Evolutionary rescue in a consumer-resource system: Adaptation and persistence depend on the affected ecological traits

Abstract:

With evolutionary rescue, a population that is declining due to an environmental change subsequently adapts to its environment, avoiding extinction. Previous work has shown that negative density-dependence reduces the probability of evolutionary rescue. However, these models make several simplifying assumptions, either that selection acts strictly on intrinsic growth rates, r, or carrying capacity, K. Here, we simulate evolutionary rescue in a consumer-resource model, making explicit several ecological traits that may differentially affect r and K, and considering a self-replenishing resource. We demonstrate that the effect of density-dependence on rescue is indeed negative but weak, and declines with an increasing degree of maladaptation. The effect of density-dependence is also shown to be trait-dependent, since an environmental perturbation that reduces fitness may also increase or decrease resource competition. Furthermore, the relationship between Malthusian fitness following an environmental change and rescue probabilities depends strongly on the affected ecological trait, such that measurements of the rate of decline alone immediately following an environmental change are likely insufficient to predict rescue outcomes. Reductions in the conversion efficiency of resources into offspring are shown to be a greater hindrance to rescue than reductions in resource acquisition rates, whereas populations are able to tolerate substantial reductions in survivorship. Lastly, an abrupt environmental change that also impacts an evolving resource always hinders the evolutionary rescue of consumers, with reductions in resource

24   carrying capacity being more harmful than reductions in resource growth rates. This work

25   improves our theoretical understanding of how ecology and evolution interact to affect

26   evolutionary rescue.

Introduction:

Understanding the factors underlying the ability of populations to persist in response to

environmental change is of utmost importance, particularly in light of the rapidly accelerating

loss of biodiversity due to climate change (Berg et al. 2010, Bellard et al. 2012). One such

mechanism of persistence is through evolutionary adaptation, or specifically evolutionary rescue

(see Carlson et al. 2014; Bell 2017 for reviews), the process by which a population that is

declining due to an environmental perturbation subsequently evolves and adapts to its new

environment, avoiding extinction. Theoretical models of evolutionary rescue (hereafter 'rescue')

are characterized by the concurrent modelling of both evolutionary and demographic processes,

since successful rescue of a population is contingent on its rate of adaptation outpacing its rate

of demographic decline (Gomulkiewicz & Holt 1995). For the most part, rescue models have

considered a single isolated species evolving to avoid extinction (e.g. Gomulkiewicz & Holt

1995; Burger & Lynch 1995; Lande & Shannon 1996). These models have recovered many

important insights into the factors underlying persistence, including the positive roles of additive

genetic variance in fitness and of initial population size. Later work has extended this theory to

consider in detail the effects of population structure (Uecker et al. 2014), recombination (Uecker

& Hermisson 2016), selfing (Uecker 2017), and various other factors (Orr & Unckless 2008,

2014; Lachapelle & Bell 2012; Osmond et al. 2020; Xu 2023). The simplifying assumption of a

single species existing in isolation, however, fails to capture ecological interactions among

individuals such as predation and competition.

Ecological interactions among species may drastically influence demographic and

evolutionary outcomes, and therefore rescue probabilities. For example, predation imposes a

demographic cost on a prey species, accelerating its demographic decline if it is in need of

rescue. However, predation may also positively influence prey rescue if predators preferentially

consume maladapted individuals–accelerating the rate of prey adaptation–and the selective

benefit outweighs the demographic cost (Osmond et al. 2017). A similar concept also applies if

maladapted individuals experience greater interspecific competition (Osmond and de

Mazancourt 2013). Moreover, complex dynamics may arise in co-evolving competitors. For

example, a species with a smaller population size may have a higher chance of rescue than a

competitor with a larger population size if the resource distribution evolves towards its optimum

and away from its competitor (Van Den Elzen et al. 2017). This result is in apparent

contradiction with the classic theoretical result of a positive relationship between population size

and rescue (Gomulkiewicz & Holt 1995), thus underscoring the importance of considering a

system's ecology.

Ecological interactions among conspecifics, particularly in the form of negative density-

dependent growth, also likely reduce rescue probabilities (Chevin & Lande 2010, Osmond & de

Mazancourt 2013, Nordstrom et al. 2023 Draghi et al. 2024). However, the magnitude of this

effect remains controversial in the literature (Bell 2017). Simply defined, negative density-

dependence (hereafter 'density-dependence' for brevity) refers to the phenomenon where

population growth declines with density, and may arise due to resource limitation, predation,

pathogens, and various other factors. It was first argued by Gomulkiewicz & Holt (1995) that

density-dependence would hinder rescue, since a population experiencing density-dependence

would experience greater demographic decline than a density-independent population of

equivalent size and fitness. However, a declining population may also experience ecological

release due to relaxed density-dependence (Carlson 2014). For example, a rapidly declining

population may experience a compensatory increase in the abundance of its biotic resource,

potentially lessening or negating the negative effects of density-dependence on growth (Abrams

2002). Thus, it is possible that the effect of density-dependence is dependent on the rate of

demographic decline and underlying mechanism of density-dependence. The effect of density-

dependence due to competition for a self-replenishing resource on evolutionary rescue,

however, remains unexplored in the rescue literature.

It is also worth noting that controversy over density-dependence may be explained in

part due to differing model specifications (first described by Osmond & de Mazancourt 2013).

Much of the theoretical work arguing a substantial effect of density-dependence on rescue has

modelled selection as acting on intrinsic growth rates, r, modelling a logarithmic decline of

density-dependent effects with population size (Chevin & Lande 2010, Nordstrom et al. 2023).

Conversely, Boulding & Hay (2001) explored rescue outcomes using a model where negative

density-dependence arises through linearly declining fecundity as a function of population size,

instead concluding that negative density-dependence makes a small difference to rescue

outcomes. Thus, we lack a clear theoretical expectation for the effect of density-dependence on

rescue, and it will likely depend on specific features of the empirical system being studied.

While the above models may differ in certain mathematical specifications, all of them

share a number of simplifying assumptions that obscure the relationship between demography

and evolution. It is often either assumed that selection acts solely on intrinsic growth rates, r,

(Chevin & Lande 2010; Nordstrom et al. 2023) or population carrying capacity, K (Osmond & de

Mazancourt 2013). However, these two parameters are likely not independent in reality (Pianka

1970, Mallet 2012, Travis et al. 2023), and there is much empirical evidence to suggest that r

and K may co-vary positively or negatively due to varying life-history strategies in yeast (Wei &

Zhang 2019), bacteria (Reding-Roman et al. 2017; Marshall et al. 2023), and spider mites

(Bisschop et al. 2021). For example, selection acting to increase individual resource acquisition

rates may increase r, but decrease K by reducing total resource availability among conspecifics

at high densities (Agrawal & Whitlock 2012), thus intensifying density-dependence. However,

selection acting to increase the efficiency of resource use would lead to a greater r as well as K

by lowering the per-capita resource requirement for maintenance (Travis et al. 2023). In

contrast, r-selected models of rescue assume that the intensity of negative density-dependence

is solely a function of population size/density, and itself doesn't evolve (since K is held

constant), thus neglecting the above relationships. Currently, we lack a theoretical framework to

understand evolutionary rescue on ecological traits that may differentially affect r and K despite

its widespread empirical evidence. This represents a major gap in our understanding of the

effects of density-dependence on evolutionary rescue.

