

Evolution of parasite transmission dispersion

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April 26, 2024

Abstract

An open question in epidemiology is why transmission is often overdispersed, meaning that most new infections are driven by few infected individuals. For example, around 10% of COVID-19 cases cause 80% of new COVID-19 cases. This overdispersion in pathogen transmission is likely driven by intrinsic biological heterogeneity among hosts, i.e. variability in SARS-CoV-2 viral loads. However, host heterogeneity could also indirectly increase transmission dispersion by driving pathogen adaptation. Specifically, transmission variation among hosts could drive pathogen specialization to highly-infectious hosts. Adaptation to rare, highly-infectious hosts could amplify transmission dispersion by simultaneously decreasing transmission from common, less-infectious hosts. This study considers whether increased transmission dispersion can be, in part, an emergent property of parasite adaptation to heterogeneous host populations. We develop a mathematical model using a Price equation framework to address this question that follows the epidemiological and evolutionary dynamics of a general host-parasite system. The results predict that parasite adaptation to heterogeneous host populations drives high transmission dispersion early in epidemics. Further, parasite adaptation can maintain increased transmission dispersion at endemic equilibria as long as virulence differs between hosts in a heterogeneous population. More broadly, this study provides a framework for predicting how parasite adaptation determines transmission dispersion for emerging and re-emerging infectious disease.

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Statement of Authorship: HM conceived of the presented idea and developed the theoretical framework; HM conducted the mathematical analysis and performed numerical simulations; RR supervised the mathematical analysis; All authors wrote the manuscript and gave final approval for publication and agree to be held accountable for the work performed therein.

Data Accessibility: Code is available on Zenodo:

<https://zenodo.org/records/10203626>

Short title: Evolution of parasite transmission dispersion

Keywords: evolutionary ecology, parasite, epidemiology, adaptation, transmission dispersion, superspreading

Article Type: Perspective

Abstract word count: 200

Main text word count: 4231

Text box word count: 0

Number of references: 38

Number of figures: 9

Number of tables: 1

Number of text boxes: 0

1 Abstract

An open question in epidemiology is why transmission is often overdispersed, meaning that most new infections are driven by few infected individuals. For example, around 10% of COVID-19 cases cause 80% of new COVID-19 cases. This overdispersion in parasite transmission is likely driven by intrinsic heterogeneity among hosts, i.e. variable SARS-CoV-2 viral loads. However, host heterogeneity could also indirectly increase transmission dispersion by driving parasite adaptation. Specifically, transmission variation among hosts could drive parasite specialization to highly-infectious hosts. Adaptation to rare, highly-infectious hosts could amplify transmission dispersion by simultaneously decreasing transmission from common, less-infectious hosts. This study considers whether increased transmission dispersion can be, in part, an emergent property of parasite adaptation to heterogeneous host populations. We develop a mathematical model using a Price equation framework to address this question that follows the epidemiological and evolutionary dynamics of a general host-parasite system. The results predict that parasite adaptation to heterogeneous host populations drives high transmission dispersion early in epidemics. Further, parasite adaptation can maintain increased transmission dispersion at endemic equilibria if virulence differs between hosts in a heterogeneous population. More broadly, this study provides a framework for predicting how parasite adaptation determines transmission dispersion for emerging and re-emerging infectious disease.

2 Introduction

Transmission events are often overdispersed during epidemics meaning that the majority of infections are transmitted from a minority of infected individuals (Galvani and May 2005). For example, less than 20% of cases cause 80% of new infections in typical outbreaks of Measles and COVID-19 (Endo et al. 2020; Galvani and May 2005; Woolhouse et al 1997). This increased dispersion in parasite transmission is likely driven to a considerable extent by intrinsic biological heterogeneity among hosts (Regoes, Nowak, and Bonhoeffer 2000; VanderWaal and Ezenwa 2016). Consequently, the majority of studies on transmission dispersion focus on the direct impact of host heterogeneity (Lloyd-Smith et al 2005; VanderWaal and Ezenwa 2016; Woolhouse et al 1997). However, this perspective overlooks a dynamic and potentially important additional factor: the evolution of parasites in response to host heterogeneity.

Could parasite adaptation to diverse host types amplify transmission dispersion? This Ideas and Perspectives article proposes that host heterogeneity may not only drive transmission dispersion directly by making some hosts more infectious than others but that it could also indirectly enhance transmission dispersion by generating selection pressure for parasites that are more transmissible on some hosts than others. How this could work is that parasite could specialize on hosts that drive more onward transmission when adapting to heterogeneous host populations composed of individuals whose infectiousness and morbidity vary following infection. In natural populations, differences in symptomatic responses to large within-host parasite densities can make some hosts more infectious than others (Jones et al 2021; VanderWaal and Ezenwa 2016; Yang et al 2021), *e.g.* asymptomatic children have significantly higher SARS-CoV2 viral loads than hospitalized adults (Yonker et al 2021). Previous work has shown that distinct host types select for parasites with different virulence levels (parasite-induced mortality), which decreases the transmission of evolved parasites infecting novel host types (Kubinak and Potts 2013; White et al 2020). In the context of how parasite adaptation could impact transmission dispersion, selection could reduce the proportion of infections responsible for most new cases if parasites evolve high within-host growth rates to exploit hosts supporting high parasite densities despite low mortality that simultaneously decrease transmission in other host types by driving high mortality following infection.

This article begins by outlining a simple epidemiological model to demonstrate how host heterogeneity alone contributes to transmission dispersion. We then show how parasite adaptation could create a situation where transmission is increasingly dominated by a few, highly efficient host-parasite interactions using the Price equation framework developed in Day and Gandon 2007. The advantage of this approach is that the epidemiological and evolutionary dynamics occur on the same time scale (in contrast to other approaches such as adaptive dynamics Dieckmann, Metz, Sabelis, and Sigmund 2002.) The framework thus predicts how parasites adapt throughout epidemics rather than only providing predictions at epidemic equilibria.

The framework used here predicts that parasite adaptation to heterogeneous host populations can drive high transmission dispersion early in epidemics and can maintain increased transmission dispersion at endemic equilibria. In addition, large differences in host quality are predicted to select for parasites that drive high transmission dispersion. More broadly, this study provides a framework for predicting how parasite adaptation determines transmission dispersion for emerging and re-emerging infectious disease. This article not only highlights the complex interplay between host heterogeneity and parasite evolution but also charts a course for future research into the mechanisms underlying transmission dispersion in both emerging and re-emerging infectious diseases.

