionary diversification
25 along a continuous environmental gradient. Based on genetic analyses, it has been suggested
26 that a pair of coregonid fishes (*Coregonus* spp.) in a postglacial German lake originated by
27 sympatric speciation. Within this lake, the two species segregate vertically and show
28 metabolic adaptations to, as well as behavioral preferences for, correspondingly different
29 temperatures. We test the plausibility of the hypothesis that this diversifying process has been
30 driven by adaptations to different thermal microhabitats along the lake's temperature-depth
31 gradient. Using an adaptive dynamics model that is calibrated with empirical data and that
32 allows the gradual evolution of a quantitative trait describing optimal foraging temperature,
33 we show that under the specific environmental conditions in the lake, evolutionary branching
34 of a hypothetical ancestral population into two distinct phenotypes may have occurred. We
35 also show that the resultant evolutionary diversification yields two stably coexisting
36 populations with trait values and depth distributions that are in agreement with those currently
37 observed in the lake. We conclude that divergent thermal adaptations along the temperature-
38 depth gradient might have brought about the two species observed today.

39 INTRODUCTION

40 Understanding the emergence of biological diversity by adaptive diversification based on
41 natural selection is a major interest in evolutionary biology. The notion of adaptive speciation
42 suggests that macro-evolutionary phenomena of diversification are ultimately the outcome of
43 micro-evolutionary processes driven by natural selection in general (Orr and Smith 1998;
44 Coyne and Orr 2004), and by frequency-dependent disruptive selection in particular
45 (Dieckmann et al. 2004). Following this notion, the formation of phenotypic and ecological
46 diversity is likely to imply disruptive natural selection arising from competitive interactions.
47 Competitive interactions between individuals and populations within the same geographical
48 area – independently of whether they result from exploitation competition, interference
49 competition, or apparent competition – are thus of major importance for understanding
50 biological diversification. Empirical studies have confirmed that intraspecific competition can
51 be frequency-dependent (e.g., Swanson et al. 2003; Schluter 2003), making disruptive
52 selection on corresponding traits of natural populations more likely than previously thought
53 (Bolnick and Lau 2008). Frequency dependence occurs whenever selection pressures depend
54 on the phenotypic composition of a population, which is a direct and often inevitable
55 consequence of the way a population shapes the environment it in turn experiences.

56 The geographical conditions underlying diversifying processes have long been a focus
57 of debate (e.g., Mayr 1963; Via 2001). However, the mechanisms of ecologically based
58 sympatric, parapatric, and allopatric speciation that can drive divergence appear to be similar,
59 with disruptive or divergent natural selection on ecologically important traits serving as the
60 driving force of diversification, and with the evolution of reproductive isolation occurring as a
61 consequence of divergent selection on those traits (Schluter 2000). There is now mounting
62 empirical evidence for the operation of ecological speciation in nature (Schluter 2009; Nosil
63 2012). The process of ecologically based adaptive speciation does not only necessitate the

64 emergence of reproductive isolation during diversification, but also requires the ability of the
65 incipient species to coexist persistently (Coyne and Orr 2004). Hence, the same ecological
66 conditions and mechanisms that facilitate disruptive natural selection can cause adaptive
67 speciation through gradually divergent evolution, promote reproductive isolation between the
68 incipient species, and enable the coexistence of closely related species in sympatry.
69 Highlighting one common mode of ecological speciation, empirical work suggests that
70 competition-driven divergent resource or habitat use plays an important role in causing
71 ecological diversification (Svanbäck and Bolnick 2007) and in promoting sympatric
72 speciation (e.g., Gislason et al. 1999; Knudsen et al. 2006) within natural populations. With
73 this study, we aim to identify ecological conditions that potentially drive adaptive
74 diversifications in sympatry, as well as the underlying phenotypic traits that are subject to
75 disruptive natural selection.

76 The theory of adaptive dynamics (Metz et al. 1992, 1996; Dieckmann and Law 1996;
77 Geritz et al. 1998) has facilitated the construction of theoretical models investigating
78 ecologically based processes of evolutionary diversification (Doebeli and Dieckmann 2005).
79 Addressing this objective is aided by the assumption, often made in the theory of adaptive
80 dynamics, that there is a sufficient separation of the timescales on which ecological change
81 and evolutionary changes unfold. The resultant framework allows evaluating the potential for
82 evolutionary diversification in complex adaptive systems with an emphasis on the ecological
83 conditions promoting the corresponding selection pressures on specific adaptive traits of
84 natural populations. Frequency-dependent selection on a slowly evolving quantitative trait of
85 an asexually reproducing population leads to directional evolution along the local selection
86 gradient until an evolutionarily singular strategy is reached. This singular strategy can either
87 be a fitness maximum, and hence be locally evolutionarily stable for a single morph, or it can
88 be a fitness minimum, and hence be an ‘evolutionary branching point’, which potentially

89 leads to the splitting and subsequent divergence of two genetically distinct morphs. Adaptive
90 dynamics models have repeatedly shown that many natural ecological settings are expected to
91 imply evolution to such fitness minima, at which evolutionary branching may then occur
92 based on frequency-dependent disruptive selection (as reviewed, e.g., in Dieckmann et al.
93 2004; see also Ito and Dieckmann 2007). In accordance with the majority of empirical
94 examples of adaptive diversification, most of the existing adaptive-speciation models assume
95 ecological specialization through resource partitioning as the key driver of diversification.

96 One conclusion from advanced adaptive-speciation models is that sympatric speciation
97 is theoretically plausible and may thus be a common process in nature (e.g., Dieckmann and
98 Doebeli 1999). However, theoretical speciation models are not easily evaluated in terms of
99 biological plausibility and may lack ecological realism or accuracy in their assumptions about
100 the properties of natural systems. For instance, the parameter regions in which evolutionary
101 branching may occur in such models is usually only indirectly comparable among different
102 models, and quantitative comparisons to natural systems are often difficult to make without
103 referring to one particular natural system. Data-based models of adaptive diversification,
104 utilizing empirically motivated and quantitative ecological assumptions, are therefore needed
105 to evaluate the importance of these processes in nature.

106 The main purpose of this study is to develop such an empirically motivated and data-
107 based model for adaptive diversification in a specific natural system that has already been
108 well investigated (*Coregonus* spp. in Lake Stechlin, Germany). The model aims to describe
109 the diversification of a single ancestral population into two stably coexisting populations by
110 physiological adaptation to different thermal microhabitats along a temperature-depth
111 gradient in a temperate freshwater lake. Our model does not consider the genetic architecture
112 of the quantitative trait involved in this process and is not meant to examine the evolution of
113 reproductive isolation. We rather examine whether the ecological conditions in our study lake

114 allow for the evolution of an ancestral population toward a fitness minimum and the
115 subsequent evolutionary branching.

116 **MATERIALS AND METHODS**

117 **Model system**

118 Temperate freshwater fish occupying postglacial lakes are among the best model systems for
119 studying adaptive diversification. Several taxa in these systems exhibit an ecological diversity
120 consistent with processes of adaptive speciation (Schluter 2000), including the coregonids
121 (reviewed in Hudson et al. 2007). In the deep postglacial Lake Stechlin in Germany
122 (maximum depth 69m), a pair of closely related coregonids coexists: common vendace
123 (*Coregonus albula* L.) and endemic dwarf-sized Fontane cisco (*Coregonus fontanae* Schulz
124 and Freyhof). The two species are easily distinguished by differential spawning times and
125 show distinct morphological characteristics (Schulz and Freyhof 2003; Helland et al. 2009).
126 Their sympatric evolution has been suggested based on mtDNA and microsatellite analyses
127 (Schulz et al. 2006). Recent AFLP analyses could neither confirm nor reject this hypothesis
128 (Mehner et al. 2010a). The species differ in their average population depths within the pelagic
129 zone, with *C. fontanae* being found deeper in the water column than *C. albula* throughout the
130 year. This difference in depth distribution is associated with a difference in mean experienced
131 water temperature (Helland et al. 2007; Mehner et al. 2010b).

132 Despite the depth segregation, diet compositions of the species are rather similar, with
133 a clear dominance of planktonic food (Helland et al. 2008). Hence, mechanisms reducing
134 competition and thus potentially driving divergence between the species are not significantly
135 related to diet. Instead, an important factor promoting ecological divergence between the
136 species is a difference in temperature-dependent metabolic costs of swimming (Ohlberger et
137 al. 2008a). This directly influences the competitive abilities of the two populations via their

138 efficiency of foraging at a specific temperature, and thus depth. Furthermore, the temperature
139 preferences of the two species correspond to the temperatures at which their net costs of
140 swimming are minimized (Ohlberger et al. 2008c), which underscores the role of temperature
141 as the predominant environmental factor shaping the divergence between the two coregonids.
142 This setting offers a unique opportunity for investigating whether the observed conditions
143 allow for ecological and evolutionary diversification of pelagic fish populations along the
144 temperature-depth gradient of the lake.

145 **Model description**

146 We modeled asexual fish populations competing for the same zooplankton resource with a
147 depth-dependent carrying capacity along a temperature-depth gradient $T(x)$, where T
148 denotes temperature and x denotes depth. The fish populations can adapt to different ambient
149 temperature conditions, i.e., to different locations along the depth axis, through adaptations in
150 a one-dimensional quantitative trait, which specifies the temperature-dependent metabolic
151 optimum, or optimum foraging temperature. For a fish morph i , with $i = 1, \dots, n$, this trait
152 value is denoted by T_i . For describing the evolutionary dynamics of the trait values, we
153 assume asexual reproduction, a low mutation probability μ , and a small standard deviation
154 σ of mutational steps, so that evolution follows the canonical equation of adaptive dynamics
155 (Dieckmann and Law 1996) in conjunction with fitness-based conditions for evolutionary
156 branching (Geritz et al. 1998), as specified in detail below. For a didactical introduction to,
157 and further details on, the adaptive dynamics approach, see, e.g., Dieckmann (2004).

158 The fish population dynamics are deterministic and structured with regard to depth x ,
159 with $0 < x < x_{\max}$. Exploitation competition for zooplankton, considered as the predominant
160 cause of competition among fish in our model, is logistic and occurs at each depth. The
161 foraging efficiency $r(T, T_i)$ of a fish morph i , with $i = 1, \dots, n$, is assumed to drop with

162 temperature T on both sides around the morph-specific optimum foraging temperature T_i .
163 The maximum zooplankton density is described by the depth-dependent carrying-capacity
164 density $K(x)$, which declines monotonically with depth. The gain in biomass density $B_i(x)$
165 of morph i at depth x , which includes fecundity, is proportional to the potential consumption
166 rate and the equilibrium zooplankton density. Biomass loss, which includes mortality, arises
167 from maintenance costs $m(T)$, which monotonically decrease with temperature. Therefore,
168 the per capita growth rates $f_i(x)$, in terms of biomass, are given by the difference between
169 per capita resource intake rates and per capita maintenance costs. We denote by f_i the
170 average growth of morph i across all depths x . We further assume the rapid redistribution of
171 individuals along the temperature-depth gradient following foraging dynamics, which in our
172 model can be chosen from a continuum between random foraging and optimal foraging by
173 varying a parameter α . Individuals keep adjusting their depth according to their potential
174 consumption rate, their potential predation risk, and resource availability.