The use of mechanistic consumer-resource models represent a promising alternative

framework to investigate the effects of density-dependence on rescue probabilities. Briefly,

consumer-resource models explicitly track both consumer and resource population dynamics

(see Sakarchi & Germain 2024 for a review). Consumer growth is a function of resource

acquisition rates, the efficiency of their conversion into offspring, resource abundance, and

mortality rates, whereas resource growth is often modelled using logistic growth minus their

consumption by consumers. Selection can thus act on a consumer's resource acquisition rates,

conversion efficiencies, or mortality rates, each with differing effects of the consumer's intrinsic

growth rate and carrying capacity (Travis et al. 2023). While consumer-resource models

typically lack analytical tractability due to the explicit modelling of both consumer and resource

dynamics as well as evolution, they facilitate a more mechanistic treatment of the effects of

density-dependence on rescue. Moreover, we are able to investigate the effects of an

environmental perturbation that also negatively impacts the growth of a focal species' resource,

which to our knowledge remains unexplored in the literature.

In this paper, we investigate the effects of density-dependence on evolutionary rescue in

a consumer following an abrupt environmental perturbation. Using a simulation-based

approach, we investigate how density-dependence in the form of resource competition affects

rescue in a consumer population experiencing viability or fecundity selection, the effects of an

environmental change that also impacts a co-evolving resource, and the dependence of rescue

outcomes on which life history traits are impacted in the consumer and resource. We find that

density-dependence tends to weakly reduce rescue probabilities, generally agreeing with

previous work on the topic. However, the relative effect of density-dependence declines as the

magnitude of the environmental perturbation increases. We also show that the relationship

between the rate of population decline following an environmental change and rescue

probability depends substantially on the affected ecological consumer trait, with reductions in

conversion efficiencies being most difficult to rescue from. Conversely, we find that populations

are best able to rescue from decreases in survivorship, an effect previously observed by Draghi

et al. (2024). Increases in mortality also generate a negative relationship between mortality and

fecundity, a form of demographic compensation (Villelas et al. 2015), which has also been

observed in species at range edges (Sheth & Angert 2018). Lastly, we show that an

environmental perturbation that also negatively impacts resource growth always lessens

consumer rescue probabilities, with decreases in resource carrying capacity generally being

more costly to consumer rescue.

147 Methods

148 Model

150 **Table 1.** The definition of terms

Term	Meaning
N_t, R_t	Consumer and resource abundance at time-step t
λ_t	Consumer growth rate, or absolute fitness, at time-step t
λ_*	Consumer growth rate immediately following an environmental change
a	Consumer resource acquisition rate
C	Consumer resource conversion efficiency
S	Consumer survivorship
r_N	Consumer intrinsic growth rate
K_N	Consumer carrying capacity
r_R	Resource intrinsic growth rate
K_R	Resource carrying capacity
ϕ	Refers to any ecological trait (a, C, S, r_R, or K_R)
z_ϕ	Phenotypic trait value determining ecological trait ϕ
θ_ϕ	Optimum trait value for phenotype z_ϕ
$\theta_{\phi*}$	Optimum trait value for phenotype z_ϕ following an abrupt environmental change
ω^2	Measure of the width of Gaussian performance function
μ	Base-pair mutation rate
α^2	Variance in the effect size of a single mutation
V_M	Mutational variance: The total amount of new phenotypic variation introduced by mutation in a time-step
V_E	Environmental variance influencing each phenotypic trait

Ecology

We modelled the ecological dynamics of the system using a discrete-time consumer-

resource model (MacArthur 1970; Abrams 2019). The use of discrete time instead of continuous

time permitted the modelling of an explicit life cycle for forward-time, individual-based

simulations. Assuming no evolution in the system for the time being, the ecological dynamics of

the consumers are described by the following recurrence relation,

$$N_{t+1} = N_t(aCR_t + S), \tag{1}$$

where N_t is consumer abundance at time-step t, a is its resource acquisition rate, C is its

conversion efficiency of acquired resources into offspring, R_t is resource abundance at time-

step t, and S is survivorship/viability ($1 - m$, where m is the mortality rate). We model

overlapping generations using a simple form of age-structure, with resource consumption

followed by births and then deaths. Individuals are able to reproduce and are susceptible to

mortality beginning the time-step following their birth, and individual fecundity, B, remains

constant with age (note that $B = aCR_t$). An individual's growth rate, or absolute fitness, at time-

step t, λ_t, is therefore $\lambda_t = aCR_t + S$, a function of an individual's fecundity, aCR_t, and its

probability of survival, S. While fitness in a model of overlapping generations may be measured

in many ways, including growth rates per generation or lifetime reproductive fitness, we focus on

the population's Malthusian parameter, or $\ln[\lambda]$ (Fisher 1930; Kimura 1958). The Malthusian

parameter most appropriately describes the rate of decline of the population per time-step,

which is of most relevance to evolutionary rescue. The effects of changes in consumer

ecological traits a, C, and S on consumer intrinsic growth rates, r_N, and carrying capacity, K_N,

are described in Appendix A. The ecological dynamics of the resource are described by the

following recurrence relation,

$$R_t = R_t(1 + r_R(1 - \frac{R_t}{K_R}) - aN_t), \qquad\qquad (2)$$

where r_R is the intrinsic growth rate of the resource and K_R is its carrying capacity.

Briefly, our modelling of consumer dynamics differs slightly to the model described

originally by MacArthur (1970, 1972), in that we do not explicitly consider a nutritive weight of

the resource. We instead consider nutritive weight to implicitly influence the resource conversion

efficiency, the exact value of which is unimportant given that we are only modelling a single

resource type. Our model also considers mortality rates to be density-independent. This would

be relevant in the case where most mortality is externally imposed, which we reason is likely the

case in many scenarios that lead to rescue, but is indeed not necessarily the case.

