

1 Title: Clustering of dispersal corridors in metapopulations leads to higher rates of recovery
2 following subpopulation extinction

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5 Running Head: Faster extinction recovery with clustered corridors

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24 **Abstract**

25 Understanding how spatially divided populations are affected by the physical characteristics of
26 the landscapes they occupy is critical to their conservation. While some metapopulations have
27 dispersal corridors spread relatively evenly through space in a homogeneous arrangement such
28 that most subpopulations are connected to a few neighbors, others may have corridors clustered
29 in a heterogeneous arrangement, creating a few highly connected subpopulations and leaving
30 most subpopulations with only one or two neighbors. Graph theory and empirical data from
31 other biological and non-biological networks suggest that heterogeneous metapopulations should
32 be the most robust to subpopulation extinction. Here, I used *Pseudomonas syringae* pv. *syringae*
33 B728a in metapopulation microcosms to compare the recovery of metapopulations with
34 homogeneous and heterogeneous corridor arrangements following small, medium, and large
35 subpopulation extinction events. I found that while metapopulations with heterogeneous corridor
36 arrangements had the fastest rates of recovery following extinction events of all sizes and had the
37 shortest absolute time to recovery following medium-sized extinction events, metapopulations
38 with homogeneous corridor arrangements had the shortest time to recovery following the
39 smallest extinction events.

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42 **Key words:** corridor, dispersal, extinction, metapopulation, microcosm, patch, recovery,

43 subpopulation

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47 **Introduction**

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49 A metapopulation is a collection of subpopulations occupying spatially divided habitat
50 fragments, but connected via dispersal across less adequate habitat (Hanski 1991).

51 Understanding how patchy habitats and fragmentation affect survival, dispersal, and growth
52 within metapopulations is critical to their conservation. Theoretical models make many

53 assumptions about growth and dispersal to predict metapopulation dynamics (Amarasekare 1998,

54 Parker 1999, Hanski and Ovaskainen 2000). To effectively use such models to prioritize

55 conservation efforts, we must determine which are best supported by data using experimental

56 tests of their predictions (Kareiva 1989, Holyoak and Lawler 2005).

57 Metapopulation dynamics are affected by factors beyond the within-patch survival and

58 growth of constituent subpopulations. A growing body of literature links classic metapopulation

59 theory with graph, or network, theory. In the latter framework, the metapopulation's

60 subpopulations occupy habitat fragments, which are nodes in a graph, and dispersal between any

61 two fragments can be represented by a graph edge (Urban et al. 2009). Using this network-

62 theoretic approach, we can use graph analytic metrics such as degree distributions and spanning

63 trees to better describe metapopulations. One such metric is network heterogeneity, which can

64 range from entirely homogeneous, in which all subpopulations are equally connected to their

65 neighbors, to strongly heterogeneous, as, for example, when most subpopulations are connected

66 to only one or two neighbors while a few "hub" subpopulations are more highly connected (Dale

67 and Fortin 2010, Gilarranz and Bascompte 2012). In other words, the connectivity degree of each

68 subpopulation, or the number of dispersal corridors by which it is connected to the network is

69 uniform in a homogeneous metapopulation, but right-skewed in a more heterogeneous

70 metapopulation (Urban and Keitt 2001, Molofsky and Ferdy 2005, Artzy-Randrup and Stone
71 2010).

72 Interpreting metapopulation dynamics through the lens of such network-theoretic metrics
73 allows us to draw parallels among diverse types of networks, both biological and non-biological.
74 For example, empirical work in systems as diverse as the World Wide Web (Albert et al. 2000)
75 and stream systems (Fagan 2002) supports models suggesting that heterogeneous networks are
76 more robust to local failure than their more homogeneous counterparts.

77 Theoretical modelling by Albert and colleagues (2000) demonstrated that heterogeneous
78 networks are more robust to failures and that such robustness cannot be explained simply by path
79 redundancy. Rather, heterogeneous networks have a smaller diameter, or average path length
80 between any two nodes, and this interconnectedness allows information, or in a metapopulation,
81 dispersing individuals, to cross the network more quickly. And a heterogeneous network has so
82 few nodes that are highly connected that a random local failure is more likely to happen at a
83 poorly connected node. Thus, the remaining nodes in such a network can still communicate with
84 each other, which makes the network less likely to experience global failure than would a more
85 homogeneous network experiencing a similar level of local failure.

86 In metapopulations, a subpopulation can be rescued from extinction by migration from
87 neighboring populations, and such rescues can increase the probability that the metapopulation
88 will persist through time (Brown and Kodric-Brown 1977). Subpopulation connectedness can
89 influence how important this rescue effect is to metapopulation persistence. Because highly
90 connected subpopulations are more likely to be rescued by migration from neighbors, a
91 metapopulation containing many highly connected subpopulations is expected to have a greater
92 rate of recovery and, therefore, a higher probability of persistence than a metapopulation with

93 fewer highly connected subpopulations (Eriksson et al. 2014). However, whether the
94 arrangement of dispersal corridors in space can itself affect the recovery of the metapopulation
95 has rarely been explored empirically, despite the fact that a deeper understanding of such
96 dynamics could provide critical conservation information.

97 In addition, the size of the extinction, or percentage of subpopulations that go extinct,
98 could interact with the effects of corridor arrangement on metapopulation recovery. For
99 example, Albert and colleagues found that as the fraction of nodes lost in a network rose, the
100 diameter of the network rose slowly in more homogeneous networks. However, in
101 heterogeneous networks, the diameter of the network was largely unaffected when nodes were
102 lost at random, but when highly connected nodes were targeted for removal, the diameter of the
103 network increased rapidly as the fraction of nodes lost increased.

104 Here, I used the network-theoretic concept of heterogeneity to characterize dispersal
105 corridor spatial distributions of experimental bacterial metapopulations and tested the effect of
106 that heterogeneity on recovery following subpopulation extinctions. I predicted that
107 metapopulations with heterogeneous corridor arrangements, compared to those with
108 homogeneous arrangements, would recover faster from subpopulation extinction events of all
109 sizes. However, at higher extinction sizes, I predicted the difference in recovery rate between
110 homogeneous and heterogeneous metapopulations would be larger.

