

## Extinction and ecosystem function debt across dispersal rate and behavior in a heterogeneous metacommunity model

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- 1 Title: Extinction and ecosystem function debt across dispersal rate and behavior
- 2 in a heterogeneous metacommunity model
- 3
- 4 A short running title: Extinction and ecosystem function debt
- 5 The codes can be accessed via the link:
- 6 https://datadryad.org/stash/share/z0Mrqixl9wJEzgLJN7GRi9oThNqiz7EoA0t3rm89\_t
- 7 M
- 8

# 9 Abstract

10	Aim: Habitat destruction causes "extinction debt" and is also thought to produce
11	ecosystem function debt, but theory of their magnitude and nature is limited.
12	Heterogeneous landscapes are fundamental to the maintenance of species richness and
13	ecosystem function, whilst directed or undirected dispersal behavior, such as dispersal
14	of seeds by animals or by the wind, is also important, especially after habitat
15	destruction. We therefore consider extinction and ecosystem function debt under
16	different dispersal rates and behaviors in heterogeneous landscapes.
17	Methods: We use a classic heterogeneous metacommunity model to capture the
18	dynamics of competing species in local patches linked by dispersal and varying in
19	environmental conditions. We remove one patch at a time, and measure extinction
20	debt and ecosystem function debt by the number / proportion of delayed extinctions
21	and the amount of biomass change, respectively.
22	Results: We reveal three species extinction regimes as dispersal increases: 1. species
23	most adapted to the removed habitat are most at risk; 2. similarly adapted species are
24	also at risk; 3. patch removal shifts competitive balance among the few species
25	coexisting at high dispersal, where competition is strong. We find surprisingly that
26	destruction of habitat can hasten the extinction of those species best adapted to harsh

27	environments, and that the proportion of diversity at risk from extinction actually
28	increases with dispersal because competition is intense there. Finally, there can be a
29	small ecosystem credit, but extinction debt, when dispersers reroute to potentially
30	more favorable remaining habitats (directed dispersal), especially when harsh
31	environments are removed. However, ecosystem debt occurs and can be large under
32	undirected dispersal.
33	Main conclusions: The magnitude and nature of extinction and ecosystem function
34	debts depend on species dispersal rates and behaviors, as well as the environmental
35	conditions of the disturbed habitats. Conservation actions will be more successful if
36	they consider these factors.
37	
38	Keywords: habitat loss, directed/undirected dispersal, extinction regimes,
39	harsh/benign environment, resource consumption, species conservation

# 41 Introduction

42	Habitat destruction causes species extinctions (Fahrig 2003; Jackson & Fahrig 2013;
43	Horváth et al. 2019; Chase et al. 2020) and interferes with ecosystem functions and
44	ecosystem services (Isbell et al. 2015). Some species in disturbed habitats disappear
45	immediately (Krauss et al. 2010), for example in the case of habitat conversion or
46	chemical spillages. Some species are able to remain in the habitat during and after
47	disturbance, depending on the time scales and severity of the disturbance (Ceballos et
48	al. 2015). Other species, having survived the initial disturbance, crowd into the
49	remaining habitat patches (Ewers & Didham 2006). For those species not destroyed
50	immediately, the habitat disturbance may cause future extinctions across multiple
51	trophic levels (Krauss et al. 2010). "Extinction debt" describes the delayed species
52	extinctions occurring after environmental disturbance (Tilman et al. 1994; Kuussaari
53	<i>et al.</i> 2009).
54	As well as causing extinction debt, habitat disturbance also causes ecosystem
55	function debt, which refers to the delayed loss of ecosystem functions such as
56	productivity or biomass (Haddad et al. 2015; Isbell et al. 2015). Ecosystem function
57	debt results from species loss and is therefore correlated with extinction debt (Haddad
58	et al. 2015; Isbell et al. 2015). However, it is possible for ecosystem function to

59	recover after habitat disturbance, even in the absence of species reintroductions. This
60	is because disturbance alters the competitive abilities of species, allowing productive
61	species to dominate entire habitats. We refer to the increase in ecosystem function
62	after habitat disturbance as ecosystem function credit. Extinction debt attracts the
63	most attention because it provides time windows for conservation actions to rescue
64	rare species from extinction (Hanski & Ovaskainen 2002; Malanson 2008; Wearn et
65	al. 2012; Halley et al. 2014; Highland & Jones 2014; Chen & Peng 2017; Otto et al.
66	2017; Figueiredo et al. 2019; Makishima et al. 2021; Ridding et al. 2021). However,
67	studies of ecosystem function debt/credit are limited (also mentioned by Gonzalez et
68	al. 2009), and as far as we know, none of these studies consider a heterogeneous
69	landscape.
70	The severity of extinction debt is usually measured as the number or
71	proportion of "delayed extinctions as a consequence of ecosystem perturbation" as
72	definition by Figueiredo et al. (2019), which means the number / proportion of
73	species extinct in the long run as a result of the perturbation, minus the number /
74	proportion of species extinct immediately after it. Similar definitions are used in the
75	other studies (Kuussaari et al. 2009; Jackson & Sax 2010; Halley et al. 2016;
76	Figueiredo et al. 2019). Meanwhile, a given species' extinction risk could be