Evolution

We model the evolution of these ecological traits of interest (a, C, or S in the consumer;

r_R or K_R in the resource) using a discrete-time quantitative genetic model, commonly used in the

evolutionary rescue literature (Gomulkiewicz & Holt 1995, Bürger & Lynch 1995, Boulding & Hay

2001, etc.). For most of our analyses, we only model the evolution of a single trait in the

consumer, except in the case where we additionally model an evolving resource trait. While this

methodology neglects potential correlations among traits (e.g. a negative correlation between a

and C), it allows us to investigate how rescue is influenced by which trait is affected, all else

being equal.

We model an arbitrary phenotypic trait underlying each ecological trait ϕ. Each individual

has an arbitrary phenotypic trait of value z_ϕ, the sum of the genetic contribution of n additive loci

with additive effects and a normally distributed environmental effect with mean 0 and variance 1

$(V_E = 1)$. The ecological trait value, ϕ_i, of an individual *i* with trait value $z_{\phi,i}$ is then described by

a Gaussian performance function:

$$\phi_i(z_{\phi,i}) = \phi_{max} exp[-\frac{(z_{\phi,i}-\theta_\phi)^2}{2\omega^2}], \tag{3}$$

where θ_ϕ refers to the trait optimum corresponding to the maximal value of ϕ, and ω^2 is a

measure of the width of the function, which was held constant among all traits. Mutations are

drawn from a Poisson distribution at per-locus rate μ, with the effect size being drawn from a

normal distribution with the mean corresponding to the previous allele's effect size and a

specified variance of $0.5V_E$ (i.e. a Gaussian mutation scheme; see Johnson & Barton 2005). To

simplify the model, the mutational variances and widths of the performance functions were held

constant among all evolving ecological traits across simulations, and we assumed that the

optimal phenotypic value underlying the ecological trait is 0 prior to the environmental change.

We discuss specific parameter choices in the *Simulations* section below.

Simulations

General Features

We ran individual-based forward-time stochastic simulations using SLiM v4.01 (Haller &

Messer 2022). Briefly, we initialized our simulations with the consumers and resources at

demographic equilibrium. We then allowed our simulations to run for a period of 2,500

generations to reach mutation-selection balance for the ecological trait of choice, which varied

depending on the experiment. In the next time-step, we introduced an abrupt change in

environment by altering the initial optimal trait value, θ_ϕ, of one of the ecological traits ϕ (a, C, or

S in the consumer; r_R or K_R in the resource), to $\theta_{\phi*}$, such that fitness following the

environmental change was less than replacement, even at low density. To avoid extinction, the

population must reduce its genetic load by rapidly evolving toward its new optimal trait value,

$\theta_{\phi*}$. Adaptation may have occurred via either standing genetic variation preceding the

environmental change or de novo mutations following it. If the consumer population had not

gone extinct by generation 10,000, we recorded the population as having been rescued.

The genomic architecture of each ecological trait in the consumers was modelled to be

100 genes split evenly among 10 chromosomes, with each chromosome of map length 30cM.

Genes underlying a trait on a chromosome were evenly spaced and therefore 3cM apart. We

modelled the genomic architecture of the resources to instead be composed of 10 freely

recombining genes (i.e. an oligogenic trait). The relatively simple design of the resource

genome was primarily chosen to minimize computational load, as resource abundance was on

the order of 10^4.

We introduced demographic stochasticity in the simulations during the birth and death

processes of consumers and resources and during the consumption of resources by

consumers. The total number of births was Poisson distributed, with a mean of $N_t \underline{aC} R_t$ for the

consumers and a mean of $R_t r_R (1 - \frac{R_t}{K_R})$ for the resources. Once the total number of offspring

was established, parentage was retroactively assigned for the offspring using a multinomial

distribution, with the probability of an individual being assigned parentage proportional to its

fecundity. We modelled the death process in the consumers by drawing from a binomial

distribution with the probability of survival equal to the individual's survivorship. We did not

explicitly model resource death as its net growth is already captured by the r_R term, which we

allocate to births for brevity. Resource consumption was modelled using a binomial distribution

with $n = R_t$ trials and $p = aN_t$, the per-capita resource consumption rate.

Experimental Design

Effects of density-dependence on evolutionary rescue

We first investigated the effects of density-dependence on the probability of evolutionary rescue following an abrupt environmental change that negatively impacted consumer fitness (either through a decrease in a, C, or S). How is the effect of density-dependence on rescue influenced by the magnitude of environmental change and whether it is a, C, or S that are affected? We compared rescue probabilities between the consumer-resource model (Eq. 1 & 2; hereafter the *DD* model) and an equivalent model of density-independent growth (the *DI* model; Gomulkiewicz & Holt 1995), which we modelled by fixing resource abundance at its carrying capacity (i.e. $R_t = K_R$ for $0 \leq t \leq \infty$).

We must, however, limit consumer growth in the *DI* model such that it does not grow infinitely, which was done by enforcing a carrying capacity above which the density-independent population could no longer grow. Consumers in the *DI* model thus grew exponentially at a constant rate $r_N = aCK_R + S - 1$ until they reached their carrying capacity, K_N, above which their fecundity was reduced to 0 such that they were unable to grow beyond K_N (similar to the models of Gomulkiewicz & Holt 1995, Burger & Lynch 1995, and Orr & Unckless 2008). Consumers in the *DD* model, on the other hand, grew exponentially at rate r_N when at critically low density ($N_t \rightarrow 0$), but growth rates would linearly decline as they approached K_N (following Eqs 1 & 2). Both the *DI* and *DD* models were matched for K_N.

Effects of load on a, C, or S on evolutionary rescue

We investigated whether the relationship between mean fitness immediately following an environmental change and the probability of rescue were affected by which consumer ecological trait was impacted. We ran three sets of rescue simulations, varying the magnitude of the impact of the environmental change on one ecological trait while holding the others constant.

This approach allowed us to draw pairwise comparisons between simulations with reductions in

fitness due to differing causes. In the first set, we modelled a to be negatively impacted by the

environmental change and subsequently evolve, holding C and S constant. We did the same

with C (holding a and S constant) and S (holding a and C constant). Ecological dynamics were

governed by Eqs. 1 and 2 for all sets of simulations, and we also allowed for environmental

effects to influence the phenotypes underlying the two non-evolving quantities (e.g. C and S in

the first set).

The effects of load on an evolving resource on consumer evolutionary rescue

Lastly, we investigated the effects of an abrupt environmental change that also

negatively influences the growth rates of a biotic resource, either by decreasing its intrinsic

growth rates, r_R, or carrying capacity, K_R. We ran the simulations described above where we

impose an abrupt environmental change that reduces consumer fitness (either through a

reduction in a, C, or S), except that we also modelled a simultaneous decrease in resource

intrinsic growth rates, r_R or carrying capacity, K_R. We modelled resource evolution similarly to

consumer evolution. For each ecological trait in the resource (r_R or K_R), we modelled an

underlying phenotypic trait whose value, z_ϕ, is the sum of the additive values of 10 loci and a

normally distributed environmental effect with mean 0 and variance 1. The ecological trait value,

ϕ_i, of a resource individual with trait value $z_{\phi,i}$ is described using Eq. 3.