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Methods

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116 To compare the rate of recolonization and recovery following subpopulation extinction
117 between metapopulations with different degrees of network heterogeneity in their dispersal
118 corridors I conducted a fully crossed experiment with two treatments: corridor arrangement and
119 extinction level. Corridor arrangement treatments were produced in Metapopulation Microcosm
120 Plates (MMPs), which are devices that resemble 96-well microtiter plates in size and shape but
121 with corridors connecting the wells in a desired spatial configuration. Complete methods for
122 design, construction, and assembly of MMPs are described by Kurkjian (2019). Here, 95 of the
123 96 wells were connected by 176 corridors. One unconnected well served as an uninoculated
124 control. The three corridor arrangements were: homogeneous, heterogeneous, and variable
125 (Figure 1). In the homogeneous arrangement, wells were connected in an even lattice
126 arrangement by 4.0 mm corridors. In the heterogeneous arrangement, wells were connected by
127 corridors of variable lengths. Well connectivities followed a right-skewed distribution such that
128 a few wells were highly connected whereas most were connected to only one or two other wells.
129 In the variable arrangement, the wells were connected in the even lattice arrangement of the
130 homogeneous treatment with the distribution of corridor lengths matching the heterogeneous
131 treatment.

132 The extinction levels were 10% (10 wells) extinct, 50% (48 wells) extinct, 90% (86
133 wells) extinct, and a 0% no extinction control. In each run of the experiment (Figure 2), parent
134 plates of each corridor arrangement treatment level were needle inoculated in every well with
135 *Pseudomonas syringae* pv. *syringae* B728a expressing green fluorescent protein from an
136 overnight culture at a concentration of $1.0\text{-}1.7 \times 10^6$ CFU/mL. The needles used for all
137 inoculations in this experiment transferred 0.651 ± 0.347 (SD) μL volume, meaning that each
138 well of the parent plates began with approximately $2.2\text{-}12.1 \times 10^3$ CFU/mL in each well. Parent

139 plates were incubated at 22°C for 48 hours, then subsampled into daughter plates with identical
140 corridor arrangements using a plate replicator from which a random subset of pins had been
141 removed (e.g. to create the 10% treatment, 10 pins were removed from the plate replicator). The
142 identities of the extinct wells were chosen by sorting the wells of the heterogeneous corridor
143 arrangement by their degree (number of neighbors) and choosing a random subset of each
144 degree, approximately in proportion with the overall degree distribution of the heterogeneous
145 treatment (Figure S1). The same well identities were extinguished in every corridor arrangement
146 treatment, such that if well B6 was chosen as an extinct well, it was made to go extinct in all
147 plates. Each daughter plate was incubated at 22°C for 156 hours and its fluorescence was
148 measured in a microplate reader every 12 hours. The full experiment was run six times for a
149 total of six replicates (plates) of each treatment combination. All uninoculated controls remained
150 sterile.

151 To measure recovery following extinction, I calculated the deviation of each well on each
152 treatment plate from the corresponding well on its control plate (the 0% extinction plate
153 subsampled from the same parent plate), normalized to the control value and starting inoculum
154 concentration, using the following formula:

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$$156 \quad \textit{Deviation from Control} = ((\textit{Fluor}_T/\textit{Start}) - (\textit{Fluor}_C/\textit{Start})) / (\textit{Fluor}_C/\textit{Start})$$

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158 where \textit{Fluor}_T is the fluorescence of the treatment well, \textit{Fluor}_C is the fluorescence of the
159 corresponding control well, and \textit{Start} is the concentration of the culture from which the parent
160 plate was inoculated (Figures S2 and S3). I used this quantity to calculate three recovery
161 metrics:

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163 1) Time to Begin Recovery Phase – I defined the recovery phase of each plate’s time series
164 as the portion during which the change in deviation from the control was positive (i.e. the
165 treatment plate was becoming more similar to the control). I calculated the time at which
166 this recovery phase began for each plate.

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168 2) Maximum Rate of Recovery – I fit each recovery phase with linear, quadratic, cubic, and
169 quartic regressions and chose the model which fit the data best using AICc model
170 selection. I then found the derivative of each best fit model and used it to calculate the
171 maximum rate of recovery during the recovery phase for each treatment.

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173 3) Time to Recovery – Using a planned contrast at each time step, I defined the time to
174 recovery as the first time period at which there was no significant difference between the
175 treatment and control.

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177 I calculated all three metrics for each combination of corridor arrangement and extinction level
178 both for the mean of all subpopulations and for extinction subpopulations only. To assess
179 whether variability in re-colonization affected the recovery of the metapopulations, I also
180 calculated the standard deviation of Deviation of Control of the wells of each plate.

181 Finally, with all subpopulations, I fit several multiple polynomial linear mixed effects
182 models using Deviation from Control after 36 hours (to include only time points after recovery
183 had begun for all plates) as the response variable and including run of the experiment as a
184 random effect. Hours post-extinction, hours post-extinction squared, and hours post-extinction

185 cubed were included as explanatory variables in all models. One or more of the following were
186 also included in each model as explanatory variables: corridor arrangement, extinction level, and
187 the interaction between corridor arrangement and extinction level. I chose the model with the
188 lowest AIC_c value as the best fit model. I repeated this procedure with extinct subpopulations
189 only. I assessed homogeneity of variance and normality of residuals by plotting the residuals.
190 All analyses were performed in R (R Core Team 2016).

191 In all runs of this experiment, MMPs were filled with Luria-Bertani liquid medium (LB;
192 Cold Spring Harbor Protocols 2006) containing 40 µg/mL nitrofurantoin and 15 µg/mL
193 tetracycline to select against loss of mutant plasmids. This experiment was performed using wild
194 type *Pseudomonas syringae* pv. *syringae* B728a containing the *pkln42gfp* plasmid, which
195 constitutively expresses *gfp* (Dulla and Lindow 2008). This strain is motile and doubles
196 approximately every 3 h at 15°C in this set-up. This strain was generously provided to me by the
197 Lindow Lab, UC Berkeley.

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Results

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203 Time to Begin Recovery Phase

204 Apart from the homogeneous corridor 10% extinction treatment combination, which
205 grew faster than the control during all periods, all 10% and 50% extinction metapopulations
206 began to recover (approach their controls) after 24 hours. All 90% extinction metapopulations
207 began to recover after 36 hours (Table S1a). All extinct subpopulations in the 10% and 50%

208 extinction metapopulations began to recover after 24 hours and all extinct subpopulations in the
209 90% extinction metapopulations began to recover after 36 hours (Table S1b).