77	measured by the time delay index, that is the persistence time of extinct species since
78	habitat loss (Hanski & Ovaskainen 2002). If the time delay index of a species is
79	higher, its extinction risk is lower (Grimm & Wissel 2004). Previous theoretical
80	studies have extrapolated ecosystem function debt from the number of species lost
81	based on correlations between community biomass and species richness (Isbell et al.
82	2015). Ecosystem function debt/credit could be measured more directly as the amount
83	of biomass lost in the long run minus the amount lost immediately after habitat
84	disturbance: if this figure is positive then ecosystem function credit has occurred; if it
85	is negative, then ecosystem function debt has occurred.
86	Environmental heterogeneity in space is key to maintaining and shaping
87	species richness and ecosystem function (Ben-Hur & Kadmon 2020; Davies et al.
88	2021; Thompson et al. 2021). In recent decades there have been a number of
89	theoretical advances towards understanding environmentally heterogeneous
90	landscapes, through the study of metacommunity models with discrete communities
91	of different environmental conditions, all linked by dispersal (Thompson et al. 2014;
92	Fournier et al. 2017; Thompson & Gonzalez 2017; Thompson et al. 2017; Leibold &
93	Chase 2018; Thompson et al. 2020). These models have also begun to be employed in

95	considered the effects of environmental heterogeneity on extinction debt to a limited
96	degree (see related studies in Mouquet et al. 2011; Thompson et al. 2017).
97	Species dispersal behavior, a crucial component of metacommunities before
98	and after habitat disturbance, can be modelled in different ways. Most animals are
99	active dispersers, easily orientating themselves and moving purposefully in the
100	direction of habitats with sufficient resource (Bowler & Benton 2005; Croteau 2010).
101	Most plants disperse passively, in the direction of whichever organism or abiotic
102	dispersal agent is transporting the seed. Seeds dispersed by animals benefit from
103	active dispersal, thus avoiding unsuitable habitats (Bowler & Benton 2005; Nield et
104	al. 2020; Mason et al. 2022) whereas seeds dispersed by the wind cannot avoid
105	unsuitable habitats (Zona 2017). In this study, we distinguish between these two
106	dispersal behaviors (Fig. 1A). One in which the individuals previously dispersing to
107	the disturbed habitat instead disperse to remaining habitats, in other words the
108	individuals change direction to available habitats (hereafter referred to as "directed
109	dispersal"), and another where those individuals continue to disperse to the disturbed
110	habitat and are lost (hereafter referred to as "undirected dispersal").
111	We use this metacommunity modeling approach to consider additional
112	questions fundamental to understanding the nature of extinction and ecosystem

113	function debt in a heterogeneous landscape (see Fig 1B). This includes basic
114	questions such as how extinction debt is distributed across the species adapted to
115	different environmental conditions, and how this may depend on how the lost habitat
116	fits into the distribution of habitats represented, and the rate of dispersal between
117	patches. We also ask whether the overarching effects of dispersal on extinction and
118	ecosystem function debts, and whether the basic tendency towards ecosystem function
119	debt one expects in homogeneous environments, is the same in heterogeneous
120	environments. When landscapes are homogeneous, one would anticipate, and recent
121	theory using a neutral model framework shows (Thompson et al. 2019), that higher
122	dispersal among patches would generally limit these debts. One would also anticipate
123	that patch removal can have only negative or at best no consequences for the
124	ecosystem functioning in the remaining patches. However, habitat heterogeneity
125	creates circumstances under which these ideas may not apply. Species competing
126	along an environmental gradient may do so more intensely when their dispersal rate is
127	higher, in which case their competitive balance may be more sensitive to patch
128	removal. Also, when a patch is removed, species better adapted to other patches could
129	fare better under directed dispersal, and hence ecosystem function in the remaining
130	patches may in fact improve. We answer these questions using an existing

131	metacommunity modeling approach based on consumer-resource dynamics, allowing
132	us to explicitly consider community biomass rather than drawing on correlations with
133	species richness. To answer our questions regarding the influence of dispersal, we add
134	to this metacommunity modeling approach the different potential dispersal behaviors
135	possible when habitat is destroyed. In addition, we link the observed metacommunity
136	extinction and ecosystem function debt behaviors with the effects of patch destruction
137	on heterogeneous metapopulations, by studying the behavior of a similar model of a
138	single species population in heterogeneous patches linked by dispersal.
139	Model and Methods
140	Model framework
141	Our model is comparable to the model used by Mouquet and Loreau (2003) which is
142	well-known for its predictions regarding the effects of dispersal on diversity and
143	ecosystem function in a heterogeneous metacommunity (Mittelbach & McGill 2019).
144	It captures the dynamics of a set of competing species in local patches that vary in
145	their environmental conditions (and hence dominant competitors) and are connected
146	by dispersal to form a metacommunity. We use a formulation of this model similar to
147	that used by Loreau et al. (2003), which models competition through resource
148	consumption rather than the local patch dynamics considered in the original model,

149	thus allowing clearer consideration of not just diversity, but also ecosystem function.
150	This metacommunity model has been studied extensively and is the core approach
151	used in recent developments in metacommunity theory (Leibold & Chase 2018; Ai &
152	Ellwood 2022).
153	Our model is not linked with a particular biological system, for example it could
154	describe grass competing for soil nitrogen (Loreau et al. 2003). We set varied
155	environments for habitat patches, such as surface soil temperatures of various plant
156	systems, and we allow species to differ in their traits, such as the temperature at which
157	each species has its maximum competitive ability for soil nitrogen (Tilman 1999).
158	Biomass of a species would increase if that species inhabits a patch where the soil
159	temperature is optimal. In this example, the model is attempting to capture the
160	dynamics of the biomass of different plant species on a heterogeneous landscape
161	through competition for soil nitrogen. The competitive ability of the species depends
162	on the match between the soil temperature of the patch and the optimal temperature of
163	the species.
164	The metacommunity consists of $M$ patches, each with a different value of an
165	environmental condition, connected by dispersal rate a. Each patch is numbered from
166	1 to $M$ . The environmental value (such as the surface soil temperature after

