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Linking host traits, interactions with competitors, and disease: Mechanistic foundations for disease dilution

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28

29 **ABSTRACT**

- 30 1) The size of disease epidemics remains difficult to predict, especially when parasites
31 interact with multiple species. Traits of focal hosts like susceptibility could directly
32 predict epidemic size, while other traits including competitive ability might shape it
33 indirectly in communities with a ‘dilution effect’.
- 34 2) In a dilution effect, diluter taxa can reduce disease by regulating (lowering) the density of
35 focal hosts (i.e., through competition), or by reducing encounters between focal hosts and
36 parasites. However, these dilution mechanisms are rarely grounded in focal host traits,
37 and the relative importance of host regulation vs. encounter reduction remains
38 understudied.
- 39 3) Here, we map focal host traits to disease—via these dilution mechanisms—in
40 communities with diluters. We measured two traits (competitive ability and
41 susceptibility) for eight genotypes of a focal host (*Daphnia*), tracked the densities of each
42 genotype in experimental mesocosms (+/- *Ceriodaphnia* competitor/diluters), and
43 monitored their infections with a virulent fungal parasite (*Metschnikowia*) over 6-8 host
44 generations. We disentangled the impacts of both traits on the density of infected hosts
45 and partitioned dilution mechanisms using path models.
- 46 4) Higher susceptibility directly fueled larger epidemics. Simultaneously, weaker
47 competitive ability indirectly suppressed epidemics by enabling higher densities of
48 diluters. These higher densities of diluters reduced the density of infected hosts indirectly
49 via host regulation. In contrast, encounter reduction was much weaker.
- 50 5) Our experiment strengthens the dilution effect paradigm with a predictable, traits-
51 oriented framework. Similar traits—susceptibility, competitive ability, and their
52 covariance—could help predict epidemic severity in a variety of other systems.
53 Partitioning the direct and indirect effects of diluters could also delineate how they
54 impact disease. Such trait-based insights could help broadly predict the size of epidemics
55 in diverse communities.

56

57 **KEY WORDS:** *Daphnia*, density of infected hosts, dilution effect, encounter reduction, host
58 regulation, host traits, intraspecific variation, path analysis

59

60 **INTRODUCTION**

61 What makes disease epidemics smaller or larger? Disease theory indicates that, among
62 other factors, traits of hosts can directly influence epidemic size (Anderson & May 1981; Dwyer
63 & Elkinton 1993; Strauss *et al.* 2015). One obvious trait is susceptibility, i.e., the rate at which
64 susceptible hosts become infected upon contact with parasite propagules, vectors, or infected
65 hosts. More resistant hosts should experience smaller epidemics, while more susceptible hosts
66 should experience larger ones (Dwyer & Elkinton 1993; Strauss *et al.* 2015). However, species
67 interactions, like competition and predation, can also influence epidemics (Keesing, Holt &
68 Ostfeld 2006; Strauss *et al.* 2016). Other traits like competitive ability may modulate the strength
69 of these interactions, and hence indirectly shape disease (e.g., Strauss *et al.* 2015). Thus, multiple
70 traits can govern epidemics in a community context, though both direct and indirect pathways.

71 Mechanistic dilution effect theory could help predict these community-level impacts of
72 host traits on epidemic size. Dilution effects arise broadly (Civitello *et al.* 2015) when resistant
73 ‘diluter’ taxa interfere with transmission among more competent focal hosts (Ostfeld & Keesing
74 2000), frequently via one or two mechanisms. First, diluters can *regulate* the density of focal
75 hosts via predation or competition (Keesing, Holt & Ostfeld 2006), thus inhibiting direct or
76 environmental transmission (Anderson & May 1981). These diluters indirectly shape disease by
77 decreasing the density of focal hosts. Whether such indirect effects constitute a dilution effect in
78 the strict sense seems beside the point (but see Begon 2008). Second, diluters might *reduce*
79 *encounters* between focal hosts and parasites by diverting vectors away from focal hosts (Ostfeld
80 & Keesing 2000), modifying focal host behavior, or consuming free-living parasites (Johnson *et*
81 *al.* 2010). Trait-based insights into either of these general mechanisms could help broadly predict
82 when diluters should exert the strongest impacts on disease.

83 Presently, such predictive power remains limited because few experiments link gradients
84 of focal host traits to dilution mechanisms. Intuitively, host regulation might matter more when
85 predation (Rohr *et al.* 2015) or competition (Strauss *et al.* 2015) depresses focal host densities
86 more strongly. Encounter reduction appears stronger when diluters remove parasites more
87 rapidly and strongly resist infection (Venesky *et al.* 2014; but see Wojdak *et al.* 2014). Yet
88 intraspecific variation in susceptibility among focal hosts may counter either dilution mechanism
89 by fueling uncontrollably large or inconsequentially small epidemics (Strauss *et al.* 2015). Thus,

90 traits of focal hosts matter as well. Furthermore, impacts of multiple focal host traits could easily
91 become confounded. For example, when susceptibility directly fuels epidemics, it could obscure
92 how traits like competitive ability—which frequently covary with susceptibility (Duncan,
93 Fellous & Kaltz 2011)—modulate the impacts of diluters. Therefore, stronger mechanistic
94 foundations for disease dilution require experiments that disentangle the impacts of covarying
95 focal host traits.

96 Drivers of epidemics in multi-host communities become even harder to delineate when
97 host regulation and encounter reduction operate simultaneously (e.g., Ogden & Tsao 2009; Rohr
98 *et al.* 2015; Dallas, Hall & Drake 2016; Strauss *et al.* 2016). Dilution theory rarely embraces this
99 challenge; yet hosts and diluters that encounter the same parasites also frequently compete. We
100 label this combination of encounter reduction and competitive host regulation ‘friendly
101 competition’ (Hall *et al.* 2009). Examples likely include the transmission of hantavirus (Clay *et*
102 *al.* 2009), Lyme (Ogden & Tsao 2009), *Schistosoma* (Johnson *et al.* 2009), and parasites in
103 intertidal (Thieltges *et al.* 2009), amphibian (Johnson *et al.* 2013), and plant communities
104 (Mitchell, Tilman & Groth 2002; Lacroix *et al.* 2014). In friendly competition, impacts of
105 diluters—hereafter, competitor/diluters—likely depend on the competitive ability of focal hosts
106 (Strauss *et al.* 2015). Competitor/diluters could become rare if focal hosts compete strongly, but
107 remain numerous if focal hosts compete weakly. High densities of competitor/diluters could
108 reduce disease via host regulation, encounter reduction, or both. However, the relative strength
109 of these dilution mechanisms remains understudied (but see Ogden & Tsao 2009).