210 In general, the metapopulations experiencing 10% extinction recovered to the highest
211 mean subpopulation sizes (relative to their controls), 50% extinction had intermediate
212 subpopulation sizes, and the 90% extinction treatment had the lowest subpopulations sizes.
213 Within the 10% extinction treatment, metapopulations with homogeneous corridors reached the
214 highest subpopulations sizes, while in the 50% extinction treatment, metapopulations with
215 heterogeneous corridors were highest (Figure 3a). The patterns were the same amongst the
216 extinct wells (Figure 3b). Standard deviation of the “deviation from control” response variable
217 was highest in the 90% extinction treatment, intermediate in the 50% extinction treatment, and
218 lowest in the 10% extinction treatment. Within the 10% extinction treatment, the standard
219 deviation was highest in the variable corridor metapopulations, intermediate in heterogeneous
220 corridors, and lowest in homogeneous. Within the 50% and 90% extinction treatments, the
221 standard deviation was highest in heterogeneous metapopulations, intermediate in variable
222 metapopulations, and lowest in homogeneous (Figure 4a). The patterns were generally the same
223 amongst extinct wells, with the exception of the heterogeneous 50% extinction treatment
224 combination, which had the highest standard deviation (Figure 4b).

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227 Maximum Rate of Recovery

228 All recovery phase trajectories were fit best (had the lowest AIC_c value) by a quadratic or
229 cubic model. The mean recoveries of all subpopulations were fit best by cubic models for all
230 treatment combinations except 10% and 50% extinction in homogeneous corridors and 10%

231 extinction in variable corridors, which were fit best by quadratic models (Table S2a). The
232 recoveries of extinct subpopulations were fit best by cubic models for all treatment combinations
233 except 10% extinction in homogeneous and heterogeneous corridors and 50% extinction in
234 variable corridors; those exceptions were fit best by quadratic models (Table S2b).

235 For the mean recovery of all subpopulations, the maximum rate of recovery in
236 metapopulations with heterogeneous corridors was highest in the 50% extinction treatment,
237 while in metapopulations with homogeneous and variable corridors the rate was highest in the
238 90% extinction treatment. At all extinction levels, recovery was fastest in metapopulations with
239 heterogeneous corridors (Figure 5a). Recovery amongst extinct subpopulations in
240 metapopulations with heterogeneous corridors was fastest in the 50% extinction treatment. In
241 metapopulations with homogeneous or variable corridors, recovery was fastest in the 90%
242 extinction treatment. At all extinction levels, recovery of extinct wells was fastest in
243 metapopulations with heterogeneous corridor arrangements (Figure 5b).

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246 Time to Recovery

247 In the 10% extinction treatment, metapopulations with a heterogeneous corridor
248 arrangement became statistically indistinguishable from their controls after 60 hours, whereas
249 homogeneous metapopulations took only 12 hours, and metapopulations with the variable
250 corridor arrangement took 36 hours. In the 50% extinction treatment, heterogeneous
251 metapopulations still took 60 hours to recover to the control condition, whereas homogeneous
252 metapopulations took 108 hours, and metapopulations with the variable corridor arrangement

253 took 72 hours. No metapopulations recovered to the control condition following a 90%
254 extinction (Figure 6a).

255 In metapopulations with the heterogeneous corridor arrangement, extinct subpopulations
256 took an average of 72 hours to recover from both 10% and 50% extinction treatments. In
257 metapopulations with the homogeneous corridor arrangement, extinct subpopulations took 36
258 hours and 144 hours to recover from 10% and 50% extinction treatments, respectively. In
259 metapopulations with variable corridor arrangements, extinct subpopulations took 48 hours and
260 120 hours to recover from 10% and 50% extinction treatments, respectively (Figure 6b).

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263 Polynomial Regression

264 All polynomial regression models of all subpopulation recoveries containing an
265 interaction term fit the data equivalently well and better than all models that did not contain an
266 interaction term (Table S3a). Tukey's Honestly Significant Difference showed that all corridor
267 arrangement by extinction level treatment combinations were significantly different from all
268 others ($p_{\text{HSD}} < 0.001$) except 50% extinction in homogeneous corridors from 50% extinction in
269 heterogeneous corridors ($p_{\text{HSD}} = 0.149$), 90% extinction in variable corridors from 90%
270 extinction in homogeneous corridors ($p_{\text{HSD}} = 0.102$), and 90% extinction in variable corridors
271 from 90% extinction in heterogeneous corridors ($p_{\text{HSD}} = 0.173$). The same pattern was true
272 among models of extinct subpopulation recoveries (Table S3b).

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Discussion

In this experiment, metapopulations with heterogeneous corridor arrangements had a faster maximum rate of recovery from all levels of extinction than those with other corridor arrangements, which matched my predictions. However, that greater maximum speed translated to a shorter absolute time to recovery only in the 50% extinction treatment. Following the 10% extinction treatment, metapopulations with homogeneous corridor arrangements recovered to their control condition in a shorter absolute time than did the heterogeneous metapopulations. This may be because when extinctions occur in a random subset of subpopulations, few are likely to occur in adjacent subpopulations, and therefore an extinct subpopulation is almost certain to be adjacent to an occupied subpopulation, which leads to rapid recolonization. In the homogeneous treatment, that extinct subpopulation is always only a short distance from its occupied neighbor, again leading to rapid recolonization. However, in the heterogeneous treatment, bacteria may have to travel down a longer corridor, despite being only a short straight-line distance from the adjacent extinct subpopulation. Most work on network heterogeneity, including that of Albert and colleagues (2000), considers unweighted networks, in which all edges are equal. In metapopulation ecology, however, the edges of spatial networks are weighted. Assuming that the time it takes an individual organism to travel from one patch to another is a function of the length of corridor between the patches, each dispersal corridor could be assigned a weight proportional to its length (Urban et al. 2009). Therefore, predictions that heterogeneous networks will be the most robust to node failure in unweighted networks may not translate perfectly to recovery dynamics in weighted networks.

