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

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

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

56 **Introduction**

57 Hosts and their parasites do not exist in isolation. Rather, the likelihood of infection of any
58 individual host (i.e. the probability an individual is infected in a short time interval) depends on the
59 community in which it is embedded, due to direct inter-specific competition and cross-species parasite
60 transmission (O'Regan et al. 2015). Competitors can "amplify" (i.e. increase) infection prevalence in
61 a host species if they have high infection "competence", meaning they have high susceptibility to
62 infection and transmission potential (Power and Mitchell 2004). Similarly, competitors can "dilute"
63 (i.e. decrease) infection prevalence in a host species if they have low competence. With low enough
64 competence, competitors can even create "friendly competition", where they increase the density of
65 the host species by lowering infection likelihood, despite competing for resources (Hall et al. 2009).
66 Ultimately, understanding how competitors alter infection likelihood of individual host species will
67 allow us to predict the viability of host populations and the risk of spillover to other host species (Luis
68 et al. 2018). However, non-linear interactions between density dependent disease processes often
69 make it difficult to predict how one host species will impact infection likelihood in heterospecific host
70 species (Searle et al. 2016).

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

85 and thus the likelihood of infection for each host species.

86 The likelihood of infection, however, often changes nonlinearly with propagule dose (Figure
87 1A, B). As propagule dose increases, the infectivity of each parasite propagule can decrease, leading
88 to a decelerating (antagonistic) dose-infectivity relationship. Alternatively, as propagule dose
89 increases, the infectivity of each parasite propagule can increase, leading to an accelerating
90 (synergistic) dose-infectivity relationship (Regoes et al., 2003). Further, as propagule dose increases,
91 infected host mortality and propagule excretion from infected individuals may change (Ashworth et
92 al. 1996, Dallas and Drake 2014) (Figure 1C, D). Together, these “dose relationships” (dose-
93 infectivity, dose-mortality, and dose-excretion relationships) make parasite transmission a function of
94 environmental propagule density, which is in turn a function of parasite transmission. This feedback
95 loop may create challenges for predicting how competing host species will influence infection
96 likelihood. To date, however, mechanistic models of multi-host systems typically do not incorporate
97 dose-dependent feedback loops (Bowers and Begon 1991, Begon and Bowers 1994, Greenman and
98 Hudson 2000, Cáceres et al. 2014, Strauss et al. 2015, Searle et al. 2016). Further, while some studies
99 suggest that accelerating dose-infectivity relationships are common (Regoes et al., 2002), we lack a
100 quantitative review of how common accelerating and decelerating dose-infectivity relationships are.
101 By exploring the frequency of different types of dose relationships, and the impact they have on
102 multi-host systems, we may be better able to predict the impact of heterospecific host interactions on
103 infection likelihood in individual host species.

104 Thus, we sought to answer several basic questions: First, are accelerating, linear, or
105 decelerating dose-infectivity relationships more common in published studies? To answer this
106 question, we conducted a meta-analysis of experimental dose-infectivity experiments and found that
107 parasites usually exhibit decelerating dose-infectivity relationships. Second, we asked whether the
108 impact of competing host species with varying infection competencies on disease in a focal host
109 would depend on the relationship (1) between dose and the infectivity of parasite propagules (dose-
110 infectivity relationships), (2) between dose and host excretion rates of parasite propagules (dose-
111 excretion relationships), or (3) between dose and the mortality rate of infected individuals (dose-
112 mortality relationships). Using 2-host 1-parasite models that incorporate the types of dose
113 relationships found in empirical studies, we examined how the effects of interspecific host density on

114 infection prevalence in a focal host were mediated by dose-infectivity, dose-mortality, and dose-
115 excretion relationships. We found dose relationships can increase, decrease, or even reverse the
116 impact of heterospecific host density on infection prevalence. These results indicate that dose-
117 dependency is common in host-parasite interactions, and that disease models that do not take these
118 dose relationships into account may result in inaccurate predictions of disease dynamics in dose-
119 dependent systems.

120 **Meta-Analysis Methods**

121 *Literature Review*

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

142 *Meta-Analysis*

143 We conducted an analysis to determine whether dose-infectivity relationships were linear,
144 decelerating, or accelerating. For linear dose-infectivity relationships, dose does not change per
145 propagule infectivity, and dose changes infection rate in a linear manner. Under decelerating dose-
146 infectivity relationships, the infectivity of individual parasite propagules decreases with increased
147 propagule dose. Thus, as dose increases, propagule infectivity decreases, and the infection rate
148 increases in a concave-down manner. This does not necessarily mean that parasites mechanistically
149 interfere with one another. Rather, this pattern could be the result of non-linear immune responses in
150 an individual as dose increases. Finally, under accelerating dose-infectivity relationships, the
151 infectivity of individual parasite propagules increases with increased propagule dose. Thus, as dose
152 increases, propagule infectivity increases, and the infection rate increases in a concave-up manner.
153 Accelerating dose-infectivity relationships can be created if a high parasite dose is required to
154 overwhelm host defenses.

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

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

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

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

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168 where β is per-propagule infectivity, k is the dose shape parameter, and $\beta(fP)^k$ is the host
169 infection rate. For a given study, if $k = 1$ then the infection rate increases linearly with dose, if $k < 1$

170 then the infection rate has a decelerating increase with dose, and if $k > 1$ then the infection rate has an
171 accelerating increase with dose (Figure 1A).

172 We used Bayesian inference to fit equation 1A-1C to the published data from our literature
173 review. For each study, we numerically ran our system of ODEs for the experimental run time. We
174 then estimated the values of β , k , and f most likely to generate the infection prevalence reported in
175 the studies for each dose treatment. We used vaguely informative priors to prevent β and k from
176 going below 0. If parasite dose was instantaneous (e.g. injections), we assumed that hosts contact all
177 parasites instantaneously (see Appendix S1 for details). In cases where parasite densities were
178 reported as dilutions, we relativized all parasite densities so that the lowest parasite density was
179 100/volume. This ensured that the parasite density in the experiment was never less than 1. We did
180 not let fP fall below 1, as individuals cannot contact partial propagules. As our main variable of
181 interest was k , we additionally tested whether the posterior estimate for k depended on β and f . While
182 artificially lowering β increased our estimate of k to compensate for the reduced infection rate, and
183 vice-versa, our posterior estimate of k did not depend on f (Appendix S1).