3 How host heterogeneity impacts transmission dispersion

We first introduce a basic epidemiological model with two distinct host types to demonstrate how host heterogeneity impacts parasite transmission dispersion in the absence of parasite evolution. The heterogeneous host population is composed of two host types with distinct transmission and virulence functions, both of which depend on the within-host growth rate of the parasite (ϵ). Thus, parasites with identical within-host growth rates transmit and increase host mortality at different rates in the two host types. This is modelled by assuming that susceptible and infectious hosts that are "high yield" from the perspective of the parasite (s_H, i_H) have high transmission and/or low virulence while infected such that parasite reproductive fitness from these hosts is high. Conversely, hosts that are "low yield" from the perspective of the parasite (s_L, i_L) have low transmission and/or high virulence following infection such that parasite reproductive fitness from these hosts is low. The within-host growth rate of the parasite is constant in this first model, but will be the trait under selection in the model that includes parasite adaptation. The epidemiological dynamics are given by

$$\frac{ds_H}{dt} = \lambda(1 - p) - (\bar{\beta}_H i_H(t) + \bar{\beta}_L i_L(t))s_H(t) - \delta s_H(t), \quad (1a)$$

$$\frac{ds_L}{dt} = \lambda p - (\bar{\beta}_H i_H(t) + \bar{\beta}_L i_L(t))s_L(t) - \delta s_L(t), \quad (1b)$$

$$\frac{di_H}{dt} = (\bar{\beta}_H i_H(t) + \bar{\beta}_L i_L(t))s_H(t) - (\delta + \gamma + \bar{\alpha}_H)i_H(t), \quad (1c)$$

$$\frac{di_L}{dt} = (\bar{\beta}_H i_H(t) + \bar{\beta}_L i_L(t))s_L(t) - (\delta + \gamma + \bar{\alpha}_L)i_L(t). \quad (1d)$$

where λ is the rate that new susceptible hosts enter the system - a proportion p of which are low yield and a proportion $(1 - p)$ of which are high yield - δ is the natural host mortality rate and γ is the rate that hosts recover from infection. $\bar{\alpha}_i$ is the additional mortality suffered by infected hosts (virulence) and $\bar{\beta}_i$ is the average transmission rate in each host type ($i = H, L$). Both $\bar{\alpha}_i$ and $\bar{\beta}_i$ are functions of the within-host growth rate of the parasite (ϵ).

The transmission and virulence functions differ between high and low yield hosts such that parasites with the same within-host growth rate trait value will transmit and increase host mortality at different rates in the two host types. In line with previous theory on the evolution of parasite virulence (Alizon, Hurford, Mideo, and Van Baalen 2009; Alizon and van Baalen 2005), the model assumes a trade-off between transmission and virulence such that transmission ($\bar{\beta}_i$) and virulence ($\bar{\alpha}_i$) functions increase as ϵ increases. The function for the transmission rate is

$$\bar{\beta}_i(\epsilon, c_i, x, \rho) = \rho(c_i + \epsilon^x) \quad (2)$$

where x controls the concavity of the transmission function as the within-host growth rate increases, ρ is a scaling parameter and c_i ($i = H, L$) is the transmission set point which can vary between high and low yield hosts. We assume $x < 1$ to study how concave trade-offs between transmission and within-host parasite growth rates impact selection. $c_i \geq 0$ and is higher in high yield hosts when transmission varies between the two host types. Note that transmission would occur even when $c_i > 0$ and $\epsilon = 0$ (no parasite), however this is not a practical concern in this study as evolutionary analyses always assume positive starting values of ϵ

72 and selection never drives ϵ to 0 (see Results section).

73 The function for virulence is

$$\bar{\alpha}_i(\epsilon, y_i) = y_i \epsilon \quad (3)$$

74 where y_i is the rate that virulence increases as the within-host growth rate increases. y_i is higher in low yield
75 hosts when virulence varies between the two host types.

76 The expected number of new infections produced by an infected host is defined following Gandon 2004 and
77 Gandon and Day 2009 to study how host heterogeneity impacts parasite fitness and transmission dispersion
78 over time.

$$R_e(t) = \frac{\bar{\beta}_H s_H(t)}{\delta + \gamma + \bar{\alpha}_H} + \frac{\bar{\beta}_L s_L(t)}{\delta + \gamma + \bar{\alpha}_L} \quad (4)$$

Transmission dispersion can then be defined as the variance-to-mean ratio (following Lloyd-Smith et al 2005)
in R_e from each host type

$$vmr(R_e(t)) = var(R_e(t))/R_e(t) \quad (5)$$

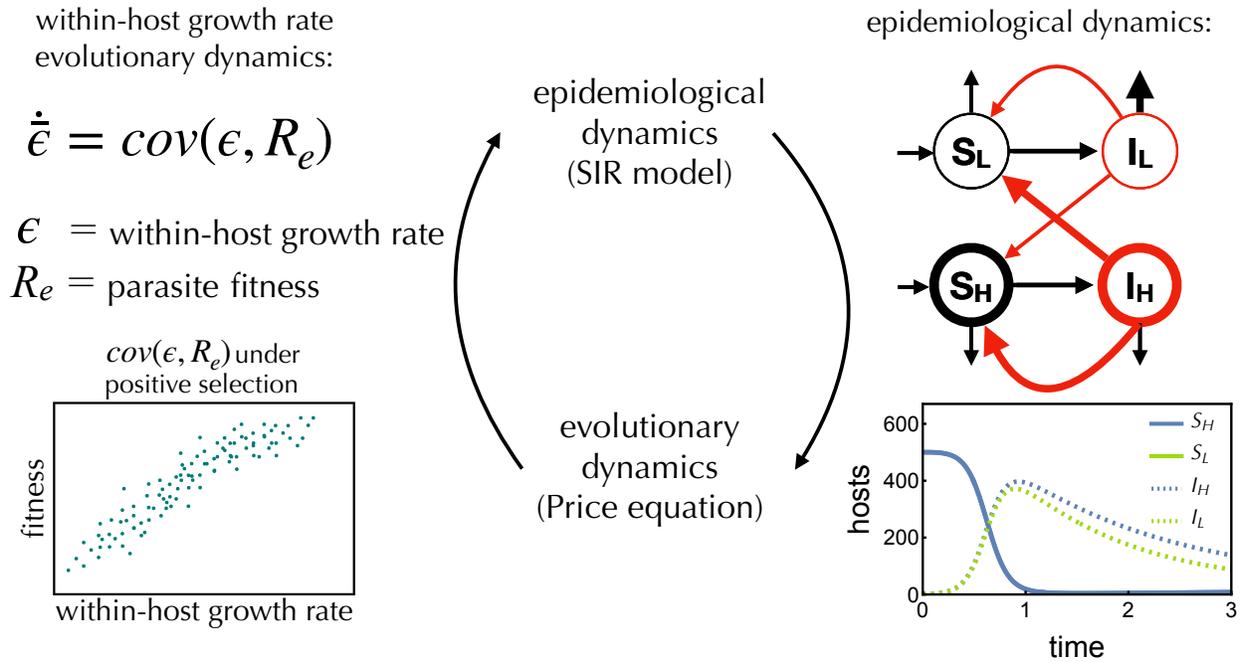


Figure 1: The modelling framework follows the epidemiological dynamics of the host population (using a SI model (Anderson and May 1991)) and the evolutionary dynamics of the parasite within-host parasite growth rate (using the Price equation (Price et al. 1970)). The epidemiological dynamics impact selection on the within-host parasite growth rate. The value of the within-host parasite growth rate impacts how quickly infected hosts transmit the parasite and die from the infection, thus impacting the epidemiological dynamics. The form of the Price equation used here ignores the impact of mutation. The plot in the bottom left shows an example of positive selection on the within-host growth rate as the trait is positively correlated with fitness. The plot in the bottom right shows an example of the epidemiological dynamics for high and low yield susceptible and infectious hosts.