298 To further examine this idea, we can compare the recovery of metapopulations with the
299 variable corridor arrangement to that of metapopulations with heterogeneous and homogeneous
300 corridor arrangements. The variable treatment had a corridor arrangement in a regular lattice
301 which matched that of the homogeneous treatment, but with a distribution of corridor lengths
302 which matched that of the heterogeneous treatment. If the differences in recovery rate and time
303 between the heterogeneous and homogeneous treatments can be explained primarily by the
304 addition of longer dispersal corridors in the heterogeneous treatment, then recovery in the
305 variable treatment should be similar to that of the heterogeneous treatment. If instead those
306 differences are due primarily to differences in the corridor arrangement, then recovery in the
307 variable treatment should be more similar to that of the homogeneous treatment. With one
308 exception, both the rates of recovery and the absolute times to recovery of the variable treatment
309 are intermediate between those of the heterogeneous and homogeneous treatments at every
310 extinction level. The exception is that the recovery rate of all subpopulations in the 10%
311 extinction treatment was slightly lower in metapopulations with a variable corridor arrangement
312 than in those with either of the other corridor treatments. This result suggests that, while some of
313 the difference between the heterogeneous and homogeneous treatments can be attributed to their
314 differences in corridor length, a portion can also be attributed directly to differences in their
315 network heterogeneity.

316 In standard deviation too, the variable corridor treatment is intermediate between the
317 homogeneous corridors, which had the lowest variation in deviation from the control, and
318 heterogeneous corridors, which had the highest variation, except for at the 10% extinction level
319 when the variable treatment slightly exceeded the heterogeneous in variability. Here again, it is
320 possible that the difference in corridor length between the homogeneous and heterogeneous

321 treatments contributes to the difference in their variability, because in order for an extinct
322 subpopulation to be recolonized individuals must travel from an occupied subpopulation. In a
323 homogeneous metapopulation that distance is always the same, while in a heterogeneous
324 metapopulation the distance, and therefore the time for an extinct subpopulation to be
325 recolonized, will vary. If this variability in corridor length and travel time was the principal
326 driver of differences in recovery between the heterogeneous and homogeneous corridor
327 treatments, we would expect the standard deviation of the variable treatment to match that of the
328 heterogeneous treatment. Its position intermediate between the heterogeneous and homogeneous
329 treatments suggests again that while corridor length contributes to this difference, network
330 heterogeneity is also important.

331 While it is unfortunate that none of the 90% extinction treatments recovered fully to the
332 level of their controls within the timeframe of the experiments (which due to the physical
333 constraints of the experimental setup had to be confined to 5 days), we can consider their rates of
334 recovery. In the 10% extinction treatment, despite the fact that the heterogeneous corridor
335 metapopulation has the highest rate of recovery, the homogeneous corridor metapopulation
336 recovers in the shortest absolute time because it begins its recovery sooner and never deviates as
337 far from its control as does the heterogeneous metapopulation. In the 50% extinction treatment,
338 the heterogeneous metapopulation again deviates the farthest from its own control, but the
339 combination of beginning its recovery at the same time as the other corridor arrangements and
340 having a faster rate of recovery, leads to a shorter absolute time to recovery. In the 90%
341 extinction treatment, by comparison, the heterogeneous metapopulation continues its pattern of
342 deviating the farthest from its control, although the difference between the greatest deviation of
343 the homogeneous and heterogeneous treatments is smaller than in the other extinction treatments.

344 The heterogeneous metapopulation also has the highest rate of recovery in the 90% extinction
345 treatment, although it is lower than the rate of recovery of the heterogeneous metapopulation in
346 the 50% extinction treatment and the difference between the rates of the heterogeneous and
347 homogeneous metapopulations in the 90% extinction treatment is much smaller than in the 50%
348 extinction treatment. This suggests that while the heterogeneous metapopulation had the shortest
349 time to recovery in the 50% extinction treatment, that pattern might not have continued in the
350 90% extinction treatments, had they had the time and resources to recover to control levels. This
351 switch in order of recovery from homogeneous soonest in the 10% extinction treatment, to
352 heterogeneous soonest in the 50% extinction treatment, and possibly back again in the 90%
353 extinction treatment, may explain why the data was fit equally well by all polynomial models
354 that contained a corridor arrangement by extinction level interaction term, meaning that all
355 recovery trajectories were different from each other with no clear trends in the main effects of
356 corridor arrangement or extinction level.

357 The effects of rescue by recolonization from neighboring subpopulations are important
358 for long-term metapopulation persistence. For example, and Gilarranz and Bascompte (2012)
359 used variations of the Hanski and Ovaskainen (2000) and Levins (1969) models to predict that
360 higher network heterogeneity in metapopulations will lead to a higher proportion of occupied
361 patches, but that that pattern will reverse at low extinction-to-colonization ratios. Highly
362 connected subpopulations are likely to be recolonized more quickly by their occupied neighbors.
363 But whether that higher rate of recolonization translates into faster recovery of the
364 metapopulation has not been explored empirically. In this experiment, extinct subpopulations
365 were recolonized fastest in metapopulations with heterogeneous corridors, and that
366 recolonization led to faster recovery of the metapopulations themselves.

367 Understanding this type of interplay between connectivity and metapopulation dynamics
368 is critical to conservation. For example, Fortuna and colleagues (Fortuna et al. 2006) found that
369 variation in the wetness of the environment affected the availability of dispersal corridors for
370 amphibians between temporary ponds, but that because of underlying connectivity amongst the
371 ponds, amphibian dispersal was unlikely to be dramatically affected by loss of individual ponds.
372 Similarly, Cowley, Johnson, and Pocock (2015) used network-theoretic metrics to model the
373 spread of oak processionary moths through oak woodlands and identify the most important
374 patches or ‘pinch points’ of invasion. Their results suggested that the patches they identified
375 would be most critical for conservation interventions to prevent invasion spread.

376 A better understanding of how the spatial arrangement of dispersal corridors on the
377 landscape might lead to different recovery outcomes depending on the size of the of the
378 extinction event could be useful in guiding efforts to plan the positions of artificial dispersal
379 corridors between habitat patches or preserve existing corridors. This experiment demonstrates
380 that heterogeneous connectivity among subpopulations can lead to faster recovery following
381 subpopulation extinction, but that greater speed of recovery may not always translate to a shorter
382 absolute time to recovery.

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460 Figure Legends

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462 Figure 1. Corridor arrangement treatments: diagrams (top), histograms of corridor lengths
463 (middle), histograms of neighbors per subpopulation (bottom).