110 Here, we disentangle the impacts of covarying focal host traits and partition the dilution
111 mechanisms operating in a multi-generational mesocosm experiment. A two-host planktonic
112 example provides tractability and captures the natural history of our study system (see Strauss *et*
113 *al.* 2016). First, we picked eight clonal genotypes of the focal host (*Daphnia dentifera*) to
114 establish a gradient of two correlated traits: susceptibility and competitive ability. Then, we
115 created epidemics of a virulent fungus (*Metschnikowia bicuspidata*) in mesocosms with and
116 without a key competitor/diluter (*Ceriodaphnia sp.*). Finally, we combined linear and path
117 models to map host traits via dilution mechanisms to disease. Although we compare two metrics
118 of epidemic size—the density of infected hosts and infection prevalence—we focus on the
119 former since it responded more clearly to diluters. Higher susceptibility directly fueled larger
120 epidemics. Simultaneously, stronger competitive ability indirectly allowed higher densities of

121 infected hosts, because the populations of diluters were constrained. Finally, the density of
122 infected hosts was primarily reduced via host regulation. In other words, the indirect effects of
123 competitor/diluters, via changes in focal host density, outweighed their direct effects on disease
124 (i.e., via encounter reduction). This trait-based framework and tractable case study brings
125 dilution theory closer to predicting the size of epidemics in multi-host communities.

126

127 **MATERIALS AND METHODS**

128 **Natural History of the Study System**

129 The focal host in this study, the cladoceran *Daphnia dentifera*, dominates grazer
130 communities in many North American freshwater lakes (Tessier & Woodruff 2002). It frequently
131 suffers autumnal epidemics caused by the virulent fungus *Metschnikowia bicuspidata* (Hall *et al.*
132 2010b; Strauss *et al.* 2016). Focal hosts consume infectious fungal spores while foraging (Hall *et al.*
133 2007) but vary in their susceptibility to infection (Hall *et al.* 2010a). Infected hosts release
134 spores after death. A second dominant cladoceran, *Ceriodaphnia sp.*, often competes (Tessier &
135 Woodruff 2002) and can reduce disease by regulating *Daphnia* density (Strauss *et al.* 2016).
136 These competitor/diluters also consume fungal spores while foraging but strongly resist
137 infection, hence reducing encounters between focal hosts and parasites (Hall *et al.* 2009; Strauss
138 *et al.* 2015). Among a set of 28 Indiana lakes (see Strauss *et al.* 2016), these two taxa constitute
139 88% of cladoceran individuals. Although higher diversity correlated with lower disease across
140 these lakes, this dilution effect was driven more specifically by higher frequencies of
141 *Ceriodaphnia* in the more diverse lakes (rather than diversity *per se*). Competitive regulation
142 appeared to reduce the density of infected hosts in these lakes, while encounter reduction
143 lowered infection prevalence more strongly. The current experiment with two-host communities
144 is inspired by these field patterns (Strauss *et al.* 2016).

145

146 **Trait Measurements**

147 We quantified indices of two important traits, susceptibility and competitive ability, for
148 eight different genotypes of the focal host (see Appendix S1 in Supporting Information for
149 details). These genotypes were selected from laboratory stocks, using limited prior information,
150 in order to spread the range of both traits. In short, we estimated an index of susceptibility (the
151 transmission coefficient, β) by fitting a mathematical model to infection assays (e.g., Hall *et al.*

2007). In these assays—replicated among genotypes—fifteen individuals were exposed to each of three parasite concentrations, maintained individually, and later inspected for signs of infection. Susceptibility was fit (bootstrapped standard errors) with maximum likelihood using the BBMLE package in R (Bolker 2008; R Core Team 2017). This parameter (β) represents the probability of a focal host becoming infected in the absence of conspecifics or competitor/diluters, given its body length (L), density of infectious spores (Z), and the duration of spore exposure (t).

We also estimated an index of competitive ability, using growth rate assays with low food resources (e.g., Hall *et al.* 2012). Mass accrual of neonates during a 5–6 day juvenile period is directly proportional to fitness (Lampert & Trubetskova 1996). In turn, competitive ability depends on fitness when resources are limiting (reviewed in Grover 1997). Therefore, we provided hosts with low resources in our assay (0.15 mg mass/L *Ankistrodesmus falcatus* daily). We dried and weighed body mass of individuals at birth (mean $N = 9.8$ among genotypes) and other individuals 5–6 days later (mean $N = 14.5$). Then, we calculated growth rate as $\ln(\text{mass accrual})/\text{time}$. Thus, this index of competitive ability represents the growth rate of an individual consuming limited resources, in the absence of conspecifics, infection, or competitor/diluters. Although we use this index to predict interspecific competition here, it also predicts intraspecific competition (i.e., clonal selection and evolution) among *Daphnia* genotypes (Strauss *et al.* 2017).

These indices of susceptibility and competitive ability provided continuous gradients of two covarying focal host traits. Next, we used these trait gradients to predict outcomes among the same genotypes in a multi-generational mesocosm experiment.

Mesocosm Experiment

The mesocosm experiment crossed focal host genotype (8 levels) with presence/absence of competitor/diluters (2 levels). All combinations of treatments were replicated 4 times in 75-L tanks. Details are presented in Appendix S1. Mesocosms began with focal hosts (15 L^{-1}), and in competition treatments, a single genotype of competitor/diluters (5 L^{-1}). Although competition treatments therefore began at slightly higher total densities (20 L^{-1}), the transient starting conditions impacted densities little over the following 6–8 generations. Instead, competitive ability structured the densities of focal hosts and diluters (see Results). After the focal host and competitor/diluter populations grew for two weeks, we began sampling by mixing and sieving 1

183 L per tank per week (80 μm mesh). After one week of sampling, we added fungal spores (5,000
184 L^{-1}) and continued sampling for seven weeks (~ 7 host generations). Removal of infected
185 individuals (via sampling only 1.7% of tank volume per week) likely did not impact epidemic
186 sizes. We tracked changes in densities of focal hosts, competitor/diluters, and infected hosts
187 using microscopes to count densities and diagnose infections (50X). Only 4 of 6,375
188 competitor/diluters examined were infected (0.06%), confirming their high resistance.

189 190 **Statistics – Linear Models**

191 For all models, we averaged time series for each tank over the 8-week (6–8 host
192 generations) duration. Even if it obscured complex temporal signals of competition or disease
193 transmission, this averaging enabled synthesis of traits, dilution mechanisms, and disease
194 metrics. Mean infection prevalence was calculated as the total number of infections summed
195 across all weeks divided by the total number of hosts sampled during the experiment (rather than
196 the temporal mean of prevalences calculated each week). This method reduced sampling error on
197 prevalence due to extremely low host densities when focal hosts were outcompeted by diluters.

198 Univariate linear models linked trait indices to mesocosm dynamics. Because several
199 patterns exhibited pronounced heteroscedasticity (e.g., see Fig. 3a), we fit the linear models with
200 generalized least squares (GLS). With GLS, we included an additional parameter to allow
201 variance to change with the independent variable, if it improved model fit via likelihood ratio
202 test. These GLS models were implemented using the NLME package in R (Pinheiro & Bates
203 2000). When focal host traits served as independent variables, we also fit complementary mixed
204 models (also using NLME) that assigned random intercepts to each focal host genotype (see
205 Appendix S1).