175 In addition to the morph indices $i = 1, \dots, n$ for the n resident morphs, we use $i = 0$ for
176 a rare mutant. The growth rate f_0 of such a rare mutant equals its invasion fitness (Metz et al.
177 1992), with its first derivative (often called selection gradient or fitness gradient) denoted by
178 g_i , and its second derivative denoted by h_i . These derivatives of invasion fitness are taken
179 with respect to the mutant trait value T_0 and are evaluated at the trait value of the resident
180 with trait value T_i . Online appendix A provides an overview of all variables, functions, and
181 parameters used in this study.

182 **Model dynamics**

183 The population dynamics are described by the change in total biomass of fish morph i over
184 time,

185
$$\frac{d}{dt} B_i = f_i B_i,$$

186 where B_i is the total biomass and f_i is the average growth rate of morph i . The total biomass
 187 is given by the integral of biomass density from zero to maximum depth,

188
$$B_i = \int_0^{x_{\max}} B_i(x) dx,$$

189 where $B_i(x)$ is the biomass density of morph i at depth x . The average growth rate of morph
 190 i is thus given by

191
$$f_i = \frac{1}{B_i} \int_0^{x_{\max}} f_i(x) B_i(x) dx,$$

192 with

193
$$f_i(x) = \lambda C(x, T_i) P_{\text{eq}}(x) - m(T(x)),$$

194 where λ is the conversion factor from zooplankton mass to fish mass (i.e., the energy-
 195 assimilation efficiency of the fish), $f_i(x)$ is the morph's per capita growth rate at depth x ,
 196 $C(x, T_i)$ is the potential consumption rate of morph i with trait T_i at depth x , $P_{\text{eq}}(x)$ is the
 197 equilibrium zooplankton density at depth x , and $m(T)$ measures the temperature-dependent
 198 maintenance costs. We thus assume that feeding under natural conditions can be
 199 approximated by a linear relationship with resource density. We also assume fast resource
 200 dynamics, so that the zooplankton density is always near its equilibrium (Online appendix B).

201 For the adaptive foraging dynamics, we assume that an individual fish with trait T_i
 202 adjusts its depth according to its potential consumption rate $C(x, T_i)$, the equilibrium
 203 zooplankton density $P_{\text{eq}}(x)$, and a foraging probability $F(x)$ that accounts for the depth-
 204 dependent risk of predation by piscivorous predators,

205
$$B_i(x) = B_i \frac{[C(x, T_i) P_{\text{eq}}(x) F(x)]^\alpha}{\int_0^{x_{\max}} [C(x', T_i) P_{\text{eq}}(x') F(x')]^\alpha dx'},$$

206 where α is the degree of foraging optimality (with $\alpha = 0$ representing random foraging and
207 $\alpha \rightarrow \infty$ representing optimal foraging).

208 The expected evolutionary dynamics of the trait value T_i describing the metabolic
209 temperature optimum of morph i , is given by the canonical equation of adaptive dynamics
210 (Dieckmann and Law 1996),

$$211 \quad \frac{d}{dt} T_i \propto B_i g_i,$$

212 where B_i is the equilibrium population size and g_i is the first derivative of a mutant's invasion
213 fitness, i.e., the selection gradient. More details on the evolutionary analysis are provided
214 under the corresponding heading below.

215 **Parameter estimation**

216 Data sources for all parameter estimates are provided in Online appendix A. Fig. 1 and Online
217 appendix B provide details on the estimation of functions describing the temperature-depth
218 gradient, observed zooplankton density, potential consumption rate, foraging efficiency,
219 capture success, maintenance costs, and foraging probability. These functions and parameters
220 were estimated based on observational data from Lake Stechlin in conjunction with various
221 laboratory measurements on the two coregonids. The foraging optimality α was estimated
222 based on the distribution patterns of the two coregonids in Lake Stechlin. This parameter
223 measures the degree to which individuals forage at their temperature optimum. In order to
224 estimate the degree of foraging optimality in the natural system, we ran our model without
225 evolutionary dynamics for the empirically determined temperature optima as fixed trait values
226 and compare the resulting average depths with the measured year-round average depths of the
227 Lake Stechlin coregonids (Helland et al. 2007). A figure provided in the online appendix
228 (Fig. B1) shows, as a function of α , the sum of absolute values of the deviations of the two
229 modeled population depths from the two observed average population depths in the natural

230 system. A foraging optimality α of about 6 was found to offer the best approximation for the
231 foraging behavior of these fish. We thus use this α -value in our further analyses.

232 **Evolutionary analysis**

233 We use a pairwise invasibility analysis to investigate the evolutionary dynamics in our system
234 under the specific ecological conditions encountered in Lake Stechlin. Pairwise invasibility
235 analysis assumes that any mutant introduced to the system first occurs at very low numbers
236 and that the resident population has come sufficiently close to its demographic equilibrium at
237 the time a mutant is introduced (Van Tienderen and De Jong 1986; Metz et al. 1992).

238 To evaluate the potential for directional evolution on the adaptive trait of an
239 established resident population, we need to calculate the invasion fitness of a mutant with a
240 trait value that slightly differs from that of the resident (Metz et al. 1992). This invasion
241 fitness is given by the sum of the mutant's growth rates across all depths, evaluated at the
242 demographic equilibrium of the resident population (Online appendix B). In case of positive
243 invasion fitness, the mutant generically replaces the resident (Geritz et al. 2002) and the
244 resident population's trait value shifts accordingly. This directional selection on the trait of a
245 monomorphic resident population persists as long as the selection gradient

$$246 \quad g_i = \left. \frac{\partial}{\partial T_0} f_0 \right|_{T_0=T_i}$$

247 remains positive or negative. The point at which directional evolution comes to a halt is
248 referred to as an 'evolutionarily singular strategy' (Metz et al. 1996).

249 Once evolution has reached such a singular strategy, selection becomes either
250 stabilizing or disruptive, depending on the local shape of the fitness landscape described by
251 f_0 as a function of T_0 . If the singular strategy is located at a local fitness maximum, no more
252 invasion of any nearby mutant morph is possible, so that the singular strategy is evolutionarily
253 stable. If the singular strategy is instead located at a local fitness minimum, it is evolutionary

254 unstable. This means that all nearby mutant morphs can invade the system and establish a
255 second resident population, so the system becomes dimorphic. A strategy at which selection
256 becomes disruptive and to which directional evolution can nevertheless converge is known as
257 an ‘evolutionary branching point’ (Metz et al. 1992, 1996; Geritz et al. 1998). Evolutionary
258 branching can occur if the second derivative of the invasion fitness,

259
$$h_i = \left. \frac{\partial^2}{\partial T_0^2} f_0 \right|_{T_0=T_i},$$

260 is positive. An evolutionary branching point implies that a mutant can invade and stably
261 coexist with the resident population. In other words, an evolutionary branching point can give
262 rise to a protected dimorphism, in which each of the two morphs can invade the other.

263 The direction of dimorphic evolution after evolutionary branching is determined
264 analogously to the monomorphic case. This is achieved by testing a system with two
265 established residents for invasibility by a mutant. If a mutant close to one of the residents has
266 positive invasion fitness, it successfully invades the system and replaces that resident. Such
267 dimorphic directional evolution then proceeds until a strategy pair is reached at which the two
268 selection gradients vanish. At this point, further evolutionary branching may occur, if at least
269 one of the populations is situated at a fitness minimum. Otherwise, evolution comes to a halt.

270 **Sexual reproduction**

271 We incorporate sexual reproduction into the model following standard procedures
272 (Roughgarden 1979; Bulmer 1980). In the asexual model, the phenotype is normally
273 faithfully inherited from parent to offspring, except for mutations occurring with small
274 probability. In contrast, in the sexual model, offspring phenotypes are always subject to
275 variation resulting from genetic segregation and recombination. Also, mating is assortative, so
276 individuals pair up for reproduction based on their similarity in trait value.

277 Specifically, an individual j with trait value T_j mates with another individual k
 278 with trait value T_k according to a Gaussian probability distribution around its trait value with
 279 standard deviation σ_a ,

$$280 \quad K_a(T_j, T_k) = \exp(-\frac{1}{2}(T_j - T_k)^2 / \sigma_a^2) / (\sqrt{2\pi}\sigma_a).$$

281 Thus, σ_a measures the width of the mating kernel K_a , and therefore determines the degree of
 282 assortment, with large σ_a corresponding to random mating and smaller σ_a corresponding to
 283 increasingly assortative mating. The probability that phenotype j reproduces with phenotype
 284 k further depends on the frequency F_k of that phenotype in the population. Assuming a
 285 discretized trait space with a finite number n of possible phenotypes, $T_i = 1, \dots, n$, the
 286 frequency F_k is given by

$$287 \quad F_k = B_k / \sum_{i=1}^n B_i.$$

288 To account for the effects of genetic segregation and recombination, the offspring trait
 289 value T_i is drawn from the following probability density,

$$290 \quad K_{sr}(T_i, T_j, T_k) = \exp(-\frac{1}{2}(T_{jk} - T_i)^2 / \sigma_{sr}^2) / (\sqrt{2\pi}\sigma_{sr}),$$

291 i.e., from a Gaussian distribution around the mid-parental trait value $T_{jk} = (T_j + T_k) / 2$, with
 292 standard deviation σ_{sr} . Thus, σ_{sr} measures the width of the segregation-recombination kernel
 293 (Roughgarden 1979), which describes the distribution of offspring traits for given parents.
 294 This approach assumes that the variation introduced by the segregation and recombination of
 295 genes is constant over time (Roughgarden 1979).

296 The probability density to be born with trait value T_i for a given maternal trait value
 297 T_j depends on the frequency distribution of phenotypes in the population, the degree of
 298 assortment, and the variation with which the offspring trait value is inherited from the parents,

$$299 \quad w(T_i, T_j) = Z_{ij}^{-1} \sum_{k=1}^n F_k K_a(T_j, T_k) K_{sr}(T_i, T_j, T_k),$$

300 where the normalization constant $Z_{ij} = \sum_{k=0}^n F_k K_a(T_j, T_k) K_{sr}(T_i, T_j, T_k)$ is chosen so that
 301 $\sum_{i=0}^n w(T_i, T_j) = 1$. Hence, the probability to be born with trait value T_i from all possible
 302 matings is given by

$$303 \quad w(T_i) = \sum_{j=1}^n F_j w(T_i, T_j).$$

304 To determine whether the evolved trait distribution in the sexual model is unimodal
 305 (representing a single species) or bimodal (representing two specialist species), we directly
 306 compare the biomass densities of the phenotypic class adapted to $T_i = 8.1^\circ\text{C}$ (corresponding
 307 to the evolutionarily singular strategy in the asexual model) to the biomass densities of the
 308 phenotypic classes adapted to $T_i = 5^\circ\text{C}$ and $T_i = 10^\circ\text{C}$ (corresponding to the two specialists at
 309 the evolutionary endpoint of the asexual model). When the intermediate class is less frequent
 310 than both the warm-adapted and the cold-adapted classes, we consider the outcome of the
 311 sexual model to represent two distinct morphs. As for the asexual model, we use numerical
 312 analyses to investigate the sexual model.