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465 Figure 2. Experimental design: For each run of the experiment, all wells of the master-parent
466 plate were filled from an overnight culture of *Pseudomonas syringae*. All wells of the master-
467 parent plate were then sub-sampled into parent plates with homogeneous, heterogeneous, and
468 variable corridor arrangements. These parent plates were incubated at 22°C for 48 hours. To
469 create the different extinction levels, each parent plate was subsampled to daughter plates with
470 identical corridor arrangements with either 10%, 50%, or 90% of plate replicator pins removed.
471 The daughter plates were incubated 22°C for 156 hours and each was read in a plate reader every
472 12 hours.

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474 Figure 3. Deviation of treatment plates from control plates with heterogeneous, homogeneous,
475 or variable corridor arrangements following 10%, 50%, or 90% extinction of a) all
476 subpopulations or b) extinct subpopulations only. Error bars are standard error of the mean
477 (n=6). Thick lines are best fit lines for each recovery trajectory.

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479 Figure 4. Standard deviation of “deviation from control” response variable for heterogeneous,
480 homogeneous, or variable corridor arrangements following 10%, 50%, or 90% extinction of a)
481 all subpopulations or b) extinct subpopulations only.

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483 Figure 5. Maximum rate of recovery, measured as change in “deviation from control” response
484 variable per hour, for each combination of corridor arrangement and extinction level for a) all
485 subpopulations and b) extinct subpopulations only. This recovery rate maximum was calculated
486 by finding the maximum derivative of the best-fit model [see Table S2] during the recovery
487 phase.

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489 Figure 6. Hours to recovery for each corridor arrangement by extinction level treatment
490 combination for a) all subpopulations and b) extinct subpopulations only.

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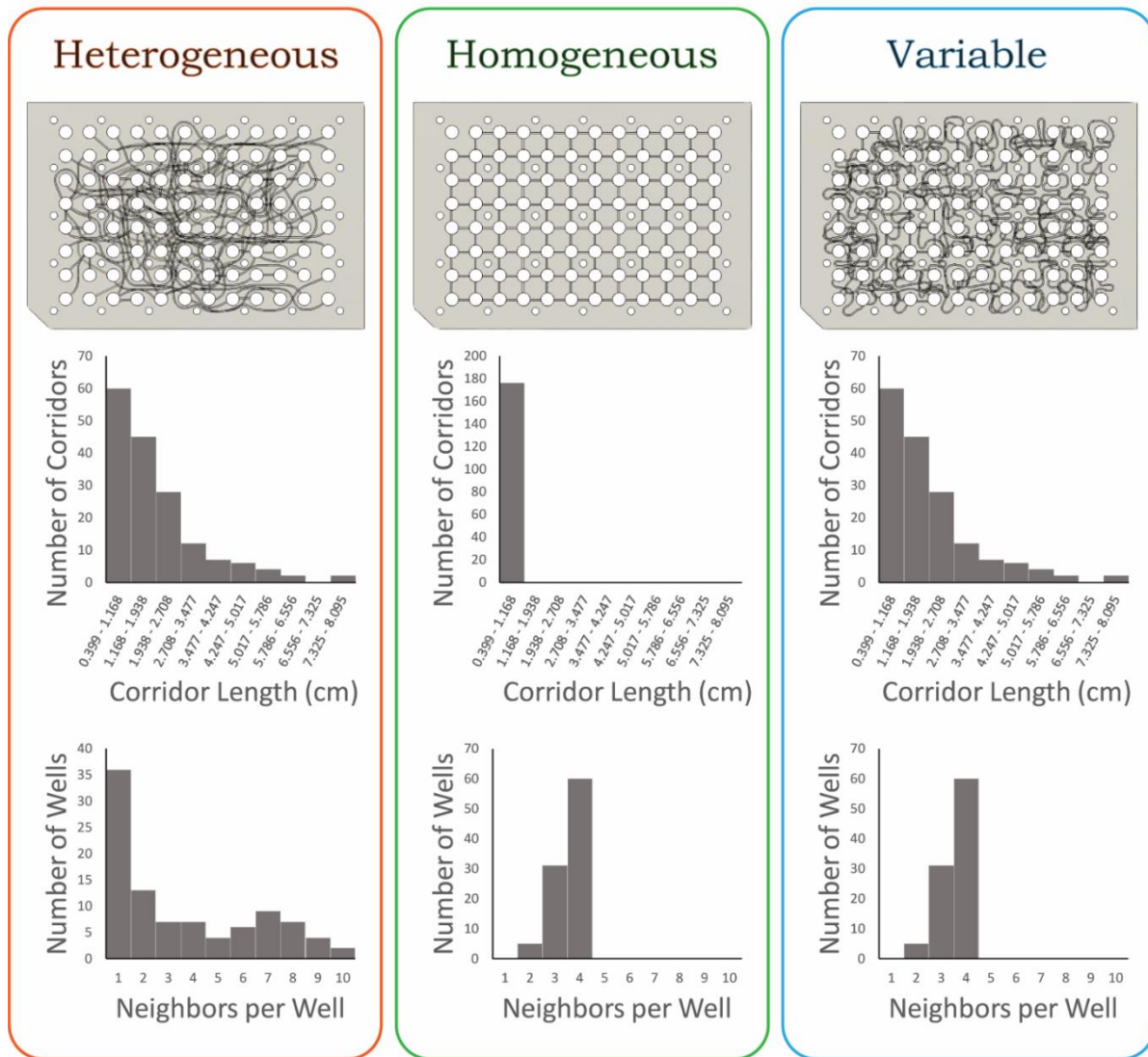
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506 Figure 1