206 Two sets of linear models evaluated specific linkages between host traits and mean
207 densities or prevalence. The first set tested whether susceptibility (β) directly predicted variation
208 in epidemic size (i.e., mean density or prevalence of infected hosts). It also evaluated whether
209 presence of competitor/diluters (denoted C) modulated these relationships (as $\beta \times C$ interactions).
210 The second set of models mapped competitive ability of focal hosts to the density of
211 competitor/diluters, linked densities of diluters and focal hosts, and evaluated how each density
212 impacted each metric of disease. In other words, this second suite of models mapped the indirect
213 effect of competitive ability on disease, mediated through potential dilution mechanisms.

214 All significant relationships between traits, mean densities, and metrics of disease then
215 became the scaffolding for path models. Because we detected strong impacts competitor/diluters
216 on the density but not prevalence of infections (see Results), we focus our path models on the
217 density of infected hosts.

218

219 **Statistics – Path Models**

220 While the univariate models facilitated a close inspection of each relationship (see Figs.
221 1-3), they also raised two specific questions better suited for path analysis. First, susceptibility
222 and competitive ability covaried, and univariate models suggested that both traits might shape
223 the density of infected hosts. Were both traits actually important, or was one relationship merely
224 a correlational shadow, masked by the other? Path analysis accounted for the covariation
225 between traits and disentangled their simultaneous impacts on disease. Second, did diluters shape
226 disease more strongly through host regulation or encounter reduction? Path analysis partitioned
227 these dilution mechanisms by evaluating the direct versus indirect pathways between the
228 densities of competitor/diluters and infected hosts. We interpreted host regulation as the indirect
229 effects of diluters on infected hosts, mediated by changes in the density of focal hosts (i.e., via
230 competition). In contrast, we interpreted encounter reduction as the direct effects of diluters on
231 infected hosts (not mediated by the density of focal hosts).

232 We fit hierarchical path models using the lavaan package in R (Rosseel 2012) and a
233 maximum likelihood estimator (MLM) that was robust to non-normal standard errors. Mesocosm
234 tank served as the unit of replication ($n = 64$). However, the trait measurements were replicated
235 by focal host genotype ($n = 8$). Therefore, we specified a two-level hierarchical structure with the
236 lavaan survey package (Oberski 2014). Unfortunately, collinearity among parameters prevented
237 the fit of a comprehensive model that included both traits, density of focal hosts, and density of
238 diluters. This undesirable collinearity likely arose due to the covariation among traits and the
239 ‘small’ sample size at the genotype level of replication ($n = 8$). Given this constraint, we fit two
240 complementary hierarchical models. The first model (which excluded the density of focal hosts)
241 disentangled the impacts of each trait on disease. The second model included only one trait
242 (susceptibility) but partitioned the strength of indirect host regulation vs. direct encounter
243 reduction. Tables S2-S4 in Appendix S1 present model fit statistics and parameter estimates.

244

RESULTS

Focal hosts varied in both traits (Fig. 1). Susceptibility, β , ranged $1.8\text{--}5.2 \times 10^{-6}$ (L spore⁻¹ mm⁻²) among the eight genotypes. Hereafter, we rank genotypes by this trait (i.e., the genotype with lowest susceptibility becomes “G1”). The second trait, juvenile growth rate on low resources (the index of competitive ability), ranged $0.13\text{--}0.17$ (day⁻¹). These traits covaried positively but non-significantly (Pearson’s $P = 0.13$). Nevertheless, this covariance became an essential link in the path models. Focal host genotypes also drove divergent outcomes in mesocosms. Appendix S1 presents time series for each genotype: G2 and G8 as illustrative examples (Fig. S1), G1, G3, and G4 (Fig. S2), and G5, G6, and G7 (Fig. S3). However, rather than focus on each genotype individually here, we summarize their mean responses along continuous gradients of their traits.

Linear Model Results

Variation in susceptibility shaped the size of epidemics (Fig. 2). Higher susceptibility fueled both higher mean densities of infected hosts (β effect, $P = 0.0046$; Fig. 2a) and higher infection prevalence (β effect: $P = 0.0008$; Fig. 2b). Mere presence of competitor/diluters did not effect either metric of epidemic size via main effect or interaction (all $P > 0.2$).

Competitive ability of focal hosts—the second trait—governed diluter densities and hence potential dilution mechanisms (Fig. 3). Strongly competing focal hosts constrained competitor/diluters to lower mean densities ($P < 0.0001$; Fig. 3a). In turn, higher densities of competitor/diluters regulated densities of focal hosts ($P = 0.0011$; Fig. 3b; this test includes tanks without any diluters). However, densities of focal hosts and competitor/diluters only significantly impacted one metric of disease. The mean density of infected hosts appeared to be reduced by higher densities of competitor/diluters ($P = 0.0005$; Fig. 3c) and elevated by higher densities of focal hosts (Hd effect: $P = 0.0048$; Fig. 3d). A path model distills the causal structure underlying this result below. In contrast, infection prevalence was not significantly impacted by the density of competitor/diluters ($P = 0.27$; Fig. 3e) or focal hosts (Hd effect: $P = 0.58$; Fig. 3f). Presence of diluters (included as a covariate with focal host density) was not a significant predictor for either metric of disease (both $P > 0.9$). Analyses using the density of focal hosts from week 2 only (when spores were added) mirrored all of these results (see Fig. S4 in Appendix S1).

276 Path Model Results

277 Both path models fit well (see Appendix S1 for diagnostic statistics and parameter
278 estimates). The first model disentangled the impacts of susceptibility and competitive ability on
279 the density of infected hosts (Fig. 4). The traits covaried positively but not significantly ($P =$
280 0.14). Nevertheless, each trait shaped disease through a unique pathway. Higher susceptibility
281 directly elevated disease ($P = 0.004$). In contrast, higher competitive abilities indirectly increased
282 disease by constraining the density of competitor/diluters ($P = 0.015$). In turn, higher densities of
283 diluters reduced the density of infected hosts ($P = 0.006$). Thus, diluters impacted disease more
284 strongly when focal hosts competed weakly, because diluters were more numerous.

285 The second path model partitioned host regulation vs. encounter reduction as drivers of
286 the density of infected hosts (Fig. 5). Intraspecific variation in susceptibility still strongly
287 impacted the size of epidemics ($P = 0.004$). Additionally, higher total densities of focal hosts led
288 to higher densities of infections ($P < 0.001$). However, higher densities of competitor/diluters did
289 not directly lead to a lower density of infected hosts ($P = 0.37$). This weak effect may seem
290 surprising, since it appeared significant when tested univariately (see Fig. 3c). Instead, in this
291 path model, higher densities of competitor/diluters suppressed densities of focal hosts ($P =$
292 0.002), which in turn lowered disease. This causal pathway defines host regulation. Using
293 standardized effect sizes, this indirect effect accounted for 71% of the total effect of diluters on
294 disease. In contrast, the direct effect, i.e., encounter reduction, accounted for only 29%. In other
295 words, the impacts of diluters consuming shared resources (i.e., competition) proved much
296 stronger than the impacts of diluters consuming parasites.