184 We further tested whether experiments in our meta-analysis best fit a sigmoidal dose-
185 infectivity relationship, where per-propagule infectivity first increases with dose, and then decreases.
186 This would match a pattern where a minimal infective dose is necessary to overcome an individual's
187 immune system and establish an infection, but further increases in parasite dose yield diminishing
188 returns and decrease per propagule infectivity. We thus reran our analysis replacing the k in eq. 1A-
189 1C with

$$k = \max(k0 - fP * k1, 0) \quad \text{eq. 2}$$

190 Such that k decreased with dose (fP), though never becomes negative. Using Bayesian
191 inference, we then estimate values of β , $k0$, $k1$, and f for each experiment. This formulation has the
192 benefit that if $k1$ is high enough, our model creates a humped relationship between dose (fP) and the
193 infection rate ($\beta(fP)^{\max(k0 - fP * k1, 0)} S$), a pattern observed in some dose-infectivity experiments
194 (Strauss et al. 2019). We considered a sigmoidal dose-infectivity relationship to best fit an experiment
195 if the model DIC was lower than that for our constant k model, and if the 95% confidence interval of
196 k fell above 1 for low dose and fell below 1 for higher experimental dose.

197 In addition to infection prevalence, studies in our meta-analysis sometimes reported changes

199 in mortality or propagule excretion from infected hosts with propagule dose. However, studies were
200 inconsistent in the metrics they used to measure mortality and parasite load (e.g. mortality could be
201 measured as proportion of dead individuals, time until death, or visible damage to individuals). We
202 noted general trends but did not analyze the dose relationships of these metrics, as the metrics used
203 were too variable.

204 **Meta-analysis Results**

205 We found that the majority of published dose-infectivity relationships are decelerating ($k < 1$), where
206 increasing propagule dose lowers per-propagule infectivity (Figure 2). The 95% confidence intervals
207 of k values fell below 1 for 79/98 host-parasite combinations (decelerating), overlapped 1 for 12/98
208 host-parasite combinations (linear), and fell above 1 for 7/98 host-parasite combinations
209 (accelerating). We found no support for sigmoidal dose-infectivity relationships. While ΔDIC values
210 gave strong support for our non-constant k compared to our constant k model in 12 out of 98 studies
211 ($\Delta\text{DIC} > 10$) and weak support in 3 out of 98 studies ($10 > \Delta\text{DIC} > 5$), in 0 studies out of 98 did the
212 95% confidence interval of k fall above 1 for low propagule densities and fall below 1 for higher
213 experimental propagule densities.

214 **Model Exploration of Dose Relationships: Methods**

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

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251 the total effect of competitor density on focal host viability. To increase the density of the competitor,
252 we increase its intrinsic growth rate, r_2 , from 0 to $2r_1$.

253 *Dose-Infectivity Relationships*

254 For all analyses, we measure the impact of competing host density on model dynamics under
255 three dose-infectivity relationships: when $k_1 = 0.5$ (decelerating dose-infectivity relationship), when
256 $k_1 = 1.0$ (linear dose-infectivity relationship), and when $k_1 = 1.5$ (accelerating dose-infectivity
257 relationship, Figure 1A). For our main results, we assume that $k_1 = k_2$, but we explore asymmetric
258 dose-infectivity relationships in Appendix S3.

259 In our model, k alters both the shape of dose-infectivity relationships, and the magnitude of
260 parasite transmission. As k increases, the infection rate, $\beta_i(f_iP)^{k_i}S_i$, increases in an exponential
261 manner, thus increasing infection likelihood. Thus, to solely examine how the shape of dose-response
262 relationships alters infection likelihood, we vary β as we vary k such that disease prevalence in the
263 focal host in the absence of the competing host is always 0.5 at equilibrium. If we vary k in our model
264 without altering β , then increasing k always increases parasite transmission. The full relationship
265 between k and β is

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

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

273 *Dose-Excretion and Dose-Mortality Relationships*

274 In our meta-analysis, we found four additional effects of propagule dose across multiple host-

276 parasite combinations. As propagule dose increased (1) propagule excretion could decrease, (2)
277 propagule excretion could increase, (3) infected host mortality rate could increase, and (4) propagule
278 excretion and host mortality could concurrently increase. (In some cases, we interpreted higher
279 parasite load within-hosts as higher propagule excretion.) Thus, we ran our model under these four
280 scenarios concurrently with decelerating, linear, and accelerating dose-infectivity relationships.

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

$$283 \quad x_{i,dose} = x_i \left(\frac{1}{2} + \frac{f_i P}{2 f_i P_1} \right)^\gamma \quad eq. 9$$

284 where $f_i P_1$ is the propagule dose at equilibrium when $N_2 = 0$ using equations 3-7. We use this
285 parameterization because it guarantees that the excretion rate of host i is equal to x_i when at
286 equilibrium in the absence of the competing host. This simplifies our analysis because it means the
287 dose-excretion relationship only affects prevalence in host i when the competing host is present.

288 Models without dose-excretion relationships are equal to models with dose-excretion relationships if
289 $\gamma = 0$. In addition to models without dose-excretion relationships, we explore dose-excretion models
290 where $\gamma = -3$ (exponential decrease in excretion with dose) and $\gamma = 0.5$ (decelerating increase in
291 excretion with dose, Figure 1B).