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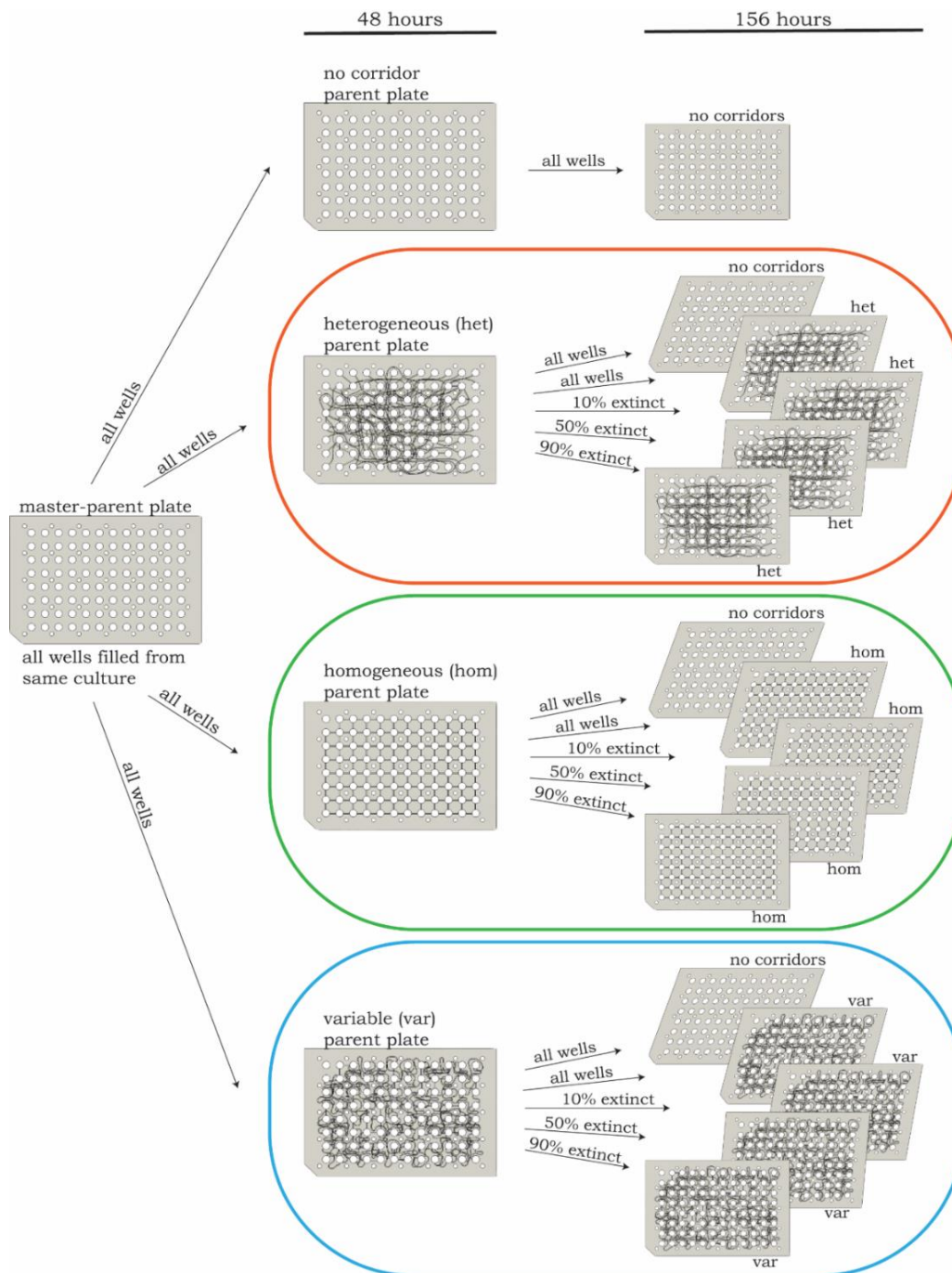
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515 Figure 2



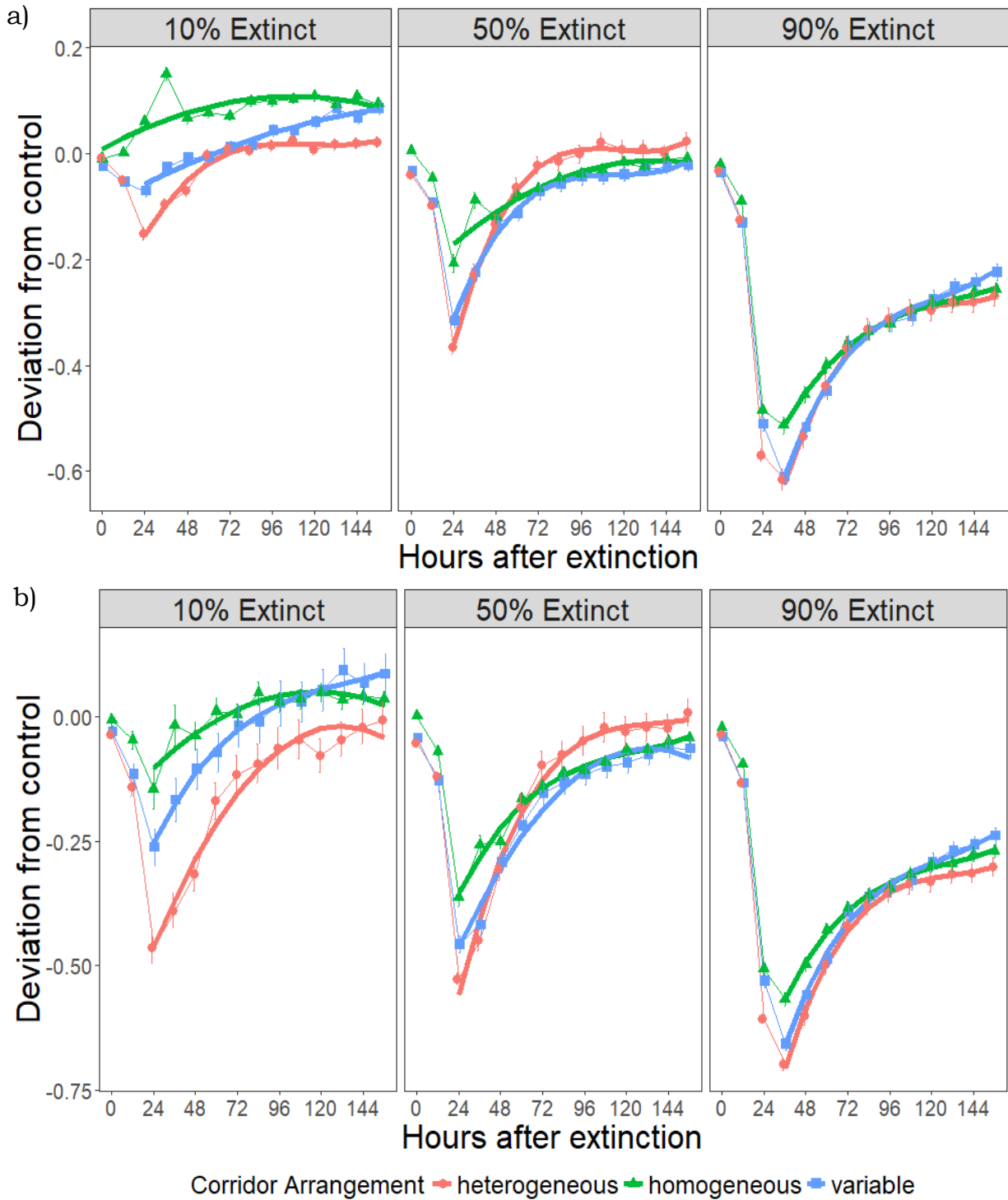
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520 Figure 3