298 DISCUSSION

299 Predicting the size of epidemics remains a central challenge in disease ecology. Host
300 traits like susceptibility can directly fuel epidemics. However, other traits—including
301 competitive ability—may govern epidemic size when other ‘diluter’ taxa can reduce disease.
302 Here, we evaluated a mechanistic, trait-based framework for ‘friendly competition’, a form of
303 local disease dilution combining competitive host regulation and encounter reduction. We
304 measured susceptibility and competitive ability for eight focal host genotypes. Then we
305 challenged each genotype with experimental epidemics, with and without diluters, in multi-
306 generational mesocosms. Finally, we disentangled the impacts of covarying traits and partitioned

307 host regulation vs. encounter reduction using path models. Higher susceptibility directly fueled
308 larger epidemics, both in terms of the density and prevalence of infections. Infection prevalence
309 did not respond significantly to diluters. However, higher densities of diluters strongly reduced
310 the density of infected hosts. Competitive ability—the second trait—indirectly shaped this metric
311 of disease by governing the density of diluters. Finally, diluters reduced the density of infected
312 hosts primarily via host regulation. In other words, their indirect effects on disease (mediated by
313 changes in focal host density) outweighed their direct effects. This traits-based framework
314 strengthens mechanistic foundations for dilution effects and brings us closer to predicting the
315 size of epidemics in diverse communities.

316 Intraspecific variation in susceptibility strongly shaped epidemic size - both the density
317 and prevalence of infections. Though seemingly obvious, few empirical examples link
318 individually-measured traits like susceptibility to epidemic size at the population-level (but see
319 Dwyer & Elkinton 1993; Strauss *et al.* 2015). In this plankton system, clonal variation in
320 susceptibility of the focal host enabled such a test. Infection prevalence responded clearly to
321 variation in susceptibility, but not the density of diluters. In contrast, the density of infected hosts
322 responded to both. Yet in the final path model, susceptibility exerted a larger standardized effect
323 on the density of infected host than the net effect of competitor/diluters. Thus, variation in
324 susceptibility of focal hosts remained essential for predicting the size of epidemics, even in
325 communities with diluters. Previous trait-based frameworks for disease dilution have focused
326 almost exclusively on inter- (rather than intra-) specific variation in susceptibility (but see
327 Pulkkinen 2007; Strauss *et al.* 2015). Such interspecific differences are essential for identifying
328 key diluter taxa (e.g., LoGiudice *et al.* 2003; Johnson *et al.* 2013; Lacroix *et al.* 2014). However,
329 as illustrated here, intraspecific variation in susceptibility can exert even stronger impacts on
330 disease than presence of key diluters. Furthermore, traits like susceptibility frequently evolve
331 during epidemics (Penczykowski, Forde & Duffy 2011). Thus, future theory should further
332 explore the impacts of intraspecific variation on the community ecology of disease, especially
333 when relevant host traits evolve (Decaestecker *et al.* 2013; Strauss *et al.* 2017).

334 The second trait—competitive ability—directly governed host density and indirectly
335 governed disease via host regulation. Both of these impacts manifested along a continuous trait
336 gradient and 6-8 generations of multi-species feedbacks. Specifically, competitor/diluters
337 constrained the density of weakly competing focal hosts, thereby indirectly lowering the density

338 of infections (see Begon 2008). However, these weakly competing focal hosts were driven
339 extinct in some tanks. From the perspective of the focal host, this risk of extinction emphasizes a
340 darker side of competition during epidemics (see also Dallas, Hall & Drake 2016). Moreover,
341 because diluters impacted disease primarily through host regulation (rather than encounter
342 reduction), the dilution effect here was tightly linked to the density cost of competition. Both
343 consequences of competition—disease dilution and risk of extinction—may frequently remain
344 undetected in shorter experiments. However, among experiments that last multiple generations,
345 competitive host regulation frequently becomes a dominant driver of disease (Mitchell, Tilman
346 & Groth 2002; Johnson *et al.* 2012a; Dallas, Hall & Drake 2016). Thus, long-term, trait-based
347 perspectives on competition in other systems might also anticipate dilution via host regulation
348 and the potential density cost suffered by focal hosts.

349 Despite their correlation, both susceptibility and competitive ability of focal hosts
350 influenced epidemic size independently. This biological outcome—and the statistical power of
351 path analysis which revealed it—matter because correlated traits present a general challenge for
352 mechanistic community-disease theory. Multiple traits frequently differ *inter*-specifically
353 between hosts and diluters or amplifiers of disease. For example, susceptibility to trematodes and
354 pace of life covary among amphibian taxa (Johnson *et al.* 2012b); competence for Lyme and
355 production of tick vectors covary among mammals (Randolph & Dobson 2012); susceptibility to
356 virus and production of aphid vectors covary among grasses (Lacroix *et al.* 2014); and
357 susceptibility and encounter rates with chytrid spores covary among tadpoles (Venesky *et al.*
358 2014). When traits that promote disease correlate positively (e.g., competitive ability and
359 susceptibility as here; reviewed in Duncan, Fellous & Kaltz 2011), they can mask each others'
360 potential impacts. Here, we addressed this challenge by partitioning impacts of both traits with
361 path analysis. If important traits correlate negatively, their net impacts also challenge simple
362 prediction, because they can pull epidemic size in opposite directions (see Randolph & Dobson
363 2012). In both scenarios, community theory for disease must continue to grapple with
364 covariation among key traits – both within and among species.

365 The statistical partition of variation in the second path model showed that the strength of
366 host regulation exceeded encounter reduction. How general is this result? Here, it likely reflects
367 the length of our experiment, metric of disease considered, and traits of diluters. As noted above,
368 host regulation became more important than encounter reduction during other multi-generational

369 experiments (Mitchell, Tilman & Groth 2002; Johnson *et al.* 2012a; Dallas, Hall & Drake 2016)
370 and models (Ogden & Tsao 2009). In contrast, shorter experiments might only allow effects of
371 encounter reduction to manifest. Interestingly, host regulation sometimes reduces the density but
372 not prevalence of infections (Johnson *et al.* 2012a; Strauss *et al.* 2016). This can occur when host
373 density correlates strongly with the density but not prevalence of infections (as it did here). In
374 contrast, infection prevalence (which was unrelated to diluters in this experiment) can remain
375 sensitive to encounter reduction, even when it is decoupled from host density (Strauss *et al.*
376 2016). Thus, the partition of dilution mechanisms can also depend on how strongly the chosen
377 metric of disease scales with host density. Finally, it seems likely that certain traits of diluters
378 could increase the strength of encounter reduction relative to host regulation. Here, we focused
379 on traits of focal hosts. However, the partition of dilution mechanisms could also depend on
380 whether diluters reduce host density (Rohr *et al.* 2015), or how rapidly they remove parasites
381 (Venesky *et al.* 2011). More partitions in other systems should test these hypotheses and
382 delineate when host regulation vs. encounter reduction matter more.