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

$$294 \quad m_{i,dose} = m_{min} + (m_i - m_{min}) \left(\frac{f_i P}{f_i P_1} \right)^\rho \quad eq. 10$$

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

301 Model Exploration of Dose Relationships: Results

302 Confirming previous models (Cáceres et al. 2014), infection prevalence in the focal host is
303 influenced by both the density of the competing host and the rate at which it releases parasite
304 propagules when infected (Figure 4B). Analytical solutions to our model show that increases in
305 competitor density increase focal host infection prevalence and propagule density (i.e. amplify
306 disease) when the competitor is a larger source of parasite propagules, and lower focal host infection
307 prevalence and propagule density (i.e. dilute disease) when the competitor is a smaller source of
308 parasite propagules than the focal host (Appendix S4: Section S1). A host is a large “source” of
309 propagules if it has a high propagule excretion rate, and/or if it removes few propagules from the
310 environment. Our numerical simulations match this result: increases in competitor density decrease
311 disease prevalence in the focal host when competitor propagule excretion is lower than the focal host
312 (Competitor Excretion < 100, light blue lines in Figure 4B), and increase disease prevalence in the
313 focal host when competitor propagule excretion is higher than the focal host (Competitor Excretion >
314 100, light blue lines in Figure 4B). Thus, our model confirms pre-existing multi-host theory in the
315 absence of dose-relationships.

316 Dose-Infectivity Relationships

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

328 Accelerating and decelerating dose-infectivity relationships alter the impact of competitor
329 density on infection prevalence and propagule density by creating feedback loops between propagule
330 dose and per-propagule infectivity. Decelerating dose-infectivity relationships create negative

331 feedback loops. If a competing host releases fewer parasite propagules than the focal host, this lowers
332 propagule density in the environment, which lowers propagule dose. Lowering propagule dose
333 increases per-propagule infectivity, thus buffering the impact of competing host density on infection
334 prevalence, and in turn propagule density/dose. On the other hand, accelerating dose-infectivity
335 relationships create positive feedback loops. If a competing host releases fewer parasite propagules
336 than the focal host, this lowers propagule density in the environment, which lowers propagule dose.
337 Lowering propagule dose decreases per-propagule infectivity, thus accelerating the impact of
338 competing host density on infection prevalence, and in turn propagule density/dose. (The converse
339 can also happen if competing hosts increase parasite dose.) Thus, infection prevalence is generally
340 more sensitive to changes in competitor density under accelerating dose-infectivity relationships than
341 under decelerating dose-infectivity relationships.

342 Dose-Excretion Relationships

343 Our literature survey showed that propagule excretion from infected hosts can increase or
344 decrease with propagule dose (Data S1). Increasing dose may decrease propagule excretion if
345 parasites face within-host competition, where initial crowding may limit the production of parasite
346 propagules. On the other hand, increasing propagule dose may increase propagule excretion if high
347 doses overwhelm the host's immune system.

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

354 Conversely, under increasing dose-excretion relationships, this creates positive feedback
355 loops: increasing propagule dose increases propagule excretion, which in turn increases propagule
356 dose. Similarly, decreasing propagule dose decreases propagule excretion, which in turn decreases
357 propagule dose. This positive feedback loop increases the impact of competitor density on infection
358 prevalence (Figure 4C vs. 4B). Because positive feedback loops destabilize systems, adding both a
359 positive dose-excretion relationship and a positive dose-infectivity relationship to our system causes

360 the system to shift from 0% infection prevalence to 100% infection prevalence with small changes to
361 system parameters (Figure 4C). Our analytical solutions support these results (see Appendix S4:
362 Section S2). We again find that, qualitatively, changes in prevalence match changes to environmental
363 propagule density (Figure 4).

364 Dose-Mortality Relationships

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

370 Dose-mortality relationships represent negative feedback loops. As dose increases, the
371 infectious period of infected hosts shrinks due to increased mortality, lowering transmission and thus
372 dose. As dose decreases, the infectious period of infected hosts increases due to reduced mortality,
373 lowering transmission and thus dose. As with negative feedback loops created by decelerating dose-
374 infectivity and negative dose-excretion relationships, the negative feedback loops created by dose-
375 mortality relationships decrease the ability of competitor hosts to influence infection likelihood. We
376 see this reflected in environmental propagule density; low-competence competitor hosts lower
377 environmental propagule density less under dose-mortality relationships, and competent competitor
378 hosts raise propagule density less (Figure 5D-F vs. 4E).

379 However, dose-mortality relationships can reverse the impact that competitors have on
380 infection prevalence. This is because increasing propagule dose both increases infection prevalence
381 by increasing the rate at which susceptible individuals become infected ($\beta_i(f_i P)^{k_i}$), and additionally
382 decreases infection prevalence by increasing the mortality rate of infected hosts ($m_i \left(\frac{f_i P}{f_i P_1}\right)^\rho$). The
383 combined effects of dose-dependent mortality and infection rate depend on the values of the shape
384 parameters k_i and ρ . If infection rate changes with parasite dose faster than mortality ($\rho < k_i$),
385 increasing competitor density will increase infection prevalence when the competitor is a large source
386 of propagules, as expected, and vice versa (Figure 5A-C). In contrast, if mortality changes with
387 parasite dose faster than infection rate changes with parasite dose ($\rho > k_i$), then we see a reverse in

388 whether competitor density increases or decreases infection prevalence — increasing the density of
389 competitors that are large sources of parasite propagules decreases infection prevalence and
390 increasing the density of competitors that are small sources of propagules increases infection
391 prevalence (Figure 5A-C). This pattern occurs because if $\rho > k$, then mortality increases with dose
392 faster than infectivity. When $\rho \approx k$, changes in mortality and infectivity approximately cancel each
393 other out as dose changes, so competitor density will have little effect on infection prevalence (Figure
394 5A-C, see Appendix S4: Section S3 for full analysis). Combining positive dose-excretion
395 relationships with dose-mortality relationships does not qualitatively change the impact of either dose-
396 relationship on prevalence and propagule patterns (Appendix S3).

397 Friendly Competition

398 Confirming previous theory, in the absence of dose-relationships competitors with weak inter-
399 specific competition and low competence increase the density of the focal host (i.e. friendly
400 competition), while competitors with strong inter-specific competition and high competence decrease
401 the density of the focal host (Figure 6B). Note that in our model, if the effect of inter-specific
402 competition on the focal host is greater than zero, increasing competitor density will always
403 eventually drive the focal host to extinction. Thus, “Friendly Competition” in our model does not
404 represent a monotonic positive effect of competing host density on focal host density, but rather a
405 humped relationship. In these circumstances, increasing competitor density initially increases focal
406 host density by decreasing the infection rate. However, as competitor density increases, the negative
407 effect of direct competition on focal host density eventually outweighs the positive effects of the
408 removal of infectious propagules.