where

$$\begin{aligned}
\text{var}(R_e(t)) &= \sum_j i_j/i(R_e(t) - R_{ej}(t))^2 \\
&= i_H(t)/(i_H(t) + i_L(t))(R_e(t) - \frac{\bar{\beta}_H(s_H(t) + s_L(t))}{\delta + \gamma + \bar{\alpha}_H})^2 \\
&\quad + i_L(t)/(i_H(t) + i_L(t))(R_e(t) - \frac{\bar{\beta}_L(s_H(t) + s_L(t))}{\delta + \gamma + \bar{\alpha}_L})^2
\end{aligned} \tag{6}$$

79 In line with previous work (Anderson, Gupta, and Ng 1990; Mandal, Sarkar, and Sinha 2011; Wonham,
80 Lewis, Rencławowicz, and van den Driessche 2006; ?), the first model demonstrates that transmission
81 dispersion increases as difference in host quality increases (Fig.2). As expected, transmission dispersion is
82 equal to zero when the host population is composed entirely of low yield or high yield hosts. The variance to
83 mean ratio is high when high yield hosts contribute more infections than low yield hosts.

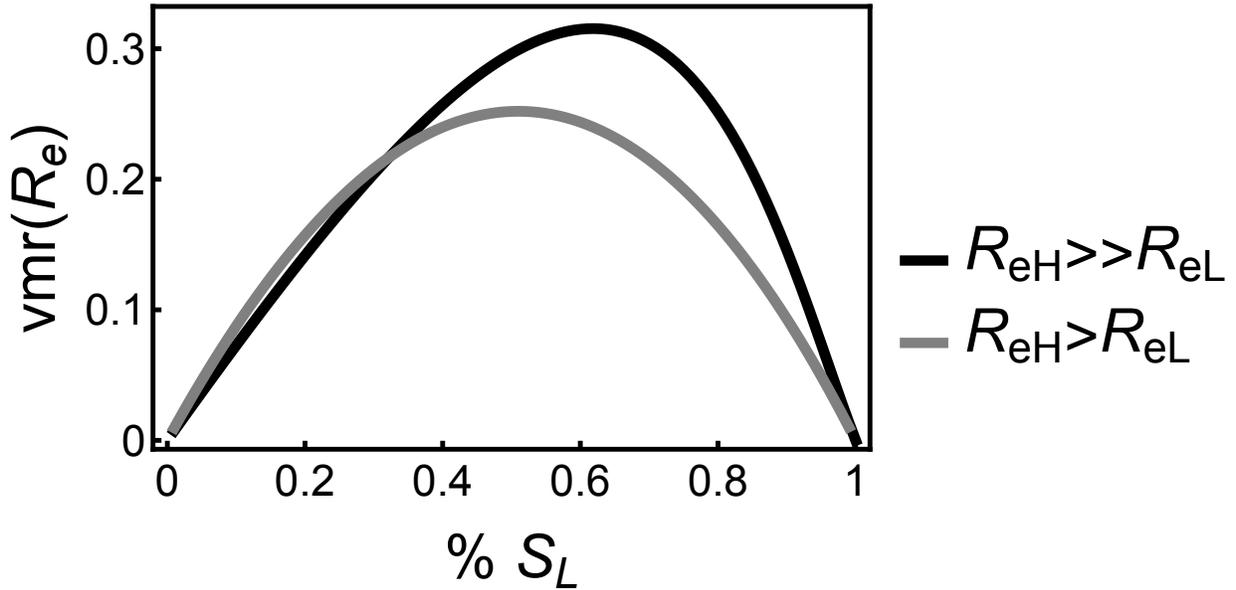


Figure 2: Large differences in host quality drive increased transmission dispersion ($vmr(R_e)$). Transmission dispersion is high when the host population is roughly equally split between high and low yield hosts ($\% S_L \approx 0.5$.) and equal to zero when the host population is entirely high or low yield hosts. In the case where the high yield host is slightly more productive from the perspective of the parasite ($R_{eH} > R_{eL}$), $c_H = 0.1, c_L = 0, y_H = 0.1, y_L = 0.2$, while in the case where the high yield host is much more productive from the perspective of the parasite ($R_{eH} \gg R_{eL}$), $c_H = 1, c_L = 0, y_H = 0.1, y_L = 1$. For both cases: $x = 0.5, \lambda = 50, \delta = 0.02, \gamma = 0.6, \rho = 10^{-2}, \epsilon = 0.25$.

84 4 A model to study how parasite adaptation impacts transmission dispersion

85 In this next section we outline a model to study how parasite adaptation impacts transmission dispersion
86 over time. To do so we use a modelling framework that was developed in Day and Gandon 2007 that
87 follows the epidemiological and evolutionary dynamics of a host-parasite system. In this framework, the

88 epidemiological dynamics are linked to the evolutionary dynamics through the Price equation and thus
 89 the impact of parasite adaptation on epidemiological dynamics is explicitly considered and vice versa (Fig.
 90 1). To get at our question, we use an extension of this original model that considers a heterogeneous host
 91 population originally developed to investigate the evolutionary dynamics of parasite adapting to partially
 92 vaccinated host populations (Gandon and Day 2007). In our case, we consider the evolution of a polymorphic
 93 parasite population infecting a heterogeneous host population in which the number of infections resulting
 94 from each host type can vary due to pre-determined biological factors. The trait under selection in this study
 95 is the within-host growth rate of the parasite (ϵ).

96 Following Day and Gandon 2007, this study uses a form of the Price equation that ignores the impact of
 97 mutation to track the mean within-host parasite growth rate (ϵ) in the parasite population in each host type

$$\frac{d\bar{\epsilon}_H}{dt} = cov(\epsilon, r_{HH}) + \frac{i_L}{i_H}(\bar{r}_{LH}(\bar{\epsilon}_L - \bar{\epsilon}_H) + cov(\epsilon, r_{LH})), \quad (7a)$$

$$\frac{d\bar{\epsilon}_L}{dt} = cov(\epsilon, r_{LL}) + \frac{i_H}{i_L}(\bar{r}_{HL}(\bar{\epsilon}_H - \bar{\epsilon}_L) + cov(\epsilon, r_{HL})). \quad (7b)$$

98 where r_{ij} terms specify the per-capita rate of production of new infections in host type j from host type i (*i.e.*
 99 fitness). $cov(\epsilon, r_{ij})$ is the covariance between the within-host growth rate and fitness for each transmission
 100 scenario. r_{ij} terms are given by

$$r_{HH} = \bar{\beta}_H s_H(t) - (\delta + \gamma + \bar{\alpha}_H), \quad (8a)$$

$$r_{HL} = \bar{\beta}_L s_H(t), \quad (8b)$$

$$r_{LL} = \bar{\beta}_L s_L(t) - (\delta + \gamma + \bar{\alpha}_L), \quad (8c)$$

$$r_{LH} = \bar{\beta}_H s_L(t). \quad (8d)$$

101 The first term in equations (4a) and (4b) describes the impact of selection on $\bar{\epsilon}_H$ and $\bar{\epsilon}_L$ from infections
 102 that pass exclusively within one host type (*e.g.* i_H infects s_H). The second and third terms in (4a) and
 103 (4b) describe the impact of transmission between host types (*e.g.* i_H infects s_L) and are thus weighted by
 104 the relative sizes of the two host type populations. The second term in (4a) and (4b) expresses the impact
 105 transmission between host types has on $\bar{\epsilon}_H$ and $\bar{\epsilon}_L$ in the absence of selection. This term will have an impact
 106 on trait values when $\bar{\epsilon}_H$ and $\bar{\epsilon}_L$ differ. The third term in (4a) and (4b) accounts for any selection that occurs
 107 during the transmission process between host types, *e.g.* parasites with high ϵ enjoy high transmission and
 108 are over-represented among strains that infect s_L from i_H . This term accounts for the fact that transmission
 109 between host types can impact ϵ trait values even if $\bar{\epsilon}_H$ and $\bar{\epsilon}_L$ are equal.

