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Running Head: Dose relationships and prevalence

Dose relationships can exacerbate, mute, or reverse the impact of heterospecific host density on infection prevalence

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Abstract– The likelihood an individual becomes infected depends on the community in which it is embedded. For environmentally transmitted parasites, host community composition can alter host density, the density of parasites that hosts encounter in the environment, and the dose to
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30 which hosts are subsequently exposed. While some multi-host theory incorporates some of these
31 factors (e.g., competition among hosts), it does not currently consider the nonlinear relationships
32 between parasite exposure dose and per-propagule infectivity (dose-infectivity relationships),
33 between exposure dose and infected host mortality (dose-mortality relationships), and between
34 exposure dose and parasite propagule excretion (dose-excretion relationships). This makes it
35 difficult to predict the impact of host species on one another's likelihood of infection. To
36 understand the implications of these non-linear dose relationships for multi-host communities,
37 we first performed a meta-analysis on published dose-infectivity experiments to quantify the
38 proportion of accelerating, linear, or decelerating dose-infectivity relationships; we found that
39 most experiments demonstrated decelerating dose-infectivity relationships. We then explored
40 how dose-infectivity, dose-mortality, and dose-excretion relationships might alter the impact of
41 heterospecific host density on infectious propagule density, infection prevalence, and density of a
42 focal host using two-host, one-parasite models. We found that dose relationships either decreased
43 the magnitude of the impact of heterospecific host density on propagule density and infection
44 prevalence via negative feedback loops (decelerating dose-infectivity relationships, positive
45 dose-mortality relationships, and negative dose-excretion relationships), or increased the
46 magnitude of the impact of heterospecific host density on infection prevalence via positive
47 feedback loops (accelerating dose-infectivity relationships and positive dose-excretion
48 relationships). Further, positive dose-mortality relationships resulted in hosts that traditionally
49 decrease disease (e.g. low-competence, strong competitors) increasing infection prevalence, and
50 vice versa. Finally, we found that dose-relationships can create positive feedback loops that
51 facilitate friendly competition (i.e., increased heterospecific density has a positive effect on focal
52 host density because the reduction in disease outweighs the negative effects of interspecific
53 competition). This suggests that without taking dose relationships into account, we may
54 incorrectly predict the effect of heterospecific host interactions, and thus host community
55 composition, on environmentally transmitted parasites.

56 **Key Words:** Dilution, Amplification, Infection, Prevalence, Infection risk

57 **Introduction**

58 Hosts and their parasites do not exist in isolation. Rather, the likelihood of infection of
59 any individual host (i.e. the probability an individual is infected in a short time interval) depends
60 on the community in which it is embedded, due to direct inter-specific competition and cross-

61 species parasite transmission (O'Regan et al. 2015). Competitors can “amplify” (i.e. increase)
62 infection prevalence in a host species if they have high infection “competence”, meaning they
63 have high susceptibility to infection and transmission potential (Power and Mitchell 2004).
64 Similarly, competitors can “dilute” (i.e. decrease) infection prevalence in a host species if they
65 have low competence. With low enough competence, competitors can even create “friendly
66 competition”, where they increase the density of the host species by lowering infection
67 likelihood, despite competing for resources (Hall et al. 2009). Ultimately, understanding how
68 competitors alter infection likelihood of individual host species will allow us to predict the
69 viability of host populations and the risk of spillover to other host species (Luis et al. 2018).
70 However, non-linear interactions between density dependent disease processes often make it
71 difficult to predict how one host species will impact infection likelihood in heterospecific host
72 species (Searle et al. 2016).

73 When parasite transmission requires infectious propagules to move through the
74 environment (environmentally transmitted parasites, Box 1), competing host species alter the
75 likelihood of infection by changing the density of parasite propagules within the environment,
76 and thus the dose of propagules that each host encounters. Many virulent parasites transmit via
77 the environment, including water borne parasites such as cholera and schistosomiasis, and orally
78 transmitted parasites such as tapeworms (Wardle and Mcleod 1952, Reidl and Kloese 2002,
79 Steinmann et al. 2006). Host species that both compete for resources and become infected by the
80 same pathogen influence the spread of environmentally transmitted parasites in three ways. First,
81 infected individuals excrete parasite propagules into the environment (Wardle and Mcleod 1952),
82 but host species differ in the number of propagules they shed. Second, hosts (and non-host
83 organisms) remove parasite propagules from the environment upon infection, and possibly by
84 consuming them (Burge et al. 2016). Third, competing host species can alter one another's
85 density via interspecific competition, changing the number of individuals available to transmit
86 and remove propagules (Strauss et al. 2015). Altogether, this means that competing host species
87 determine the dose of parasite propagules that each individual contacts, and thus the likelihood
88 of infection for each host species.

89 The likelihood of infection, however, often changes nonlinearly with propagule dose
90 (Figure 1A, B). As propagule dose increases, the infectivity of each parasite propagule can
91 decrease, leading to a decelerating (antagonistic) dose-infectivity relationship. Alternatively, as

92 propagule dose increases, the infectivity of each parasite propagule can increase, leading to an
93 accelerating (synergistic) dose-infectivity relationship (Regoes et al., 2003). Further, as
94 propagule dose increases, infected host mortality and propagule excretion from infected
95 individuals may change (Ashworth et al. 1996, Dallas and Drake 2014) (Figure 1C, D).
96 Together, these “dose relationships” (dose-infectivity, dose-mortality, and dose-excretion
97 relationships) make parasite transmission a function of environmental propagule density, which
98 is in turn a function of parasite transmission. This feedback loop may create challenges for
99 predicting how competing host species will influence infection likelihood. To date, however,
100 mechanistic models of multi-host systems typically do not incorporate dose-dependent feedback
101 loops (Bowers and Begon 1991, Begon and Bowers 1994, Greenman and Hudson 2000, Cáceres
102 et al. 2014, Strauss et al. 2015, Searle et al. 2016). Further, while some studies suggest that
103 accelerating dose-infectivity relationships are common (Regoes et al., 2002), we lack a
104 quantitative review of how common accelerating and decelerating dose-infectivity relationships
105 are. By exploring the frequency of different types of dose relationships, and the impact they have
106 on multi-host systems, we may be better able to predict the impact of heterospecific host
107 interactions on infection likelihood in individual host species.

108 Thus, we sought to answer several basic questions: First, are accelerating, linear, or
109 decelerating dose-infectivity relationships more common in published studies? To answer this
110 question, we conducted a meta-analysis of experimental dose-infectivity experiments and found
111 that parasites usually exhibit decelerating dose-infectivity relationships. Second, we asked
112 whether the impact of competing host species with varying infection competencies on disease in
113 a focal host would depend on the relationship (1) between dose and the infectivity of parasite
114 propagules (dose-infectivity relationships), (2) between dose and host excretion rates of parasite
115 propagules (dose-excretion relationships), or (3) between dose and the mortality rate of infected
116 individuals (dose-mortality relationships). Using 2-host 1-parasite models that incorporate the
117 types of dose relationships found in empirical studies, we examined how the effects of
118 interspecific host density on infection prevalence in a focal host were mediated by dose-
119 infectivity, dose-mortality, and dose-excretion relationships. We found dose relationships can
120 increase, decrease, or even reverse the impact of heterospecific host density on infection
121 prevalence. These results indicate that dose-dependency is common in host-parasite interactions,
122 and that disease models that do not take these dose relationships into account may result in

123 inaccurate predictions of disease dynamics in dose-dependent systems.