167	normalization) of the first patch is $E_1=1$ , then the adjoining patch is defined as $E_j=E_{j-1}$ -
168	1/(M-1) where j is from 2 to M. There are S species in each patch initially, numbered
169	from 1 to S. The optimal niche value of the first species is $H_l=1$ , and remaining
170	species are defined as $H_i = H_{i-1} - 1/(S-1)$ where <i>i</i> is from 2 to <i>S</i> . The species competitive
171	ability in each patch is determined by the match between species' optimal niche value
172	and the environmental value of a patch. Under these definitions, in the case of
173	maximal potential diversity (our focus), where there are as many species as patches
174	(i.e. $M=S$ ), the first species is the best competitor in the first patch, the second species
175	is the best competitor in the second patch, and so on. We refer to the patches towards
176	the center of the range of environmental conditions as the habitats with benign
177	environments, and the patches with extreme environmental values, such as 0 or 1, as
178	the habitat with harsh environments. Patches with benign environments are suitable
179	for all species, whereas patches with harsh environments are suitable only for those
180	species adapted to that patch. Note that those patches are not necessarily spatially
181	central in our model, in which all patches are assumed equally connected to focus on
182	environmental heterogeneity between patches rather than dispersal limitation. Hence,
183	species with extreme optimal environmental values close to 0 or 1 are best adapted to
184	patches with harsh environments, whereas species with mid-range environmental

185 values are best adapted to benign environments.

186 Within patches, species compete for a resource, with species' resource  
187 consumption and therefore competitive ability in the patch determined by the match  
188 between their niche value and the environmental value of the patch. The dynamics of  
189 the biomass of the resource in patch *j*, R<sub>j</sub>, follow (Loreau *et al.* 2003):  
190 
$$\frac{dR_j(t)}{dt} = I_j - l_jR_j(t) - R_j(t)\sum_{i=1}^{S} C_{ij}P_{ij}(t)$$
 (1)  
191 where  $I_j$  is the rate of input, and  $l_j$  is the rate of loss of resource in patch *j*. Species *i*  
192 consumes the resource in patch *j* at a rate defined by  $C_{ij}$  that depends on the difference  
193 between the optimal niche value of species *i*  $H_i$ , and the environmental condition of  
194 patch *j*,  $E_j$  (modified from Gonzalez *et al.* 2009):

195 
$$C_{ij} = \frac{1.5 - |H_i - E_j|}{b}$$
 (2)

196 where *b* determines the overall magnitude of resource consumption of all species. The

197 dynamics of  $P_{ij}(t)$ , the biomass of species *i* in patch *j* at time *t*, follow:

198 
$$\frac{dP_{ij}(t)}{dt} = eC_{ij}R_j(t)P_{ij}(t) - mP_{ij}(t) + a(\frac{1}{M-1}\sum_{k=1,k\neq j}^M P_{ik}(t) - P_{ij}(t))$$
(3)

199 where e is the rate of conversion of resource into new biomass, and m is the rate of

- 200 loss of biomass of each species in each patch. The first term describes the
- 201 consumption and conversion of biomass by species; the second term describes the
- 202 decrease in species biomass as species die; the third term describes species

203 immigration from other patches; the fourth term describes species emigration to other 204 patches. The parameter a is the per-capita rate at which individuals are moving out 205 of a patch, and the proportion of dispersers coming into any given patch is then  $\frac{1}{M-1}$ , 206 where M is the number of patches (Plitzko & Drossel 2015). 207 Our model assumes that the environment is constant within each patch, but 208 environmental variation occurs between patches (Ai & Ellwood 2022). In Loreau et 209 al. (2003), the environmental value of each patch was the function of time and the 210 initial environmental value of each patch, making the  $C_{ij}$  of each patch also a function 211 of time. The Loreau et al. (2003) study considered spatial insurance effects, and how 212 the diversity and functioning of the metacommunity emerges from the spatial 213 dynamics of new species coming into and proliferating in local communities as the 214 local environment changes. 215 Simulated patch removal and dispersal behaviors 216 We ran each simulation to reach the first approximate equilibrium (when the total 217 biomass of all species is a fixed point and the biomass of each species becomes 218 saturated, see Fig. S1), and then removed *a patch* from the metacommunity and ran 219 the model until the second approximate equilibrium. We carried out this patch 220 removal and subsequent simulation individually for each patch. We included two

221	different dispersal behaviors that would have major consequences for the structure of
222	the metacommunity once the habitat is lost. Under directed dispersal, when we
223	remove a patch, we remove all of its connections to other patches, lowering the
224	number of patches $M$ by 1 in the third term of Eq. 3 (since the same dispersers are
225	now spread over fewer patches), and maintain dispersal between the remaining
226	patches. Under undirected dispersal, we do not remove a patch and its connections
227	from the system. Instead, we assume that the patch was destroyed permanently for
228	any species, but dispersal to it is still occurring through the existing connections
229	(Fig.1A). Species dispersing to the destroyed patch would die immediately.
230	Model parameters
231	Model parameters were: $M=50$ , $S=50$ , $e=0.2$ , $m=0.2$ , $I_j=165$ , $l_j=10$ , $P_{ij}(0)=10$ for all $i$
232	and <i>j</i> . Some of these parameters were set as they were in other studies (Loreau <i>et al</i> .
233	2003; Gonzalez et al. 2009; Shanafelt et al. 2015; Thompson et al. 2017). If the total
234	biomass of a species across the whole metacommunity fell below five, we considered
235	the species extinct in the metacommunity. This metacommunity scale extinction
236	cutoff is comparable to, but leads to simpler and faster code than, the patch-scale
237	cutoff of 0.1 used in previous studies (Loreau et al. 2003; Shanafelt et al. 2015), as it
238	is equal to $0.1$ times the number of patches $M$ . Furthermore, the small population size