313 RESULTS

314 To investigate the evolutionary dynamics of our asexual model and to answer the question
 315 whether evolutionary branching may occur under the specific ecological conditions specified
 316 by the data-based parameter estimates and empirically motivated structural model

317 assumptions, we apply the methods of evolutionary analysis described above. We visualize
318 our results using a ‘pairwise invasibility plot’ (PIP) and a ‘trait-evolution plot’ (TEP; Geritz et
319 al. 1998). Furthermore, to investigate the dynamics of the sexual model, we evaluate how the
320 evolutionary outcome is determined by the widths of the mating kernel (degree of assortment)
321 and the segregation-recombination kernel (offspring distribution).

322 Fig. 2 shows the PIP for all trait-value combinations of a resident and a mutant morph
323 with temperature optima between 4°C and 12°C for a foraging optimality α of 6. A
324 monomorphic population starting with any trait value is subject to directional selection and
325 therefore evolves until the singular strategy is reached at about 8.1°C. At this point, the
326 population experiences a fitness minimum and selection becomes disruptive, so that
327 evolutionary branching can occur.

328 Fig. 3 provides a TEP to visualize the subset of trait combinations for which the two
329 morphs are able to coexist in a protected dimorphism (grey area). It also depicts the direction
330 of selection gradients (arrows), the evolutionary isoclines along which one of the selection
331 gradients vanishes (continuous and dotted lines), and the expected course of dimorphic
332 directional evolution after evolutionary branching (dashed lines). Within the area of
333 coexistence, the two morphs evolve to a point (large filled circles) at which the two isoclines
334 intersect. The resultant trait combinations are 5.0°C and 10.0°C. At these trait combinations,
335 both selection gradients vanish and directional selection for the two morphs thus ceases. Since
336 both isoclines are at a local fitness maximum at their intersection, the resulting dimorphism is
337 locally evolutionarily stable. Hence, no secondary evolutionary branching can take place in
338 the system, and the two morphs are recognized as the model-predicted evolutionary outcome.

339 Fig. 4 compares the model-predicted trait values and biomass distributions with the
340 empirically observed trait values and biomass distributions. Fig. 4A shows time series of the
341 evolving monomorphic and dimorphic optimum foraging temperatures, and compares the

342 latter with the observed temperature preferences of the two coregonids in Lake Stechlin. The
343 initial trait value of the monomorphic population does not alter the evolutionary outcome,
344 since the evolutionary branching point is globally convergence stable (i.e., a monomorphic
345 population evolves towards this point irrespective of its initial value). The temperature optima
346 at the endpoint of dimorphic evolution (5.0°C and 10.0°C) are in good agreement with the
347 measured temperature preferences of the two coregonids (4.2°C and 9.0°C) (Ohlberger et al.
348 2008c). Fig. 4B shows the depth distributions of the two morphs after dimorphic evolution
349 has come to a halt and compares their average depths with the year-round average depths
350 observed for the coregonids in Lake Stechlin. This comparison shows that the model-
351 predicted average depths of the populations at the endpoint of dimorphic evolution match
352 very well those observed in the field (Helland et al. 2007).

353 Fig. 5 displays the evolutionary outcome for the sexual model in dependence on the
354 functions describing assortative mating and segregation/recombination. The diagram shows
355 that speciation becomes more likely as the standard deviations of these two kernels decrease.
356 For low degrees of assortment (high σ_a), and for a wide offspring trait distribution around the
357 mid-parental trait value (high σ_{sr}), the population remains monomorphic, centered around a
358 trait value of about 8°C. The threshold for the width of the assortative-mating kernel above
359 which diversification cannot occur is ca. 0.9, which means that a focal individual choosing
360 between two potential mates is 65% as likely to choose a mate whose trait value differs by
361 1°C than to choose a mate with its own trait value. The threshold for the width of the
362 segregation-recombination kernel above which diversification cannot occur is ca. 0.6, which
363 means that the offspring trait value has a probability of ca. 90% to differ by less than 1°C
364 from the mid-parental trait value. Our results show that the two thresholds for the assortative-
365 mating kernel and the segregation-recombination kernel are fairly independent of each other.

366 Finally, Fig. 6 shows a trait-evolution plot for the sexual model for a given degree of
367 assortment ($\sigma_a = 0.2$) and a given variance of the offspring trait distribution ($\sigma_{sr} = 0.2$). This
368 setting yields a dimorphic outcome, with the highest biomass densities at trait values around
369 5°C and 10°C, which is in accordance with the values we have identified above for the
370 asexual model and with the values that have been reported for the empirically observed
371 thermal-preference traits of the two species (Ohlberger et al. 2008c).

372 A main target parameter for a sensitivity analysis of our model is the foraging
373 optimality α , since the evolutionary dynamics are much affected by this parameter. An α -
374 value of less than about 1 results in a monomorphic evolutionarily stable strategy at an
375 intermediate temperature optimum, whereas values larger than about 10 may lead to
376 secondary evolutionary branching, and thus to the evolutionary establishment of higher
377 degrees of polymorphism. It is therefore reassuring to confirm that even considerable
378 variations in α (over the range of 1 to 10, compared with the value of $\alpha = 6$ estimated from
379 the empirical distribution patterns of the two coregonids in the lake; Fig. B1) do not have a
380 qualitative effect on the evolutionary outcome in our model. We note, however, that, as
381 higher values of α approximate the foraging behavior of the fish reasonably well (Fig. 1B),
382 the ecological conditions in the lake might favor multiple evolutionary branching. In that
383 case, other factors such as the degree of assortment or the distribution of offspring trait values
384 could have prevented a second diversification event among the studied coregonids. All other
385 model parameters are based on empirical data from Lake Stechlin and the species pair, are
386 direct system characteristics, or have been taken from other literature on coregonids (Online
387 appendix A). In order to account for uncertainty in the empirically derived parameters and to
388 check the robustness of our results to changes in these parameters, we performed a univariate
389 sensitivity analysis by increasing or decreasing each parameter value by 10% and evaluating
390 the resulting trait values at the evolutionary endpoint of the asexual model (Table C1). All

391 considered parameter perturbations allow for evolutionary branching and result in trait values
392 for the two morphs that are similar to those predicted by the non-perturbed model – which
393 means that they are also similar to the experimentally determined trait values of the two
394 coregonids. The trait value of the cold-adapted phenotype (4.6-5.5°C) was most strongly
395 affected by changes in the minimum temperature at lake bottom (T_{\min}), i.e., by the lower
396 thermal limit of the habitat available to the fish. The trait value of the warm-adapted
397 phenotype (9.1-10.8°C) was most strongly affected by changes in the maximum temperature
398 at which the fish forage (T_{\max}), i.e., by the upper thermal limit of their habitat (Online
399 appendix C).

400 **DISCUSSION**

401 It is widely believed that the same ecological conditions that produce disruptive selection and
402 cause adaptive speciation drive the ecological differentiation that enables the coexistence of
403 closely related species in sympatry (Schluter 2000; Coyne and Orr 2004). Previous empirical
404 studies on the species pair in Lake Stechlin had revealed that the two coexisting coregonids
405 have diverged with respect to their vertical distribution in the lake (Helland et al. 2007,
406 2009), the temperature-dependence of their metabolism (Ohlberger et al. 2008a), and their
407 associated thermal preferences (Ohlberger et al. 2008c). The concordance in the ecological,
408 physiological, and behavioral differentiation of the species with respect to temperature
409 suggests thermal specialization as the main driver of their eco-evolutionary divergence. This
410 divergence may have originated from two preexisting species through character displacement,
411 or from a single ancestral species through sympatric speciation, with the latter option being
412 supported by genetic analyses (Schulz et al. 2006; Mehner et al. 2010a).

413 We have shown by analyses of the evolutionary dynamics of the model introduced in
414 this study, that under the specific empirical conditions observed in Lake Stechlin (*i*)
415 evolutionary branching of a single ancestral population into two morphs with distinct thermal

416 specialization should have been favored by the ecological conditions, that (ii) the two model-
417 predicted populations can stably coexist, and (iii) that they have temperature optima and
418 depth distributions that closely correspond to those in the field. The model thereby supports
419 the hypothesis that the temperature-depth gradient in this system has mediated the thermal
420 specialization of the two species, thereby allowing them to occupy different thermal niches
421 along the vertical lake axis. This result clearly contrasts with the commonly observed
422 divergence of many other sympatric fish pairs, especially in newly colonized postglacial
423 lakes. The common situation in these lakes is the occurrence of limnetic-benthic species pairs
424 that have diverged into distinct ecotypes or species by exploiting either benthic food in
425 profundal or littoral habitats, or planktonic food in pelagic habitats, giving rise to an
426 ecological pattern known as trophic polymorphism (Schluter and McPhail 1993; Lu and
427 Bernatchez 1999; Knudsen et al. 2006). Our results are in line with recent findings suggesting
428 that divergent thermal adaptation of sibling taxa caused by differential selection between
429 thermal environments may underlie reproductive isolation or allow for stable coexistence in
430 several animal and plant systems (Keller and Seehausen 2012).

431 To our knowledge, we present the first data-based model of adaptive diversification
432 along a continuous environmental gradient. Based on our model analyses, we conclude that
433 the ecological conditions in Lake Stechlin are prone to evolutionary branching, and that
434 sympatric speciation of coregonids in this lake is thus an ecologically plausible scenario.
435 Since the hypothesized speciation process occurs along an environmental gradient, i.e., along
436 a spatial dimension in our model, it possesses elements of parapatric speciation. However,
437 since isolation by distance is not expected to play any significant role over the short spatial
438 ranges that characterize the vertical differentiation of the two species, the particular speciation
439 process modeled here presumably is very close to the sympatric end along the sympatric-
440 allopatric continuum of parapatric speciation. Other authors who have attempted to describe

441 the evolutionary dynamics of a particular system based on empirical data have used models in
442 which selection (acting on several traits) is density-dependent and the environment is
443 represented by a number of novel and discrete ecological niches (e.g., Gavrilets and Vose
444 2007; Gavrilets et al. 2007). The main advantage of an empirically calibrated mathematical
445 model of adaptive speciation is its ability to help identify the crucial environmental factors in
446 processes of evolutionary diversification. Previously, the theory of adaptive dynamics has
447 been successfully applied to developing various strategic models of sympatric and parapatric
448 speciation, although its usefulness for approaching evolutionary problems has occasionally
449 been challenged (e.g., Gavrilets 2005; Waxman and Gavrilets 2005). Our work here, although
450 limited by the underlying simplifying assumptions and by a focus on one specific natural
451 system, shows that adaptive dynamics theory also offers useful tools for identifying, through
452 interfacing with the relevant empirical data, key ecological factors involved in processes of
453 evolutionary diversification under natural conditions.