124 **Meta-Analysis Methods**

125 *Literature Review*

126 To find empirical dose-infectivity relationships, we conducted a literature search in
127 Google Scholar using the terms “parasite dose”, “pathogen dose”, “propagule dose”, “bacterial
128 dose”, “viral dose”, “dose-response relationship” AND “parasite” or “pathogen”, or “ID50”
129 AND “prevalence”. This search led to underrepresentation of marine systems compared to
130 terrestrial and freshwater systems, so we additionally searched for “dose” combined with well-
131 studied marine parasites. We accepted experimental studies that (a) exposed individual hosts to
132 varying parasite propagule doses/densities, (b) reported the proportion of hosts infected for each
133 propagule dose/density, and (c) found variation in the proportion of hosts infected across
134 propagule doses/densities. Our literature review included host-parasite systems across a variety
135 of habitats, host taxa, and parasite taxa (Table 1). Many experiments exposed hosts to a variety
136 of propagule densities, but were not able to measure contact rate, and thus dose. In these cases,
137 we assumed that dose scaled linearly with propagule density, though this is not always true
138 (Strauss et al. 2019). To avoid biases from model organisms, we only accepted one experiment
139 per combination of host species and parasite species, choosing the experiment with the most dose
140 treatments. We did not include experiments performed on incarcerated people due to ethical
141 concerns. For each host-parasite pair, we recorded the parasite dose used in each treatment, the
142 number of individuals per treatment, the number of individuals successfully infected in each
143 treatment, and the duration of time that individuals were exposed to parasites. Where raw data
144 was not available, we extracted the number of infected individuals from published figures.
145 Finally, we recorded whether dose altered any other aspects of infection, such as host mortality
146 or the number of parasite propagules released from each individual.

147 *Meta-Analysis*

148 We conducted an analysis to determine whether dose-infectivity relationships were
149 linear, decelerating, or accelerating. For linear dose-infectivity relationships, dose does not
150 change per propagule infectivity, and dose changes infection rate in a linear manner. Under
151 decelerating dose-infectivity relationships, the infectivity of individual parasite propagules
152 decreases with increased propagule dose. Thus, as dose increases, propagule infectivity
153 decreases, and the infection rate increases in a concave-down manner. This does not necessarily

154 mean that parasites mechanistically interfere with one another. Rather, this pattern could be the
155 result of non-linear immune responses in an individual as dose increases. Finally, under
156 accelerating dose-infectivity relationships, the infectivity of individual parasite propagules
157 increases with increased propagule dose. Thus, as dose increases, propagule infectivity increases,
158 and the infection rate increases in a concave-up manner. Accelerating dose-infectivity
159 relationships can be created if a high parasite dose is required to overwhelm host defenses.

160 To determine whether the dose-infectivity relationships in our literature review were
161 better represented by accelerating, decelerating, or linear relationships, we derived an equation
162 that described the proportion of individuals infected for a given dose of parasites. We model an
163 experiment where N individuals are exposed to parasite propagules at density P . The dose that
164 individuals consume is fP , where f is the parasite contact rate. Parasites are removed from the
165 experiment when they contact individuals, at a rate fPN . We assume that the length of the
166 experiment is sufficiently short such that total host density is constant, infected individuals do
167 not recover from infection, and infected individuals do not release new parasite propagules into
168 the environment. In the model, the changes in susceptible host density (S), infected host density (I), and P are

$$\frac{dS}{dt} = -\beta(fP)^k S \quad \text{eq. 1A}$$

$$\frac{dI}{dt} = \beta(fP)^k S \quad \text{eq. 1B}$$

$$\frac{dP}{dt} = -fNP \quad \text{eq. 1C}$$

173
174 where β is per-propagule infectivity, k is the dose shape parameter, and $\beta(fP)^k$ is the
175 host infection rate. For a given study, if $k = 1$ then the infection rate increases linearly with dose,
176 if $k < 1$ then the infection rate has a decelerating increase with dose, and if $k > 1$ then the
177 infection rate has an accelerating increase with dose (Figure 1A).

178 We used Bayesian inference to fit equation 1A-1C to the published data from our
179 literature review. For each study, we numerically ran our system of ODEs for the experimental
180 run time. We then estimated the values of β , k , and f most likely to generate the infection
181 prevalence reported in the studies for each dose treatment. We used vaguely informative priors to
182 prevent β and k from going below 0. If parasite dose was instantaneous (e.g. injections), we

183 assumed that hosts contact all parasites instantaneously (see Appendix S1 for details). In cases
184 where parasite densities were reported as dilutions, we relativized all parasite densities so that
185 the lowest parasite density was 100/volume. This ensured that the parasite density in the
186 experiment was never less than 1. We did not let fP fall below 1, as individuals cannot contact
187 partial propagules. As our main variable of interest was k , we additionally tested whether the
188 posterior estimate for k depended on β and f . While artificially lowering β increased our
189 estimate of k to compensate for the reduced infection rate, and vice-versa, our posterior estimate
190 of k did not depend on f (Appendix S1).

191 We further tested whether experiments in our meta-analysis best fit a sigmoidal dose-
192 infectivity relationship, where per-propagule infectivity first increases with dose, and then
193 decreases. This would match a pattern where a minimal infective dose is necessary to overcome
194 an individual's immune system and establish an infection, but further increases in parasite dose
195 yield diminishing returns and decrease per propagule infectivity. We thus reran our analysis
196 replacing the k in eq. 1A-1C with

$$k = \max(k_0 - fP * k_1, 0) \quad \text{eq. 2}$$

197 Such that k decreased with dose (fP), though never becomes negative. Using Bayesian
198 inference, we then estimate values of β , k_0 , k_1 , and f for each experiment. This formulation has
199 the benefit that if k_1 is high enough, our model creates a humped relationship between dose (fP)
200 and the infection rate ($\beta(fP)^{\max(k_0 - fP * k_1, 0)}S$), a pattern observed in some dose-infectivity
201 experiments (Strauss et al. 2019). We considered a sigmoidal dose-infectivity relationship to best
202 fit an experiment if the model DIC was lower than that for our constant k model, and if the 95%
203 confidence interval of k fell above 1 for low dose and fell below 1 for higher experimental dose.

204 In addition to infection prevalence, studies in our meta-analysis sometimes reported
205 changes in mortality or propagule excretion from infected hosts with propagule dose. However,
206 studies were inconsistent in the metrics they used to measure mortality and parasite load (e.g.
207 mortality could be measured as proportion of dead individuals, time until death, or visible
208 damage to individuals). We noted general trends but did not analyze the dose relationships of
209 these metrics, as the metrics used were too variable.

211 **Meta-analysis Results**

212 We found that the majority of published dose-infectivity relationships are decelerating ($k < 1$),
213 where increasing propagule dose lowers per-propagule infectivity (Figure 2). The 95%

214 confidence intervals of k values fell below 1 for 79/98 host-parasite combinations (decelerating),
 215 overlapped 1 for 12/98 host-parasite combinations (linear), and fell above 1 for 7/98 host-
 216 parasite combinations (accelerating). We found no support for sigmoidal dose-infectivity
 217 relationships. While ΔDIC values gave strong support for our non-constant k compared to our
 218 constant k model in 12 out of 98 studies ($\Delta\text{DIC} > 10$) and weak support in 3 out of 98 studies (10
 219 $> \Delta\text{DIC} > 5$), in 0 studies out of 98 did the 95% confidence interval of k fall above 1 for low
 220 propagule densities and fall below 1 for higher experimental propagule densities.

221 Model Exploration of Dose Relationships: Methods

222 To understand how dose-response relationships alter the impact of heterospecific host
 223 density on infection prevalence, we first built a 2-host, 1-parasite model with either linear,
 224 accelerating, or decelerating dose-infectivity relationships. Our model contains 2 host species,
 225 N_1 and N_2 , made up of susceptible classes S_1 and S_2 , and infected classes I_1 and I_2 . Growth of
 226 the susceptible classes are parameterized by their intrinsic growth rates, r_i , intra-specific
 227 competition coefficients, α_{ii} , and inter-specific competition coefficients, α_{ij} . Individuals move
 228 from S_i to I_i as a function of parasite propagules in the environment at density P , contact rate f_i
 229 and per-propagule infectivity, β_i . Propagule dose is calculated as $f_i P$, and is raised to the dose
 230 shape parameter, k_i . We treat k_i as a constant based on the results of our meta-analysis. Infected
 231 individuals then die at a rate m_i . All infected individuals excrete parasite propagules into the
 232 environment at a rate x_i . Propagules then leave the environment as a function of their
 233 degradation rate, μ , and via contact with hosts. The full model (Figure 3) is thus:

$$234 \quad \frac{dS_1}{dt} = \overbrace{N_1[r_1 - \alpha_{11}N_1 - \alpha_{12}N_2]}^{\text{growth and competition}} - \overbrace{\beta_1(f_1P)^{k_1}S_1}^{\text{infection}} \quad \text{eq. 3}$$