110

111 (4a) and (4b) can be expanded by assuming $cov(\epsilon, r_{ij}) \approx var_i(\epsilon)(\frac{dr}{d\epsilon})$, which yields

$$\frac{d\bar{\epsilon}_H}{dt} = var_H(\epsilon)\left(\frac{d\beta_H}{d\epsilon} s_H(t) - \frac{d\alpha_H}{d\epsilon}\right) + \frac{i_L(t)}{i_H(t)}(\bar{\beta}_L s_H(t)(\epsilon_L - \epsilon_H) + var_L(\epsilon)\frac{d\beta_L}{d\epsilon} s_H(t)),$$

$$\frac{d\bar{\epsilon}_L}{dt} = var_L(\epsilon)\left(\frac{d\beta_L}{d\epsilon} s_L(t) - \frac{d\alpha_L}{d\epsilon}\right) + \frac{i_H(t)}{i_L(t)}(\bar{\beta}_H s_L(t)(\epsilon_H - \epsilon_L) + var_H(\epsilon)\frac{d\beta_H}{d\epsilon} s_L(t)).$$

112 which using Eq. 2 expands to

$$\frac{d\bar{\epsilon}_H}{dt} = var_H(\epsilon)(\rho x \epsilon_H^{x-1} s_H(t) - y_H) + \frac{i_L(t)}{i_H(t)}(\bar{\beta}_L s_H(t)(\epsilon_L - \epsilon_H) + var_L(\epsilon)\rho x \epsilon_L^{x-1} s_H(t)), \quad (9a)$$

$$\frac{d\bar{\epsilon}_L}{dt} = var_L(\epsilon)(\rho x \epsilon_L^{x-1} s_L(t) - y_L) + \frac{i_H(t)}{i_L(t)}(\bar{\beta}_H s_L(t)(\epsilon_H - \epsilon_L) + var_H(\epsilon)\rho x \epsilon_H^{x-1} s_L(t)). \quad (9b)$$

113 5 Results

114 The goal of this study is to determine how parasite adaptation to heterogeneous host populations impacts
 115 transmission dispersion. That is, does parasite adaptation skew the contribution that infections from each
 116 host type make to parasite fitness (measured as an increase in $\text{vmr}(R_e)$)? To do so, we use a mathematical
 117 modelling framework that follows epidemiological dynamics coupled to the evolutionary dynamics of the
 118 parasite trait under study: the within-host growth rate (ϵ). The framework differs from adaptive dynamics in
 119 that the epidemiological and evolutionary dynamics proceed simultaneously such that the epidemiological
 120 dynamics need not be at an equilibrium for new parasite strains to emerge, (*i.e.* no time-scale separation
 121 between the epidemiological and evolutionary dynamics.) The epidemiological dynamics follow susceptible
 122 and infectious host densities (Eq. 1a - 1d) which determine parasite reproductive fitness (measured as R_e , Eq.
 123 6) and drive the adaptation of the within-host growth rate (Eq 5a, 5b). The epidemiological and evolutionary
 124 dynamics are linked such that changes in the within-host growth rate impact the transmission and virulence
 125 of infected hosts and thus the epidemiological dynamics. The host population is heterogeneous as there
 126 are two distinct host types: (1) a high yield host that has high transmission and/or low virulence following
 127 infection such that parasite reproductive fitness from these hosts is high and (2) a low yield host that has low
 128 transmission and/or high virulence following infection such that parasite reproductive fitness from these
 129 hosts is low. We predict how different host population compositions (*e.g.* proportion of low yield hosts,
 130 differences in transmission set points) impact parasite evolution, which changes the reproductive fitness
 131 of the parasite from both host types and thus the relative contribution both hosts make to parasite fitness,
 132 measured as transmission dispersion (Eq. 9.) The model predicts that transmission dispersion is highest
 133 when the host population is mostly composed of low yield hosts and when the difference in quality between
 134 the two host types is large.

135 5.1 How transmission dispersion changes over the course of an epidemic

136 Susceptible host density is high early in an epidemic and quickly drops as hosts become infected (Fig.3A).
 137 Infected host density peaks relatively early in an epidemic and then drops before eventually rebounding
 138 and settling to an endemic equilibrium (Fig.3B). Infected host density drops as the influx of new infections
 139 decreases from the drop in susceptible host availability and as hosts recover and die from the infection.
 140 Low yield infected hosts maintain lower densities than high yield infected hosts when they experience
 141 higher virulence than high yield hosts. Parasite adaptation changes the dynamics that occur in the absence
 142 of evolution by driving an earlier decrease in susceptible host density (Fig.3C) and maintaining lower
 143 equilibrium host densities (Fig.3D).

144 Two selective forces act on the within-host growth rate (ϵ): (1) susceptible host density selects for increased
 145 within-host growth rate - with strength proportional to susceptible host density (2) virulence selects for
 146 decreased ϵ - with strength proportional to the rate that virulence increases as ϵ increases (y_i). Early in an
 147 epidemic, there is strong selection for high ϵ as susceptible hosts are abundant (Fig. 4A, 3C). Selection for

148 high ϵ is approximately the same in both host types during this period ($\epsilon_H \approx \epsilon_L$). As susceptible hosts
149 are depleted, selection for increased ϵ disappears. Low ϵ is then adaptive as negative selection from the
150 cost of virulence outweighs the weak positive selection from the few remaining susceptible hosts (Fig. 4B,
151 3D). Selection for higher ϵ resumes as the susceptible host population is replenished and continues until
152 the epidemiological and evolutionary dynamics reach an equilibrium. Note that the extent that ϵ_H and ϵ_L
153 trait values diverge while approaching the evolutionary equilibrium increases as the difference in virulence
154 between hosts increases (this can also be predicted from (Eq. 3a, 3b)).

155 The dynamics of parasite fitness (R_e) and transmission dispersion ($vmr(R_e)$) are determined by both
156 the epidemiological dynamics and the evolutionary dynamics of the within-host growth rate (Figure 5).
157 Parasite fitness in both high and low yield hosts is high early in epidemics regardless of parasite adaptation
158 as susceptible host density is high (R_{eH}, R_{eL} in Fig. 5A, 3C). Parasite adaptation increases the relative fitness
159 from high and low yield hosts by driving more new infections from high yield hosts compared to low yield
160 hosts (Fig. 5A). Parasite adaptation can thus result in increased transmission dispersion by increasing parasite
161 fitness from high yield hosts more than low yield hosts. The early peak in parasite fitness and transmission
162 dispersion is followed by a dip when susceptible hosts are rare (Fig. 5A, 3C). Transmission dispersion is
163 also low when hosts are rare as parasite fitness is low for infections of both host types (*i.e.* R_{eH} and R_{eL}
164 are both low which drives low $vmr(R_e)$, Fig. 5C). Parasite fitness and transmission dispersion both rise
165 again as susceptible host abundance increases (Fig. 5B, D). At equilibrium, transmission dispersion is lower
166 compared to early in an epidemic as host densities are relatively low (Fig. 3D, 5D).