454 The endpoint of evolutionary dynamics in our model describes a dimorphic fish
455 population consisting of one morph with a lower temperature optimum that occupies deeper
456 water layers relative to another morph with a higher temperature optimum that occupies
457 shallower water layers (with optimum foraging temperatures of 5.0°C and 10.0°C,
458 respectively). This model outcome matches the experimentally determined species-specific
459 thermal preferences and corresponding temperatures of lowest swimming costs (4.2°C and
460 9.0°C; Ohlberger et al. 2008a,c), and is in good accordance with the experienced water
461 temperatures of the coregonids in Lake Stechlin (4.0-6.0°C and 6.5-9.0°C; Mehner et al.
462 2010b). These comparisons show that the two populations that evolve in our model well
463 represent the natural populations in terms of their distributions and trait values. Interestingly,
464 Lake Stechlin and the nearby Lake Breiter Luzin are the only two lakes in Germany in which
465 sympatric pairs of coregonids occur. Both are located at the southern border of the *C. albula*

466 distribution range (Kottelat and Freyhof 2007) and are the deepest in that region (Mehner et
467 al. 2005), thus featuring the broadest temperature ranges and steepest gradients among lakes
468 worldwide in which *C. albula* occurs (Kottelat and Freyhof 2007). Besides temperature, light
469 intensity decreases continuously with lake depth. This might have facilitated a differentiation
470 in light-dependent feeding efficiencies along the depth gradient. However, there is good
471 evidence that the two species do not differ in their foraging efficiency at light intensities
472 experienced in the lake (Ohlberger et al. 2008b). Other factors that might have facilitated the
473 evolutionary divergence of an ancestral population include a species-poor environment and a
474 high level of intraspecific competition at the time of colonization, conditions that are often
475 found in newly colonized postglacial lakes, and are believed to promote ecologically based
476 adaptive diversification in fishes (Vamosi 2003; Bolnick 2004).

477 Environmental temperature gradients are common phenomena in nature, with a major
478 effect on biogeographical patterns of species distributions. For instance, the large-scale
479 biogeography and the small-scale distribution of fishes are broadly shaped by temperature
480 patterns (Somero 2002, 2005; Pörtner et al. 2007, 2010). Furthermore, it has been reported
481 that niche segregation of fishes can be facilitated by differentiation in thermal preference
482 (Larsson 2005) and that adaptive differences in thermal physiology promote ecological
483 divergence between closely related intertidal fishes (Hilton et al. 2008). However, the general
484 importance of environmental gradients for small-scale patterns of species distributions and
485 their diversification has long been a matter of debate. The traditional understanding is that
486 gradual evolution of a quantitative trait of a single population along an environmental
487 gradient would necessarily result in local adaptation to an optimum, with some smooth and
488 continuous variation around this optimum occurring due to the homogenizing effects of gene
489 flow and stabilizing selection (e.g., Kirkpatrick and Barton 1997). More recent theoretical
490 models, however, show that frequency-dependent selection due to local competition in trait

491 space and physical space naturally results in discrete clustering of an adaptive trait along a
492 continuous environmental gradient: these models demonstrate the plausibility of evolutionary
493 branching under such ecological conditions in asexual populations (Mizera and Meszéna
494 2003; Leimar et al. 2008), as well as the evolution of reproductive isolation in sexual
495 populations (e.g., Doebeli and Dieckmann 2003; Heinz et al. 2009). Interestingly, empirical
496 evidence for the maintenance and emergence of adaptive divergence along environmental
497 gradients is accumulating, in particular relating to depth gradients of aquatic environments
498 (Vonlanthen et al. 2009; Ingram 2011). The results presented here support the theory of
499 ecologically based adaptive diversifications along environmental gradients. The two
500 populations segregate in space, even though the underlying spatial range is continuous and
501 does not feature any intrinsic bimodality of environmental conditions.

502 There is ongoing debate among evolutionary ecologists whether or not adaptive,
503 sympatric speciation is a plausible and common scenario for the evolution of biological
504 diversity (e.g., Bolnick and Fitzpatrick 2007, Räsänen and Hendry 2008). Such a process
505 requires the evolution of two ecologically distinct and stably coexisting populations from a
506 common ancestor, as well as the emergence of reproductive isolation between the diverging
507 populations. The purpose of the present model was to identify potential ecological drivers and
508 physiological determinants of the diversifying process in a well-studied natural system. Our
509 results suggest that adaptive diversification would likely be promoted by natural selection
510 under the specific ecological conditions studied through our model. We found that
511 evolutionary diversification in the sexual model is likely if the probability of choosing the
512 same phenotype over a phenotype that differs in trait value by 1°C is greater than 65%
513 (corresponding to $\sigma_a = 0.9$). To our knowledge, no empirical estimates exist of mating
514 probabilities based on traits related to thermal preference/performance of fishes within a
515 similar ecological context. Yet, some degree of phenotype matching seems likely for the

516 coregonids, considering that changes in temperature preference are associated with changes in
517 habitat use, and potentially, in growth efficiency (Mehner et al. 2011). We also found that
518 diversification in the sexual model is likely if the probability that the offspring trait value
519 differs by less than 1°C from the mid-parental trait value is greater than 90% (corresponding
520 to $\sigma_{sr} = 0.6$). The resulting standing variation in a hypothetical monomorphic population
521 corresponds to a genetic coefficient of variation of 13% (Houle 1992). For comparison, a
522 value of 6% was used for the initial genetic coefficient of variation in an evolutionary model
523 parameterized for Atlantic cod (*Gadus morhua*), a value that was considered conservative
524 (i.e. low) in light of the available empirical evidence (Enberg et al. 2009).

525 Reproductive isolation between the Lake Stechlin coregonids is attained via a
526 phenological separation in terms of spawning times, with *C. albula* spawning in late autumn
527 and *C. fontanae* in spring (Schulz and Freyhof 2003). This temporal separation of
528 reproduction, which hinders gene flow between the populations, might have evolved as a by-
529 product of thermal specialization, owing to slower maturation and later spawning at lower
530 temperatures. In fact, the majority of other European populations of the *C. albula* complex are
531 autumn-spawners (Kottelat and Freyhof 2007; Mehner et al. 2010a) and generally show
532 thermal preferences far above that of *C. fontanae* in Lake Stechlin (Ohlberger et al. 2008c).
533 Another potential by-product of such a thermal adaptation is differentiation in body size, as
534 observed for the coregonids in Lake Stechlin. Mature females of *C. fontanae* are smaller (87-
535 126 mm) than those of sympatric *C. albula* (134-167 mm), with both maturing at about 1½
536 years (Schulz and Freyhof 2003). Colder water temperatures slow down the rate of energy
537 assimilation and hence of growth and maturation (Mehner et al. 2011). In consequence, cold-
538 adapted phenotypes are more likely to mature later than warm-adapted ones, which may lead
539 to a phenological separation in spawning time. This assortment may be reinforced by the lack
540 of food during winter months, which can impede maturation and thus lead to a further delay

541 in maturation until spring. Such a bimodal spawning opportunity would strengthen assortment
542 based on similarity in optimum temperature. Reproductive isolation through assortative
543 mating based on differences in spawning time and/or location, possibly in combination with
544 differentiation in body size, has been described for many species, including freshwater fishes
545 in postglacial lakes (Jonsson and Jonsson 2001; Østbye et al. 2004). It should also be noted,
546 however, that whether assortative mating preceded or followed ecological divergence remains
547 uncertain in most cases (Snowberg and Bolnick 2008).

548 While other mechanisms in principle might have brought about the temporal separation
549 in spawning times, there are no data supporting this view. Furthermore, the evolution of
550 reproductive isolation is not sufficient for explaining the stable coexistence of species if these
551 are ecologically identical (Coyne and Orr 2004). Because the two Lake Stechlin coregonids
552 have surprisingly similar diets, their differential use of thermal microhabitats appears to be
553 the only mechanism by which they have achieved ecological segregation. Whether divergence
554 in thermal-microhabitat use and the corresponding adaptations caused a shift in spawning
555 time or whether causation has worked in the opposite direction cannot be answered based on
556 present knowledge. It should also be acknowledged that the ecological conditions in the lake
557 have not been stable since the last glaciation and that past climate events might have
558 facilitated the segregation.

559 Our study highlights the potential role of temperature gradients for shaping processes
560 of ecological and evolutionary diversification through thermal adaptation. This perspective on
561 the ecologically based evolution of sympatric species is currently receiving increasing
562 scientific interest and presents a promising path for future research (Keller and Seehausen
563 2012; Svensson 2012). Prospective work may use the present model as a starting point, for
564 example, to develop a predictive evolutionary framework for evaluating the impact of future
565 temperature changes due to global warming on fish populations similar to those studied here.

566 Such an approach could provide insights into the potential implications of thermal
567 specialization for species interactions and for their adaptive responses to altered
568 environmental conditions.

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753

754 **FIGURE LEGENDS**

755 **Fig. 1:** Functions used as model input (continuous lines) together with the corresponding
756 empirical data (open circles). (A) Temperature as a function of depth [$T(x) = T_{\min} + T_{\text{diff}} e^{-\phi x^2}$].
757 (B) Zooplankton carrying capacity density as a function of depth [$K(x) = Z_{\min} + Z_{\text{diff}} e^{-\phi x}$]. (C)
758 Foraging probability as a function of depth [$F(x) = (T_{\max} - T(x))_+ / (T_{\max} - T_{\min})$]. (D) Capture
759 success as a function of depth [$n(x) = e^{-\psi x}$]; (E) Foraging efficiency as a function of
760 temperature [$r(T, T_i) = r_{\min} + (1 - r_{\min}) e^{-\nu(T-T_i)^2}$]. (F) Maintenance costs as a function of
761 temperature [$m(T) = \gamma m_0 B_{\text{mean}}^\beta e^{\omega T}$]. See Online appendices A and B for further details.

762 **Fig. 2:** Pairwise invasibility plot (PIP) for trait values between 4°C and 12°C. The grey areas
763 indicate at which trait-value combinations a mutant phenotype can invade a resident
764 phenotype, i.e., where the mutant's invasion fitness is positive. The evolutionarily singular
765 strategy at which the selection gradient vanishes is indicated by the small filled circle. Based
766 on this PIP, we expect that a monomorphic population continuously evolves along the main
767 diagonal until it reaches the evolutionarily singular strategy at about 8.1°C (which implies
768 that the singular strategy is convergence stable). At this branching point, the population
769 experiences a fitness minimum and selection accordingly becomes disruptive (which implies
770 that the singular strategy is locally evolutionarily unstable).