$$235 \quad \frac{dI_1}{dt} = \overbrace{\beta_1(f_1P)^{k_1}S_1}^{\text{infection}} - \overbrace{m_1I_1}^{\text{mortality}} \quad \text{eq. 4}$$

$$236 \quad \frac{dS_2}{dt} = \overbrace{N_2[r_2 - \alpha_{22}N_2 - \alpha_{21}N_1]}^{\text{growth and competition}} - \overbrace{\beta_2(f_2P)^{k_2}S_2}^{\text{infection}} \quad \text{eq. 5}$$

$$237 \quad \frac{dI_2}{dt} = \overbrace{\beta_2(f_2P)^{k_2}S_2}^{\text{infection}} - \overbrace{m_2I_2}^{\text{mortality}} \quad \text{eq. 6}$$

$$238 \quad \frac{dP}{dt} = \overbrace{\frac{Propagule\ release}{x_1I_1 + x_2I_2}} - \overbrace{\frac{degradation}{\mu P}} - \overbrace{\frac{Host\ Contact}{f_1(N_1)P - f_2(N_2)P}} \quad \text{eq. 7}$$

239 For all the analyses we present in the main text, we assume the focal host species and

240 competing host species have identical parameter values except for their population growth rates
241 $s(r_i)$ and propagule excretion rates (x_i) (see Appendix S2 for all parameters). However, we
242 repeated the analyses for scenarios where the two host species have unequal competitive abilities
243 ($\alpha_{12} \neq \alpha_{21}$), susceptibility to infection ($\beta_1 \neq \beta_2$), and shape parameters ($k_1 \neq k_2$). Our results are
244 qualitatively the same in all scenarios; see Appendix S3 for details.

245 *Testing the Impact of Heterospecific Host Density on Infection Prevalence*

246 We use our model to test whether increasing the density of a “competitor” host species,
247 N_2 , will increase or decrease (1) the infection prevalence in our “focal” host species, $\frac{I_1}{N_1}$, (2) the
248 parasite propagule density in the environment, P , and (3) the density of the focal host species, N_1
249 . Biological reasons for considering these three variables are the following. Responses in the
250 propagule density in the environment allow us to measure the effect of competitor density on
251 likelihood of infection in the focal host as well as get a general sense for how competitor density
252 will alter spillover risk for other unmodelled hosts. The likelihood of spillover will likely scale
253 positively, though not linearly, with propagule dose, and is relevant for spillover of infection
254 from multi-host communities into human or agricultural systems. Responses in infection
255 prevalence will allow us to relate our model to disease indexes observed by field ecologists; we
256 say a competitor host species dilutes or amplifies disease in the focal host when infection
257 prevalence in the focal host is lower or higher, respectively, in the presence of the competitor
258 host species. Responses in the focal host density (N_1) allow us to measure the total effect of
259 competitor density on focal host viability. To increase the density of the competitor, we increase
260 its intrinsic growth rate, r_2 , from 0 to $2r_1$.

261 *Dose-Infectivity Relationships*

262 For all analyses, we measure the impact of competing host density on model dynamics
263 under three dose-infectivity relationships: when $k_1 = 0.5$ (decelerating dose-infectivity
264 relationship), when $k_1 = 1.0$ (linear dose-infectivity relationship), and when $k_1 = 1.5$
265 (accelerating dose-infectivity relationship, Figure 1A). For our main results, we assume that k_1
266 $= k_2$, but we explore asymmetric dose-infectivity relationships in Appendix S3.

267 In our model, k alters both the shape of dose-infectivity relationships, and the magnitude
268 of parasite transmission. As k increases, the infection rate, $\beta_i(f_i P)^{k_i} S_i$, increases in an
269 exponential manner, thus increasing infection likelihood. Thus, to solely examine how the shape

270 of dose-response relationships alters infection likelihood, we vary β as we vary k such that
 271 disease prevalence in the focal host in the absence of the competing host is always 0.5 at
 272 equilibrium. If we vary k in our model without altering β , then increasing k always increases
 273 parasite transmission. The full relationship between k and β is

$$274 \quad \beta_i = \frac{m_i}{\left(\frac{x_i \frac{2r_i - m_i}{4\alpha_{ii}}}{f_i \frac{2r_i - m_i}{\mu + f_i \frac{2r_i - m_i}{2\alpha_{ii}}}} \right)^{k_i}} \quad eq. 8$$

275 (See Appendix S2 for full derivation.) This ensures that varying the dose shape parameter k does
 276 not affect the equilibrium level of disease in the focal host when the second host is absent.
 277 Whether a competitor increases disease in a focal host often depends on the ability of the
 278 competitor to become infected and excrete parasite propagules (i.e., host competency). Thus, we
 279 ran our model while varying competitor excretion rates. We additionally ran a scenario where the
 280 focal host cannot maintain parasite transmission, and the infection prevalence in the absence of
 281 the competing host is 0 (Appendix S3).

282 *Dose-Excretion and Dose-Mortality Relationships*

283 In our meta-analysis, we found four additional effects of propagule dose across multiple
 284 host-parasite combinations. As propagule dose increased (1) propagule excretion could decrease,
 285 (2) propagule excretion could increase, (3) infected host mortality rate could increase, and (4)
 286 propagule excretion and host mortality could concurrently increase. (In some cases, we
 287 interpreted higher parasite load within-hosts as higher propagule excretion.) Thus, we ran our
 288 model under these four scenarios concurrently with decelerating, linear, and accelerating dose-
 289 infectivity relationships.

290 To model changes in the excretion rate with increasing dose, we replace propagule
 291 excretion rate, x_i , with dose-dependent propagule excretion rate, $x_{i,dose}$, given by

$$292 \quad x_{i,dose} = x_i \left(\frac{1}{2} + \frac{f_i P}{2f_i P_1} \right)^\gamma \quad eq. 9$$

293 where $f_i P_1$ is the propagule dose at equilibrium when $N_2 = 0$ using equations 3-7. We use this
 294 parameterization because it guarantees that the excretion rate of host i is equal to x_i when at

295 equilibrium in the absence of the competing host. This simplifies our analysis because it means
296 the dose-excretion relationship only affects prevalence in host i when the competing host is
297 present. Models without dose-excretion relationships are equal to models with dose-excretion
298 relationships if $\gamma = 0$. In addition to models without dose-excretion relationships, we explore
299 dose-excretion models where $\gamma = -3$ (exponential decrease in excretion with dose) and $\gamma = 0.5$
300 (decelerating increase in excretion with dose, Figure 1B).

301 To increase infected host mortality with dose, we replaced infected host mortality, m_i ,
302 with a dose dependent mortality, $m_{i,dose}$, given as

$$303 \quad m_{i,dose} = m_{min} + (m_i - m_{min}) \left(\frac{f_i P}{f_i P_1} \right)^\rho \quad eq. 10$$

304 where $f_i P_1$ is once again the propagule dose at equilibrium when $N_2 = 0$ using equations 3-7,
305 and m_{min} is the minimum mortality of infected individuals. Thus, the mortality rate of host i is
306 equal to m_i when at equilibrium in the absence of the competing host, and so dose-mortality
307 relationships do not alter infection prevalence in the absence of the competing host. In our
308 model, host mortality is independent of dose for $\rho = 0$, increasing at a decelerating rate with
309 dose for $\rho = 0.5$, increasing linearly with dose for $\rho = 1$, and increasing at an accelerating rate
310 with dose for $\rho = 1.5$ (Figure 1C).