167 5.2 Impact of host composition on transmission dispersion

168 The composition of the host population impacts the extent that parasite adaptation increases transmission
169 dispersion. For example, host populations that are mostly composed of low yield hosts select for parasites
170 that drive high transmission dispersion (Fig. 6.) Parasites infecting and adapting to host populations with
171 only a few high yield hosts cause high transmission dispersion as the small proportion of high yield hosts
172 are responsible for an outsized proportion of new cases. Further, parasite adaptation drives larger increases
173 in transmission dispersion as the difference in quality between high and low yield hosts increases (Fig. 7).
174 Both differences in transmission and virulence can drive increased $vmr(R_e)$: transmission dispersion is high
175 when low yield hosts have lower transmission ($c_H > c_L$) and/or higher virulence ($y_H < y_L$) compared to
176 high yield hosts (Figure 7).

177 5.3 Parasite within-host growth rate variance impacts transmission dispersion

178 The phenotypic variance of the within-host growth rate in the parasite population ($\text{var}\epsilon$) controls how quickly
179 parasite adaptation occurs (Figure 8). Early in an epidemic, high phenotypic variance drives increased peak
180 within-host growth rates as parasites can quickly adapt to exploit high host densities (Figure 8). When
181 phenotypic variance is low, parasite adaptation proceeds more slowly such that host densities drop, reducing
182 selection pressure for high within-host growth rates, before high trait values evolve. High phenotypic variance
183 also drives high transmission dispersion ($vmr(R_e)$) early in the epidemic through the high values of the
184 within-host growth that result in larger increases in parasite fitness from high yield hosts (R_{eH}) relative
185 to fitness from low yield hosts (R_{eL}) (Fig. 9 A,B,C). Conversely, low phenotypic variance prevents high
186 within-host growth rates from evolving and thus maintains low transmission dispersion by keeping R_{eH} and
187 R_{eL} relatively close to one another.

188 The impact of phenotypic variance on parasite adaptation differs at the endemic equilibrium compared to
 189 early in an epidemic: intermediate phenotypic variance maximizes equilibrium transmission dispersion (Fig.
 190 9 F). Similar to early in an epidemic, increasing phenotypic variance increases the within-host growth rate
 191 and transmission dispersion for small values of phenotypic variance (Fig. 9 D, F). However, further increases
 192 in phenotypic variance decreases the within-host growth rate in low yield hosts (ϵ_L) while maintaining
 193 high within-host growth rates in high yield hosts (ϵ_H) (Fig. 9 D). When phenotypic variance is low,
 194 selection dominates to increase ϵ_L and ϵ_H to prioritize increasing fitness on high yield hosts at the cost
 195 of decreased fitness on low yield hosts (Fig. 9D). Conversely, high phenotypic variance results in within-host
 196 growth rate specialization in the two host types which maintains high reproductive fitness in both hosts and
 197 simultaneously decreases transmission dispersion (Fig. 9 D, F).

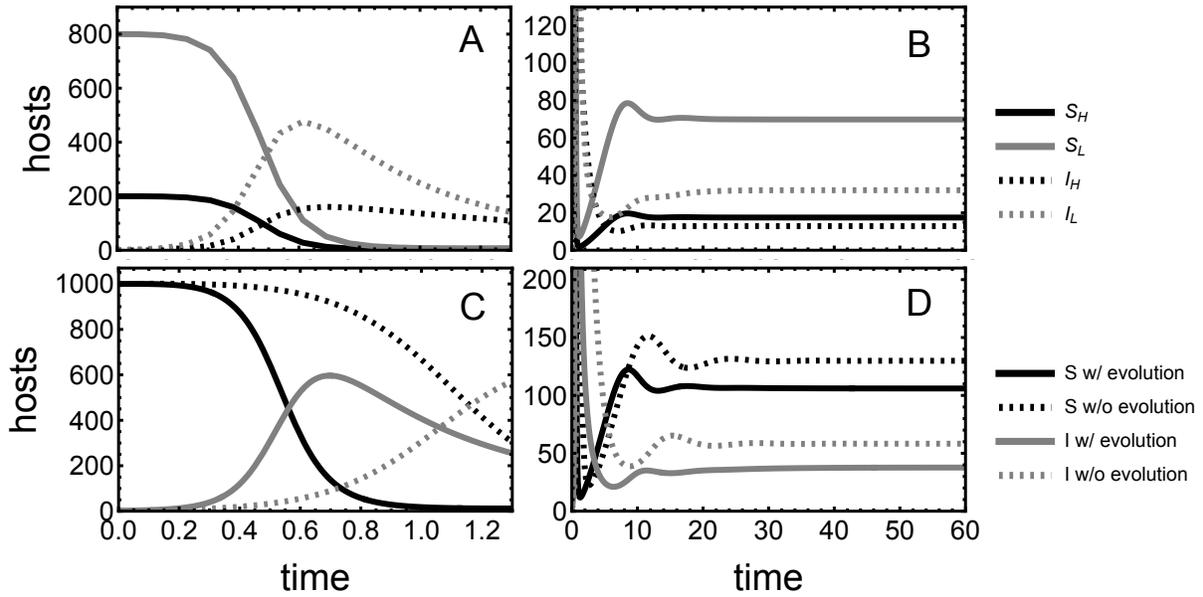


Figure 3: Epidemiological dynamics of a heterogeneous host population early (A, C) and late (B, D), with (A, B) and without evolution (C, D). A and B show the density of susceptible high yield (S_H) and low yield hosts (S_L) and infectious high yield (I_H) and low yield hosts (I_L). C and D show the total density of susceptible (S) and infectious (I_H) hosts with and without parasite evolution. In the absence of parasite evolution, the parasite within-host growth rate (ϵ) is set to 0.25 and does not change. Note that high and low yield susceptible host densities are identical when they have equal proportions in the host population ($p = 0.5$ in Eq. 1a-1d.) $c_H = 1, c_L = 0.1, y_H = 0.1, y_L = 1, x = 0.5, \lambda = 50, \delta = 0.02, \gamma = 0.6, \rho = 10^{-2}, var_H(\epsilon) = var_L(\epsilon) = 1$.

198 6 Summary and discussion

199 This study provides a framework for predicting how parasite adaptation impacts transmission dispersion for
 200 emerging and re-emerging infectious diseases. The model predicts that parasite adaptation to heterogeneous
 201 host populations can result in increased transmission dispersion. Parasite adaptation drives the greatest
 202 increases in transmission dispersion when host populations are composed of high yield hosts that have higher
 203 transmission and lower virulence compared to low yield hosts. The results predict that parasite adaptation

204 to heterogeneous host populations drives the evolution of high transmission dispersion of parasites early in
 205 epidemics. Further, parasite adaptation can maintain increased transmission dispersion at endemic equilibria.