239	effects that influence extinction risk (breakdown of mating and defense systems, and
240	demographic or environmental stochasticity effects) may be more accurately
241	considered as acting at the metacommunity scale when there is significant dispersal.
242	We studied patch removal under various consumption rates and magnitudes by
243	varying $b$ (which consumption rates are inversely proportional to) from 10, 20, 21, 22,
244	23, 24, and dispersal rates from 0.01, 0.02, 0.03, 0.04, 0.05, 0.06, 0.07, 0.08, 0.09, 0.1,
245	0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9, 1. These <i>b</i> values were chosen to confirm that the
246	behavior of the model was consistent with $b$ close to values causing collapse of the
247	metacommunity (b above 24).
248	Metapopulation simulations
249	To improve our overall understanding of community-level observations, we also
249 250	To improve our overall understanding of community-level observations, we also designed a metapopulation model to look at the effects of patch removal at the
249 250 251	To improve our overall understanding of community-level observations, we also designed a metapopulation model to look at the effects of patch removal at the species-level. The metapopulation is a version of the model with three patches (i.e. the
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<ul> <li>249</li> <li>250</li> <li>251</li> <li>252</li> <li>253</li> <li>254</li> <li>255</li> </ul>	To improve our overall understanding of community-level observations, we also designed a metapopulation model to look at the effects of patch removal at the species-level. The metapopulation is a version of the model with three patches (i.e. the M=3, $S=1$ case of our model). This kind of metapopulation model is able to eliminate the interaction between species, instead focusing on the match between the environmental condition of the removed patch and the optimal niche of the species. A metapopulation with three patches is the simplest topology, as it is still a

257	a patch would be more obvious in this kind of metapopulation than in
258	metapopulations with many patches. We considered two scenarios, one in which the
259	species was best adapted to the patch with harsh environmental condition, and one in
260	which the species was best adapted to the patch with benign environmental condition.
261	These scenarios correspond to the patches with harsh and benigh environment in the
262	metacommunity model, where each patch has the best adapted species. The
263	environmental condition $E$ for the three patches was 1, 0.98 and 0.96 (these were the
264	first three patches in the metacommunity model) in both scenarios, while the species
265	optimal niche value was $H=1$ in the first scenarios and $H=0.98$ in the second scenarios
266	(these were the first and second species in the metacommunity model). We removed
267	patches with different environmental conditions and assessed the extinction risk and
268	ecosystem function debt in the metapopulation model, allowing a deeper insight into
269	the metacommunity results. For the metapopulation simulations we used $I=67.6$ ,
270	<i>a</i> =0.01 and <i>b</i> =10.
271	Experimental Design

272 Each simulation ran for  $4 \times 10^{6}$  generations (specifically, we set per-capita death rate

273 as *m*=0.2, and simulated the model from t = 0 to  $t = 2 \times 10^{7}$ , which corresponds to

274 2×10^7×0.2=4×10^6 generations) to reach approximate equilibrium (Fig. S1), at

275 which point a patch was removed, before continuing to run for 4×10<sup>6</sup> generations to 276 reach a second approximate equilibrium post habitat destruction. We explained patch 277 removal under directed and undirected dispersal behavior in the section Simulated 278 patch removal and dispersal behaviors. 279 We also ran each simulation without patch removal for  $8 \times 10^{6}$  generations, to 280 compare with the patch removal case, and to identify whether species go extinct due 281 to metacommunity dynamics or habitat loss. For a given species, 1) if it goes extinct 282 before the first approximate equilibrium, this extinction is caused by metacommunity 283 dynamics; 2) if it goes extinct during the first to second approximate equilibrium, but 284 the persistence time is shorter when we remove a patch than when we don't, habitat 285 loss has hastened the extinction of the species already destined for extinction due to 286 metacommunity dynamics; 3) if it goes extinct during the first to second approximate 287 equilibrium under patch removal but not without it, then the extinction is driven by 288 habitat loss; 4) if it does not go extinct during the first to second approximate 289 equilibrium, then it's extinction risk is not substantially affected by metacommunity 290 dynamics or habitat loss. We only focused on the species which go extinct due to 291 habitat loss, or whose extinction is hastened by habitat loss, i.e. species falling into 292 categories 2 and 3 above. Revealing the nature of these extinctions due to patch

293 removal allowed us to specify extinction regimes across dispersal and resource

294	consumption rates.
	consumption rates.

295	We recorded the biomass of all species in each patch at the first approximate
296	equilibrium of each simulation to see the distributions of community composition and
297	ecosystem function along heterogeneous environments. We also recorded the
298	persistence times of each species across the whole metacommunity after patch
299	removal to measure the risk intensity of each extinct species due to patch removal as
300	the inverse of persistence time (unit as generations^-1, as recommended by Grimm
301	and Christian in 2004). To measure the magnitude of extinction debt, we calculated
302	the number of extinct species as the total number of extinct species at the second
303	approximate equilibrium minus the number of species going extinct immediately after
304	patch removal. The proportion of species going extinct was calculated as the number
305	of extinct species divided by the number of species before removal but after the initial
306	approximate equilibrium. To study total ecosystem function debt, we calculated the
307	total biomass change as the mean biomass across the M-1 remaining patches at the
308	second approximate equilibrium minus the mean biomass of each patch before
309	removal, and we also calculated the immediate biomass change as the mean biomass
310	in the M-1 remaining patches after patch removal minus the mean biomass before