In sum, we had three parameter sets overall: one without resource evolution, one where

an evolving resource experiences a reduction in r_R, and one where an evolving resource

experience a reduction in K_R. Within each parameter set, we varied which consumer ecological

trait was affected (a, C, or S).

Model Parameters

We modelled a slow growing consumer population, with a maximum consumer growth

rate ($r_N = aCL + S - 1$) of $r_N = 0.03$ (Cortés 2016). The ecological parameters of our slow

growing consumer population were: $a_{max} = 10^{-4}$, $C_{max} = 10^{-2}$, $S_{max} = 0.99$, $r_{R,max} = 0.5$, and

$K_{R,max} = 10^5$. Individuals have high survival rates but low fecundity, characteristics that are

commonly associated with K-selected species, for which we expect density-dependence to be

most relevant. While our effect sizes may be sensitive this choice, the qualitative relationships

between our variables, in terms of their effects on r_N and K_N, are not. The sole exception to this

rule is the dependence of $\frac{dK_N}{da}$ on R_t ('resource over-consumption', see *Appendix A*), but we do

not simulate this case. Consumer evolution was modelled using the following parameters: $\mu =$

10^{-4}, $\alpha^2 = 0.5$, $\omega^2 = 10$. Mutation rates and effect sizes were modelled to result in a relatively

high mutational variance, V_m, according to empirical estimates (Johnson & Barton 2005); $V_m =$

$10^{-2}V_E$ here. Following an abrupt environmental change, rescue was therefore driven primarily

by a few mutations of large effect, given that the variance in the effect size of a heterozygous

mutation is $\alpha^2 = 0.5V_E$. Prior to the environmental change, however, these large effect

mutations were highly deleterious and rapidly removed via selection, such that the amount of

standing genetic variation was relatively low. In fact, much of the phenotypic variation for our

traits prior to the environmental change was due to environmental variation (i.e. $V_P \approx V_E$).

For each replicate simulation, we recorded population sizes (N_t, R_t), mean consumer

trait values ($\bar{a}_t$, $\bar{C}_t$, and $\bar{S}_t$), and mean resource trait values ($\overline{r_{R,t}}$ and $\overline{K_{R,t}}$) over time in a CSV file.

Subsequent analyses of the recorded data and plotting were conducted in R 4.4.2. We

compared the probability of rescue between the sets of the simulations for each change in

optimal trait value by running a χ^2 contingency test for each pair of simulations.

Results:

The effects of density-dependence on evolutionary rescue

We ran simulations to compare rescue probabilities between a model of density-dependent and density-independent growth. Following an abrupt environmental perturbation that reduces consumer fitness, density-independent populations (hereafter *DI* populations) immediately experience maximal fecundity, whereas density-dependent populations (hereafter *DD* populations) must decline to critically low densities to experience maximal fecundity.

For the case where an abrupt environmental change negatively impacts consumer survivorship, S, we find the greatest absolute difference in rescue probabilities for a relatively weak environmental perturbation, $\theta_{S*} = 1.75 V_E$ (Fig. 1). Despite both *DI* and *DD* populations exhibiting the canonical U-shaped trajectory of population size over time associated with evolutionary rescue, 69.7% [66.7, 72.4] of the *DI* populations experience rescue, compared to 56% [52.8, 58.9] in the DD populations. Indeed, we observe the relaxation of density-dependence as *DD* consumer populations decline, with density-dependent populations experiencing a steep rise in fecundity (Fig. 1C) due to increased resource availability (Fig. S1) following a decrease in fitness. However, there is still a lag to this release, as *DI* populations must first decline to a sufficiently low size, whereas *DI* populations are immediately able to benefit from maximal fecundity.

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Figure 1. The effects of density-dependence on evolutionary rescue following a decrease in
survivorship S. Consumer population size over time with $\theta_{S*} = 1.75 V_E$ in (A) 1000 replicate *DI Model*
simulations (blue) and (B) 1000 replicate *DD Model* simulations (red). The blue (red) line represents
median population size, N_t, in the rescued *DI* (*DD*) populations across the 1000 replicates. The darker
blue (red) shaded region represents the 25th and 75th percentile N_t, and the lighter blue (red) shaded
region represents the 10th and 90th percentile N_t. The grey lines represent median N_t for the populations
that went extinct. (C) Per-capita fecundity over time in the *DD* and *DI* simulations. The green (grey) line
represents median fecundity, B, for *DD* (*DI*) populations. The darker green shaded region represents 25th
and 75th percentile fecundity across the *DD* populations, and the lighter green shaded region represents
the 10th and 90th percentile. (D) The fraction of replicates that were rescued in the *DI* versus *DD*
simulations, with 95% confidence intervals calculated using the Agresti-Coull method.

Surviving *DD* populations also exhibit slower demographic recovery compared to *DI*

populations (*cf.* Fig. 1A & 1B, see Fig. S2). As *DI* populations rescue, resource abundance

declines, and density-dependent reductions in growth rate return. Thus, while rescue

probabilities following an abrupt environmental change may be similar between *DI* and *DD*

populations, *DI* populations recover faster. Additionally, as the magnitude of decline in

survivorship increases, the absolute difference in probability of rescue between *DI* and *DD*

populations declines (Fig. 2C), suggesting that the relative effect of density-dependence to

reduce rescue probabilities decreases as the magnitude of environmental change increases.

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Figure 2. Rescue probabilities in *DD* versus *DI* populations. The fraction of simulated populations that
successfully underwent evolutionary rescue following an abrupt decline in (A) resource acquisition rates,
(B) resource conversion efficiency, and (C) survivorship. Rescue fractions for *DD* populations are in red,
and rescue fractions for *DI* populations are in blue. 95% confidence intervals were estimated using the
Agresti-Coull method. Abrupt environmental changes were imposed by shifting the optimal phenotypic
value underlying the ecological trait θ. General model parameters can be found in the *Model Parameters*
section of the *Methods*.

Reductions in resource acquisition rates, a, result in similar probabilities of rescue

between *DI* and *DD* populations across all simulated magnitudes of environmental change (Fig.