311 **Model Exploration of Dose Relationships: Results**

312 Confirming previous models (Cáceres et al. 2014), infection prevalence in the focal host
313 is influenced by both the density of the competing host and the rate at which it releases parasite
314 propagules when infected (Figure 4B). Analytical solutions to our model show that increases in
315 competitor density increase focal host infection prevalence and propagule density (i.e. amplify
316 disease) when the competitor is a larger source of parasite propagules, and lower focal host
317 infection prevalence and propagule density (i.e. dilute disease) when the competitor is a smaller
318 source of parasite propagules than the focal host (Appendix S4: Section S1). A host is a large
319 “source” of propagules if it has a high propagule excretion rate, and/or if it removes few
320 propagules from the environment. Our numerical simulations match this result: increases in
321 competitor density decrease disease prevalence in the focal host when competitor propagule
322 excretion is lower than the focal host (Competitor Excretion < 100, light blue lines in Figure 4B),
323 and increase disease prevalence in the focal host when competitor propagule excretion is higher

324 than the focal host (Competitor Excretion > 100 , light blue lines in Figure 4B). Thus, our model
325 confirms pre-existing multi-host theory in the absence of dose-relationships.

326 Dose-Infectivity Relationships

327 Accelerating dose-infectivity relationships increase the strength of dilution/amplification,
328 while decelerating dose-infectivity relationships decrease the strength of dilution/amplification.
329 Analytical solutions to our model show that the absolute value of the relationship between
330 competitor density and infection prevalence increases as k increases. This means that, for
331 accelerating dose-infectivity relationships (high k), as competitor density increases, there is a
332 large change in infection prevalence; for decelerating dose-infectivity relationships (low k), there
333 is a smaller change in infection prevalence (Appendix S4: Section S1). These analytical results
334 are matched by our numerical results, which also show that decelerating dose-infectivity
335 relationships lead to a smaller change in infection prevalence due to competitor density than
336 accelerating dose-infectivity relationships (Figure 4B). We find that, qualitatively, changes in
337 prevalence match changes in environmental propagule density (Figure 4E).

338 Accelerating and decelerating dose-infectivity relationships alter the impact of competitor
339 density on infection prevalence and propagule density by creating feedback loops between
340 propagule dose and per-propagule infectivity. Decelerating dose-infectivity relationships create
341 negative feedback loops. If a competing host releases fewer parasite propagules than the focal
342 host, this lowers propagule density in the environment, which lowers propagule dose. Lowering
343 propagule dose increases per-propagule infectivity, thus buffering the impact of competing host
344 density on infection prevalence, and in turn propagule density/dose. On the other hand,
345 accelerating dose-infectivity relationships create positive feedback loops. If a competing host
346 releases fewer parasite propagules than the focal host, this lowers propagule density in the
347 environment, which lowers propagule dose. Lowering propagule dose decreases per-propagule
348 infectivity, thus accelerating the impact of competing host density on infection prevalence, and
349 in turn propagule density/dose. (The converse can also happen if competing hosts increase
350 parasite dose.) Thus, infection prevalence is generally more sensitive to changes in competitor
351 density under accelerating dose-infectivity relationships than under decelerating dose-infectivity
352 relationships.

353 Dose-Excretion Relationships

354 Our literature survey showed that propagule excretion from infected hosts can increase or

355 decrease with propagule dose (Data S1). Increasing dose may decrease propagule excretion if
356 parasites face within-host competition, where initial crowding may limit the production of
357 parasite propagules. On the other hand, increasing propagule dose may increase propagule
358 excretion if high doses overwhelm the host's immune system.

359 Under decreasing dose-excretion relationships, increases in competing host density have
360 less of an impact on focal host infection prevalence (Figure 4A vs. 4B); this occurs because of
361 negative feedback loops. Under these negative feedback loops, increasing propagule dose
362 decreases propagule excretion, which in turn decreases propagule dose. Similarly, decreasing
363 propagule dose increases propagule excretion, which in turn increases propagule dose. This
364 creates smaller changes in prevalence as competing host density increases, compared to a
365 scenario with fixed excretion.

366 Conversely, under increasing dose-excretion relationships, this creates positive feedback
367 loops: increasing propagule dose increases propagule excretion, which in turn increases
368 propagule dose. Similarly, decreasing propagule dose decreases propagule excretion, which in
369 turn decreases propagule dose. This positive feedback loop increases the impact of competitor
370 density on infection prevalence (Figure 4C vs. 4B). Because positive feedback loops destabilize
371 systems, adding both a positive dose-excretion relationship and a positive dose-infectivity
372 relationship to our system causes the system to shift from 0% infection prevalence to 100%
373 infection prevalence with small changes to system parameters (Figure 4C). Our analytical
374 solutions support these results (see Appendix S4: Section S2). We again find that, qualitatively,
375 changes in prevalence match changes to environmental propagule density (Figure 4).

376 Dose-Mortality Relationships

377 In some host-parasite combinations, increasing propagule dose increases infected host
378 mortality (dose-mortality relationship). This could occur if parasites damage the host upon
379 contact. Alternatively, if hosts die when parasites reach a certain density within the host,
380 increasing propagule dose could decrease the amount of time it takes for parasites to reach that
381 density, thus decreasing time until host death.

382 Dose-mortality relationships represent negative feedback loops. As dose increases, the
383 infectious period of infected hosts shrinks due to increased mortality, lowering transmission and
384 thus dose. As dose decreases, the infectious period of infected hosts increases due to reduced
385 mortality, lowering transmission and thus dose. As with negative feedback loops created by

386 decelerating dose-infectivity and negative dose-excretion relationships, the negative feedback
387 loops created by dose-mortality relationships decrease the ability of competitor hosts to influence
388 infection likelihood. We see this reflected in environmental propagule density; low-competence
389 competitor hosts lower environmental propagule density less under dose-mortality relationships,
390 and competent competitor hosts raise propagule density less (Figure 5D-F vs. 4E).

391 However, dose-mortality relationships can reverse the impact that competitors have on
392 infection prevalence. This is because increasing propagule dose both increases infection
393 prevalence by increasing the rate at which susceptible individuals become infected ($\beta_i(f_iP)^{k_i}$),
394 and additionally decreases infection prevalence by increasing the mortality rate of infected hosts
395 ($m_i\left(\frac{f_iP}{f_iP_1}\right)^\rho$). The combined effects of dose-dependent mortality and infection rate depend on the
396 values of the shape parameters k_i and ρ . If infection rate changes with parasite dose faster than
397 mortality ($\rho < k_i$), increasing competitor density will increase infection prevalence when the
398 competitor is a large source of propagules, as expected, and vice versa (Figure 5A-C). In
399 contrast, if mortality changes with parasite dose faster than infection rate changes with parasite
400 dose ($\rho > k_i$), then we see a reverse in whether competitor density increases or decreases
401 infection prevalence — increasing the density of competitors that are large sources of parasite
402 propagules decreases infection prevalence and increasing the density of competitors that are
403 small sources of propagules increases infection prevalence (Figure 5A-C). This pattern occurs
404 because if $\rho > k$, then mortality increases with dose faster than infectivity. When $\rho \cong k$, changes
405 in mortality and infectivity approximately cancel each other out as dose changes, so competitor
406 density will have little effect on infection prevalence (Figure 5A-C, see Appendix S4: Section S3
407 for full analysis). Combining positive dose-excretion relationships with dose-mortality
408 relationships does not qualitatively change the impact of either dose-relationship on prevalence
409 and propagule patterns (Appendix S3).

410 Friendly Competition

411 Confirming previous theory, in the absence of dose-relationships competitors with weak
412 inter-specific competition and low competence increase the density of the focal host (i.e. friendly
413 competition), while competitors with strong inter-specific competition and high competence
414 decrease the density of the focal host (Figure 6B). Note that in our model, if the effect of inter-
415 specific competition on the focal host is greater than zero, increasing competitor density will

416 always eventually drive the focal host to extinction. Thus, “Friendly Competition” in our model
417 does not represent a monotonic positive effect of competing host density on focal host density,
418 but rather a humped relationship. In these circumstances, increasing competitor density initially
419 increases focal host density by decreasing the infection rate. However, as competitor density
420 increases, the negative effect of direct competition on focal host density eventually outweighs the
421 positive effects of the removal of infectious propagules.

