

Environmentally transmitted parasites: host-jumping in a heterogeneous environment

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2 *Abstract.* Groups of chronically infected reservoir-hosts contaminate resource patches by
shedding a parasite’s free-living stage. Novel-host groups visit the same patches, where they
4 are exposed to infection. We treat arrival at patches, levels of parasite deposition, and
infection of the novel host as stochastic processes, and explicitly derive the expected time
6 elapsing until a host-jump (initial infection of a novel host) occurs. At stationarity, mean
parasite densities are independent of reservoir-host group size. But within-patch
8 parasite-density variances increase with reservoir group size. The probability of infecting a
novel host declines with parasite-density variance; consequently larger reservoir groups
10 extend the mean waiting time for host-jumping. Larger novel-host groups increase the
probability of a host-jump during any single patch visit, but also reduce the total number of
12 visits per unit time. Interaction of these effects implies that the waiting time for the first
infection increases with novel-host group size. If the reservoir-host uses resource patches in
14 any non-uniform manner, reduced spatial overlap between host species increases the waiting
time for host-jumping.

16 **1 Introduction**

Many ecologically and epidemiologically important pathogens are transmitted through the
18 abiotic environment [Miller et al. 2006, Breban et al. 2009, Cizauskas et al. 2014a].

Analyses of environmentally transmitted parasites emphasize how the dynamics differs from
20 infections spread through direct contact between hosts

[Bani-Yaghoub et al. 2012, Garira et al. 2014]; these distinctions have significant
22 implications for epidemiological invasion [Rohani et al. 2009, Breban et al. 2010] and for
the likelihood a parasite can “jump” to a novel host species

24 [Woolhouse et al. 2005, Caraco et al. 2014].

Our study of environmental transmission focuses on host-jumping, the initial infection of
26 a novel-host population or species. Our assumptions best match intestinal parasites of
humans, non-human primates and gregarious herbivores, ordinarily transmitted via the
28 fecal-environment-oral route [Hutchings et al. 2001, Cizauskas et al. 2014b]. But other
macroparasites, as well as certain bacteria and viruses, use this mode of transmission.

30 Therefore, our model’s predictions should apply broadly to group-living hosts. We envision
a reservoir population and a novel-host population inhabiting the same environment.

32 Chronically infected reservoir individuals shed the parasite at resource patches, exposing
the novel host to infection. We ask how each species’ social group size affects the random
34 waiting time until the parasite first jumps to the novel host.

2 Background

36 Climate change and increased species-introductions have driven biogeographic range
expansion of many parasites [Crowl et al. 2008, Jolles et al. 2008]. Ecological disturbance,

38 especially human landscape alteration, has placed new host-parasite combinations in
common environments [Patz et al. 2000]. These processes increase opportunities for
40 parasites to jump to novel hosts [Cooper et al. 2012].

We motivate our general analysis by citing Hausfater and Meade’s [1982] study of yellow
42 baboons (*Papio cynocephalus*) infested by intestinal nematodes. A population of
approximately 200 individuals was partitioned into groups; individuals associated
44 consistently with the same group. Each evening a group would enter a spatially distinct
cluster of yellow-barked acacia (*Acacia xanthophloea*). Group members would rest in the
46 trees, safe from predators, for about 10 hours, and then spend time socializing/foraging
beneath the trees.

48 After one or two nights at a sleeping site, the group would move to another location,
leaving a substantial fecal accumulation beneath the trees. Soil from the sleeping area
50 yielded ova and larvae of nematodes at densities exceeding those in samples from outside
sleeping sites [Hausfater and Meade 1982]. Intestinal nematodes of baboons are transmitted
52 environmentally [Olsen 1974], and the same acacia clusters used by yellow baboons provided
feeding/resting sites to other mammals, including vervet monkeys (*Cercopithecus aethiops*).
54 Our model addresses host jumping when different species occupy locations where a parasite
may be deposited by one host species and acquired by a second.