409 Positive feedback loops facilitate friendly competition. Our model shows that dose-
410 relationships that create positive feedback loops (accelerating dose-infectivity relationships, positive
411 dose-excretion relationships) increase the parameter space where competing hosts can increase focal
412 host density (Figure 6, green vs. light blue in all panels, and B,E,H,K vs C,F,I,L). Alternatively, dose-
413 relationships that create negative feedback loops (decelerating dose-infectivity relationships, all dose-
414 mortality relationships, negative dose-excretion relationships) decrease the parameter space where
415 competing hosts can increase focal host density (Figure 6, dark blue vs. light blue in all panels, A-C
416 vs. D-L, and B,E,H,K vs. A,D,G,J). This is because friendly competition occurs when competing

417 hosts strongly dilute disease. As we see in Figure 4, dose-relationships that create positive feedback
418 loops increase the strength of dilution.

419 **Discussion**

420 Parasite dose underlies every aspect of infectious disease transmission, and can transform
421 interactions between hosts who share parasites. Our study shows that the effect of parasite dose on
422 per-propagule infectivity, host mortality, and propagule excretion can strengthen, weaken, or even
423 reverse the impact of heterospecific host density on disease in a focal host. Our meta-analysis
424 indicates that most dose-infectivity relationships are decelerating (Figure 2), and thus may decrease
425 the impact of heterospecific host density on infection prevalence and infectious propagule density via
426 negative feedback loops (Figure 4). Dose-excretion relationships can create positive or negative
427 feedback loops, increasing or decreasing the impact of heterospecific hosts on infection prevalence
428 and propagule density (Figure 4). Further, dose-mortality relationships can make the impact of
429 heterospecific hosts on infection prevalence negatively correlated with the effects on propagule
430 density (Figure 5). Finally, our results show that positive feedback loops created by accelerating dose-
431 infectivity relationships and positive dose-infectivity relationships can facilitate friendly competition,
432 even in the face of high interspecific competition. Together, these results suggest that dose
433 relationships could fundamentally alter how interspecific host interactions influence disease
434 dynamics, and that models that ignore dose relationships may mislead us in our efforts to understand
435 and predict how changes in host communities will alter disease patterns.

436 *Dose-response feedback loops*

437 Dose-response relationships create feedback loops that can increase or decrease the extent that
438 competing hosts alter disease prevalence, parasite propagule density, and density of focal hosts (Table
439 2). The transmission of a parasite within an ecosystem increases with (1) parasite dose, (2) the
440 probability that each parasite in that dose will infect a host, (3) the rate of propagule excretion from
441 hosts once they are infected, and (4) the lifespan of those infected hosts. If increasing dose increases
442 any of these factors, then propagule dose and parasite transmission enter a positive feedback loop. If
443 increasing dose decreases any of these factors, then propagule dose and parasite transmission enter a
444 negative feedback loop (feedback loops in Figure 3). Ultimately, through these feedback loops, dose-
445 response relationships can strengthen, weaken, or reverse predictions for whether a host will amplify

446 or dilute disease based purely on their competence.

447 *Dose-infectivity relationships*

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

455 As dose increases, the per-propagule probability of infection decreases under decelerating
456 dose-infectivity relationships. This creates a negative feedback loop between dose and the infection
457 rate that should weaken the ability of competing hosts to increase or decrease disease, and should
458 weaken the ability of hosts to increase one another's density via dilution in the face of interspecific
459 competition (Figure 4,5,6). This information can help us interpret experiments. For example, in our
460 meta-analysis we found decelerating dose-infectivity relationships for *Daphnia dentifera* infected by
461 *Metschnikowia bicuspidata* (Dallas and Drake 2014), a model system for the dilution/amplification
462 effect in two-host experiments (Hall et al. 2009, Strauss et al. 2015, Searle et al. 2016). Mechanistic
463 models of this system have thus far assumed mass-action infection processes and would most likely
464 be improved by implementing decelerating dose-infectivity relationships. Further, if dose-infectivity
465 relationships are usually decelerating, then changes to parasite dose due to competing hosts will have
466 the largest impact on infection rate, and thus infection prevalence, at low doses (Figure 1A). Knowing
467 this will help us identify natural systems where host community composition will likely alter infection
468 prevalence.

469 While our meta-analysis found that most experimental dose-infectivity relationships are
470 decelerating (Figure 2), many dose-infectivity relationships exhibit a minimal infective dose (Ward
471 and Akin 1984), a feature not possible under a purely decelerating dose-infectivity relationships. A
472 decelerating dose-infectivity relationship that nevertheless has a minimal infective dose could fit a
473 piecemeal function that is 0 below the minimal infective dose and decelerates above the minimal
474 infective dose, or a sigmoidal function where per-propagule infectivity increases at low doses and

475 decreases at higher doses. Mechanistically, a dose-infectivity relationship that both accelerates or
476 decelerates depending on propagule dose could be possible because infection is determined by
477 interactions between parasites and many host defenses, and defenses such as the immune system may
478 respond non-linearly to propagule dose (Van Leeuwen et al. 2019, Stewart Merrill et al. 2019). We
479 tested for this latter possibility, but found no evidence for sigmoidal dose-infectivity relationships in
480 our meta-analysis. Nonetheless, our results explain how a sigmoidal dose-infectivity relationship
481 would affect the relationship between focal infection prevalence and competitor density or between
482 parasite density and competitor density: at low doses, changes in dose will create positive feedback
483 loops, while at high doses, changes in dose will create negative feedback loops.