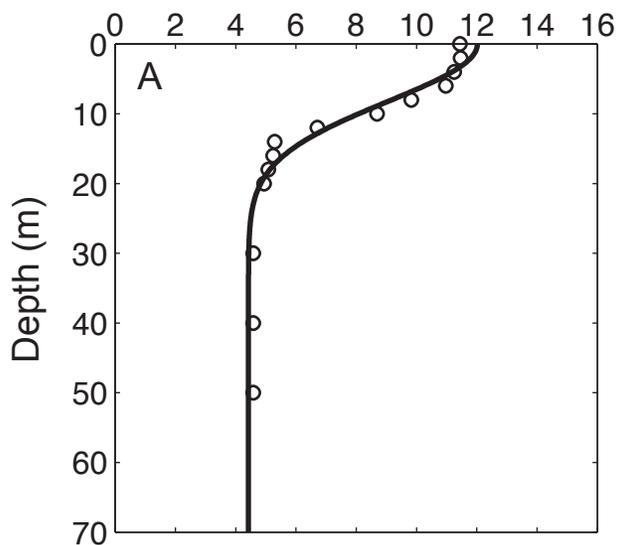
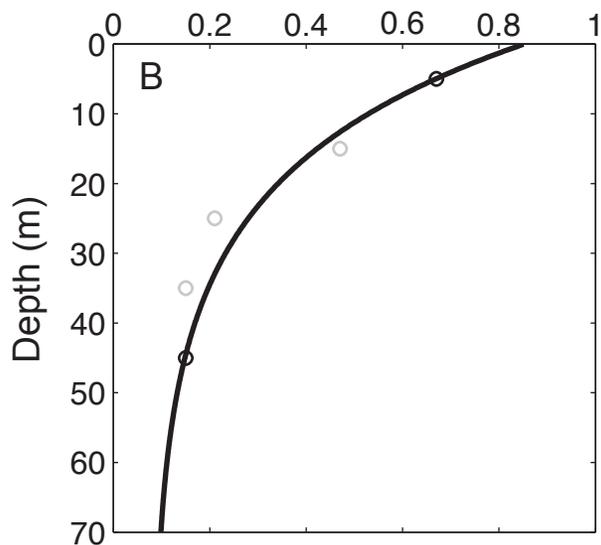
771 **Fig. 3:** Trait-evolution plot (TEP) for trait values between 4°C and 12°C. The parts of this
772 plot that lie above and below the diagonal are fully equivalent, since the numbering of the two
773 considered morphs is arbitrary. The grey areas represent trait-value combinations for which
774 the two morphs can stably coexist, i.e., for which they form a protected dimorphism. Arrows
775 indicate the direction of the selection gradient within this coexistence area. The continuous
776 and dotted lines show the evolutionary isoclines, along which the selection gradient for one

777 morph vanishes. As explained in Fig. 2, evolutionary branching can occur at the small filled
778 circle. The dashed lines show the evolutionary dynamics after branching, which converge to
779 the protected dimorphism at which the two evolutionary isoclines intersect (large filled
780 circles). Since this intersection point is located along the parts of the isoclines that correspond
781 to a fitness maximum (continuous lines), as opposed to a fitness minimum (dotted lines), the
782 resulting dimorphism is not only protected, but also locally evolutionarily stable. The two
783 thus established resident morphs have trait values of 5.0°C and 10.0°C.

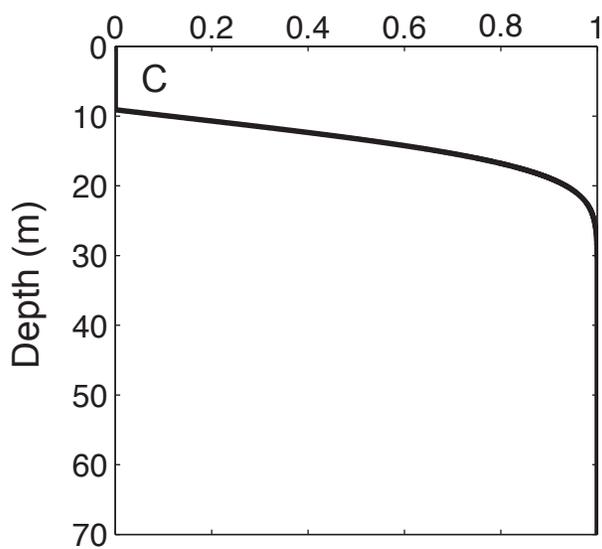
784 **Fig. 4:** Comparison of model predictions and observations of the natural system. (A)
785 Evolution of trait values over time (continuous and dashed lines) in comparison with the trait
786 values observed for the natural populations (dotted lines). The monomorphic evolution (left
787 part) is represented by the PIP in Fig. 2, whereas the dimorphic evolution after branching
788 (right part) is represented by the TEP in Fig. 3. The initial optimum foraging temperature for
789 the monomorphic population (here set to 4°C) can be randomly chosen, since they do not
790 affect the evolutionary outcome of the model. The evolutionary outcome for the optimum
791 foraging temperatures of the two resulting morphs (5.0°C and 10.0°C) is in good agreement
792 with the experimentally determined temperature preferences of the Lake Stechlin coregonids.
793 These are shown as point estimates with 95% confidence intervals on the right side of the plot
794 for *C. fontanae* (dashed line) and *C. albula* (continuous line) (Ohlberger et al. 2008c). Units
795 along the horizontal axis can be considered arbitrary, as they are freely adjustable through the
796 choice of mutational parameters. (B) Depth distributions at the evolutionary endpoint
797 (continuous and dashed curved lines) for the two populations. The resultant model-predicted
798 average depths (continuous and dashed horizontal lines) are compared with the average
799 depths observed for the Lake Stechlin coregonids (dotted lines), showing very good
800 agreement (at 17 m and 24 m, respectively; Helland et al. 2007).

801 **Fig. 5:** Evolutionary outcomes of the sexual model (white: monomorphic; gray: dimorphic)
802 depending on the widths of the assortative-mating kernel (σ_a) and of the segregation-
803 recombination kernel (σ_{sr}). The number of morphs represented by the resulting trait
804 distribution for a given combination of σ_a and σ_{sr} is determined by comparing the biomass
805 density of the single phenotype representing the evolutionarily singular strategy (trait value
806 8.1°C) to the sum of the biomass densities of the two phenotypes representing the
807 evolutionary endpoint of the asexual model (trait values 5°C and 10°C).

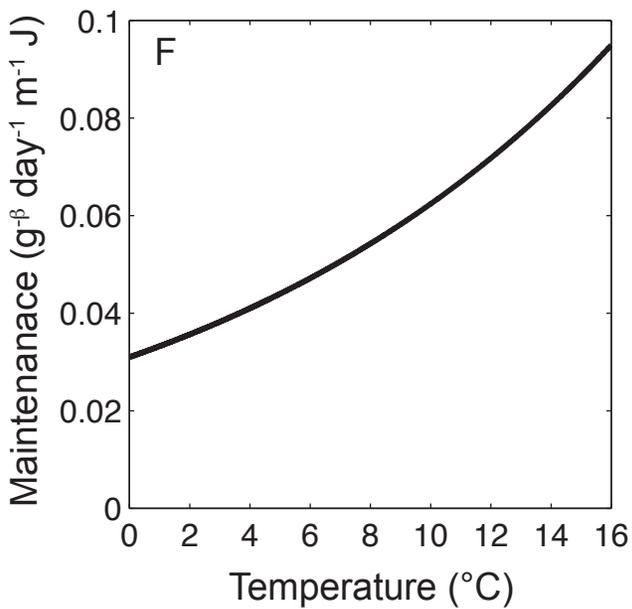
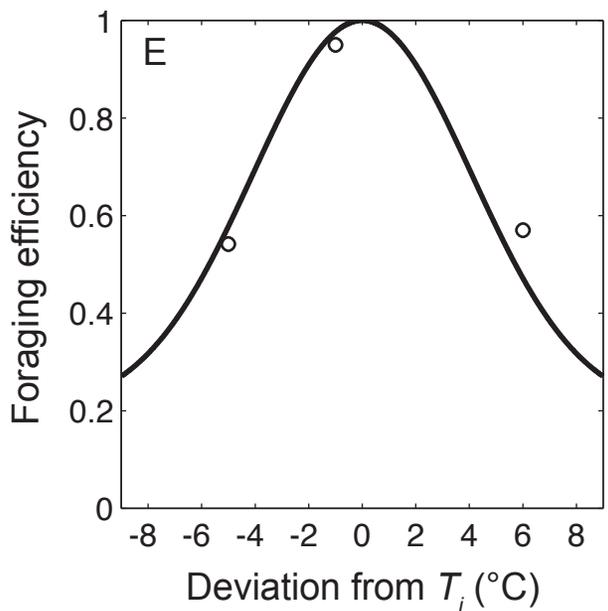
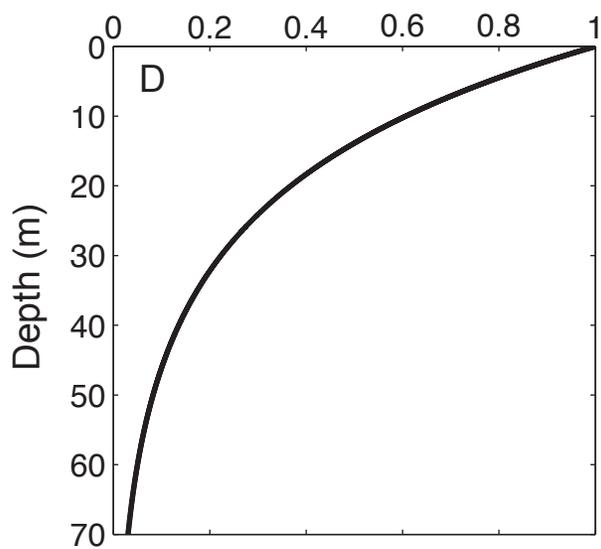
808 **Fig. 6:** Trait-evolution plot for the sexual model, illustrating the distribution of biomass
809 density across all phenotypes (dark grey = high biomass density, light grey = low biomass
810 density). The parameters for the two kernels are $\sigma_a = 0.2$ and $\sigma_{sr} = 0.2$, and the resolution in
811 trait space was set to a value of 0.1. The sexual model yields a dimorphic outcome similar to
812 the asexual model, with the highest biomass densities at trait values around 5°C and 10°C.