383 Our trait-centered framework for friendly competition could be readily expanded. First,
384 parallel experiments could incorporate traits of diluters (Venesky *et al.* 2014) or impacts of
385 predation. Should diluters that consume parasites faster always reduce disease, or only when
386 susceptibility of focal host falls within a certain range (Strauss *et al.* 2015)? When size-selective
387 predators mediate competition between focal hosts and diluters (Strauss *et al.* 2016), do traits
388 like body size become more important than ‘competitive ability’ as measured here? Yet other
389 traits might matter at the metacommunity scale, where much dilution effect research focuses
390 (Ostfeld & Keesing 2000; Johnson *et al.* 2013). Maintenance of diluters in a metacommunity
391 could depend less of local competitive ability and more on dispersal ability or risk of extinction
392 (Joseph *et al.* 2013). Thus, expanding a traits-based framework for friendly competition to a
393 metacommunity scale might predict the sizes of local epidemics and the emergence of a dilution
394 effect across sites. Finally, eco-evolutionary perspectives could grapple with feedbacks between
395 trait diversity in the focal host population (Decaestecker *et al.* 2013), trait-driven impacts on
396 disease and dilution, and rapid evolution imposed by competitor/diluters or parasites (Strauss *et*
397 *al.* 2017). All of these expansions promise exciting frontiers.

398 In summary, intraspecific variation among focal host traits helped predict epidemic size
399 through direct and indirect, dilution-mediated pathways. Using path models, we disentangled

400 how variation in two general, correlated traits—susceptibility and competitive ability—shaped
401 epidemics. Higher susceptibility directly fueled larger epidemics, while stronger competitive
402 ability constrained diluters and indirectly allowed higher densities of infections. The reduction of
403 the density of infected hosts by diluters was driven primarily by competitive host regulation. The
404 second dilution mechanism—encounter reduction—was relatively weak. This empirically
405 evaluated framework provides mechanistic trait-based foundations for dilution effect theory.
406 Such theory brings disease ecologists closer to predicting the size of epidemics in diverse
407 communities.

408

409 **AUTHORS' CONTRIBUTIONS**

410 ATS, SRH, MAD, and CEC designed the study. ATS led trait measurement assays. AMB set up
411 the mesocosm experiment with assistance from ATS and SRH. AMB led sampling. ATS wrote
412 the first draft of the manuscript, and all authors contributed to revisions.

413

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420 **DATA ACCESSIBILITY**

421 All data and scripts have been archived on Dryad Digital Repository: doi:10.5061/dryad.1f7sk

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533

534 SUPPORTING INFORMATION

535 Additional supporting information may be found in the online version of this article.

536 Appendix S1. Supplementary methods and results.

537 Figure S1. Mesocosm dynamics of two illustrative focal host genotypes varying in key traits.

538 Figures S2 & S3. Mesocosm dynamics of all other focal host genotypes.

539 Table S1. Comparisons between GLS and mixed models.

540 Figure S4. Density of focal hosts during week 2 (when parasites were added).

541 Table S2. Test statistics and cutoff criteria for path models.

542 Table S3. Parameter estimates for the first path model (Fig. 4).

543 Table S4. Parameter estimates for the second

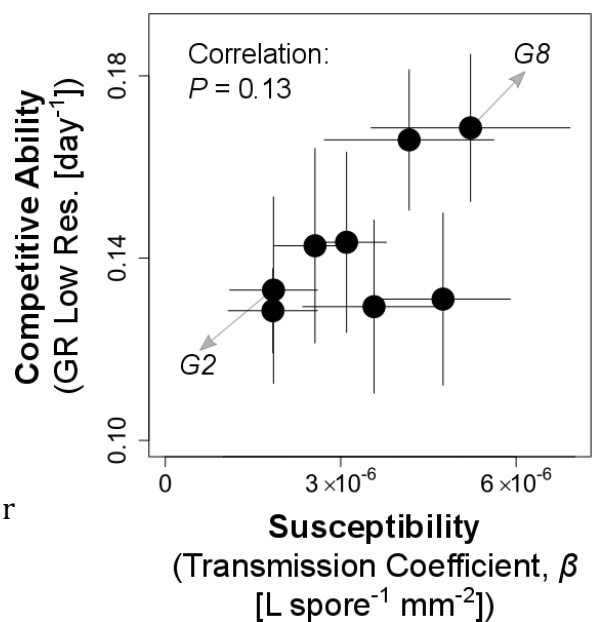
544 path model (Fig. 5).

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547 FIGURES & CAPTIONS

548 **Figure 1.** Two key traits covary among eight
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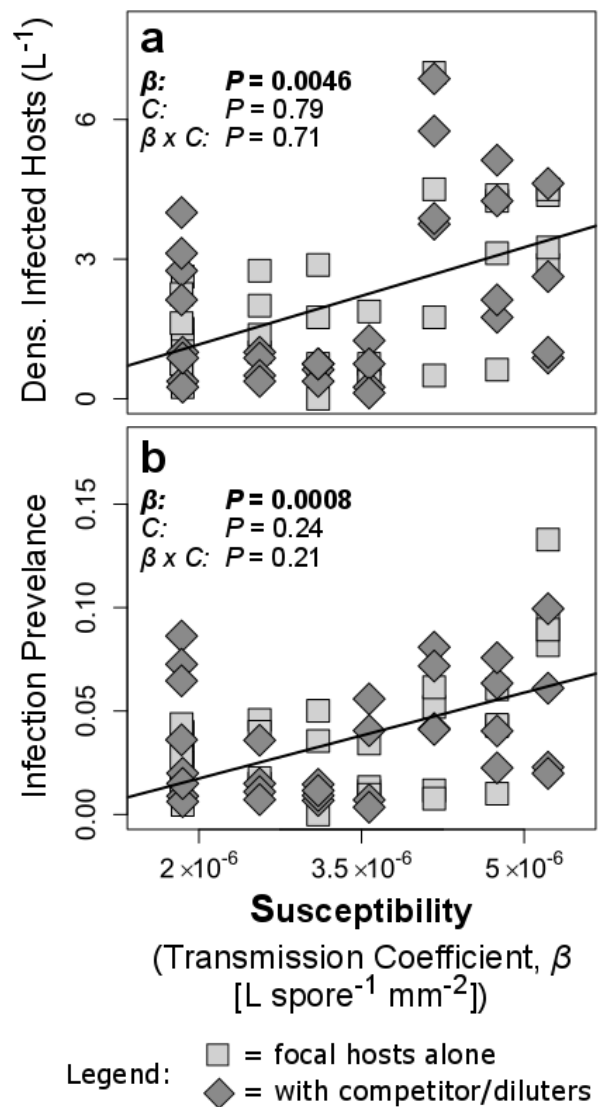


550 *focal host genotypes*. Susceptibility is indexed as a transmission coefficient (β ; measured with
 551 infection assays). Growth rate of juveniles on low resources represents an index of competitive
 552 ability. The traits covary positively but non-significantly ($P = 0.13$). However, both traits and
 553 their covariation become foundations for linear (Figs. 2–3) and path models (Figs. 4–5).
 554 Genotypes are named according to variation in susceptibility (along x axis; Figs. S1-S3 in
 555 Appendix S1 present each genotype’s time series in the mesocosm experiment). Error bars are
 556 bootstrapped standard errors.