422 Positive feedback loops facilitate friendly competition. Our model shows that dose-
423 relationships that create positive feedback loops (accelerating dose-infectivity relationships,
424 positive dose-excretion relationships) increase the parameter space where competing hosts can
425 increase focal host density (Figure 6, green vs. light blue in all panels, and B,E,H,K vs C,F,I,L).
426 Alternatively, dose-relationships that create negative feedback loops (decelerating dose-
427 infectivity relationships, all dose-mortality relationships, negative dose-excretion relationships)
428 decrease the parameter space where competing hosts can increase focal host density (Figure 6,
429 dark blue vs. light blue in all panels, A-C vs. D-L, and B,E,H,K vs. A,D,G,J). This is because
430 friendly competition occurs when competing hosts strongly dilute disease. As we see in Figure 4,
431 dose-relationships that create positive feedback loops increase the strength of dilution.

432 Discussion

433 Parasite dose underlies every aspect of infectious disease transmission, and can transform
434 interactions between hosts who share parasites. Our study shows that the effect of parasite dose
435 on per-propagule infectivity, host mortality, and propagule excretion can strengthen, weaken, or
436 even reverse the impact of heterospecific host density on disease in a focal host. Our meta-
437 analysis indicates that most dose-infectivity relationships are decelerating (Figure 2), and thus
438 may decrease the impact of heterospecific host density on infection prevalence and infectious
439 propagule density via negative feedback loops (Figure 4). Dose-excretion relationships can
440 create positive or negative feedback loops, increasing or decreasing the impact of heterospecific
441 hosts on infection prevalence and propagule density (Figure 4). Further, dose-mortality
442 relationships can make the impact of heterospecific hosts on infection prevalence negatively
443 correlated with the effects on propagule density (Figure 5). Finally, our results show that positive
444 feedback loops created by accelerating dose-infectivity relationships and positive dose-
445 infectivity relationships can facilitate friendly competition, even in the face of high interspecific
446 competition. Together, these results suggest that dose relationships could fundamentally alter

447 how interspecific host interactions influence disease dynamics, and that models that ignore dose
448 relationships may mislead us in our efforts to understand and predict how changes in host
449 communities will alter disease patterns.

450 *Dose-response feedback loops*

451 Dose-response relationships create feedback loops that can increase or decrease the
452 extent that competing hosts alter disease prevalence, parasite propagule density, and density of
453 focal hosts (Table 2). The transmission of a parasite within an ecosystem increases with (1)
454 parasite dose, (2) the probability that each parasite in that dose will infect a host, (3) the rate of
455 propagule excretion from hosts once they are infected, and (4) the lifespan of those infected
456 hosts. If increasing dose increases any of these factors, then propagule dose and parasite
457 transmission enter a positive feedback loop. If increasing dose decreases any of these factors,
458 then propagule dose and parasite transmission enter a negative feedback loop (feedback loops in
459 Figure 3). Ultimately, through these feedback loops, dose-response relationships can strengthen,
460 weaken, or reverse predictions for whether a host will amplify or dilute disease based purely on
461 their competence.

462 *Dose-infectivity relationships*

463 Most dose-infectivity relationships in our meta-analysis decelerate (Figure 2). Previously,
464 the vast majority of dose-response experiments showed that infection probability increases in a
465 sigmoidal pattern with $\log(\text{dose})$ (Smith et al. 1997, Regoes et al. 2003). However, this pattern
466 can be created by accelerating, linear, or decelerating dose-infectivity relationships (Figure 1B).
467 In fact, the null assumption for most studies has been that parasite propagules behave
468 independently of one another, creating a linear dose-infectivity relationship (Zwart et al. 2009).
469 Our analysis suggests decelerating dose-infectivity relationships are what we expect to see in
470 most systems.

471 As dose increases, the per-propagule probability of infection decreases under decelerating
472 dose-infectivity relationships. This creates a negative feedback loop between dose and the
473 infection rate that should weaken the ability of competing hosts to increase or decrease disease,
474 and should weaken the ability of hosts to increase one another's density via dilution in the face
475 of interspecific competition (Figure 4,5,6). This information can help us interpret experiments.
476 For example, in our meta-analysis we found decelerating dose-infectivity relationships for
477 *Daphnia dentifera* infected by *Metschnikowia bicuspidata* (Dallas and Drake 2014), a model

478 system for the dilution/amplification effect in two-host experiments (Hall et al. 2009, Strauss et
479 al. 2015, Searle et al. 2016). Mechanistic models of this system have thus far assumed mass-
480 action infection processes and would most likely be improved by implementing decelerating
481 dose-infectivity relationships. Further, if dose-infectivity relationships are usually decelerating,
482 then changes to parasite dose due to competing hosts will have the largest impact on infection
483 rate, and thus infection prevalence, at low doses (Figure 1A). Knowing this will help us identify
484 natural systems where host community composition will likely alter infection prevalence.

485 While our meta-analysis found that most experimental dose-infectivity relationships are
486 decelerating (Figure 2), many dose-infectivity relationships exhibit a minimal infective dose
487 (Ward and Akin 1984), a feature not possible under a purely decelerating dose-infectivity
488 relationships. A decelerating dose-infectivity relationship that nevertheless has a minimal
489 infective dose could fit a piecemeal function that is 0 below the minimal infective dose and
490 decelerates above the minimal infective dose, or a sigmoidal function where per-propagule
491 infectivity increases at low doses and decreases at higher doses. Mechanistically, a dose-
492 infectivity relationship that both accelerates or decelerates depending on propagule dose could be
493 possible because infection is determined by interactions between parasites and many host
494 defenses, and defenses such as the immune system may respond non-linearly to propagule dose
495 (Van Leeuwen et al. 2019, Stewart Merrill et al. 2019). We tested for this latter possibility, but
496 found no evidence for sigmoidal dose-infectivity relationships in our meta-analysis. Nonetheless,
497 our results explain how a sigmoidal dose-infectivity relationship would affect the relationship
498 between focal infection prevalence and competitor density or between parasite density and
499 competitor density: at low doses, changes in dose will create positive feedback loops, while at
500 high doses, changes in dose will create negative feedback loops.

501 *Dose-excretion relationships*

502 While dose-infectivity and dose-mortality relationships mostly cause negative feedback
503 loops, dose-excretion relationships can cause both positive and negative feedback loops, either
504 increasing or decreasing disease amplification and dilution. To cause a negative feedback loop,
505 parasite propagule excretion must decrease with dose. This could potentially occur if increasing
506 dose lowers the within host growth rate of the parasite (Regoes et al. 2002). Or in cases where
507 hosts only excrete parasites at host death, dose may decrease excretion rates if it simultaneously
508 decreases host lifespan, limiting the amount of time that parasites have to grow (Ebert et al.

509 2000). To cause a positive feedback loop, parasite propagule excretion must increase with dose.
510 This is most likely for macroparasites that do not reproduce in certain hosts, and thus excretion is
511 limited by parasite dose (Johnson et al. 2012). Ultimately, dose-excretion relationships might be
512 the most important dose-response relationship to measure in future experiments, as we do not
513 have strong prior assumptions about whether these relationships should be positive or negative.

514 *Dose-mortality relationships*

515 Increasing dose generally decreases infected host lifespan (Appendix S5). This creates a
516 negative feedback loop between dose and the infection rate which should weaken the ability of
517 competing hosts to dilute or amplify disease, and should prevent friendly competition (Figure
518 5,6). Further, we found that while infection prevalence is generally positively related with
519 propagule density, dose-mortality relationships can reverse this relationship (Figure 5).

520 Traditionally, we assume that competing hosts are more likely to decrease infection prevalence if
521 they remove many propagules from the environment, if they have a low transmission rate or
522 susceptibility, and if they are strong competitors (Cáceres et al. 2014, Strauss et al. 2015).

523 Competing hosts with these traits reduce disease because they lower environmental propagule
524 density, lowering dose and infection rate, and ultimately lowering infection prevalence.

525 However, dose-mortality relationships can make infection rate and infection prevalence
526 negatively correlated, and thus challenge our assumptions of which hosts should reduce infection
527 prevalence in a community. If host mortality increases at a faster rate with propagule dose than
528 infection rate does, then infection rate will be negatively correlated with prevalence — thus the
529 low competence, strongly competing hosts that might otherwise be expected to decrease disease
530 will actually increase disease prevalence over some range of densities. This scenario is

531 potentially common, as many systems display positive dose-mortality relationships (for instance,
532 Ashworth et al. 1996; Agnew and Koella 1997; Blair and Webster 2007; De Roode et al. 2007).