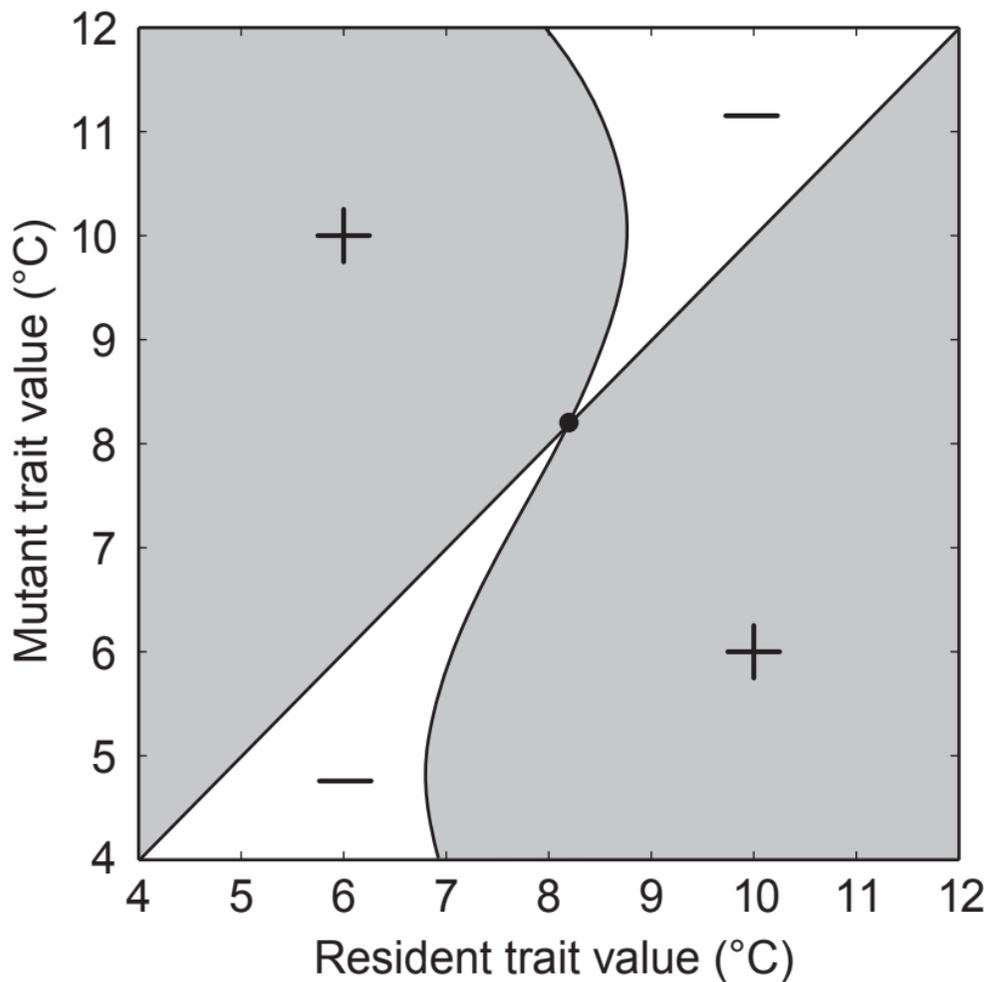
Temperature ($^{\circ}\text{C}$)Zooplankton density (g m^{-1})

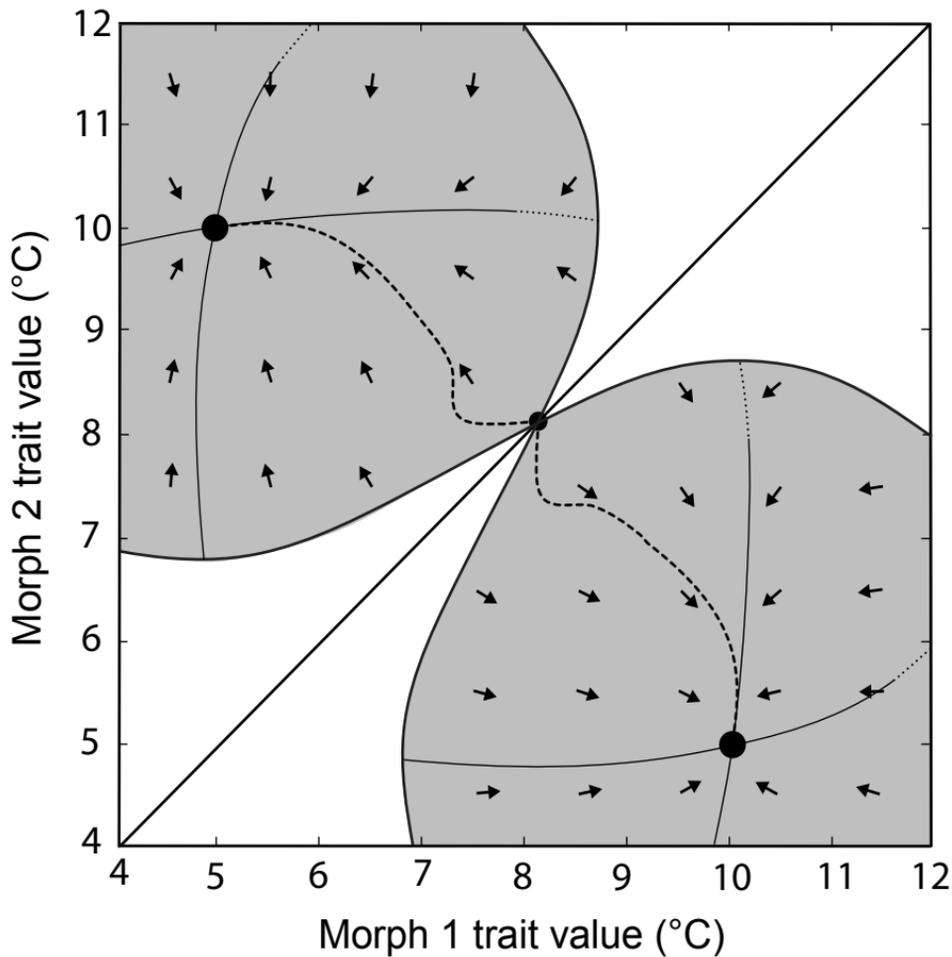
Foraging probability

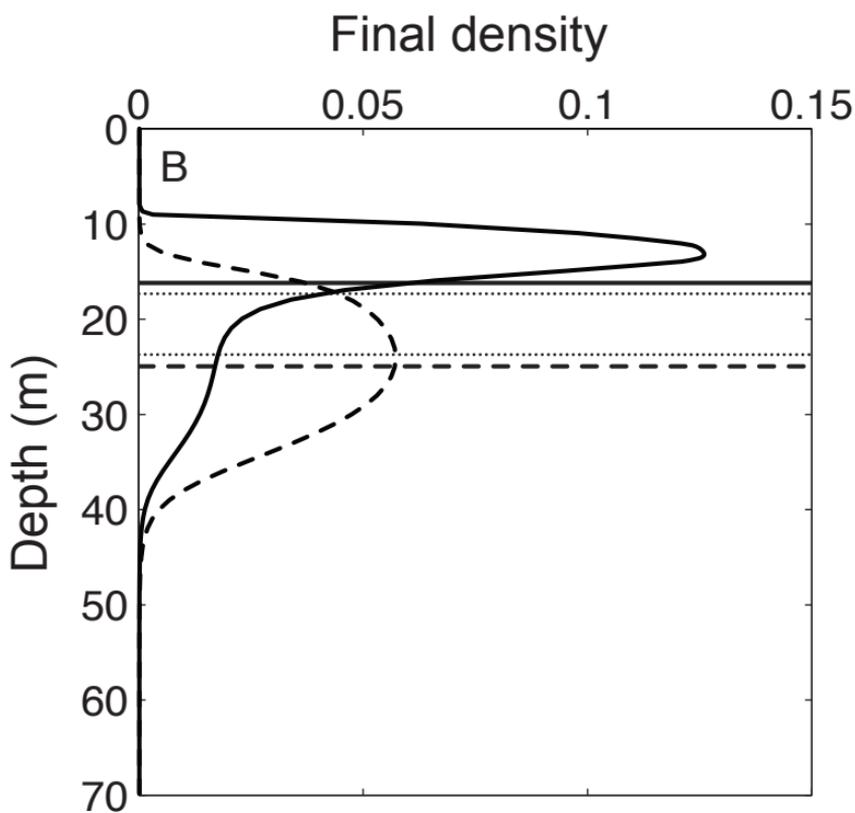
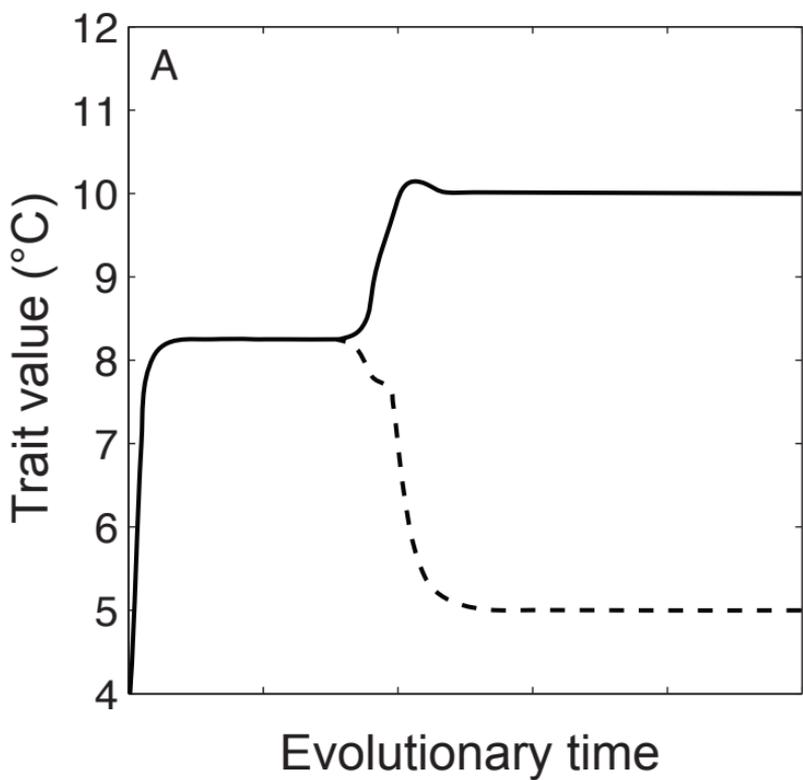