484 *Dose-excretion relationships*

485 While dose-infectivity and dose-mortality relationships mostly cause negative feedback loops,
486 dose-excretion relationships can cause both positive and negative feedback loops, either increasing or
487 decreasing disease amplification and dilution. To cause a negative feedback loop, parasite propagule
488 excretion must decrease with dose. This could potentially occur if increasing dose lowers the within
489 host growth rate of the parasite (Regoes et al. 2002). Or in cases where hosts only excrete parasites at
490 host death, dose may decrease excretion rates if it simultaneously decreases host lifespan, limiting the
491 amount of time that parasites have to grow (Ebert et al. 2000). To cause a positive feedback loop,
492 parasite propagule excretion must increase with dose. This is most likely for macroparasites that do
493 not reproduce in certain hosts, and thus excretion is limited by parasite dose (Johnson et al. 2012).
494 Ultimately, dose-excretion relationships might be the most important dose-response relationship to
495 measure in future experiments, as we do not have strong prior assumptions about whether these
496 relationships should be positive or negative.

497 *Dose-mortality relationships*

498 Increasing dose generally decreases infected host lifespan (Appendix S5). This creates a
499 negative feedback loop between dose and the infection rate which should weaken the ability of
500 competing hosts to dilute or amplify disease, and should prevent friendly competition (Figure 5,6).
501 Further, we found that while infection prevalence is generally positively related with propagule
502 density, dose-mortality relationships can reverse this relationship (Figure 5). Traditionally, we assume
503 that competing hosts are more likely to decrease infection prevalence if they remove many propagules

504 from the environment, if they have a low transmission rate or susceptibility, and if they are strong
505 competitors (Cáceres et al. 2014, Strauss et al. 2015). Competing hosts with these traits reduce disease
506 because they lower environmental propagule density, lowering dose and infection rate, and ultimately
507 lowering infection prevalence. However, dose-mortality relationships can make infection rate and
508 infection prevalence negatively correlated, and thus challenge our assumptions of which hosts should
509 reduce infection prevalence in a community. If host mortality increases at a faster rate with propagule
510 dose than infection rate does, then infection rate will be negatively correlated with prevalence — thus
511 the low competence, strongly competing hosts that might otherwise be expected to decrease disease
512 will actually increase disease prevalence over some range of densities. This scenario is potentially
513 common, as many systems display positive dose-mortality relationships (for instance, Ashworth et al.
514 1996; Agnew and Koella 1997; Blair and Webster 2007; De Roode et al. 2007). Further, it is when
515 decelerating dose-infectivity relationships, which our meta-analysis shows to be common (Figure 2),
516 are combined with dose-mortality relationships that we see expected low-competence hosts increase
517 disease, and vice versa (Figure 5). Indeed, highly competent hosts with positive dose-mortality
518 relationships and decelerating dose-infectivity relationships have been shown to dilute disease (Ebert
519 et al. 2000, Dallas and Drake 2014, Searle et al. 2016). Arguably, infection prevalence is only
520 indirectly important, and what matters is that competent hosts increase infection rates, and low-
521 competence hosts decrease infection rates, regardless of infection prevalence. However, infection
522 prevalence is important in that we can readily measure it, and thus use it as a proxy for infectious
523 disease severity in ecosystems. Thus, infectious disease ecologists should factor in dose-mortality
524 relationships when trying to infer infection processes from infection prevalence.

525 *Future directions*

526 Pairing multi-host empirical studies with mechanistic dose models will allow us to uncover the
527 mechanisms driving disease patterns in multi-host communities. Mechanistic models paired with
528 empirical data have generated valuable insights into the processes driving disease in multi-host
529 communities, such as when inter-host interactions are simultaneously amplifying and diluting disease
530 (Luis et al. 2018), or the relative contributions of competition and host competency to disease dilution
531 (Strauss et al. 2015). Pairing mechanistic dose models with empirical data will allow us to answer
532 many open questions about the real-world importance of dose relationships, such as (a) do dose

533 relationships often alter biodiversity-disease relationships in natural populations? (b) Are decelerating
534 dose-infectivity relationships truly common in natural populations? And (c) do dose effects alter
535 infection prevalence most strongly via infectivity, host-mortality, or propagule excretion? Overall, an
536 improved understanding of dose response relationships will enable us to better understand the impact
537 of host species interactions on disease risk, and thus make more informed conservation and public
538 health decisions.

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

544 **Supporting Information**

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

546 **Open Research**

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

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625

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.

629 **Table 1.** Categories of host/parasite interactions included in literature review. We found 98
630 host/parasite combinations across 63 studies. We consider “Environmental” parasites to be parasites
631 where host contact is not required for transmission, and where parasites are not transmitted via
632 vectors. Parasites in the “Other” taxa category include cercozoan, myxozoan, platyzoan, and
633 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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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

636 **Figure 1:** Dose relationships can take a variety of forms. X-axis shows propagule dose, and Y-axis
637 shows (A) the infection rate (dose-infectivity relationship), (B) the proportion of individuals
638 becoming infected after exposure to that dose (dose-infectivity relationship, cont.), (C) the rate at
639 which parasite propagules are excreted from infectious individuals (dose-excretion relationship), and
640 (D) the mortality rate of infected individuals (dose-mortality relationship). The shape of each dose
641 relationship is described by a shape parameter (k for dose-infectivity relationships, eq. 1, γ for dose-
642 excretion relationships, eq. 9, and ρ for dose-mortality relationships, eq. 10). If k , ρ , or γ is greater
643 than 1, the dose relationship has an accelerating increase. If k , ρ , or γ is equal to 1, the dose
644 relationship has a linear increase. If k , ρ , or γ is between 1 and 0, the dose relationship has a
645 decelerating increase. If k , ρ , or γ is equal to 0, the dose relationship is static. If k , ρ , or γ is less than
646 0, the dose relationship has an exponential decrease. Lines are shown for parameter values included in
647 model results, based on the literature review results.