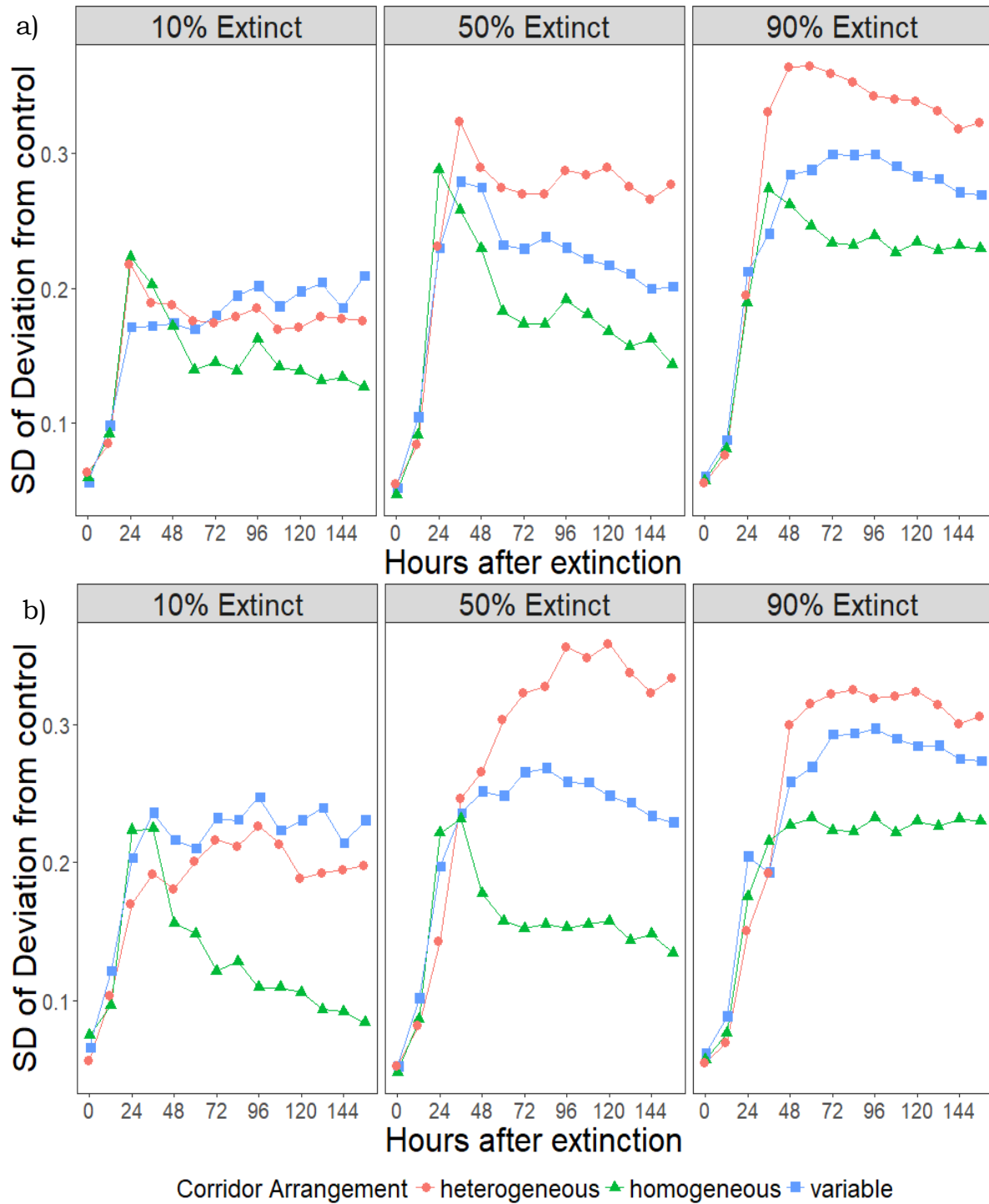


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524 Figure 4

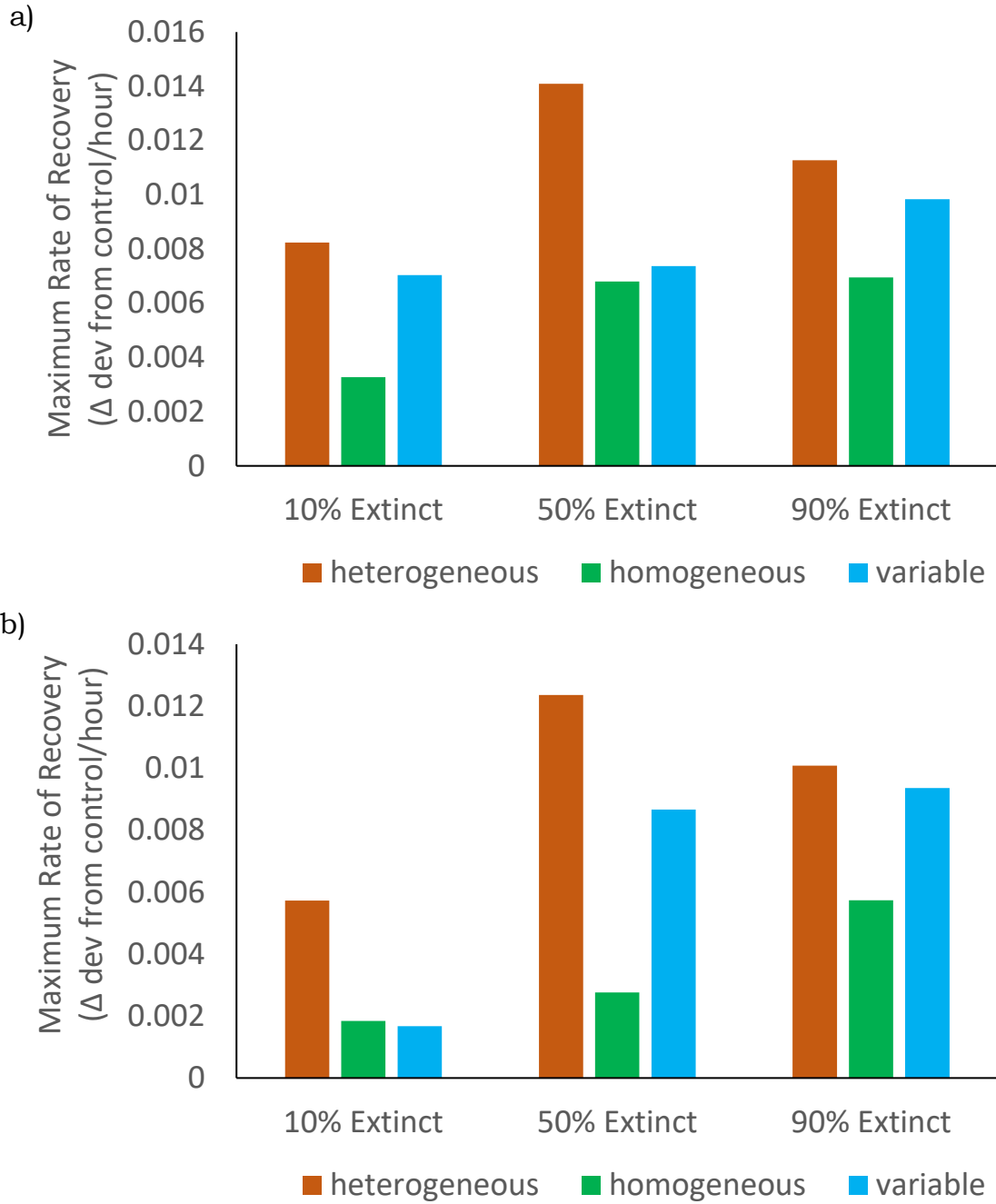


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528 Figure 5