206 The results of the current study strengthen the idea that differences in host transmission can drive
 207 transmission dispersion but also predict that differences in virulence can increase transmission dispersion by
 208 impacting parasite adaptation. That is, epidemiological studies have previously shown that differences in
 209 transmission potential among hosts (often measured as parasite load) is a source of heterogeneity that is
 210 associated with increased transmission dispersion (Chen et al. 2021). This study suggests that heterogeneity
 211 in host transmission potential can also indirectly increase transmission dispersion by selecting for parasites
 212 that drive more infections from higher yield hosts than lower yield hosts. Thus given that these results
 213 raise the possibility that differences in (host) virulence can select for parasites that enhance transmission
 214 dispersion, more effort should be made to experimentally disentangle the relationship between virulence
 215 and transmission dispersion.

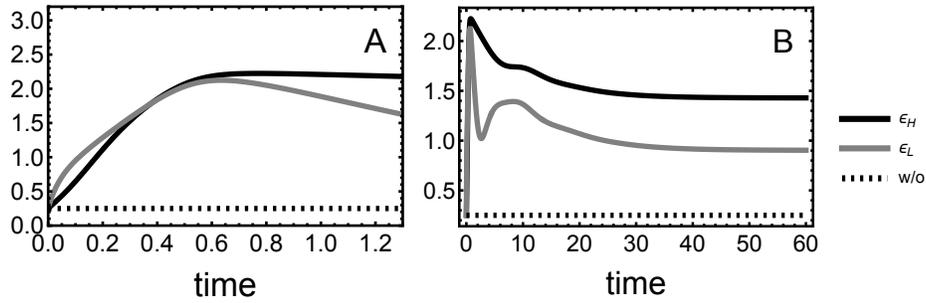


Figure 4: Evolutionary dynamics of the within-host parasite growth rate (ϵ) early (A) and late (B). Figure shows the trait value of the within-host parasite growth rate in high yield (ϵ_H) and low yield hosts (ϵ_L). Dotted black line shows the value of the within-host growth rate in the absence of adaptation ($\epsilon = 0.25$). $c_H = 1, c_L = 0.1, y_H = 0.1, y_L = 1, x = 0.5, \lambda = 50, \delta = 0.02, \gamma = 0.6, \rho = 10^{-2}, var_H(\epsilon) = var_L(\epsilon) = 1$.

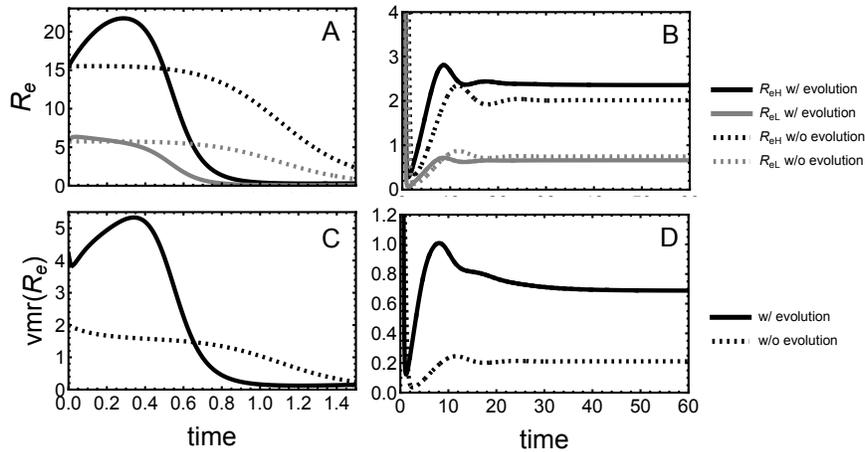


Figure 5: Parasite adaptation drives higher parasite fitness and transmission dispersion. Both parasite fitness and transmission dispersion are highest early in epidemics when susceptible host density is also high. Figure shows parasite fitness of high and low yield hosts (R_{eH}, R_{eL}) and transmission dispersion ($vmr(R_e)$) over time. $c_H = 1, c_L = 0.1, y_H = 0.1, y_L = 1, x = 0.5, \lambda = 50, \delta = 0.02, \gamma = 0.6, \rho = 10^{-2}, var_H(\epsilon) = var_L(\epsilon) = 1$.

216 The variation of virulence across host types is predicted to determine whether parasite adaptation drives
 217 increased transmission dispersion, however virulence manifests in many different ways in nature. An obvious
 218 question is thus how different virulence modes could impact the predictions made here. Virulence in this
 219 study is modelled as an increase in host mortality following infection that increases with the parasite within-
 220 host growth rate. Parasites can enjoy a high growth rate on high yield hosts that are more tolerant to infection
 221 (*i.e.* high yield hosts that do not experience high mortality despite being infected by parasites with high
 222 growth rates.) Transmission dispersion can result from some hosts dying more quickly than others. Thus
 223 alternative forms of virulence that also vary between hosts and shorten the duration of the infection could

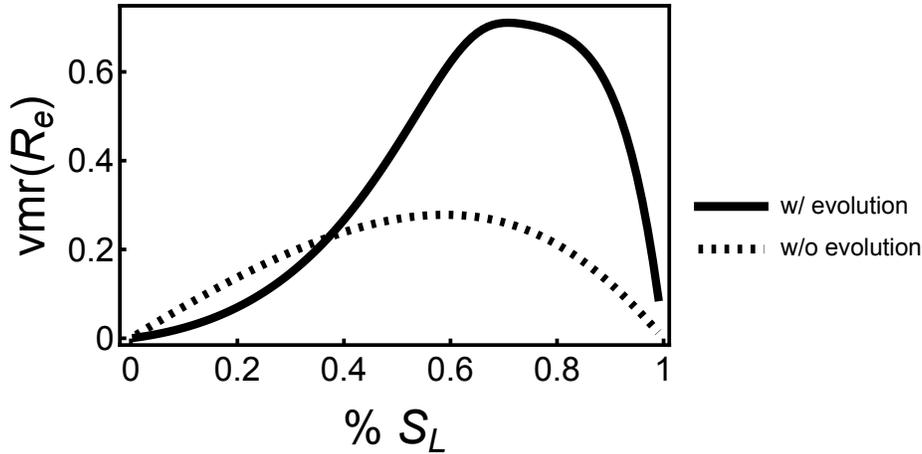


Figure 6: Transmission dispersion is highest when parasites adapt to host populations that are mostly composed of low yield hosts. Figure shows the variance-to-mean ratio of $R_e(t)$ at the endemic equilibrium as the percentage of susceptible low yield hosts (s_L) in the system varies. $c_H = 1, c_L = 0.1, y_H = 0.1, y_L = 1, x = 0.5, \lambda = 50, \delta = 0.02, \gamma = 0.6, \rho = 10^{-2}, var_H(\epsilon) = var_L(\epsilon) = 1$.