2A), suggesting no discernible effect of density-dependence. In contrast, with a reduction in

368 resource conversion efficiencies, C, *DD* populations experienced lower rescue probabilities than

369 *DI* populations for most magnitudes of environmental change, but they appear to converge as

370 the strength of environmental change increases (Fig. 2B). Declining populations that experience

371 a decrease in a benefit from maximum resource availability (i.e. $R_t = K_R$) at higher densities

372 compared to populations that experience a decrease in C (Fig. S3), because, in the former

373 case, resource growth benefits from both decreased consumer size (N_t) and per-capita

374 consumption (a) (see the last term of Eq. 2). Thus, following a decrease in a, adapting

375 populations experience ecological release sooner than populations experiencing a decrease in

376 C, which is beneficial to evolutionary rescue. In *Appendix B*, we also show mathematically that a

377 consumer population will always experience greater density-dependence due to a reduction in C

378 versus a.

The effect of genetic load on a, C, or S on evolutionary rescue

380   Next, we explored how the probability of consumer rescue depends on whether the

381 abrupt environmental change impacted resource acquisition rates a, resource conversion

382 efficiencies C, or survivorship S in the *DD* model. Our simulations demonstrate that genetic load

383 on C leads to a marginally lower probability of rescue than equivalent load on a (Fig. 3 inset),

384 agreeing with our analysis in *Appendix B*. On the other hand, populations appear to be

385 substantially more resilient to abrupt declines in S (Fig. 3 blue dots) than to reductions in a or C.

386 For example, an immediate fitness decline of 0.33 due to a decrease in S results in similar

387 rescue probabilities to an immediate decline in fitness of 0.04 due to a decrease in C (S: 0.081

388 [0.066, 0.010] vs. C: 0.091 [0.075, 0.110]). This suggests that the population growth rate

389 immediately after the environmental change, λ_*, alone is an insufficient predictor of rescue

390 probabilities. Instead, the relationship between a population's growth rate immediately following

391 an environmental change, or, equivalently, its initial rate of decline, and rescue probability is

highly dependent on whether it is fecundity or mortality that are impacted, and to some extent

how fecundity is impacted (Fig. 3).

We briefly explored to what extent the differences in observed rescue probability were a

result of density-dependence acting on fecundity rather than mortality, an assumption of our

model. We simulated mean time to extinction in the absence of evolution, complemented with

numerical approximations (see *Appendix C*). We find that the median time to extinction does not

appear to vary significantly among ecological traits (Fig. S4), suggesting that the differences in

rescue between fecundity and mortality are primarily driven by differences in selection rather

than density-dependence.

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Figure 3. Rescue probabilities in *DD* populations as a function of the Malthusian parameter
following an environmental change, $\ln(\lambda_*)$**.** The fraction of *DD* populations that were rescued following
an abrupt environmental change that negatively impacted resource resource acquisition rates (red dots),
resource conversion efficiency (green dots), or survivorship (blue dots), plotted as a function of mean
fitness immediately following the environmental change, $\ln(\lambda_*)$. Each dot represents the fraction of
simulated populations that were rescued from 1000 replicates, and bars represent 95% confidence
intervals estimated using the Agresti-Coull method. For our resource acquisition rates and conversion
efficiency simulations, we modelled $\theta_{a*} = \theta_{C*} = \{3, 4, 4.75, 5, 5.25, 5.5, 5.75, 6\}$, and for our survivorship
simulations, we modelled $\theta_{S*} = \{1, 1.5, 1.75, 2, 2.5, 3\}$. General model parameters can be found in the
Model Parameters section of the *Methods*.

We also observe a significant difference in rescue probabilities depending on whether

the environmental change impacts resource acquisition rates or conversion efficiencies, which

both underlie an individual's fecundity (Fig. 3 inset). Furthermore, as the average probability of

rescue declines, populations experiencing a decline in acquisition rates have greater relative

probabilities of rescue than populations experiencing a decline in conversion efficiencies. For

the case where $\ln(\lambda_*) = -0.025$, we observe that 84.9% [82.5, 87.0] of the acquisition rate

simulations undergo rescue, whereas 72.6% [69.7, 75.2] of conversion efficiency simulations

undergo rescue. This reflects an approximately 16.9% higher probability of rescue. For the case

where $ln(\lambda_*) = -0.039$, we instead observe a 32.9% higher probability of rescue in the

acquisition rate simulations compared to the conversion efficiency simulations.

Since we simulate above a 'slow-growing' consumer population, we also investigated

whether the differences in rescue probabilities between affected ecological traits translated to a

'fast-growing' consumer population (i.e. one with a higher r_N). To do so, we ran simulations with

a different consumer parameter set, where $r_N = 0.18$. We found that the results were indeed

qualitatively similar to the results presented in the main text, with load on survivorship and

conversion efficiencies being the least and most costly to evolutionary rescue, respectively (Fig.

S5).

The effects of genetic load on a co-evolving resource on
consumer evolutionary rescue

Lastly, we explored the effects of an abrupt environmental change that also negatively

impacts an evolving resource. We consider the simple case of a resource that experiences an

abrupt decline in either its intrinsic growth rates, r_R, or carrying capacity, K_R, simultaneously with

a consumer that experiences an abrupt decline in either its acquisition rates, a, conversion

efficiencies, C, or survivorship, S.

Figure 4. Probability of rescue when a co-evolving resource is affected. The probability of rescue for various magnitudes of environmental change when (A) a is affected, (B) C is affected, or (C) S is affected. The blue lines represent rescue probabilities when the resource itself is unaffected. The green lines represent rescue probabilities when the resource intrinsic growth rates, r_R, are also reduced ($\theta_{r_R^*} = 5$), and the red lines represent the case when resource carrying capacity, K_R, is reduced ($\theta_{K_R^*} = 2.58$). $\theta_{r_R^*}$ and $\theta_{K_R^*}$ were chosen such that mean fitness immediately following the environmental change, $\underline{\lambda}_*$, was equal among the two cases. General model parameters can be found in the *Model Parameters* section of the *Methods*.

We find that resource load decreases the probability of consumer rescue across all

simulated parameters, with a reduction in K_R resulting in a lower probability of rescue in

consumers than a reduction in r_R (Fig. 4). With a reduction in K_R, maximum resource

abundance is reduced substantially, and consumers therefore experience lower fecundity even

at low densities (Fig. S5). On the other hand, with a reduction in r_R, maximum resource

abundance is unaffected, but is attained at a slower rate than when the resource itself is

unaffected (Fig. S5).

Discussion

In this article, we show that the effect of density-dependence on evolutionary rescue is

weak, but negative, in a consumer of a self-replenishing resource. We also make the previously

unappreciated point that an environmental perturbation may itself reduce the intensity of

density-dependence on a declining population. We have also shown that the relationship

between the probability of consumer rescue and the population rate of decline varies strongly

with the affected ecological trait, such that the rate of population decline immediately following

an environmental change alone may be an insufficient predictor of extinction probabilities. In the

case of density-independent mortality, we show that increases in mortality are most likely to

lead to rescue. Conversely, declines in resource conversion efficiencies lead to the lowest

chance of rescue, as individuals are unable to fully utilize increased resource availability. Lastly,

an environmental change that also negatively impacts an evolving resource will always hinder

consumer rescue.