533 Further, it is when decelerating dose-infectivity relationships, which our meta-analysis shows to
534 be common (Figure 2), are combined with dose-mortality relationships that we see expected low-
535 competence hosts increase disease, and vice versa (Figure 5). Indeed, highly competent hosts
536 with positive dose-mortality relationships and decelerating dose-infectivity relationships have
537 been shown to dilute disease (Ebert et al. 2000, Dallas and Drake 2014, Searle et al. 2016).

538 Arguably, infection prevalence is only indirectly important, and what matters is that competent
539 hosts increase infection rates, and low-competence hosts decrease infection rates, regardless of

540 infection prevalence. However, infection prevalence is important in that we can readily measure
541 it, and thus use it as a proxy for infectious disease severity in ecosystems. Thus, infectious
542 disease ecologists should factor in dose-mortality relationships when trying to infer infection
543 processes from infection prevalence.

544 *Future directions*

545 Pairing multi-host empirical studies with mechanistic dose models will allow us to
546 uncover the mechanisms driving disease patterns in multi-host communities. Mechanistic models
547 paired with empirical data have generated valuable insights into the processes driving disease in
548 multi-host communities, such as when inter-host interactions are simultaneously amplifying and
549 diluting disease (Luis et al. 2018), or the relative contributions of competition and host
550 competency to disease dilution (Strauss et al. 2015). Pairing mechanistic dose models with
551 empirical data will allow us to answer many open questions about the real-world importance of
552 dose relationships, such as (a) do dose relationships often alter biodiversity-disease relationships
553 in natural populations? (b) Are decelerating dose-infectivity relationships truly common in
554 natural populations? And (c) do dose effects alter infection prevalence most strongly via
555 infectivity, host-mortality, or propagule excretion? Overall, an improved understanding of dose
556 response relationships will enable us to better understand the impact of host species interactions
557 on disease risk, and thus make more informed conservation and public health decisions.

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560 conceptualized this project. PAC conducted the literature review and meta-analysis, ran
561 numerical simulations, and wrote the manuscript. PAC and MHC created the model. MHC
562 conducted analytical solutions of the model. All authors contributed to the editing of manuscript.

563 **Supporting Information**

564 Additional supporting information may be found online at: [link to be added in production]

565 **Open Research**

566 Code (Clay et al. 2021) for simulations, meta-analysis, and meta-analysis data are available from
567 Dryad: <https://doi.org/10.5061/dryad.3tx95x6fz>

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Box 1: Defining Environmental Transmission

We categorize parasites as environmentally transmitted if they must travel through the environment when transmitting between hosts. We consider “the environment” to be any space that is not in or on a host or vector. In these systems, infected hosts release parasite propagules into the environment. Susceptible hosts come in contact with a *dose* of parasite propagules, based on the density of parasite propagules in the environment, and the rate at which hosts come in contact with those propagules (e.g. in the case of water borne pathogens, propagule dose will increase if propagule density in the water increases, or if the host drinks more water). Susceptible hosts then have some probability of becoming infected based on the dose of propagules they contact.

652

653

654 **Table 1.** Categories of host/parasite interactions included in literature review. We found 98
655 host/parasite combinations across 63 studies. We consider “Environmental” parasites to be
656 parasites where host contact is not required for transmission, and where parasites are not
657 transmitted via vectors. Parasites in the “Other” taxa category include cercozoan, myxozoan,
658 platyzoan, and trypanosome parasites.

Category	No. combinations
Environment	
Freshwater	14
Marine	13
Terrestrial	71
Transmission	
Direct	3
Environmental	86
Vector borne	9
Host taxa	
Ciliate	1

Human	9
Invertebrate	46
Plant	26
Non-human vertebrate	16
Parasite taxa	
Bacteria	13
Fungi	13
Nematode	3
Oomycete	25
Protist	7
Trematode	3
Virus	30
Other	4

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660

Table 2. Summary of model outcomes, compared to model with no linear dose-infectivity, static dose-excretion, and static dose-mortality relationships. Dose-relationships can increase or decrease the magnitude of the impact of heterospecific host density on infection prevalence in the focal host or propagule density or can reverse the trend entirely. Dose-relationships can also facilitate or prevent friendly competition. There are no qualitative synergies between dose-relationships, when dose has an impact on multiple aspects of transmission, so we only describe outcomes for individual dose-relationships.

Scenario	Infection prevalence	Propagule density	Friendly competition	Mechanism
Decelerating dose-infectivity relationship	decrease	decrease	prevent	negative feedbacks between dose and per-propagule infectivity
Accelerating dose-infectivity relationship	increase	increase	facilitate	positive feedbacks between dose and per-propagule infectivity
Negative dose-excretion relationship	decrease	decrease	prevent	negative feedbacks between dose and propagule excretion rate
Positive dose-excretion relationship	increase	increase	facilitate	positive feedbacks between dose and propagule excretion rate
Positive dose-mortality relationship ($\rho \leq k$)	decrease	decrease	prevent	negative feedbacks between dose and infected host lifespan; infected host mortality changes with dose slower than infection rate
Positive dose-mortality relationship ($\rho > k$)	reverse	decrease	prevent	negative feedbacks between dose and infected host lifespan; infected host mortality changes with dose faster than infection rate

661 **Figure 1:** Dose relationships can take a variety of forms. X-axis shows propagule dose, and Y-
662 axis shows (A) the infection rate (dose-infectivity relationship), (B) the proportion of individuals
663 becoming infected after exposure to that dose (dose-infectivity relationship, cont.), (C) the rate at
664 which parasite propagules are excreted from infectious individuals (dose-excretion relationship),
665 and (D) the mortality rate of infected individuals (dose-mortality relationship). The shape of each
666 dose relationship is described by a shape parameter (k for dose-infectivity relationships, eq. 1, γ
667 for dose-excretion relationships, eq. 9, and ρ for dose-mortality relationships, eq. 10). If k , ρ , or
668 γ is greater than 1, the dose relationship has an accelerating increase. If k , ρ , or γ is equal to 1,
669 the dose relationship has a linear increase. If k , ρ , or γ is between 1 and 0, the dose relationship
670 has a decelerating increase. If k , ρ , or γ is equal to 0, the dose relationship is static. If k , ρ , or γ
671 is less than 0, the dose relationship has an exponential decrease. Lines are shown for parameter
672 values included in model results, based on the literature review results.

673 **Figure 2:** Most empirical dose-infectivity relationships are decelerating. Values on the x-axis
674 show Bayesian estimates of the dose shape parameter (k) values from published dose-infectivity
675 relationships, with bars showing 95% confidence intervals of the posterior distribution. If an
676 interval overlaps the 1 line, then we do not reject the null hypothesis that infection rate increases
677 linearly with dose, which implies that dose does not alter per-propagule infectivity. If intervals
678 lie below one, then per-propagule infectivity decreases with dose, and dose-infectivity
679 relationships have a decelerating increase. If the interval lies above the 1 line, then per-propagule
680 infectivity increases with dose, and dose-infectivity relationships have an accelerating increase.