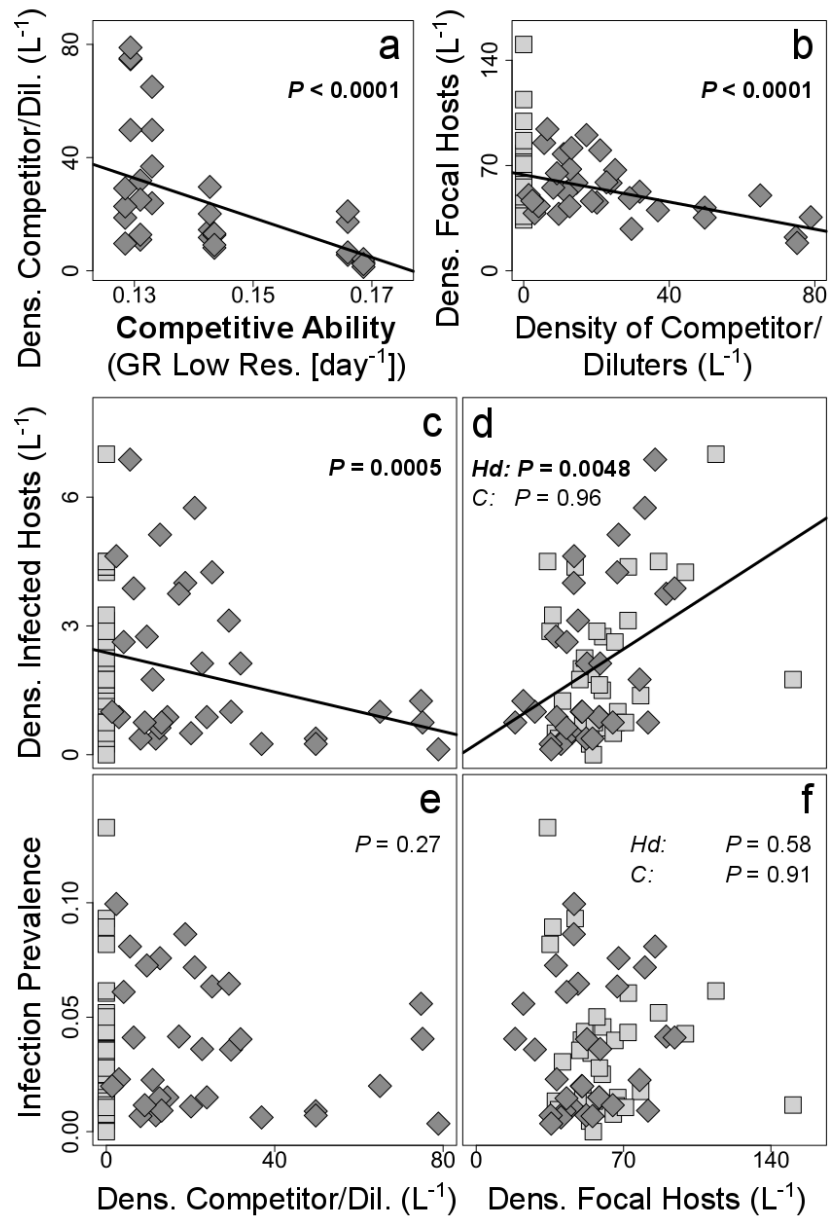
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558 **Figure 2.** *Variation in susceptibility*
 559 *predicts the size of epidemics*. Points are
 560 temporal averages for each mesocosm tank.
 561 Higher susceptibility fuels both **a)** higher
 562 mean densities of infected hosts and **b)**
 563 higher mean infection prevalence (β effects;
 564 solid lines). Neither metric of epidemic size
 565 is effected by the mere presence of
 566 competitor/diluters (C), or its interaction
 567 with susceptibility ($\beta \times C$). P values are fits
 568 of linear models. Key: squares = focal hosts
 569 alone; diamonds = with competitor/diluters.

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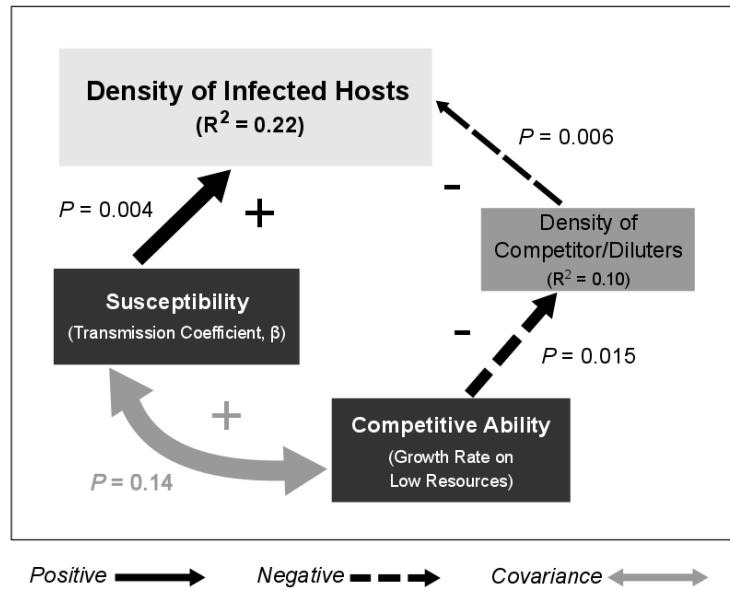


571 **Figure 3.** Variation in
 572 competitive ability
 573 structures the densities of
 574 diluters and focal hosts,
 575 and both correlate with
 576 the density of infected
 577 hosts. **a)** Genotypes of
 578 focal hosts with higher
 579 competitive abilities
 580 constrain
 581 competitor/diluters to
 582 lower densities. **b)** Higher
 583 densities of diluters
 584 reduce the density of
 585 focal hosts. In turn, the
 586 density of infected hosts
 587 is both **c)** lowered by
 588 higher densities of
 589 competitor/diluters and **d)**
 590 elevated by higher
 591 densities of focal hosts. In
 592 contrast, infection
 593 prevalence is sensitive to
 594 neither densities of **e)**
 595 competitor/diluters nor **f)** focal hosts. *P* values are fits of linear models. Key: *C* = presence of
 596 competitor/diluters; *Hd* = density of focal hosts; squares = focal hosts alone; diamonds = with
 597 competitor/diluters.
 598



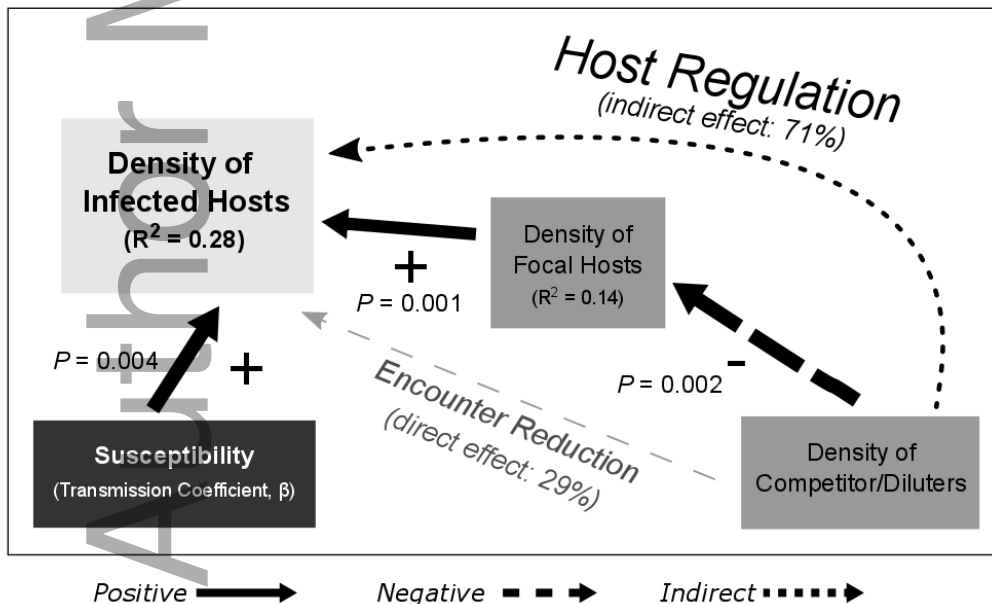
Legend: \square = focal hosts alone \blacklozenge = with competitor/diluters