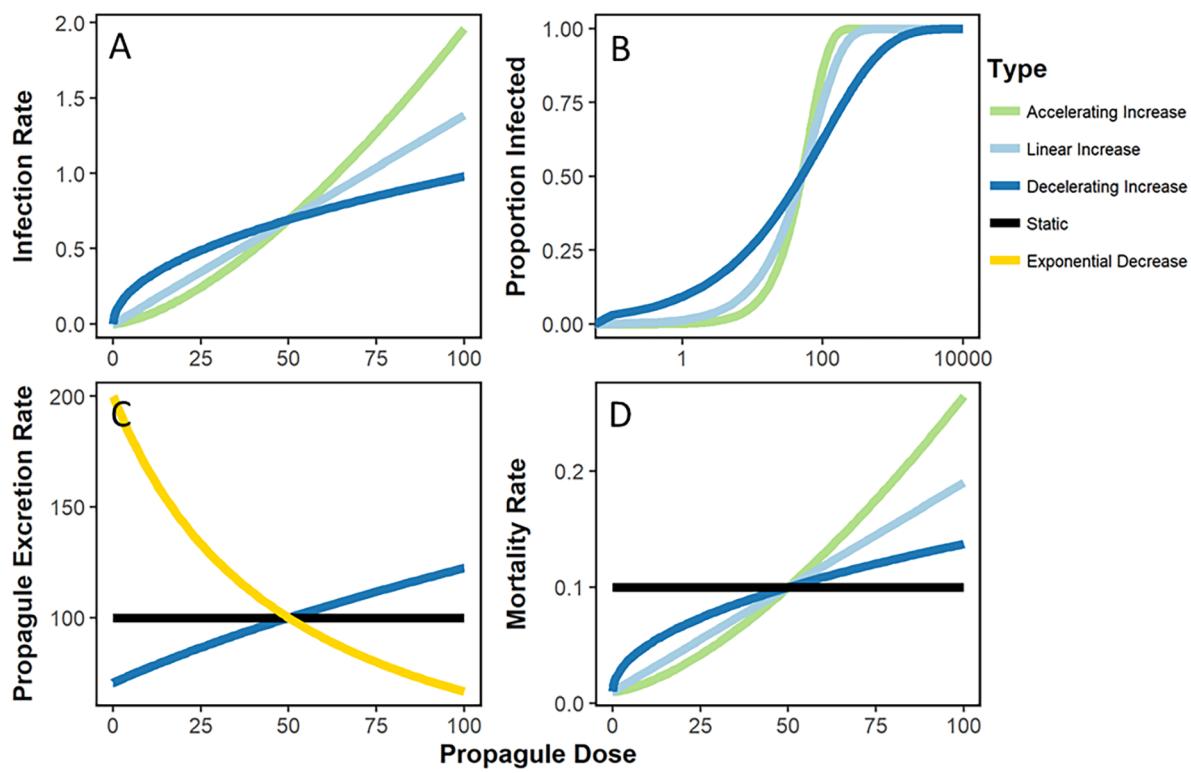
648 **Figure 2:** Most empirical dose-infectivity relationships are decelerating. Values on the x-axis show
649 Bayesian estimates of the dose shape parameter (k) values from published dose-infectivity
650 relationships, with bars showing 95% confidence intervals of the posterior distribution. If an interval
651 overlaps the 1 line, then we do not reject the null hypothesis that infection rate increases linearly with
652 dose, which implies that dose does not alter per-propagule infectivity. If intervals lie below one, then
653 per-propagule infectivity decreases with dose, and dose-infectivity relationships have a decelerating
654 increase. If the interval lies above the 1 line, then per-propagule infectivity increases with dose, and
655 dose-infectivity relationships have an accelerating increase. **Figure 3:** Schematic of equations 3-9.
656 Black lines represent dose-independent processes and blue lines represent dose-dependent processes.
657 Dashed green lines connect environmental propagule density to dose dependent processes to visualize
658 feedback loops. S_1 and I_1 represent susceptible and infected individuals of species 1, S_2 and I_2
659 represent susceptible and infected individuals of species 2, and P represents environmentally
660 transmitted parasite propagules. (a) All hosts give birth as a function of intraspecific and interspecific
661 density and competition (eq. 3, 5). (b) Susceptible individuals become infected at a rate determined by
662 parasite dose (eq. 4, 6). (c) Infected individuals excrete parasite propagules into the environment as a
663 function of dose (eq. 7, 8). (d) Propagules degrade over time (eq. 7). (e) Finally, infected individuals
664 die as a function of parasite dose (eq. 4, 6, 9).

665 **Figure 4:** Negative dose-excretion relationships or decelerating dose-infectivity relationships
666 decrease (and positive dose-excretion relationships or accelerating dose-infectivity relationships
667 increase) the magnitude of the relationship between infection prevalence and competitor density and
668 between propagule density and competitor density. Changes in infection prevalence of the focal host
669 (Y-Axis A-C) and log propagule density (Y-Axis D-F) as competitor density increases (X-axis).
670 Panels represent models with negative dose-excretion relationships (A,D), no dose-excretion
671 relationship (B,E), or positive dose-excretion relationships (C,F). Solid lines represent competitors
672 with lower propagule excretion than the focal host species, while dashed lines represent competitors
673 with higher propagule excretion than the focal host species. Dark blue lines show decelerating dose-
674 infectivity relationships, light blue lines show linear dose-infectivity relationships, and green lines
675 show accelerating dose-infectivity relationships.