56 **3 Model**

We extend the conceptual framework for environmentally transmitted parasites in two ways.
58 First, we consider host socio-ecology; group size in each population affects the parasite’s
invasion of a novel host. Second, since host-jumping occurs relatively rarely

Symbols	Definitions
N_1	Number of reservoir (infested) host individuals, Species 1
R	Number of resource patches, indexed by r
G_1	Group size, reservoir species
λ_j	Rate at which each group of Species- j ($j = 1, 2$) arrives at the set of R patches
p_{jr}	Probability Species j uses patch r ; $r = 1, 2$
$X_{1r}(t)$	Total visits by Species 1 to patch r on $(0, t)$
$M_r(t)$	Parasite density in patch r at time t , continuous random variable
$f(M_r)$	Stationary probability density of M_r
m	Randomly increment to parasite density; mean = $G_1 \mu$, variance = $G_1 \sigma^2$
ξ	Decay rate of free-living parasite density
t_c	Characteristic time for parasite mortality dynamics
N_2	Number of novel host individuals, Species 2
G_2	Group size, novel host
γ	Susceptibility parameter, novel host
$\zeta_r(\tau)$	Probability novel host has not acquired parasite in patch r by time τ
T	Random waiting time until parasite first jumps to novel host

Table 1: Definitions of model symbols

60 [Woolhouse et al. 2005], we treat deposition of the parasite by the reservoir species, and
 acquisition of the parasite by the reservoir host, as stochastic processes. Combining these
 62 processes leads to an analytic expression for the mean waiting time until the parasite jumps
 to the novel host.

64 We dichotomize model development. To begin, we consider how the reservoir species
 generates densities of the parasite's infectious stage across resource patches. We then
 66 address the novel host's use of the same patches and consequent exposure to the parasite.
 We fix population sizes of both species, to focus links between group size and parasite
 68 densities. Table 1 collects model symbols.

3.1 Reservoir-host and parasite densities

70 We index the reservoir host as Species 1. The reservoir population contains a total of N_1
 individuals that chronically shed the parasite's infectious stage. We treat N_1 as an endemic

72 equilibrium, a constant. The N_1 reservoir hosts are distributed uniformly across groups of
size G_1 , where $G_1 \in \{1, 2, \dots, N_1\}$. Reservoir group size G_1 can range from solitary to the
74 entire set of parasitized hosts; uniformity of G_1 implies that group size is a species
characteristic [Trainor and Caraco 2006]. For convenience we treat N_1/G_1 as an integer,
76 and ignore Species-1 individuals free of parasitism.

3.1.1 Reservoir patch use

78 The environment includes a set of R discrete resource patches. For clarity and simplicity, we
set $R = 2$, but generalization is straightforward. A patch might offer food, drinking water,
80 refuge from predators, or a place to rest.

Infested hosts shed the parasite during patch visits. Each reservoir-host group visits the
82 set of R patches as an independent Poisson process, with the same constant probabilistic
rate λ_1 . Since there are N_1/G_1 such groups, the reservoir population as a whole visits the
84 set of patches at combined probabilistic rate $\lambda_1 N_1/G_1$. Visits are sufficiently brief, relative
to the time between visits, that we ignore the possibility of simultaneous patch occupation.
86 λ_1 does not depend on group size. This means that each individual in the reservoir-host
population enters a resource patch at the same average rate whether groups contain g_1 or
88 $2g_1$ members. But in the latter case the population is structured into half as many groups,
so that the population-scale rate $\lambda_1 N_1/G_1$ is halved. We do not anticipate that frequencies
90 of resting, sleeping and drinking (and perhaps feeding) will depend on group size.

Each reservoir-host group has the same resource preferences, defined by the probability
92 distribution of visits across the R patches. p_{1r} represents the probability that any given visit
by the reservoir species occurs at patch r ; $r = 1, 2$. $X_{1r}(t)$ counts the cumulative number of
94 visits by reservoir-host groups to patch r by time t ; $X_{1r}(t = 0) = 0$. Then each $X_{1r}(t)$ is an

independent Poisson variable [Ross 1983] with expectation $E[X_{1r}(t)] = p_{1r}\lambda_1 N_1 t / G_1$.