599 **Figure 4.** Both covarying focal
 600 host traits simultaneously
 601 govern the density of infected
 602 hosts. Higher susceptibility
 603 directly fuels larger epidemics
 604 directly (see Fig. 2a). In
 605 contrast, stronger competitive
 606 ability enables epidemics
 607 indirectly by limiting the
 608 density of diluters (see Fig. 3a).



609 In turn, higher densities of
 610 diluters reduce the density of
 611 infected hosts. These impacts of diluters could be due to host regulation, encounter reduction, or
 612 both (partitioned in Fig. 5). Key: solid = positive coefficients; dashed = negative coefficients;
 613 two-headed arrow = covariance between traits; arrow weights = standardized effect sizes.

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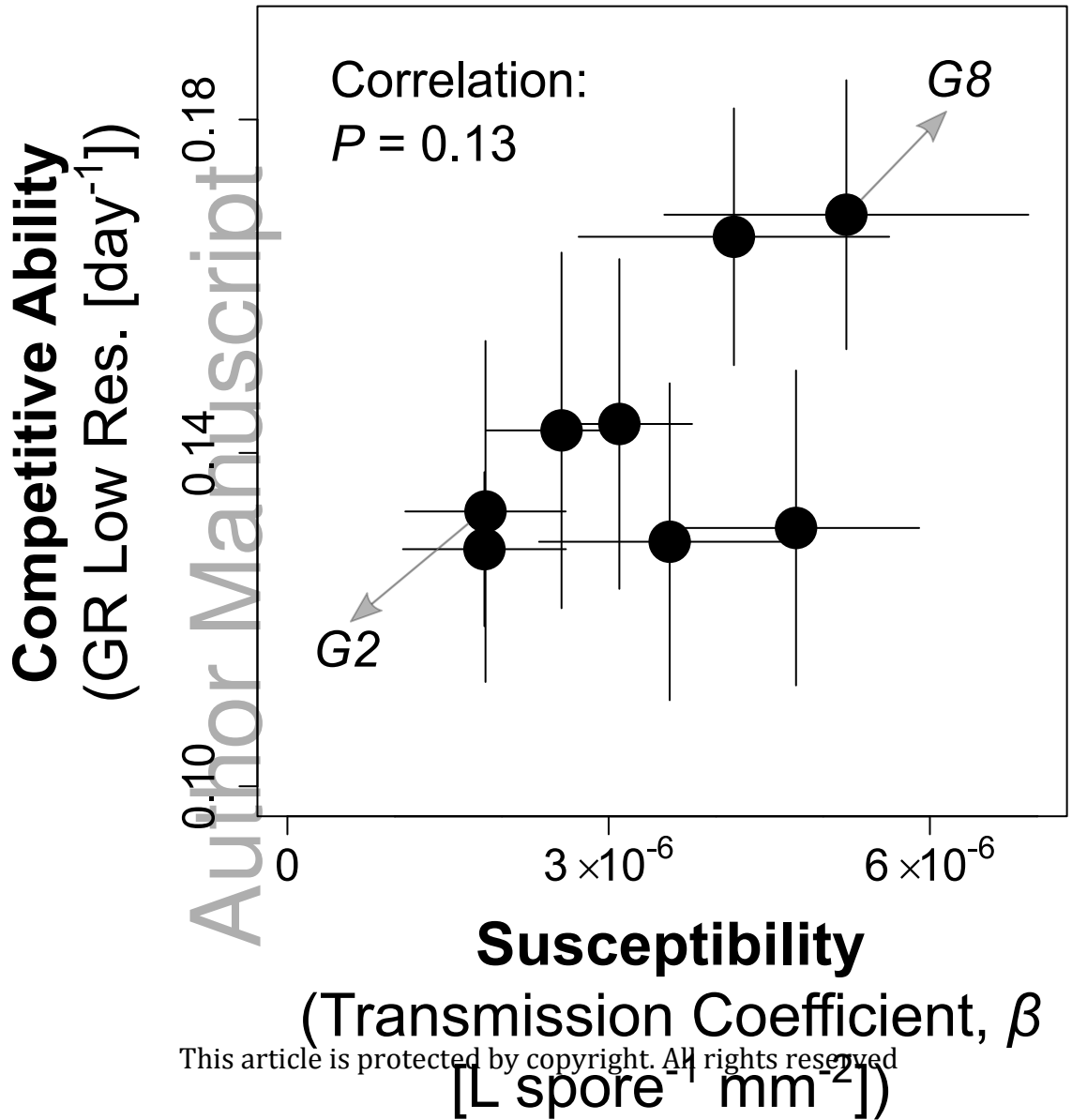