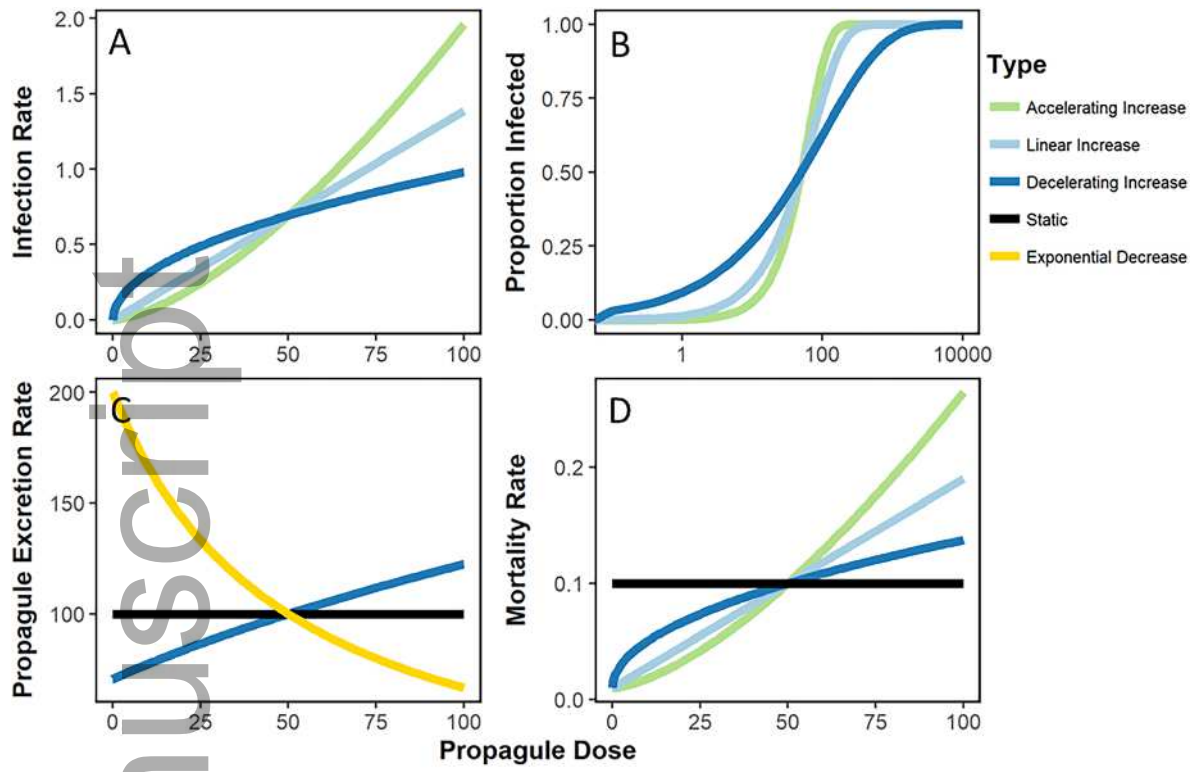
681 **Figure 3:** Schematic of equations 3-9. Black lines represent dose-independent processes and blue
682 lines represent dose-dependent processes. Dashed green lines connect environmental propagule
683 density to dose dependent processes to visualize feedback loops. S_1 and I_1 represent susceptible
684 and infected individuals of species 1, S_2 and I_2 represent susceptible and infected individuals of
685 species 2, and P represents environmentally transmitted parasite propagules. (a) All hosts give
686 birth as a function of intraspecific and interspecific density and competition (eq. 3, 5). (b)
687 Susceptible individuals become infected at a rate determined by parasite dose (eq. 4, 6). (c)
688 Infected individuals excrete parasite propagules into the environment as a function of dose (eq. 7,
689 8). (d) Propagules degrade over time (eq. 7). (e) Finally, infected individuals die as a function of
690 parasite dose (eq. 4, 6, 9).

691 **Figure 4:** Negative dose-excretion relationships or decelerating dose-infectivity relationships

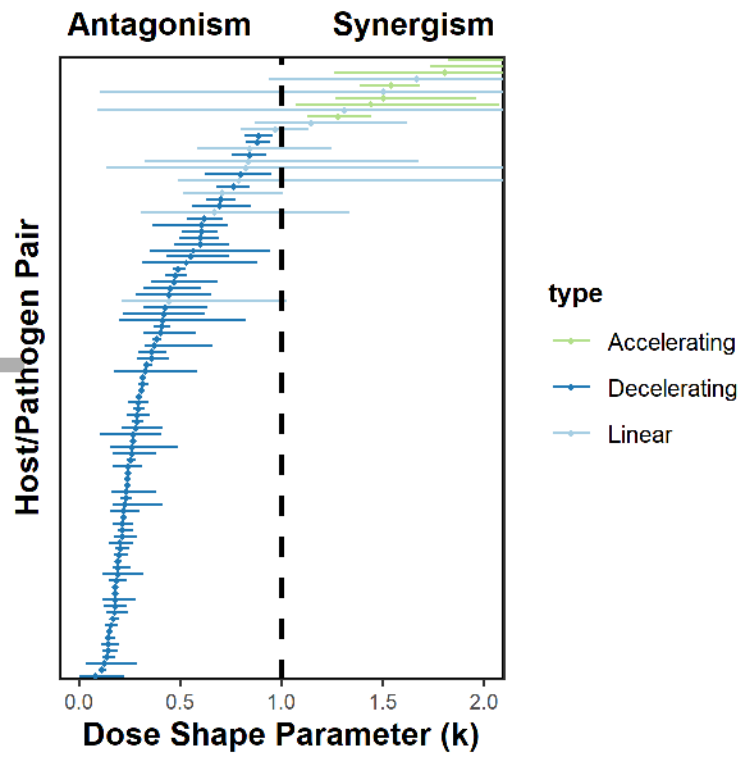
692 decrease (and positive dose-excretion relationships or accelerating dose-infectivity relationships
693 increase) the magnitude of the relationship between infection prevalence and competitor density
694 and between propagule density and competitor density. Changes in infection prevalence of the
695 focal host (Y-Axis A-C) and log propagule density (Y-Axis D-F) as competitor density increases
696 (X-axis). Panels represent models with negative dose-excretion relationships (A,D), no dose-
697 excretion relationship (B,E), or positive dose-excretion relationships (C,F). Solid lines represent
698 competitors with lower propagule excretion than the focal host species, while dashed lined
699 represent competitors with higher propagule excretion than the focal host species. Dark blue
700 lines show decelerating dose-infectivity relationships, light blue lines show linear dose-
701 infectivity relationships, and green lines show accelerating dose-infectivity relationships.

702 **Figure 5:** Decelerating dose-mortality relationships decrease (and accelerating dose-mortality
703 relationships increase) the magnitude of the relationship between infection prevalence and
704 competitor density and between propagule density and competitor density. Changes in infection
705 prevalence of the focal host (Y-Axis A-C) and log propagule density (Y-Axis D-F) as competitor
706 density increases (X-axis). Panels represent models with decelerating dose-mortality
707 relationships (A,D), Linear dose-mortality relationships (B,E), or Accelerating dose-mortality
708 relationships (C,F). Solid lines represent competitors with lower propagule excretion than the
709 focal host species, while dashed lined represent competitors with higher propagule excretion than
710 the focal host species. Dark blue lines show decelerating dose-infectivity relationships, light blue
711 lines show linear dose-infectivity relationships, and green lines show accelerating dose-
712 infectivity relationships.

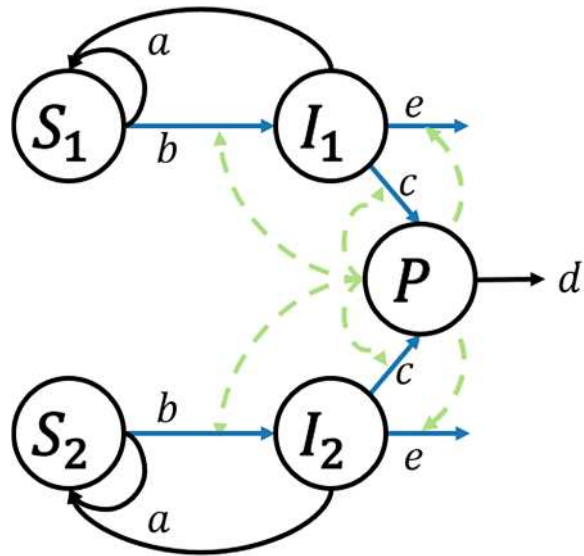
713 **Figure 6:** Positive dose-mediated feedbacks loops facilitate friendly competition. Regions of
714 parameter space show whether focal host density can increase with density of competing hosts
715 (friendly competition), with competitor propagule excretion rate (x_2) on the X-axis and
716 interspecific competition (α_{12} and α_{21}) on the Y-axis. Dark blue indicates friendly
717 competition for all dose-infectivity relationships, light blue indicates friendly competition if per-
718 propagule infectivity increases linearly or accelerates with dose, green indicates friendly
719 competition only if per-propagule infectivity accelerates with dose, and black indicates no
720 friendly competition. Panels indicate different dose-mortality relationships ($\rho=0$ for none, $\rho=0.5$
721 for decelerating, $\rho=1.0$ for linear, $\rho=1.5$ for accelerating) and different dose-excretion
722 relationships ($\gamma=-3$ for negative, $\gamma=0$ for none, $\gamma=0.5$ for positive).