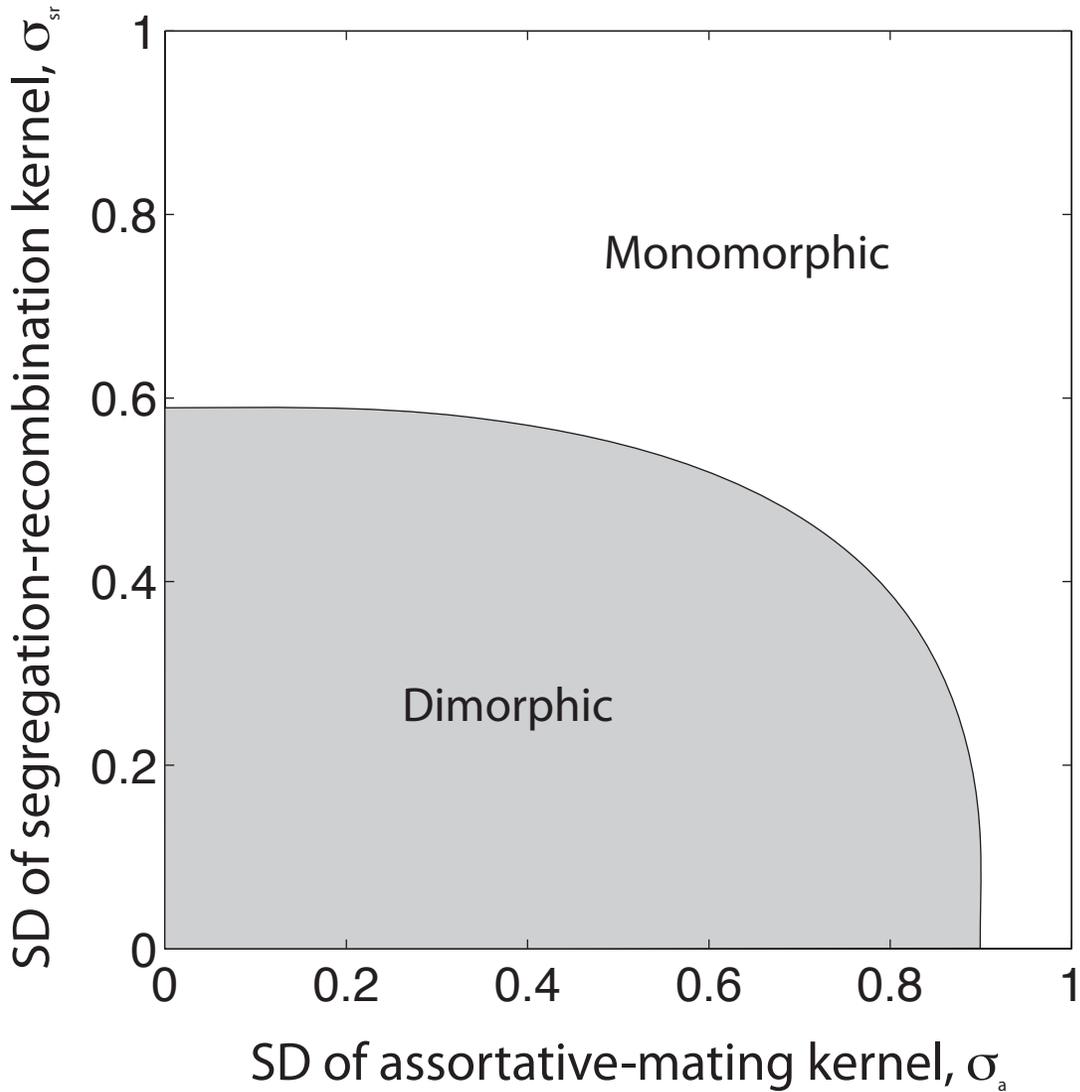
Capture success

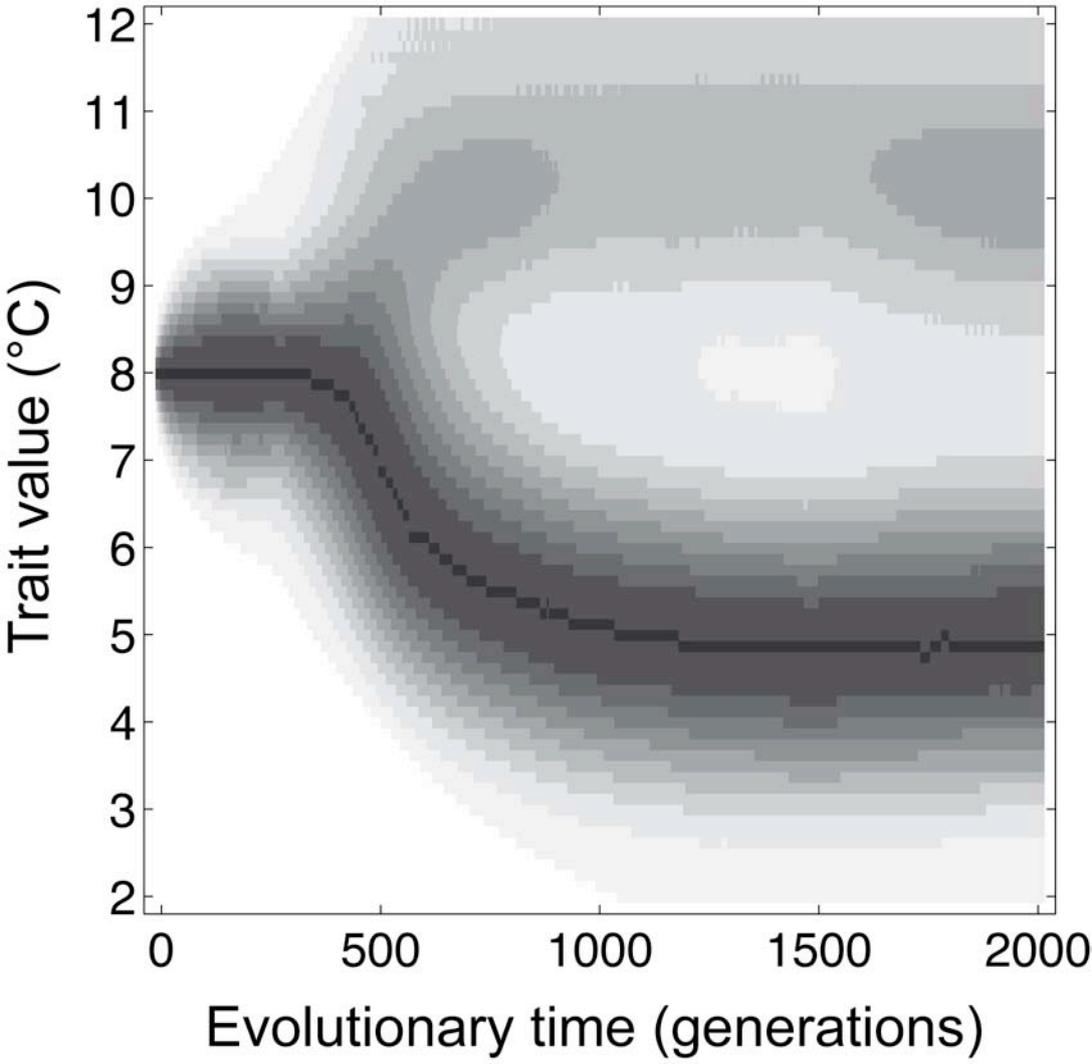












1 **ONLINE APPENDIX A: MODEL VARIABLES, FUNCTIONS, AND PARAMETERS**

2 **Table A1: Model variables and functions**

3	Notation	Unit	Description
4	x	m	Depth
5	$T(x)^*$	°C	Temperature at depth x
6	T_i	°C	Temperature optimum of morph i (evolving trait value)
7	i	n.a.	Morph index ($i = 1, \dots, n$: residents; $i = 0$: rare mutant)
8	$P(x)$	g m^{-1}	Zooplankton density at depth x
9	$P_{\text{eq}}(x)$	g m^{-1}	Equilibrium zooplankton density at depth x
10	$K(x)^*$	g m^{-1}	Carrying capacity of zooplankton at depth x
11	$C(x, T_i)^*$	s^{-1}	Potential consumption rate of morph i with trait T_i at depth x
12	$r(T, T_i)^*$	n.a.	Foraging efficiency of morph i with trait T_i at temperature T
13	$n_i(x)^*$	n.a.	Capture success of morph i at depth x
14	$m(T)^*$	$\text{g s}^{-1} \text{m}^{-1}$	Maintenance-cost density at temperature T
15	$B_i(x)$	g m^{-1}	Biomass of morph i at depth x
16	B_i	g	Total biomass of morph i
17	b_i	g	Average biomass of an individual of morph i
18	$f_i(x)$	$\text{g s}^{-1} \text{m}^{-1}$	Growth-rate density of morph i at depth x
19	f_i	g s^{-1}	Invasion fitness of morph i
20	g_i	$\text{g s}^{-1} \text{°C}^{-1}$	First derivative of invasion fitness at trait value of morph i
21	h_i	$\text{g s}^{-1} \text{°C}^{-2}$	Second derivative of invasion fitness at trait value of morph i

22 * Model input functions calibrated from empirical data as specified in Online appendix B

24 **Table A2: Model parameters**

25	Notation	Unit	Value (\pm SE)*	Source	Description
26	α	n.a.	6.0	1	Degree of foraging optimality
27	x_{\max}	m	68	2	Depth of lake bottom
28	T_{\min}	$^{\circ}\text{C}$	4.42 (\pm 0.14)	3	Minimum temperature at lake
29					bottom
30	T_{diff}	$^{\circ}\text{C}$	7.60 (\pm 0.26)	3	Temperature difference between
31					lake surface and lake bottom
32	T_{\max}	$^{\circ}\text{C}$	9.0	3	Temperature below which the fish
33					forage
34	ϕ	m^{-2}	0.0073 (\pm 0.0006)	3	Measure of how quickly
35					temperature decreases with depth
36	Z_{\min}	g m^{-1}	0.057	4	Minimum zooplankton density at
37					lake bottom
38	Z_{diff}	g m^{-1}	0.78	4	Difference in zooplankton density
39					between lake surface and lake
40					bottom
41	φ	m^{-1}	0.047	4	Measure of how quickly the
42					observed zooplankton density
43					decreases with depth
44	r_p	day^{-1}	0.20	5	Renewal rate of zooplankton
45	r_{\min}	n.a.	0.20	7	Minimum foraging efficiency

46	v	$^{\circ}\text{C}^{-2}$	0.030	7	Measure of how quickly
47					foraging efficiency decreases
48					around T_i
49	C_{\max}	day^{-1}	0.040	8	Maximum consumption rate
50	ψ	m^{-1}	0.050	9	Measure of how quickly capture
51					success decreases with depth
52	λ	n.a.	0.6	10	Conversion factor from
53					zooplankton mass to fish mass
54	γ	g J^{-1}	0.00018	6	Conversion factor from energy to
55					mass
56	m_0	$\text{g}^{-\beta} \text{day}^{-1} \text{m}^{-1} \text{J}$	0.82 (± 0.13)	7	Metabolic scaling factor
57	B_{mean}	g	10	4,7	Mean biomass of fish
58	β	n.a.	0.93 (± 0.03)	7	Metabolic scaling exponent
59	ω	$^{\circ}\text{C}^{-1}$	0.070 (± 0.006)	7	Temperature coefficient of
60					metabolism

61 Sources: (1) Figure B1; (2) Mehner et al. 2005; (3) Mehner et al. 2010; (4) Helland et al.
62 2007, I. P. Helland, unpublished data; (5) Schulz et al. 2004; (6) Gjelland 2008; (7) Ohlberger
63 et al. 2007; (8) Binkowski and Rudstam 1994; (9) Ohlberger et al. 2008; (10) Ohlberger et al.
64 2011

65 * Standard errors could be provided for only some of the empirically estimated parameters,
66 either due to low numbers of data points or because estimates were taken from the literature

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1 **ONLINE APPENDIX B: MODEL FUNCTIONS AND THEIR ESTIMATION FROM** 2 **EMPIRICAL DATA**

3 **Temperature-depth gradient**

4 The temperature-depth gradient $T(x)$ (Fig. 1A) was estimated based on data on average year-
5 round depth-dependent temperatures in Lake Stechlin (Mehner et al. 2010). The function

$$6 \quad T(x) = T_{\min} + T_{\text{diff}} e^{-\phi x^2}$$

7 was fit to the data through least-square optimization. The parameters T_{\min} and T_{diff} describe,
8 respectively, the minimum temperature (which is asymptotically attained at the lake bottom,
9 $x = x_{\max}$) and the temperature difference between lake surface and lake bottom. The

10 parameter ϕ measures how quickly temperature decreases with depth. Lake Stechlin is a
11 stratified lake that freezes irregularly. During periods of ice cover, the stratification pattern
12 reverses. Since ice cover on average lasts only about one month (Kirillin et al. 2012), we used
13 weekly water-temperature records throughout the year to calculate average temperature as a
14 function of depth.