311	removal. Under both directed and undirected dispersal, biomass change was given by
312	the total biomass change minus the immediate biomass change upon patch removal.
313	The proportion of biomass change was the biomass change over the mean biomass of
314	each patch before removal. With these data, we analyzed how the magnitude of
315	extinction and ecosystem function debt changed with dispersal rate and the
316	environmental condition of each removed patch under directed and undirected
317	dispersal behaviors.
318	We used the forward Euler method to simulate the differential equations with
319	<i>dt</i> =0.01. The simulation code was written in Java, and data were analyzed in R
320	version 4.1.3 (R 2022).
320 321	version 4.1.3 (R 2022). Results
<ul><li>320</li><li>321</li><li>322</li></ul>	version 4.1.3 (R 2022). Results Extinction regimes under different dispersal and consumption rates
<ul><li>320</li><li>321</li><li>322</li><li>323</li></ul>	version 4.1.3 (R 2022).   Results   Extinction regimes under different dispersal and consumption rates   Dispersal and consumption rates determine which species go extinct after habitat loss.
<ul> <li>320</li> <li>321</li> <li>322</li> <li>323</li> <li>324</li> </ul>	version 4.1.3 (R 2022).   Results   Extinction regimes under different dispersal and consumption rates   Dispersal and consumption rates determine which species go extinct after habitat loss.   We distinguish three extinction regimes (Fig. 2A) outside of a low consumption rate
<ul> <li>320</li> <li>321</li> <li>322</li> <li>323</li> <li>324</li> <li>325</li> </ul>	version 4.1.3 (R 2022).   Results   Extinction regimes under different dispersal and consumption rates   Dispersal and consumption rates determine which species go extinct after habitat loss.   We distinguish three extinction regimes (Fig. 2A) outside of a low consumption rate   zone where no species can coexist before patch removal (white area in Fig.2A).
<ul> <li>320</li> <li>321</li> <li>322</li> <li>323</li> <li>324</li> <li>325</li> <li>326</li> </ul>	version 4.1.3 (R 2022).   Results   Extinction regimes under different dispersal and consumption rates   Dispersal and consumption rates determine which species go extinct after habitat loss.   We distinguish three extinction regimes (Fig. 2A) outside of a low consumption rate   zone where no species can coexist before patch removal (white area in Fig.2A).   Under low dispersal and medium to high consumption rates (green area in Fig.
<ul> <li>320</li> <li>321</li> <li>322</li> <li>323</li> <li>324</li> <li>325</li> <li>326</li> <li>327</li> </ul>	version 4.1.3 (R 2022). <b>Results</b> <i>Extinction regimes under different dispersal and consumption rates</i> Dispersal and consumption rates determine which species go extinct after habitat loss. We distinguish three extinction regimes (Fig. 2A) outside of a low consumption rate zone where no species can coexist before patch removal (white area in Fig.2A). Under low dispersal and medium to high consumption rates (green area in Fig. 2A), removing a patch causes the species best adapted to it to go extinct under both

329	species best adapted to the patch being removed are at risk. However, that risk
330	(measured as the inverse of persistence time) varies and is highest under the removal
331	of the harsh environment patches (Fig. 2B and 2C). The extinction risk pattern in this
332	regimes makes intuitive sense given the pattern of biomass found before patch
333	removal, in which the biomass of each species is quite low other than in the patch it is
334	best adapted to (Fig. S2A).
335	Under medium dispersal rates and medium to high consumption rates (blue area
336	in Fig. 2A), removing a patch can cause both the species adapted to it and the next
337	most closely adapted species to go extinct under both directed and undirected
338	dispersal (Fig. 2D-G). This makes sense given the pattern of biomass found before
339	patch removal (Fig. S2B and S2C) in this extreme, where species have substantial
340	biomass in patches similar to the ones they are best adapted to, and hence removal of
341	such patches could have a large impact on their metacommunity biomass. In addition,
342	species adapted to harsh environments already destined for extinction may go extinct
343	faster due to patch removal under directed dispersal (Fig. 2D), even when patches
344	with very benign environments are removed (see Fig. S3, where the persistence time
345	of species with optimal niche value as 0.16 in the case without removal is $5.8 \times 10^5$
346	generations, but it is $1.34 \times 10^5$ when patch with environmental condition as 0.51 is

347	removed). In this regime, the species best adapted to the removed patch is at the
348	greatest risk, and the risk intensity decreases with the difference between the species
349	preferred condition and the environment of the removed patch (Fig. 2D-G).
350	Under high dispersal and high consumption rates (dark brown area in Fig. 2A),
351	the metacommunity is dominated by two species before patch removal (Fig. S2D),
352	removing a patch may cause one of the remaining species to go extinct (Fig. 2H)
353	under directed or undirected dispersal with high consumption rates (Fig. S4), or all of
354	them to go extinct under undirected dispersal with relatively low consumption rates
355	(Fig. 2I). In cases where the metacommunity has an odd number of patches and is
356	dominated by only one species before patch removal (Fig. S5), that species persisted
357	after patch removal, in both the directed and undirected dispersal cases. Hence in this
358	regime, extinction risk is sensitive to the patch structure and dispersal context. See
359	Fig. S6 as the rescaled version of Fig.2.
360	The magnitude of extinction debt and ecosystem function debt or credit under various
361	dispersal rates
362	We consider the dependence of extinction debt on dispersal rate, measuring extinction
363	debt as the number (Fig. 3A, B) and the proportion (Fig. 3C, D) of species going
364	extinct in the whole metacommunity. Generally, the mean number of extinct species

365	first increased and then decreased with dispersal rate in both the directed and
366	undirected dispersal cases for all consumption rates (Fig. 3A, B). However, the
367	proportion of species going extinct increased with dispersal rate for all three
368	maximum consumption rates under both directed and undirected dispersal, because
369	total species richness before patch removal declines with dispersal at high dispersal
370	(Fig. 3C, D). Under directed dispersal behavior, habitat loss can result in some small
371	benefits to the mean ecosystem function of the remaining patches for all consumption
372	and dispersal rates (y-axis are positive in Fig. 3E, G). This benefit had a mostly
373	positive relationship with dispersal rate under all consumption rates, especially when
374	measured as a proportional increase in biomass (Fig. 3G). Under undirected dispersal,
375	there was always an ecosystem function debt. Mean and proportional biomass
376	decreases were largely more substantial under higher dispersal rates (Fig. 3F, H).
377	The magnitudes of extinction debt and ecosystem function debt or credit with
378	environmental conditions of the removed patch
379	Variation in extinction debt with environmental conditions of the removed patch
380	occurred in regime 2 (in the other two regimes a single species went extinct regardless
381	of which patch was removed). Under both directed and undirected dispersal,
382	removing a patch with benign environmental conditions caused more species to go