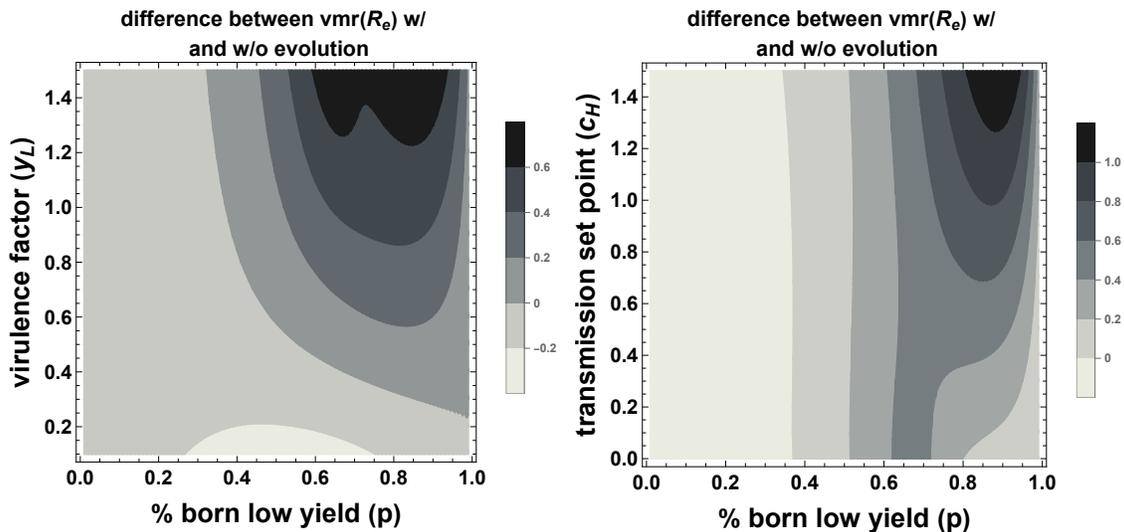


Figure 7: The difference between transmission dispersion ($vmr(R_e)$) with and without parasite evolution is greatest when most hosts are born low yield (high p) and either high yield hosts are much higher quality than low yield hosts from the parasite's perspective ($y_L \ll y_H, c_H \gg c_L$). $y_H = 0.1, y_L = 1, c_H = 0.5, c_L = 0$. All other parameters are the same as in Figure 3.

224 lead to similar phenomena as predicted here. For example, hosts could vary in the severity of the symptoms
 225 they experience post-infection (*e.g.* lethargy). Hosts experiencing severe symptoms may then decrease
 226 contact with other hosts and thus decrease the likelihood that the infection is spread. Similarly, hosts could
 227 vary in how quickly they experience symptoms after becoming infectious. Hosts that quickly experience
 228 symptoms may also be less likely to spread the infection through decreased contact with other hosts. Finally,
 229 the results of the current study could apply to the heterogeneous distribution of treatment against infection
 230 that decrease virulence such that increased transmission dispersion could be adaptive if transmission is not
 231 impacted.

232 The phenotypic variance in the within-host parasite growth rate impacts transmission dispersion by
 233 determining selection strength (Fig. 8)). Low variance in the within-host growth rate constrains parasite
 234 adaptation by limiting the range of trait values that natural selection can act upon which keeps the within-host
 235 growth rate and transmission dispersion low. Conversely, high variance in the within-host growth rate selects
 236 for high within-host growth rates in high yield hosts but low within-host growth rates in low yield hosts and
 237 thus slightly lower transmission dispersion compared to intermediate variance. The current study assumes
 238 that the variance in the within-host growth rate is the same in both host types, however in nature high yield
 239 hosts would likely support higher variance as they are infected for longer periods of time and maintain
 240 higher parasite loads. Relaxing this assumption and assuming that selection on the within-host growth rate
 241 in low yield hosts is subject to low variance would likely drive higher transmission dispersion as selection for
 242 low within-host growth rates in low yield hosts would be weaker than selection for high within-host growth
 243 rates in high yield hosts.

244 Often highly infectious hosts experience high virulence due to the burden of carrying large parasite loads
 245 (De Roode, Yates, and Altizer 2008; Fraser, Hollingsworth, Chapman, de Wolf, and Hanage 2007). However,
 246 observations of hosts that tolerate high parasite loads while experiencing little or no increase in virulence
 247 are also common (Soper 1939; VanderWaal and Ezenwa 2016). This study combines a bit of both ideas by
 248 assuming that virulence increases as the within-host parasite growth rate increases but also assumes that a
 249 subset of hosts tolerate those increases in growth rate well. Further, this study assumes that the hosts that

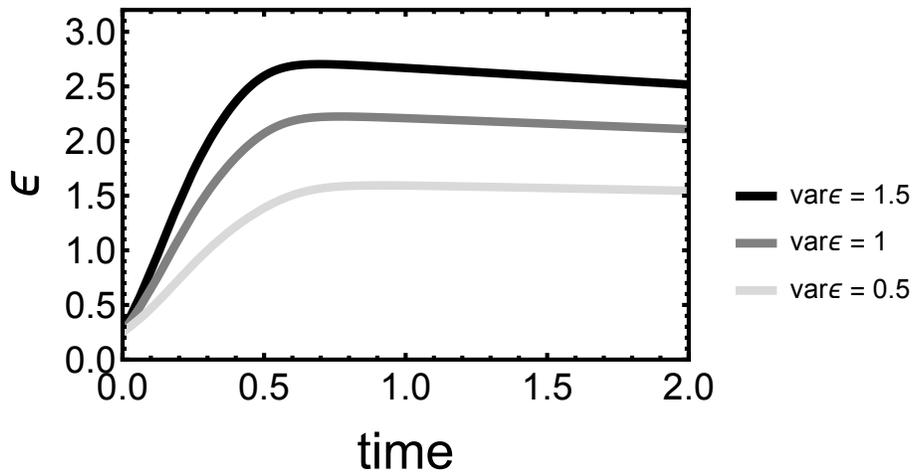


Figure 8: Phenotypic variance in the population of the within-host growth rate ($\text{var}\epsilon$) controls how quickly adaptation occurs. Plots show the dynamics of within-host growth rate (ϵ) adaptation for different values of $\text{var}\epsilon$ and demonstrate that ϵ adapts more quickly and reaches higher values when $\text{var}\epsilon$ is high. All parameters are the same as in Figure 3.