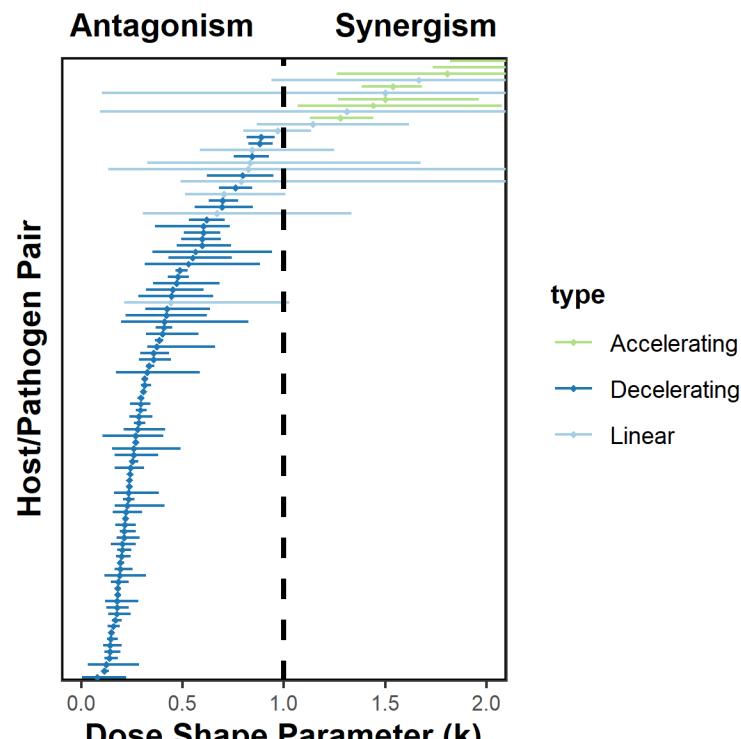
676 **Figure 5:** Decelerating dose-mortality relationships decrease (and accelerating dose-mortality
677 relationships increase) the magnitude of the relationship between infection prevalence and competitor
678 density and between propagule density and competitor density. Changes in infection prevalence of the
679 focal host (Y-Axis A-C) and log propagule density (Y-Axis D-F) as competitor density increases (X-
680 axis). Panels represent models with decelerating dose-mortality relationships (A,D), Linear dose-
681 mortality relationships (B,E), or Accelerating dose-mortality relationships (C,F). Solid lines represent
682 competitors with lower propagule excretion than the focal host species, while dashed lines represent
683 competitors with higher propagule excretion than the focal host species. Dark blue lines show
684 decelerating dose-infectivity relationships, light blue lines show linear dose-infectivity relationships,
685 and green lines show accelerating dose-infectivity relationships.

686 **Figure 6:** Positive dose-mediated feedbacks loops facilitate friendly competition. Regions of
687 parameter space show whether focal host density can increase with density of competing hosts
688 (friendly competition), with competitor propagule excretion rate (x_{-2}) on the X-axis and interspecific
689 competition (α_{-12} and α_{-21}) on the Y-axis. Dark blue indicates friendly competition for all dose-
690 infectivity relationships, light blue indicates friendly competition if per-propagule infectivity
691 increases linearly or accelerates with dose, green indicates friendly competition only if per-propagule
692 infectivity accelerates with dose, and black indicates no friendly competition. Panels indicate different
693 dose-mortality relationships ($\rho=0$ for none, $\rho=0.5$ for decelerating, $\rho=1.0$ for linear, $\rho=1.5$ for

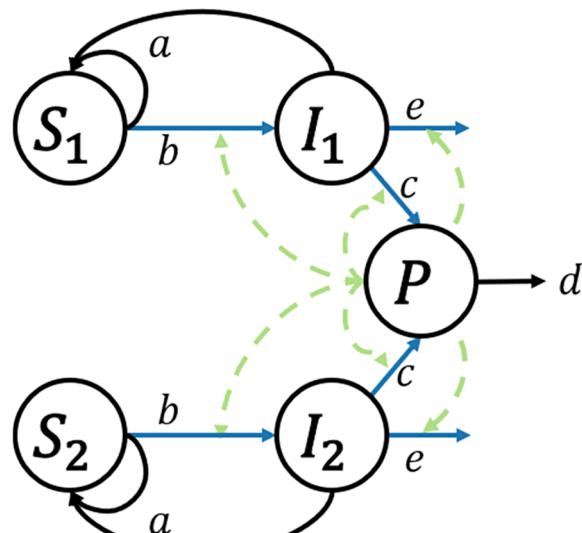
694 accelerating) and different dose-excretion relationships ($\gamma=-3$ for negative, $\gamma=0$ for none, $\gamma=0.5$ for
695 positive).