Theoretical work on the effect of density-dependence on evolutionary rescue has mostly

considered the evolutionary process as distinct from demographic processes (e.g. Boulding &

Hay 2001; Chevin & Lande 2010; Nordstrom et al. 2023), representing a major simplification

(which these authors are aware of). In reality, these processes are intertwined, and the outcome

of rescue is likely to be dependent on the interaction between demography and evolution, given

the substantial empirical evidence of varying relationships between r and K (e.g. Wei & Zhang

2019; Marshall et al. 2023). Moreover, the relationship between evolutionary adaptation and

demography is highly dependent on the evolving ecological trait (Abrams 2019; Bisschop et al.

2022). In the context of rescue, the rate of population decline immediately following an

environmental change is thus likely an insufficient estimator of future population growth and

extinction, as this potentially neglects downstream ecological interactions experienced by the

population (Draghi et al. 2024), and ecological interactions between individual fitness and

population growth.

We find an appreciable difference in the probability of rescue when resource acquisition

rates are affected versus conversion efficiencies, two traits underlying fecundity. As populations

suffer from decreases in resource conversion efficiencies, individuals must acquire more

resources in order to maintain birth rates, thus intensifying density-dependence substantially.

Conversely, as populations suffer from decreases in resource acquisition rates, they are able to

experience more rapid ecological compensation. Interestingly, we also find a stark difference in

rescue probabilities depending on whether it is fecundity or mortality affected. This has been

observed before in the rescue literature by Draghi et al. (2024), who provide a convincing

explanation. Briefly, an increase in mortality will accelerate the turnover of individuals in the

population, hence reducing generation time and hastening adaptation. Conversely, a decrease

in birth rates will reduce the turnover of individuals, lengthening generation time and slowing

adaptation. More simply, individuals who experience lower fitness due to lower survivorship will

be replaced by fitter individuals more rapidly than individuals who experience lower fitness due

to lower fecundity. Given that we do not observe any meaningful difference in median times to

extinction among ecological traits, it is likely that our observed differences are indeed a result of

the differential rates of adaptation as described above.

However, it remains unclear in the first place what specific trait may be affected by an

environmental change. There is some evidence to suggest that the dependence of evolutionary

adaptation through a specific ecological trait varies with respect to its environment. Wei &

Zhang (2019) show that quantitative trait loci (QTLs) in the yeast *S. cerevisiae* tend to show

concordant effects on r and K (r/K trade-ups) in lower quality environments, but negative

effects (r/K trade-offs) in higher-quality environments. They propose an explanatory model

characterized by trade-offs between ATP production rates and utilization efficiency in the

context of cell maintenance and division. Briefly, they propose a non-monotonic relationship

between intrinsic growth rates, r, and resource usage per cell per generation, which we define

as ψ (they define this as C, which we have already used to denote conversion efficiency). Under

low quality environments, r is low, and the evolution of increased r decreases ψ, thus

increasing both r and K. This aligns with the predictions of our model, where an increase in

conversion efficiencies, C, will increase both r and K. Conversely, in a high quality environment

where r is already high, ψ increases with increasing r. Here, the population is instead

dominated by competition for the acquisition of resources, which indeed aligns with the

consumer-resource expectation that selection for resource acquisition rates, a, may result in r/K

trade-offs.

While their model specifications may differ from ours in some ways, evolutionary rescue

experiments in *S. cerevisiae* may be a possible avenue to test the predictions of our model.

Empiricists could bias rescue towards potential life history strategies by controlling

environmental quality. Declining yeast populations subject to an environmental stressor may

perhaps be more likely to adapt through increases in resource conversion efficiency, C, in a low

quality environment, or through resource acquisition rates, a, in a high quality environment.

Consumer-resource models offer useful mechanistic insight into the relationship

between ecological traits and fitness, with ecological traits having the potential to affect both

consumer intrinsic growth rates and carrying capacities differently. Despite this, our model has a

number of limitations. To begin, we model a resource with the capacity to replenish itself, such

that resource abundance is able to increase as consumer populations decline. While this is

likely the case in many natural systems, as in the case of any herbivore and plant system, these

results do not apply to the case of a consumer of a finite, non-replenishing resource, such as

bacterial populations consuming medium. Moreover, the assumption of logistic resource growth

is arbitrary, and we do not consider other forms used in the literature, such as a chemostat

model (e.g. MacArthur 1972, Abrams 2019).

We make a number of additional simplifying assumptions. First, we assume that

consumer mortality is density-independent, a common assumption in consumer-resource

models (e.g. Abrams 2019). However, mortality itself may be an increasing function of

consumer abundance. When consumer density is high, it is likely that consumers with relatively

lower acquisition rates may be in worse condition, and are therefore more susceptible to

mortality. Moreover, we assume that ecological traits are independent among consumers.

Additionally, it is plausible that consumer ecological traits have similar underlying genetic and

phenotypic structures, which we neglect. Instead, we assume that only a single trait in the

consumer or resource is affected by the environmental change. This approach allows us to

investigate, all else being equal, the effect of an ecological trait on rescue, but neglects the case

where multiple co-varying traits may be concurrently affected. We also consider the simple case

of intraspecific competition for a non-substitutable resource, without direct interference among

competitors, as well as type 1 functional responses in the consumer. We also do not explore the

effects of interspecific competition among species on the probability of rescue. The relaxation of

these assumptions would be a fruitful avenue of future theoretical investigation.

We have here provided an improved theoretical framework to understand how

evolutionary rescue is influenced by the complex interplay between evolution and demography.

Given that evolutionary rescue is essentially a race between demographic decline and

evolutionary adaptation, the use of mechanistic models that provide increased biological realism

is an important step forward in our treatment of populations at risk of extinction (Urban 2024), an

increasingly relevant problem given the rate of climate change (Berg 2010). The work presented

here extends our understanding of the various ways through which environmental pressures

may impose ecological costs on evolving species as well as species they interact with, and the

resultant influence on species persistence.

Appendix A: Density-dependence in a consumer-resource model.