15 **Carrying-capacity density of zooplankton**

16 The carrying-capacity density $K(x)$ of zooplankton was estimated based on the observed
17 average zooplankton densities in the shallowest and deepest parts of Lake Stechlin, where
18 predation by the coregonids can reasonably be assumed to be negligible. The function

$$19 \quad K(x) = Z_{\min} + Z_{\text{diff}} e^{-\varphi x}$$

20 was fitted to the data (Helland et al. 2007; I. P. Helland, unpublished data) through least-
21 square optimization (Fig. 1B). The parameters Z_{\min} and Z_{diff} , respectively, describe the
22 minimum zooplankton density (which is asymptotically attained at the lake bottom, $x = x_{\max}$)
23 and the difference in zooplankton density between lake surface and lake bottom, while the
24 parameter φ measures how quickly zooplankton density decreases with depth.

25 **Dynamics of zooplankton density**

26 Changes in the depth-dependent zooplankton density $P(x)$ are determined by logistic growth
27 and Lotka-Volterra consumption,

$$28 \quad \varepsilon \frac{dP(x)}{dt} = r_p P(x) [1 - P(x) / K(x)] - P(x) \sum_{i=1}^n C(x, T_i) B_i(x),$$

29 where $K(x)$ is the carrying-capacity density of zooplankton at depth x , r_p is the renewal rate
30 of zooplankton, and ε is a parameter separating the timescales of zooplankton dynamics and
31 fish population dynamics. The last term in the equation above is the total zooplankton-
32 consumption density at depth x , calculated as the sum of the zooplankton-consumption
33 densities of all morphs at depth x .

34 **Equilibrium zooplankton density**

35 The zooplankton dynamics in the lake are assumed to be much faster than the fish population
36 dynamics ($\varepsilon \rightarrow 0+$). This leads to a depth-dependent equilibrium zooplankton density,

$$37 \quad P_{\text{eq}}(x) = \left(K(x) - \frac{1}{r_p} \sum_{i=1}^n C(x, T_i) B_i(x) \right)_+,$$

38 where $(X)_+$ equals X for $X > 0$ and 0 for $X \leq 0$. The zooplankton dynamics are thus
39 assumed to show no temporal variation due to abiotic environmental changes or behavioral
40 responses such as diel vertical migration.

41 **Potential consumption rate**

42 The potential consumption rate $C(x, T_i)$ of a fish at depth x with trait T_i is calculated
43 according to

$$44 \quad C(x, T_i) = C_{\text{max}} r(T(x), T_i) n(x),$$

45 where C_{max} is the maximum consumption rate of an individual fish under optimal light and
46 temperature conditions, $r(T, T_i)$ is the foraging efficiency of a fish with trait T_i as a function
47 of temperature T , and $n(x)$ is the capture success as a function of depth x .

48 **Foraging efficiency**

49 The temperature-dependent foraging efficiency (Fig. 1E) was estimated based on empirical
50 data on the temperature-dependent swimming performance in the Lake Stechlin coregonids.
51 Specifically, it was fitted to the temperature-dependent costs of transport in *C. albula*, that is,
52 the energy expenditure per unit distance during swimming (Ohlberger et al. 2007), with the
53 assumption that the feeding rate of the fish scales proportionally with the distance covered
54 when foraging. It was further assumed that foraging efficiency drops symmetrically around
55 the temperature at which foraging efficiency is optimal (Ohlberger et al. 2008a),

$$56 \quad r(T, T_i) = r_{\min} + (1 - r_{\min}) e^{-\nu(T-T_i)^2}.$$

57 The same functional relationship between foraging efficiency and temperature was used for
58 *C. fontanae*.

59 **Capture success**

60 The depth-dependent capture success (Fig. 1D) was estimated based on measurements of the
61 feeding efficiency of the Lake Stechlin coregonids at different light levels (Ohlberger et al.
62 2008b). According to these data, the decrease in capture success with decreasing light
63 intensity was well represented by an exponential relationship,

$$64 \quad n(x) = e^{-\psi x}.$$

65 **Maintenance costs**

66 The maintenance costs (Fig. 1F) are described as a function of temperature and biomass
67 according to the energetic models by Ohlberger et al. (2007),

$$68 \quad m(T) = \gamma m_0 B_{\text{mean}}^{\beta} e^{\omega T},$$

69 where γ converts energy to fish mass, m_0 is the metabolic scaling factor, B_{mean} is the mean
70 biomass of a coregonid, β is the metabolic scaling exponent, and ω is the temperature
71 coefficient of metabolism.

72 **Foraging probability**

73 The depth-dependent foraging probability (Fig. 1C) was estimated based on observations in
74 Lake Stechlin showing that the coregonids rarely forage in the epilimnion at temperatures
75 above 9°C, although food availability is highest near the surface (Helland et al. 2007; Mehner
76 et al. 2010). This might be explained by the higher predation risk from perch (*Perca*
77 *fluviatilis*) in the epilimnion (Mehner et al. 2007), a species that is known to become almost
78 inactive at temperatures below about 9-10°C (Karås and Thoresson 1992). Foraging animals
79 commonly alter their behavior according to the spatial structure of prey availability and
80 predation risk as a consequence of a behavioral trade-off between maximizing food and
81 safety, also referred to as the ‘landscape of fear’ (Brown and Kotler 2004; Searle et al. 2008).
82 Consequently, we assumed that the foraging probability decreases from 1 in the coldest zones
83 at the lake bottom to 0 at $T_{\max} = 9^{\circ}\text{C}$,

$$84 \quad F(x) = (T_{\max} - T(x))_+ / (T_{\max} - T_{\min}),$$

85 where $(X)_+$ again equals X for $X > 0$ and 0 for $X \leq 0$.

86 **Invasion fitness of a rare mutant**

87 Since any mutant invading a system with n resident morphs is at first very rare, the
88 population of mutants initially has a negligible biomass relative to the resident population.
89 Hence, the depth-dependent growth rate of a mutant morph can be calculated based on the
90 biomasses and consumption rates of the residents at equilibrium,

$$91 \quad f_0(x) = C(x, T_0)P_{\text{eq}}(x) - m(T(x)).$$

92 The invasion fitness of the mutant is then given by

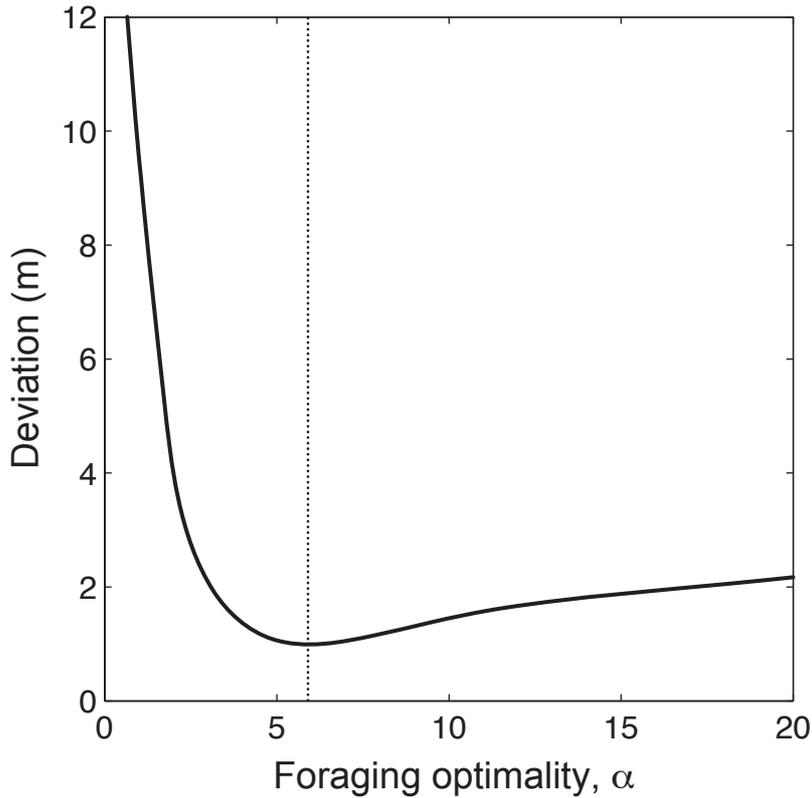
$$93 \quad f_0 = \frac{1}{B_0} \int_0^{x_{\max}} f_0(x)B_0(x)dx ,$$

94 where $B_0(x)$ is the mutant population’s density at depth x and B_0 is its total biomass. In
95 turn, $B_0(x)/B_0$ depends on the mutant’s potential consumption rate, its foraging probability,

96 and the equilibrium resource distribution (which, as the mutant is rare, depends only on the
 97 biomass distribution of the resident morphs),

98
$$\frac{1}{B_0} B_0(x) = B_i \frac{[C(x, T_0) P_{eq}(x) F(x)]^\alpha}{\int_0^{x_{max}} [C(x', T_0) P_{eq}(x') F(x')]^\alpha dx'}$$

99



100

101 **Fig. B1:** Estimation of foraging optimality α . The foraging optimality is a measure of the
 102 degree to which individuals of a population forage according to their physiological optimum,
 103 the potential consumption rate, and the predation risk. The figure shows, as a function of α ,
 104 the sum of absolute values of the deviations between the two model-predicted and the two
 105 observed average population depths of the coregonids. The least deviation occurs for a
 106 foraging optimality of $\alpha = 6$, as indicated by the dotted vertical line.

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1 **ONLINE APPENDIX C: SENSITIVITY ANALYSIS**

2 **Table C1: Sensitivity analysis for all model parameters ($\pm 10\%$) listed in Table A2**

Parameter	Change	Resultant trait values ($^{\circ}\text{C}$)*	
		Morph 1	Morph 2
All	$\pm 0\%$	5.0	10.0
α	+10%	5.0	10.1
	-10%	5.0	9.9
x_{\max}	+10%	5.0	10.0
	-10%	5.0	10.0
T_{\min}	+10%	5.4	10.0
	-10%	4.6	10.0
T_{diff}	+10%	5.0	10.0
	-10%	5.0	10.0
T_{\max}	+10%	5.1	10.8
	-10%	5.0	9.1
ϕ	+10%	5.0	10.0
	-10%	5.0	10.0
Z_{\min}	+10%	5.0	10.0
	-10%	5.0	10.0
Z_{diff}	+10%	5.0	10.1
	-10%	5.0	9.9
φ	+10%	5.1	10.0
	-10%	4.9	10.0
r_{P}	+10%	5.0	10.0
	-10%	5.0	10.0

k	+10%	5.0	10.0
	-10%	5.0	10.0
r_{\min}	+10%	5.0	9.9
	-10%	5.0	10.1
v	+10%	5.0	9.9
	-10%	5.0	10.1
C_{\max}	+10%	4.9	10.1
	-10%	5.0	9.9
ψ	+10%	5.0	10.0
	-10%	4.9	10.0
λ	+10%	4.9	10.1
	-10%	5.0	9.9
γ	+10%	5.0	10.0
	-10%	5.0	9.9
m_0	+10%	5.0	9.9
	-10%	4.9	10.1
B_{mean}	+10%	4.9	10.0
	-10%	5.0	9.9
β	+10%	5.1	9.8
	-10%	4.9	10.1
ω	+10%	5.0	10.0
	-10%	5.0	10.0

3 * Evaluated at the evolutionary endpoint of the asexual model