383	extinct than removing a patch with harsh environmental conditions (Fig. 4A, B).
384	Under directed dispersal behavior and at low dispersal rates, a small
385	ecosystem function credit occurred no matter which patch was removed (Fig. 4C). At
386	medium to high dispersal, larger ecosystem function credit occurred only when harsh
387	patches were removed (Fig. 4E, G). Under undirected dispersal, there was always an
388	ecosystem function debt (Fig. 4D, F, H). In most scenarios, the relationship between
389	change in average biomass of remaining patches and the position on the
390	environmental gradient of the lost patch was U-shaped (Fig. 4C-H)-meaning that
391	change was always the least positive or the most negative when patches with benign
392	environments were removed. In other words, the removal of the benign patches had
393	either the least benefit, or caused the largest debt, depending on whether there were
394	ecosystem credits or debts.
395	Verifying extinction debt and ecosystem function credit mechanisms in
396	metapopulations
397	Under directed dispersal, habitat loss benefitted the mean biomass in the remaining
398	patches in the metapopulation analysis. The poorer the match between the species and
399	the environment in the removed patch, the higher the mean biomass for each of the
400	remaining patches (Fig.5A). We also hypothesized that extinction risk is often highest

401	when harsh habitat patches are removed in our metacommunity model because the
402	species in those patches were on average more poorly adapted to remaining patches.
403	In support of this, we found that the metapopulation experienced the greatest risk of
404	extinction upon removal of the best patch for the species when that best patch was a
405	harsh environment rather than a benign environment (Fig. 5B). Under undirected
406	dispersal, the population went extinct regardless of which patch was removed.
407	Discussion
408	Existing theory of extinction and ecosystem function debt is limited, especially for
409	heterogeneous landscapes. Here we contribute to the foundations of this theory
410	through the study of the effects of patch removal in a metacommunity model with
411	patches differing in environmental conditions. We find a number of behaviors that
412	make intuitive sense, such as the likelihood of highly adapted species going extinct,
413	and patches with benign environment causing more extinctions when removed. We
414	also find more surprising behaviors, such as the exacerbation of already existing
415	extinction risk of species adapted to harsh environments, even when patches with very
416	different environmental conditions were removed, an increase in the proportion of
417	species going extinct with dispersal, with the possibility of ecosystem credit in the
418	remaining patches under one of the two types of dispersal we explored.

420

dispersal

Trends in extinction risk across species, with different patches removed, and with

421	In our model, the number of species is equal to the number of patches, and each
422	species has a patch to which it is best adapted. For example, the species with $H_i = 0.5$
423	is best adapted to the patch with $E_j = 0.5$ . The diagonal in each panel of Fig. 2
424	indicates the species best adapted to each patch. As in previous studies, species
425	sorting under low dispersal rates allowed each species to dominate its preferred patch
426	(Mouquet & Loreau 2003; Suzuki & Economo 2021; Ai & Ellwood 2022). Hence,
427	removing a patch tends to cause the species better adapted to this patch to go extinct,
428	resulting in the diagonal risk zones in Fig. 2B-G. Species with central niches have
429	numerous relatively suitable habitats and are therefore less at risk than species with
430	extreme niches and few suitable habitats. We show that this can be understood in
431	terms of the principles applying to metapopulations, in that removing the best patch
432	for that species creates less of a risk if that patch is benign than if that patch is a harsh
433	environmental condition relative to the gradient experienced by that metapopulation
434	(Fig. 5B). At medium dispersal rates, species transit quickly between patches, and
435	hence patches with similar environmental conditions play important roles in
436	maintaining species even though they are sub-optimal patches. Regardless of which

437	patch is removed, the species adapted to harsh environments are at greater risk of
438	extinction; their low competitive ability prevents them from increasing their
439	productivity in any of the patches, and habitat loss reduces their biomass and further
440	exacerbates their extinction.
441	At high dispersal rates and with an even number of patches, mass effects generate
442	two dominant species (with $H_i$ around 0.5, see Fig. S2) that are well-matched regional
443	best competitors because they are equal in the number of patches they are better
444	adapted to. Patch removal can then disrupt this balance and cause extinction of one of
445	the species, or both if they both require large numbers of patches to persist, further
446	leading to horizontal risk zones (Fig.2H and I). This constitutes a high proportion of
447	species going extinct at high dispersal (Fig. 3C, D), since the regional species richness
448	is so low at high dispersal, due to increased competition for patches (Mouquet &
449	Loreau 2003; Ai & Ellwood 2022).
450	Ecosystem debts under undirected dispersal but credits under directed dispersal
451	Under directed dispersal in metapopulations, the harsher the removed patch, the
452	higher the mean biomass became for remaining patches (Fig. 5A). Dispersers which
453	previously dispersed to the less productive patches dispersed to more productive
454	patches after patch removal, which increased the biomass. Similar patterns could be