250 experience low virulence also transmit at a higher rate. In this way we have focused on the two extremes in
 251 the disease ecology literature: hosts that transmit a lot because of the combination of low virulence and high
 252 transmission and hosts that only transmit a little because they simultaneously have high virulence and low
 253 transmission. In reality, host populations will also be composed of intermediate host types, *e.g.* hosts that
 254 have high virulence and high transmission, hosts are resistant to infection naturally or through treatments
 255 that result in low virulence and low transmission. We focused on the two extreme host cases because it
 256 was most likely to enhance the impact of heterogeneity on the evolution of increased parasite transmission
 257 dispersion. Less extreme forms of host heterogeneity have been considered in other studies. For example,
 258 Gog, Hill, Danon, and Thompson (2021) assumed that a proportion of the host population is more vulnerable
 259 to the infection but mixes less with other hosts while all other hosts are less vulnerable to the infection but
 260 mix more readily. Similarly, Gandon and Lion (2022) studied how high and low contact rates as well as high
 261 and low vulnerability to infection following vaccination impacted the speed that parasites evolve. Parasites
 262 that adapt to these more intermediate host types will likely drive more modest increases in transmission
 263 dispersion compared to the host populations studied here. Nevertheless, future work should study the effect

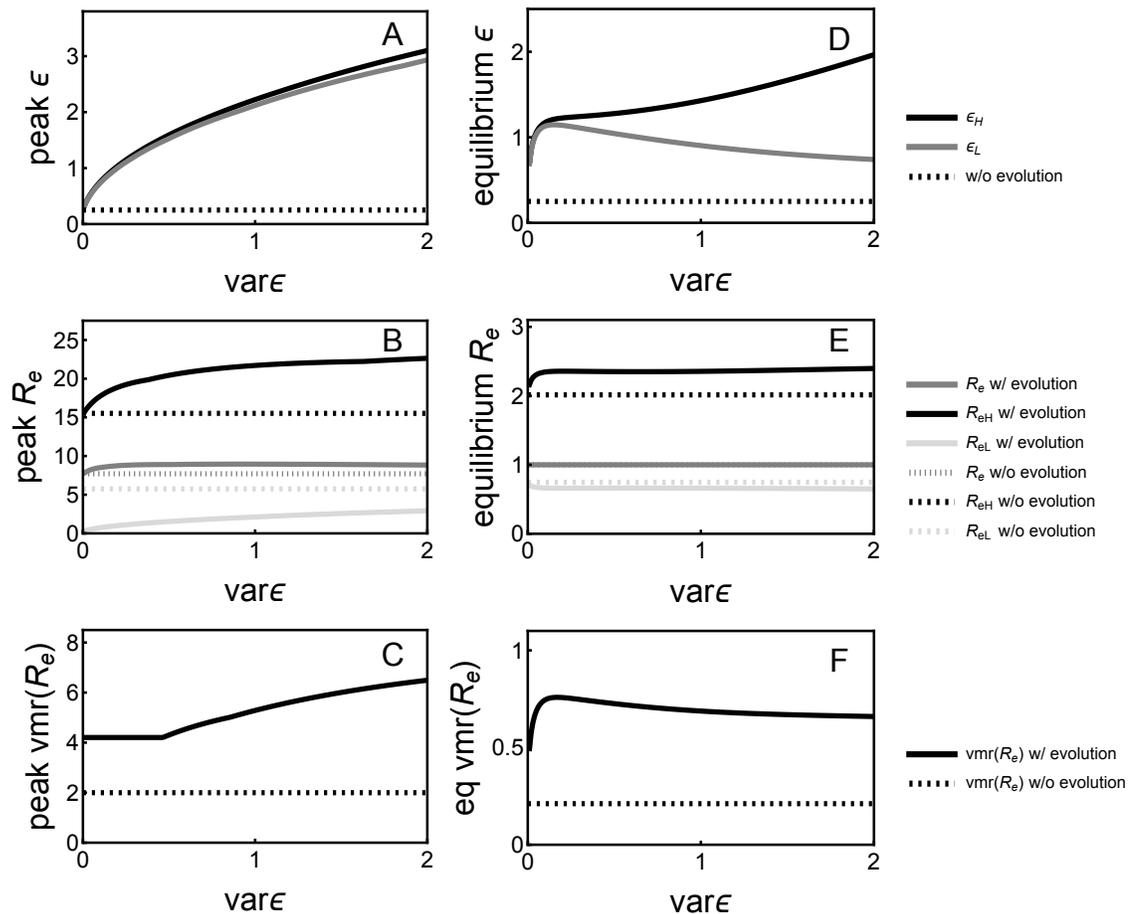


Figure 9: Variance in the within-host growth rate ($\text{var}\epsilon$) impacts parasite adaptation and transmission dispersion. Left plots show how $\text{var}\epsilon$ impacts peak within-host growth rates (ϵ), R_e and transmission dispersion ($\text{vmr}(R_e)$) that occur early in the epidemic (see Figure 5). Right plots show how $\text{var}\epsilon$ impacts the within-host growth rate (ϵ), R_e and transmission dispersion ($\text{vmr}(R_e)$) at the endemic equilibrium (see Figure 5). All parameters are the same as in Figure 3.

264 of parasite adaptation to intermediate host types on transmission dispersion.

265 A key question is how common the host compositions studied here are in nature. Empirical studies on
266 host heterogeneity tend to focus on one trait: either transmission (often measured as parasite load or contact
267 rate) or virulence (this is measured many ways, e.g. symptoms, fertility, death rate). However one study that
268 is relevant to the assumptions of this model showed that the distribution of SARS-Cov2 viral loads among
269 symptomatic/pre-symptomatic and asymptomatic cases were similar (Chen et al. 2021). In other words, viral
270 load distributions are not clearly associated with host virulence and thus a subset of asymptomatic hosts could
271 have very high viral loads. Thus while there is some evidence that high and low yield hosts similar to those
272 modelled in this study exist, the exact compositions of host populations are not well documented. That is, the
273 percentages of hosts that are high yield and low yield as well as the distributions of intermediate traits such
274 as transmission found in host populations are often not known. Thus more empirical research is necessary to
275 determine how the compositions of host traits relevant to parasite spread vary in host populations.

276 The predictions made in this study should be tested experimentally. Successfully validating theory
277 requires controlled experiments where the factor-of-interest can be manipulated to compare empirical results
278 to theoretical predictions. Experiments to test whether parasite adaptation to heterogeneous host populations
279 can drive increased transmission dispersion will require a disease system with two host types that differ in
280 transmission and/or virulence. Many disease systems fit one of these criteria in that host types have been
281 identified that are capable of causing a disproportionate number of new infections either through increased
282 shedding of the parasite (Bassetti et al. 2005; Capparelli et al. 2009; Dougan and Baker 2014; Hughes and
283 Randolph 2001; Wang et al. 2017) or through decreased virulence (e.g. long infectious periods) (Curtis et al.
284 1999; Ferreira et al. 2011). One disease system that could be well-suited to test the predictions made in this
285 study is *Daphnia magna* and its bacterial parasite, *Pasteuria ramosa*. The *Daphnia-Pasteuria* system is ideal in
286 that both spore load (i.e. transmission potential) and virulence (measured as reductions in post-infection
287 lifespan or fecundity) can vary between males and females (Gipson and Hall 2018) and across age classes
288 (Izhar and Ben-Ami 2015). Thus it may be possible to use existing *Daphnia* lines that meet the high and low
289 yield host classifications used in this study and test the predictions made here.

290 Most new infections are transmitted from relatively few infected individuals. This increased transmission
291 dispersion is largely attributed to differences among hosts, thus most research to date has focused on
292 the importance of variability in host populations. The current study presents an additional evolutionary
293 mechanism that could enhance this phenomena whereby parasite specialization on highly infectious hosts
294 comes at the cost of transmitting less effectively from less-infectious hosts. The model predicts that in this
295 way, parasite adaptation can further skew transmission events so that most new infections are transmitted
296 from a minority of infectious hosts. Further, this study presents a framework for predicting how parasite
297 adaptation determines transmission dispersion for emerging and re-emerging infectious disease.

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