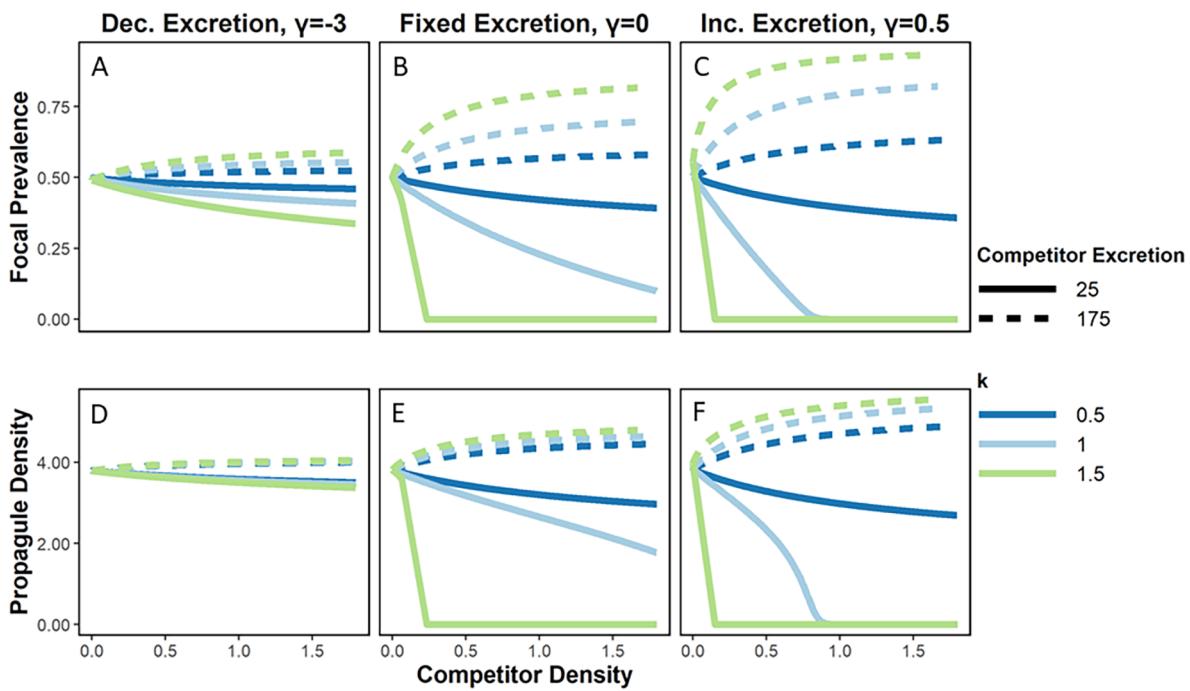
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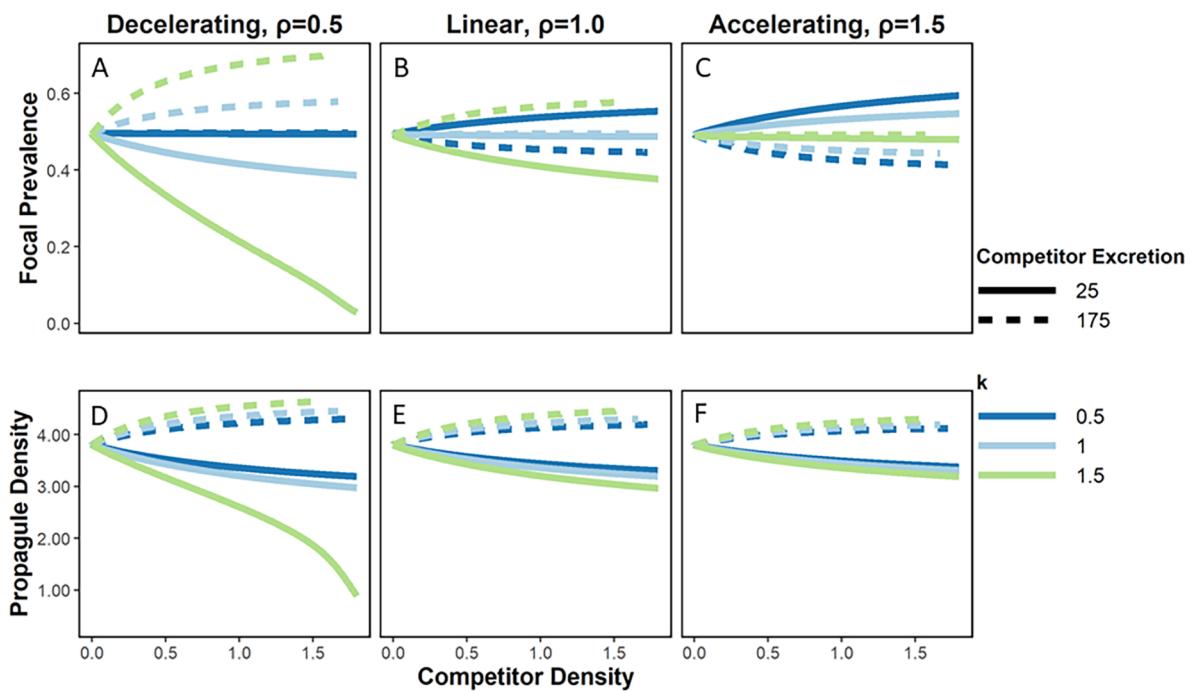
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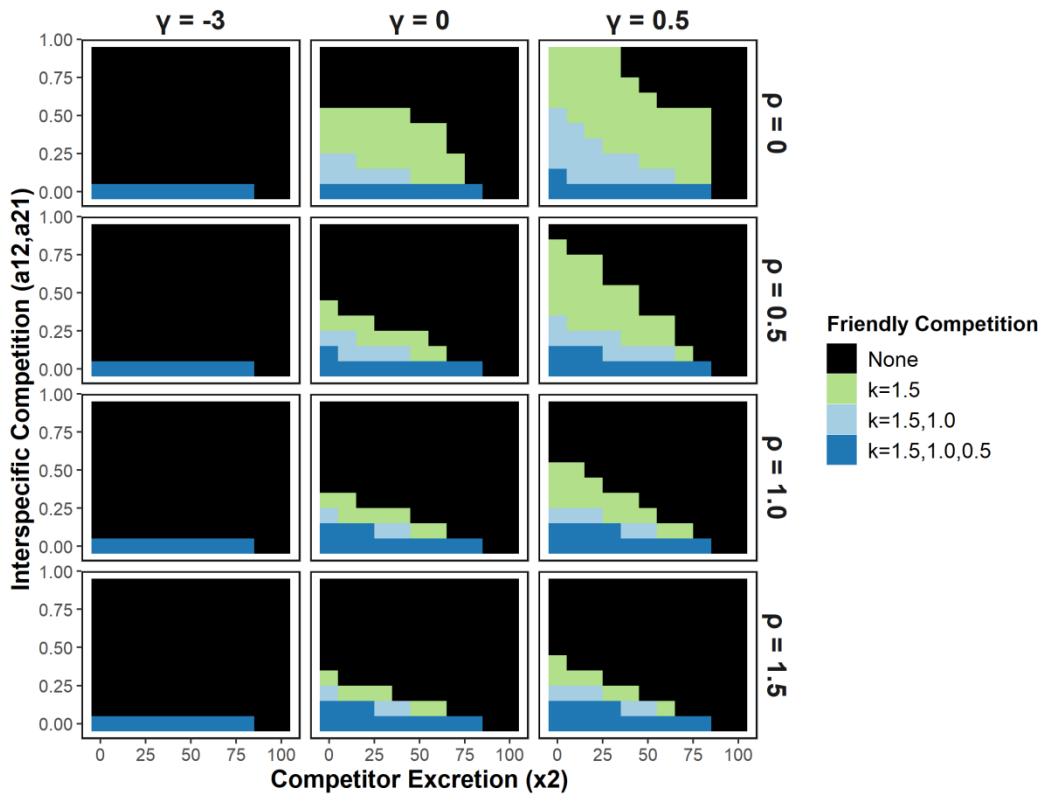
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ecy_3422_f4.tif



ecy_3422_f5.tif



ecy_3422_f6.tif