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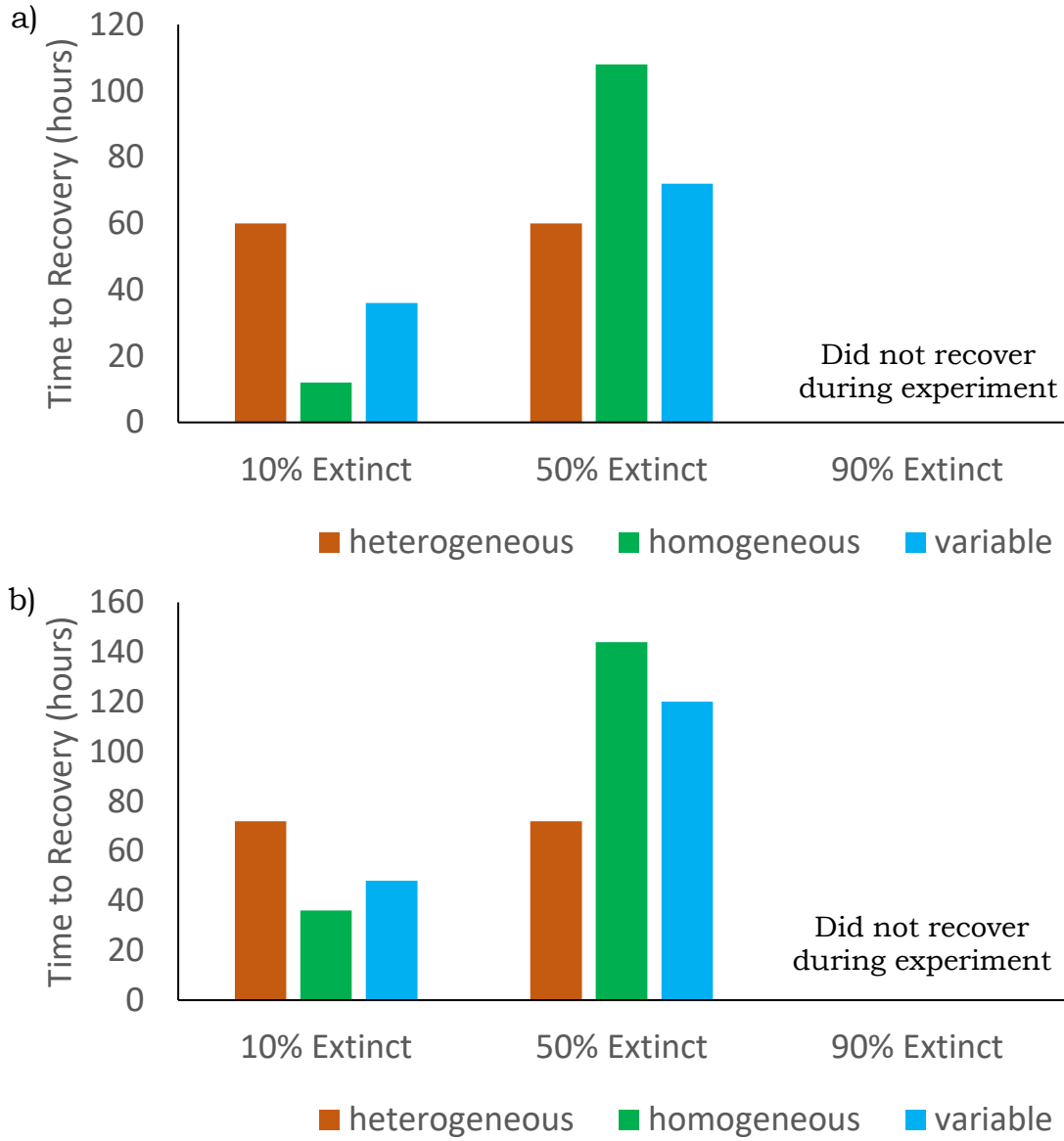
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534 Figure 6



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