455	seen in the metacommunities, especially when harsh patches were removed (Fig. 4C,
456	E and G). However, while the mean biomass increased greatly in the metapopulation,
457	it increased only slightly in the metacommunity (see y-axis of Fig. 3E, G and 5A), and
458	in addition, removing a patch caused the whole biomass of the metacommunity to
459	decrease, whereas it increased in the metapopulation (Fig. 5A). This is because in the
460	metacommunity, the removed patch may have been the worst habitat for some
461	species, but the best habitat for other species, hence removing a patch would cause the
462	mean biomass of some species to increase and others to decrease. Moreover,
463	removing any patch from a metacommunity caused species to lose habitat, thus
464	increasing extinction, whereas in a metapopulation only the removal of the best patch
465	would cause it to go extinct.
466	Under undirected dispersal, biomass decreased in remaining patches after patch
467	removal (Fig. 3F, H). This was driven partly by the fact that a higher proportion of
468	species went extinct at high dispersal rates under undirected dispersal than under
469	directed dispersal (Fig. 3C, D). Also, and perhaps more importantly, under undirected
470	dispersal none of the individuals dispersing to destroyed habitats contribute to
471	ecosystem function.
472	Comparison with existing theory of extinction and ecosystem function debt

473	Tilman et al. (1994), one of the most influential theoretical studies of extinction
474	debt, concluded that when habitats are lost, extinction risk is greatest for the system's
475	best competitor. Implemented within the competition-colonization trade-off
476	framework, Tilman's model revealed that habitat destruction weakens the colonization
477	rate of all species, but it especially impacts the system's best competitor because of its
478	lower colonization rate. Whereas Tilman's model relied on the strength of trade-offs
479	between competition and colonization, our model focuses on heterogeneous
480	metacommunities, and hence the "best competitor" varies across habitats according to
481	their environmental condition.
482	The specific assumptions of perfect trade-offs in Tilman's model have been
483	criticized as being unrealistic (Loehle & Li 1996; Banks 1997; Malanson 2008).
484	Moreover, relaxing the assumptions leads to very different results, such as species
485	losses occurring more quickly than predicted by the model with perfect trade-off.
486	Moreover, this included not just competitive species, but all types of species could go
487	extinct due to habitat destruction (Loehle & Li 1996; Banks 1997). In addition, recent
488	studies (Li et al. 2020; Liao et al. 2022; Zhang et al. 2023) relaxed the strict
489	competition-colonization trade-offs through weakening relative competition strength
490	or violating the strict hierarchical competition by considering intransitive competition,

491	and found that the species loss oscillated with disturbance extent rather than following
492	a simple monotonic relationship as predicted. Even in these studies, the best
493	competitor suffered the most impact as the disturbance extent increases (see Fig.2 in
494	Liao <i>et al.</i> 2022).
495	Mouquet et al. (2011) investigated extinction debt in heterogeneous
496	metacommunities, concluding that less competitive species at the regional scale are
497	more strongly affected by habitat destruction, especially at high dispersal rates. Here
498	we noted that these regionally less competitive species, adapted to harsh
499	environments, had relatively low biomass before habitat loss (Fig. S2B, S2C). We
500	found that in fact, because of the metacommunity dynamics, these species went
501	extinct even without habitat loss, but it took a long time. Habitat loss can accelerate
502	their extinction but it is not the root cause. In nature, it is difficult to identify the
503	drivers of species extinctions because most systems are not at a steady state, and so
504	species extinctions might result from community dynamics, or extrinsic factors such
505	as habitat loss.
506	Implications of our results for biodiversity conservation
507	Extinction debt provides a window for species restoration and landscape
508	management (Kuussaari et al. 2009; Wearn et al. 2012), but which recovery plans

509	should be applied depends on many factors. Huxel and Hastings (1999) concluded
510	that, "either restoring patches adjacent to occupied patches or reintroducing the
511	species into restored patches increases the efficacy of the recovery effort". Our results
512	indicate that the type of extinction regime must be identified before deciding which
513	species to protect during restoration efforts, since different species are at risk in
514	different regimes. For example, when the most adapted species is most at risk, efforts
515	should focus on reintroducing the lost habitat and its best competitors, whereas in the
516	regime in which species adapted to the neighboring patches are also at risk, restoring
517	the adjacent patches should also be a priority. In some management cases, such as
518	quantifying the size of reservation areas with limited resources, protecting habitats
519	with benign environments should be prioritized, because losing this type of habitat
520	will cause the most extinction and ecosystem function debt since a wide range of
521	species are adapted to those habitats. Our model suggests that, regardless of which
522	habitats are lost, the total biomass of the whole metacommunity is lower, so the
523	habitats with harsh environments should also be protected. Undirected dispersal may
524	cause species to continue dispersing to the destroyed habitat, meaning that restoring
525	the lost habitat immediately would be a good way of restoring species richness and
526	ecosystem function. In summary, our results suggest that restoration actions should be

527 guided by extinction regime, dispersal behaviors, and landscape heterogeneity.

# *Future Directions*

529	Metacommunity theory is the theoretical framework for this model, and there
530	are some limiting assumptions which could be relaxed in future studies. First, most
531	natural landscapes are not spatially implicit, and community composition and
532	ecosystem function depend on the spatial configuration of patches, such as distance
533	between patches, topologies of metacommunities, patch size and shape etc. (Suzuki &
534	Economo 2021; Ai & Ellwood 2022; Zhang et al. 2023). Meanwhile, some other
535	parameters in our model could also be relaxed, for example, set variable dispersal
536	ability in species, or set variable dispersal networks (Zhang et al. 2020), or consider
537	the stage structure in the dispersal process, or even associate the dispersal among
538	patches with metacommunity topologies. These kinds of assumptions would help to
539	generalize the model and the results.
540	Conclusion
541	We developed a new theory of extinction and ecosystem function debt in a
542	heterogeneous landscape. Habitat loss hastens the extinction of species adapted to
543	extreme environments, and always causes extinction debt under both directed and
544	undirected dispersal. Interestingly, habitat loss causes ecosystem function debts under