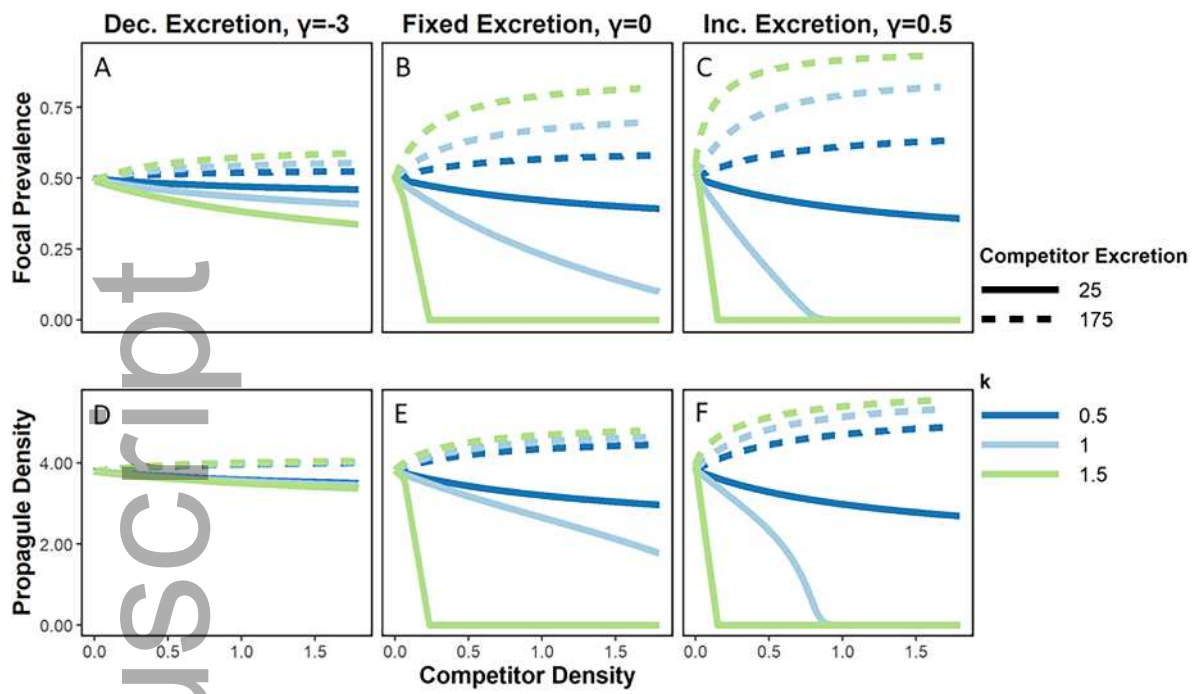
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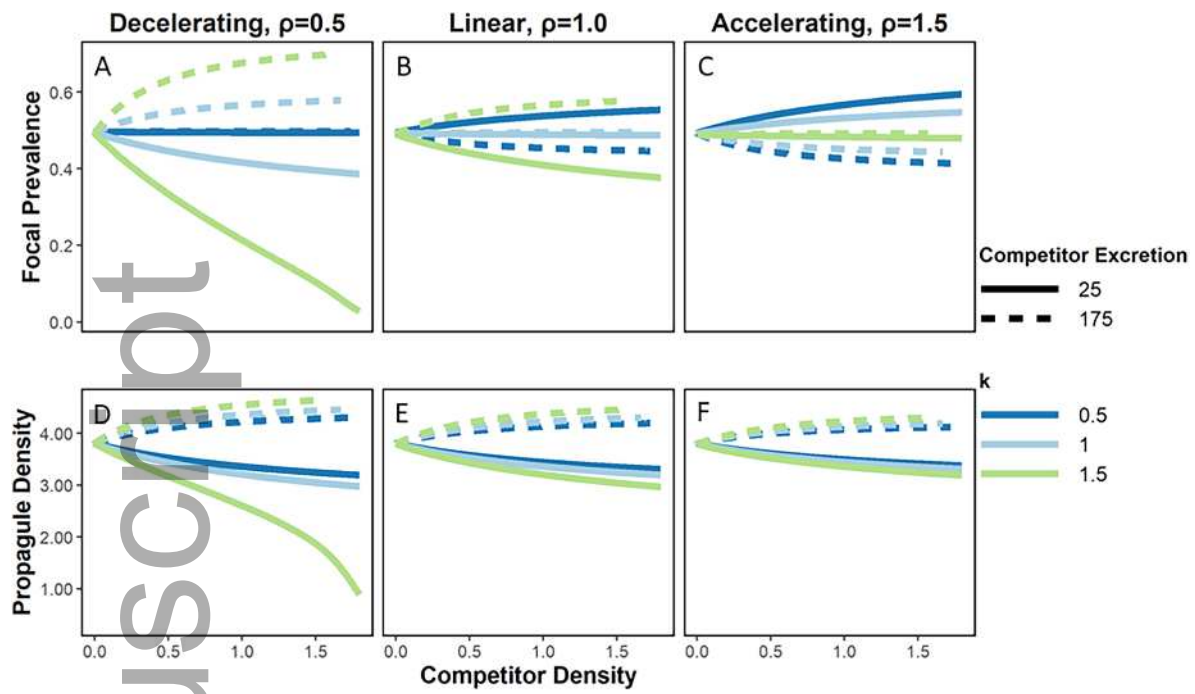
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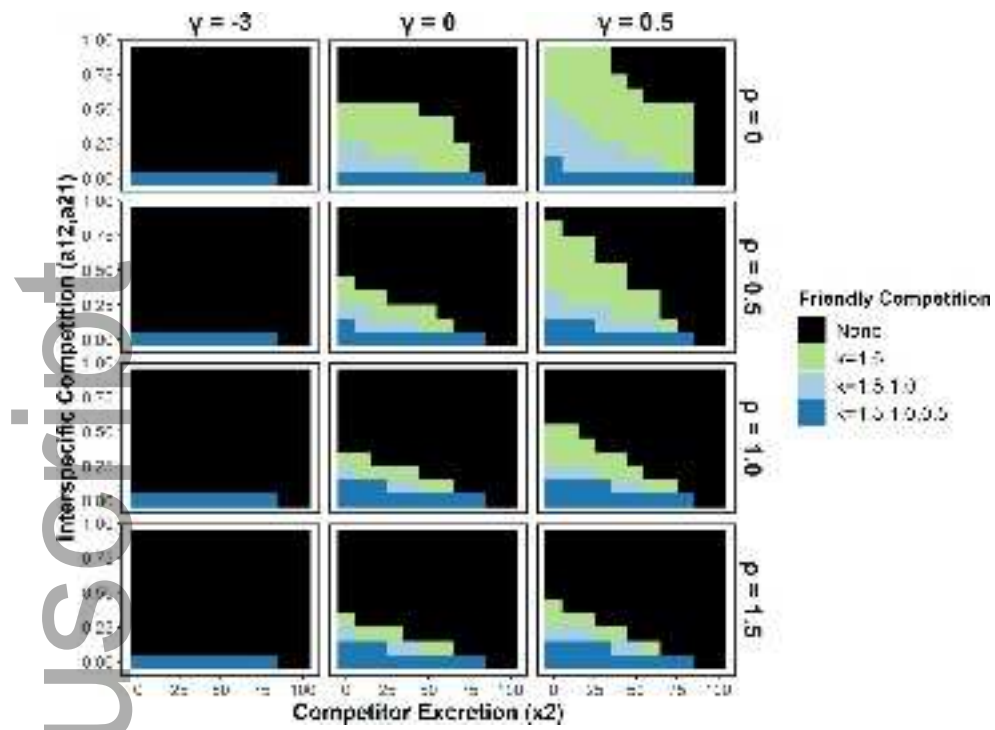
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ecy_3422_f4.tif



ecy_3422_f5.tif



ecy_3422_f6.tiff