Here we summarize the relationship of consumer ecological traits a, C, and S to consumer density-dependence, following the methodology described by Travis et al. (2023). We can find the equilibrium values of both consumers, $\widehat{N}$, and resources, $\widehat{R}$, through algebraic manipulation of Eqs. 1 and 2: $\widehat{N} = \frac{r_R(1-\frac{\widehat{R}}{K_R})}{a}$ and $\widehat{R} = \frac{1-S}{aC}$. It becomes clear that the equilibrium size of the consumer (i.e. its carrying capacity) is a function of the ratio of the per-capita rate of resource replenishment (numerator, $r_R(1-\frac{\widehat{R}}{K_R})$) and per-capita rate of consumption (denominator, a), whereas the equilibrium size of the resource is dependent solely on the three potentially evolvable parameters of the consumer. Here density-dependence in the consumer arises via competition for a finite resource pool. As consumer population size grows the total consumption rate of resources increases, and fecundity declines linearly with resource abundance. Growth would instead be considered density-independent if resource abundance were constant with respect to consumer density, leading to exponential growth.

Density-dependence in a consumer-resource model is more complex than that of a logistic model, given that consumer intrinsic growth rates, r_N and carrying capacity, K_N (Travis et al. 2023) share several underlying ecological traits. It is still, however, possible to develop intuition for these relationships by re-writing consumer ecological traits in terms of the logistic growth parameters r_N and K_N. We can first solve for r_N by considering a consumer population that is at a critically low size ($N_t \approx 0$). In this regime, resource abundance is effectively at its own carrying capacity ($R_t \approx K_R$ when $N_t \approx 0$), and consumer growth is effectively exponential. Plugging this resource abundance into Eq. 1, we see that the consumer intrinsic growth rate, r_N, is $r_N = aCK_R + S - 1$. Consumer carrying capacity may be calculated by considering consumer equilibrium size, which through rearrangement of Eq. 2 is found to be $K_N = \frac{r_R(1-\frac{\widehat{R}}{K_R})}{a}$. We assume

that resource growth occurs on a substantially shorter time-scale that consumer growth, such

that resources reach demographic equilibrium immediately following an abrupt environmental

perturbation (separation of time scales, $R_t = \hat{R}$ for $0 \leq t \leq \infty$). This assumption allows us to

account for how equilibrium resource abundance may itself change following an environmental

change that affects a consumer, given that resource abundance is a function of consumer trait

values ($\hat{R} = \frac{1-S}{aC}$). . Plugging in $\hat{R} = \frac{1-S}{aC}$ into our formula for K_N, we find that $K_N = \frac{r_R(aCK_R+S-1)}{a^2CK_R} =$

$\frac{r_R r_N}{a^2CK_R}$. It then becomes clear that both r_N and K_N share several underlying parameters and are

functions of one another. The assumption of independence between r_N and K_N thus leads to an

incomplete understanding of the effects of density-dependence on rescue from previous work.

We begin by exploring how intrinsic growth rates vary with a decrease in consumer

ecological trait values, finding that $-\frac{\partial r_N}{\partial a} = -CK_R$, $-\frac{\partial r_N}{\partial C} = -aK_R$, and $-\frac{\partial r_N}{\partial S} = -1$. Intrinsic

growth rates all decline linearly with respect to a decrease in a, C, or S. The magnitude of

decline in a and C are dependent on both K_R and each other. This is because a decrease in

resource acquisition would reduce consumer growth proportional to the rate at which individuals

were able to convert resources into offspring in the first place. A decrease in conversion

efficiencies, on the other hand, would reduce growth proportional to the total number of

resources being acquired in the first place. Survivorship is independent of density here, and has

a 1-to-1 relationship with r_N.

We now turn to exploring how consumer carrying capacity is influenced by a decline in

consumer fitness, focusing on the dependence of this relationship on resource abundance.

Considering the effects of a decrease in resource acquisition rates, a, on carrying capacity, K_N,

we recover

600
$$-\frac{\partial K_N}{\partial a} = -\frac{s(2(1-S)-aCK_R)}{a^3CK_R}. \tag{A1}$$

Given that all variables are positive, the right hand side of Eq. A1 remains negative only when

$2(1-S) > aCK_R$. Assuming a separation of time-scales and substituting in $\hat{R} = \frac{1-S}{aC}$, this

inequality simplifies to $\hat{R} > \frac{K_R}{2}$. Thus, a decrease in acquisition rates will lead to a decrease in

consumer carrying capacity, therefore intensifying density-dependence, only when the new

resource equilibrium is above half of the resource's carrying capacity. Conversely, a decrease in

acquisition rates may lead to an increase in resource carrying capacity when the new resource

equilibrium is below half of the resource's carrying capacity.

Considering the effects of a decline in resource conversion efficiency, C, on carrying

capacity, K_N,

610
$$-\frac{\partial K_N}{\partial C} = -\frac{s\hat{R}}{aCK_R},$$ (A2)

which is negative for all parameter sets, and a linear function of $\hat{R}$. Thus, the decline in carrying

capacity due to a decline in conversion efficiencies is linearly proportional to resource

abundance. If resource abundance is already low, then a decline in conversion efficiencies will

be less costly, as individuals are already acquiring a low number of resources.

Considering the effects of a decline in survivorship, S, on carrying capacity, K_R, we find

that

617
$$-\frac{\partial K_N}{\partial S} = -\frac{s}{a^2 CK_R}$$ (A3)

which is also negative for all simulations presented in the main text. However, the effects of a

decline in survivorship on carrying capacity are independent of density, unlike the case of a

decline in resource conversion efficiencies (*cf.* Eq. A3 & A4).

While the logic above provides a reasonable starting point to understanding density-

dependence in a consumer-resource model, it remains unclear how these predictions translate

to evolutionary rescue. Evolutionary rescue is a process characterized by non-equilibrium

processes, in the form of fluctuating demography and rapid evolution due to maladaptation. For

example, despite a decline in resource acquisition rates leading to a decrease in both intrinsic

growth rates, r_N, and consumer carrying capacity, K_N, when resource abundance is high,

subsequent evolution would then increase both r_N and K_N, thus accelerating growth.

Appendix B. The effects on carrying capacity of a reduction in a versus C.