545	undirected dispersal, but credits under directed dispersal. Both extinction debt and
546	ecosystem function credit/debt increase with dispersal rate. Our study indicates that
547	extinction regime, dispersal behavior, and the environmental conditions of habitats
548	should be considered before taking conservation actions to mitigate the effects of
549	habitat loss.
550	Data availability statement:
551	This theoretical study has no data; codes are available at link:
552	https://datadryad.org/stash/share/z0Mrqix19wJEzgLJN7GRi9oThNqiz7EoA0t3rm89_t
553	М.
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706	

### 708 Figure captions



710 Figure 1 Dispersal behavior and questions asked. Different shapes with dashed



717	dispersal, individuals continue to disperse to the destroyed habitat (represented by
718	lines with single arrows) and are lost. Panel B shows a diagram illustrating the patch
719	removal in a heterogeneous metacommunity and lists the questions raised when a
720	patch is removed, for example, the round patch has been removed here. Questions are
721	explained in detail in the main text. For simplicity, we show a metacommunity with
722	three patches, but our model includes 50 patches.
723	



The environmental condition of the removed patch (E<sub>i</sub>)

726	Figure 2 The regimes of extinction behavior under patch removal in the space of
727	possible consumption and dispersal rates (A), and the patterns of extinction risk
728	after patch removal in each regime (B-I). In panel A, the colored areas represent
729	distinct extinction regimes, whereas the four black dots indicate the values of
730	consumption and dispersal rates where we have shown species risk in panels B-I. In
731	the white area to the left of the colors in A, no species can persist in the
732	metacommunity even in the absence of patch removal. In B-I, squares of color
733	indicate the extinction risk (measured as the inverse of the number of generations the
734	species persists after patch removal) of the species with the corresponding
735	environmental niche value $(H_i)$ on the vertical axis, under removal of the patch with
736	the corresponding environmental condition $(E_j)$ on the horizontal axis. Hence colors
737	on the diagonal indicate a species' extinction risk when the patch it is best adapted to
738	(the patch for which $E_j = H_i$ ) is removed, and colors just to the left and right of the
739	diagonal indicate its risk when a patch close in environmental condition to its
740	environmental niche value is removed. White areas mean that species went extinct
741	before patch removal due to metacommunity dynamics, grey areas mean that species
742	did not go extinct under removal of that patch, and colorful areas mean that species
743	went extinct due to patch removal. As dispersal increases, the pattern of extinction

744	risk changes from only the species best adapted to the removed habitat being at risk
745	(first dot, B-C), to species adapted to similar habitats also being at risk $(2^{nd} \text{ and } 3^{rd})$
746	dots, D-G). In the latter, species with edge environmental niches may also be hastened
747	towards extinction when very different habitat patches are removed (2 <sup>nd</sup> dot, D). At
748	high dispersal (4 <sup>th</sup> dot, H-I), only 1-2 species persist before patch removal, and one or
749	both are vulnerable to extinction upon patch removal. Extinction regimes (panel A)
750	are the same both under directed and undirected dispersal. See text for a detailed
751	description of these regimes.





debt/credit (E-H) across dispersal rates. The left column of panels shows behaviors
under directed dispersal, and the right undirected dispersal. Line colors differ by
consumption rate as indicated in the legend in A. Extinction debt is shown as both the
number of species going extinct (A, B) and the proportion of species going extinct (C,

D). Ecosystem function debt/credit is shown as both the biomass change (E, F) and

760	proportion of biomass change (G, H). In panel E-H, positive y-axis means ecosystem
761	function credit, whereas negative means ecosystem function debt. Key trends are that
762	extinction debt generally increases with dispersal when measured as a proportion of
763	species, and that directed dispersal can result in an ecosystem function credit (but
764	extinction debt), while undirected dispersal always results in ecosystem debt. Note
765	that results differ somewhat for an odd number of patches but on the whole extinction
766	debt also increases with dispersal in that case.



768



## 770 magnitude of ecosystem function debt (C-H) on the environmental condition of

771 the removed patch. The left column of panels shows behaviors under directed

dispersal, and the right undirected dispersal. Line colors differ by consumption rate as

indicated in the legend in A. Panels A and B represent the number of species going

774	extinct for dispersal rate $a=0.05$ in the extinction regimes "additional species similarly
775	adapted also at risk". Panels C-H show ecosystem function debt/credit as the biomass
776	change for three different dispersal rates ( $a=0.01, 0.05$ , and 0.5, from top to bottom
777	row). The species cannot coexist when the dispersal rate is greater than 0.1 at low
778	consumption rate (b=21), so there are no data for that consumption rate in the last row
779	(G, H). The dotted line in panel E and G is for a change in biomass=0 since they
780	include both positive and negative biomass change. In panel C-H, positive values on
781	the y-axis mean ecosystem function credit, whereas negative values mean ecosystem
782	function debt. A key trend is that the removal of benign habitats has the largest effects.
783	There are 50 patches here. See details in the main text.



787 Figure 5 The biomass of each patch (A) and extinction risk intensity (B) with

788 patch removal in a metapopulation. Each metapopulation has three patches (square,

789	triangle, and circle with dashed line) and one species (the shape with solid line). For
790	the first scenario, i) the species (represented by a square solid line) is best adapted to
791	the square patch (see the diagram in panel A and the left diagram in panel B) and that
792	patch has a harsh environmental condition, while the triangle patch is the next-best
793	patch, and the circle patch is the worst patch for the species (see $E$ for patches and $H$
794	for species in the main text); for the second scenario, ii) the triangle patch is the best
795	patch for the species (represented by a triangle solid line) and has a benign
796	environmental condition (see the right diagram in panel B, see detailed descriptions in
797	the main text).