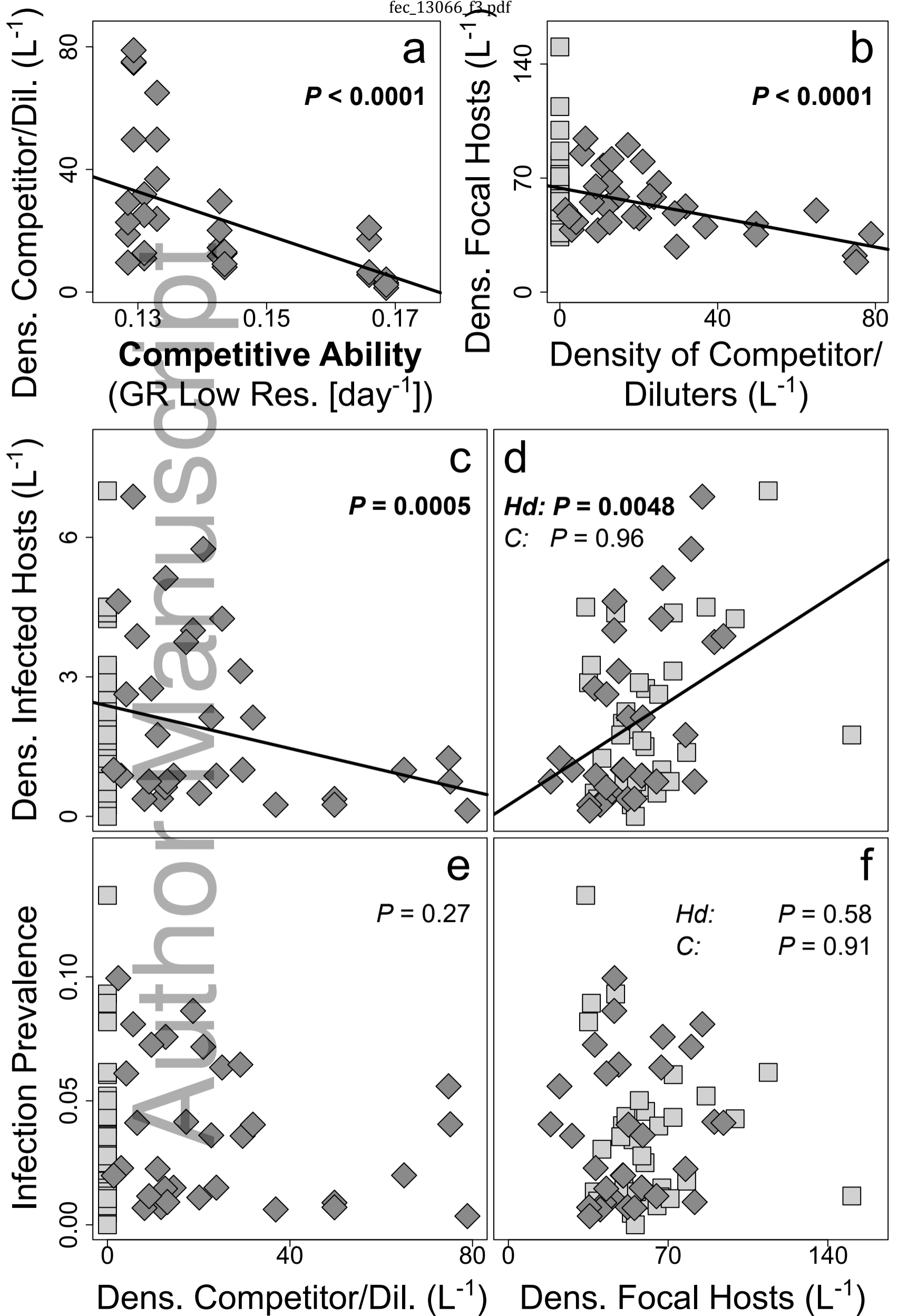
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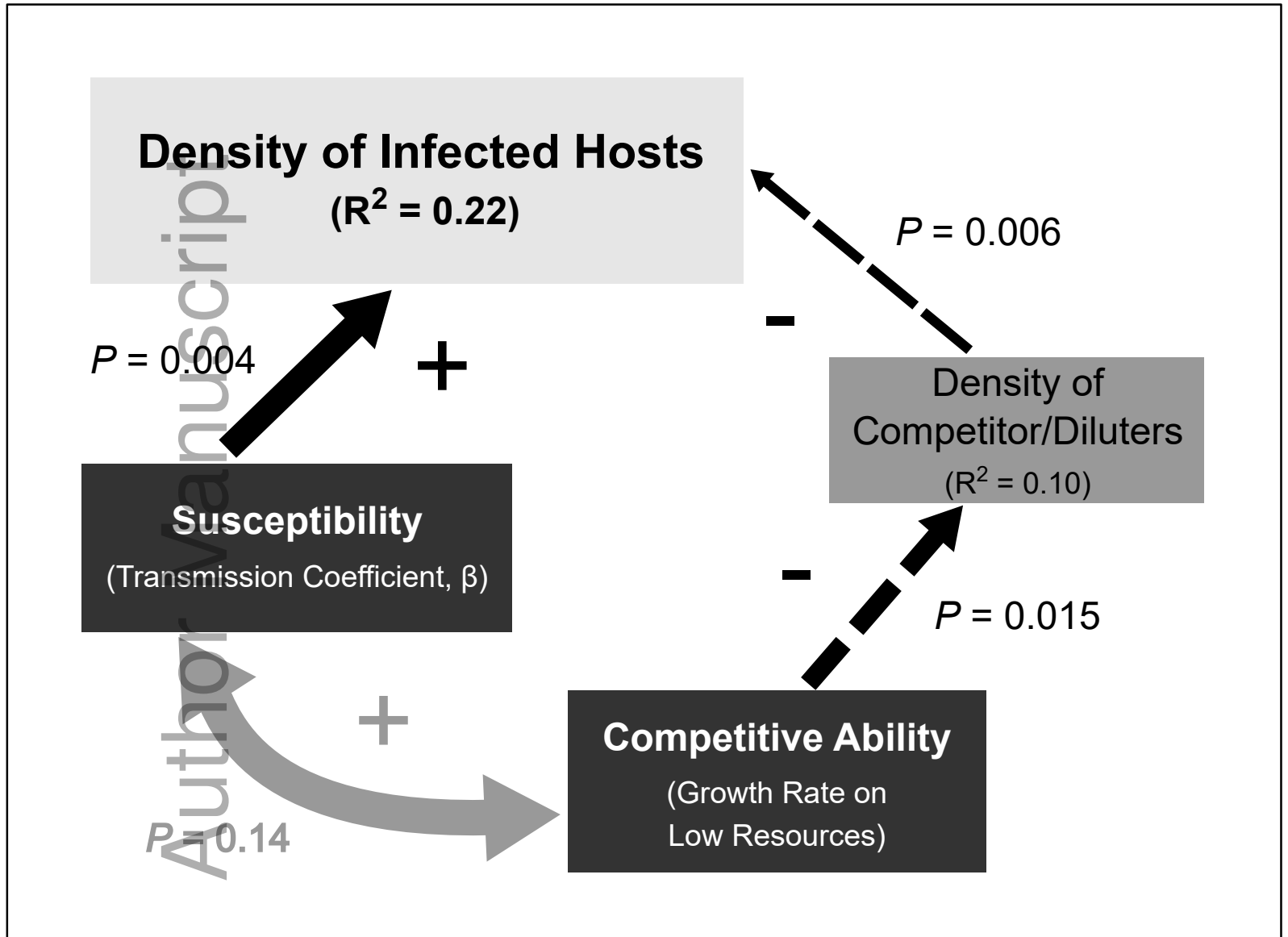
616 **Figure 5.** Partitioning two dilution mechanisms: Does host regulation or encounter reduction

617 *reduce the density of infected hosts?* Higher total densities of focal hosts lead to a higher density
618 of infected hosts (plotted in Fig. 3d). However, higher densities of competitor/diluters did not
619 directly lead to a lower density of infected hosts (despite the apparent relationship in Fig. 3c).
620 This direct effect, i.e., encounter reduction, explained a relatively small proportion (29%) of the
621 net effect of diluters on disease. Instead, higher densities of competitor/diluters suppressed
622 densities of focal hosts, which in turn lowered disease. This indirect effect, i.e., host regulation,
623 explained the majority (71%) of the impact of diluters on disease. In addition to this dilution
624 effect, variation in susceptibility remained an important driver of epidemic size. Key: solid =
625 positive coefficients; dashed = negative coefficients; dotted = indirect effect; arrow weights =
626 standardized effect sizes.

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Negative \dashrightarrow

Covariance \longleftrightarrow