631 In *Appendix A*, we found that a decrease in conversion efficiency, C, always decreases

632 consumer carrying capacity, K_N, therefore intensifying density-dependence. Here we explore

633 when a decrease in resource C will intensify density-dependence to a greater degree than a

634 decrease in a. We can approximate the effects of a decrease in a or C on K_N using a first-order

635 Taylor expansion around their initial values:

$$K_{N,a*} = K_N + \frac{\partial K}{\partial a}(a_* - a) \tag{B1a}$$

638 and

$$K_{N,C*} = K_N + \frac{\partial K}{\partial C}(C_* - C), \tag{B1b}$$

641 where a and C refer to mean acquisition rates before an environmental change (we assume no

642 individual variation here for brevity), a_* and C_* refer to mean acquisition rates and conversion

643 efficiencies immediately following an environmental change, and $K_{N,a*}$ and $K_{N,C*}$ refer to

644 consumer carrying capacities following a reduction in a or C, respectively. Thus, $K_{N,C*} < K_{N,a*}$

645 when

$$\frac{\partial K_N}{\partial C}(C_* - C) < \frac{\partial K_N}{\partial a}(a_* - a). \tag{B2}$$

649   In our simulations, we model a proportional reduction in both ecological traits a and C such that

650   they are matched for mean fitness following the environmental change. Thus, $C_* - C =$

651    $C_{max}(e^{-\frac{\theta^2}{2\omega^2}} - 1)$ and $a_* - a = a_{max}(e^{-\frac{\theta^2}{2\omega^2}} - 1)$. We can substitute the above values and simplify

652   the inequality to

654
$$-\frac{\partial K_N}{\partial C} C_{max} < -\frac{\partial K_N}{\partial a} a_{max}. \tag{B3}$$

656   Next, we substitute in the right hand sides of Eqs. A1 and A2 and get the following inequality:

658
$$a_*(1 - S)C_{max} > C_*[2(1 - S) - a_* C_* K_R]a_{max} \tag{B4}$$

660   We can simplify with consideration that resource abundance following an environmental

661   change, R_*, is $R_* = \frac{1-S}{aC}$ under the assumption of time-scale separation. Then we find that the

662   above inequality is true when $R_* < K_R$, which is always true except in the case of an extinct

663   consumer. Thus, a reduction in resource conversion efficiency will always entail a greater

664   reduction in carrying capacity, K_N, than a proportional reduction in resource acquisition rates.

Appendix C. Time to extinction in the absence of evolution.

Here we briefly discuss how we calculated the time to extinction in a consumer-resource

model, in the absence of evolution. In *Appendix A,* we use a simple heuristic argument to show

that consumer growth can be re-written in logistic growth terms r_N and K_N, where $r_N = aCK_R -$

m (where $m = 1 - S$) and $K_N = \frac{r_R r_N}{a^2 CK_R}$. Given that the mean number of births and deaths per

time-step is reasonably low in our model, we can approximate the population growth rate to be

occurring in continuous time, such that $\frac{dN}{dt} = r_N(1 - \frac{N}{K_N})$. Substituting in our values of r_N and K_N,

and performing some algebraic manipulation, we find that $\frac{dN}{dt} = aCK_R(1 - \frac{aN}{r_R}) - m$, where $R =$

$K_R(1 - \frac{aN}{r_R})$. This formula makes it clear that density-dependence in the system arises through

limited resource availability influencing consumer fecundity, which represents a balance of

resource consumption by consumers (aN) and replenishment by the resource (r_R). Returning to

the continuous-time logistic growth model, it has a well-known general solution of the form $N_t =$

$K_N \frac{N_0}{N_0 + e^{-r_N t}(K_N - N_0)}$. We can solve for the time to extinction, t_e, by plugging in $N_{t_e} = 0$, $r_N =$

$aCK_R - m$, and $K_N = \frac{r_R r_N}{a^2 CK_R}$, which yields a deterministic prediction for the mean time to

extinction in the absence of evolution. This method provides a reasonably accurate prediction,

although it exhibits a slight upward bias (Fig. S4 triangles). This is to be expected since

stochasticity will often reduce the time to extinction, as populations at low density that produce

lower offspring than expected will achieve extinction much more quickly.

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788   # Supplementary Figures:

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Figure S1. Resource abundance (R_t) over time (t) for the *Dynamic* simulations presented
in Fig. 1 of the main text. The green lines represent replicates that survived, and the dark
green line represents median population size, R_t, across the 100 replicates. The darker green
shaded region represents the 25th and 75th percentile R_t, and the lighter green shaded region
represents the 10th and 90th percentile R_t. The purple lines represent median R_t for the
populations that went extinct.
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799   **Figure S2. Time to recovery (t_R) in *DI* versus *DD* populations.** The probability density
800   functions for mean time to recovery (t_R), measured as the time it takes for a population to return
801   to its prior population size following an environmental change that reduces fitness. The data
802   used to make this plot was from 1000 replicate simulations with $\theta_{S*} = 1.75 V_E$ in both the *DI* and
803   *DD* populations, same as Fig. 1 in the main text.
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807 **Figure S3. Time to max growth ($t_{0.98K_R}$) after an environmental change.** The probability
808 density function for the mean time it takes, following an abrupt environmental change, for
809 resource abundance to be at least 98% of its carrying capacity (i.e. $R_t = 0.98\,K_R$). The red
810 distribution describes the case where $\theta_{a*} = 5$, and the blue distribution describes the case
811 where $\theta_{C*} = 5$.

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Figure S4. The median time to extinction, t_e, as a function of the Malthusian parameter
immediately following the environmental change, $\ln(\lambda_*)$. Median time to extinction across 50
replicate simulations for simulations where resource acquisition rates are affected (red),
resource conversion efficiencies are affected (green), and survivorship is affected (blue). The
round dots connected by the lines represent values from the simulations, and the triangles
represent numerical approximations under the assumption of time-scale separation between the
consumer and resource.

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Figure S5. Rescue probabilities in *DD* populations as a function of the Malthusian
parameter immediately following an environmental change, $\ln(\lambda_*)$, for a consumer
population with intrinsic growth rate $r_N = 0.18$. The fraction of *DD* populations that were
rescued following an abrupt environmental change that negatively impacted resource resource
acquisition rates (red dots), resource conversion efficiency (green dots), or survivorship (blue
dots), plotted as a function of mean fitness immediately following the environmental change, $\underline{W_*}$.
Each dot represents the fraction of simulated populations that were rescued from 100 replicates,
and bars represent 95% confidence intervals estimated using the Agresti-Coull method. For our
resource acquisition rates and conversion efficiency simulations, we modelled $\theta_{a*} = \theta_{C*} =$
$\{5, 6, 7, 8, 9\}$, and for our survivorship simulations, we modelled $\theta_{S*} = \{1, 2, 3, 4\}$. The ecological
parameters were as follows: $a_{max} = 10^{-4}$, $C_{max} = 10^{-2}$, $S_{max} = 0.9$, $r_{R,max} = 0.53$, and $K_{R,max} =$
1.6×10^5, and the evolutionary parameters were identical to those found in the *Model*
Parameters section.

869
870 **Figure S5. Resource abundance over time in an affected resource.** Following an abrupt
871 environmental change that reduces consumer resource conversion efficiencies, $\theta_{C*} = 4.75$,
872 median resource abundance when the resource itself is unaffected ($\theta_{s*} = \theta_{L*} = 0$; blue line),
873 resource intrinsic growth rates are affected ($\theta_{s*} = 5$; green line), and resource carrying capacity
874 is affected ($\theta_{L*} = 2.58$; red line). The shaded regions represent that 25th and 75th percentile R_t
875