96 3.1.2 Parasite densities

The continuous random variable $M_r(t)$ represents parasite density for patch $r = 1, 2$;

98 $M_r(t = 0) = 0$. $M_r(t)$ increases when a reservoir-host group enters patch r and sheds the parasite. $M_r(t)$ decreases between visits by the reservoir species due to parasite mortality.

100 $M_r(t)$ depends only on the dynamics of shedding and decay in patch r ; parasite density does not depend on patch area or explicit location. We treat the parasite as infectious when
102 shed by a reservoir host. This assumption holds for most microparasites, but eggs/propagules of macroparasites may require 5-15 days before producing infectious larvae.

104 Suppose that the i -th visit by reservoir-host groups occurs at time t_i ; $i \geq 1$. Then at time $t = t_i$, parasite density $M_r(t)$ increases by an amount m_i . Each m_i is a positive,
106 continuous random variable, with both mean and variance increasing with reservoir group size. The mean is $E[m] = G_1\mu$, and the variance is $V[m] = G_1\sigma^2$. That is, each
108 reservoir-host group member independently adds to the local density of the parasite's infectious stage.

110 Macro-parasitic burdens *per* host commonly exhibit statistical aggregation; the variance exceeds the mean [Hudson and Dobson 1995, Poulin and Morand 2000]. Aggregation should
112 hold even with zero-counts truncated [Shaw et al. 1998]. Although the rate at which a reservoir host sheds a parasite's infectious stage need not be proportional to that host's
114 burden, we maintain $\sigma^2 \geq \mu$ during model evaluation.

We make biologically reasonable assumptions concerning increments to parasite densities.

116 m_i does not depend on t_i . The m_i are mutually independent, identically distributed, and independent of $X_{1r}(t)$, $M_r(t)$ and r ; shedding by the reservoir host is unaffected by the local

118 parasite density. Finally, we assume that between reservoir-host visits, parasite density
 decays in a constant proportional manner, as Hausfater and Meade [1982] observed among
 120 infectious larvae of yellow baboons' endoparasites. The parasite-mortality rate ξ is the same
 in each patch. Then the parasite density in patch r at time t satisfies:

$$M_r(t) = \sum_{i=1}^{X_{1r}(t)} m_i e^{-\xi(t-t_i)} \quad (1)$$

122 Each $M_r(t)$ is an independent Markov shot-noise process
 [Lemoine and Wenocur 1986, Laio et al. 2001]. In our particular model, parasite density
 124 decays continuously, but at Poisson-process intervals the density is suddenly incremented by
 a reservoir-host group's shedding. Standard approaches obtain the mean and variance of
 126 each parasite density from the moment generating function of the shot-noise process
 [Ross 1983, Lowen and Teich 1990]. Expected values are:

$$E[M_r(t)] = (p_{1r}/\xi) \lambda_1 N_1 \mu (1 - e^{-\xi t}); \quad r = 1, 2 \quad (2)$$

128 Mean parasite densities do not depend on reservoir-host group size G_1 . The parasite-density
 variances are:

$$V[M_r(t)] = (p_{1r}/2\xi) \lambda_1 N_1 (\sigma^2 + G_1 \mu^2) (1 - e^{-2\xi t}); \quad r = 1, 2 \quad (3)$$

130 Variance in the number of patch visits and variance in the level of parasite-shedding per
 visit contribute to the overall parasite-density variances.

132 As time grows large, parasite deposition and mortality will approach stochastic
 equilibrium. Parasite densities consequently approach their stationary probability

134 distributions, with respective means and variances as $t \rightarrow \infty$:

$$E[M_r] = p_{1r}\lambda_1 N_1 \mu / \xi; \quad V[M_r] = p_{1r}\lambda_1 N_1 (\sigma^2 + G_1 \mu^2) / 2\xi \quad (4)$$

Each mean parasite density, and each variance, increases as $\lambda_1 N_1$ increases, since the rate at
136 which patches are visited will increase. As we noted, mean parasite densities are
independent of reservoir-host group size, but the variance increases with G_1 . If
138 reservoir-species groups are small ($G_1 \rightarrow 1$) parasite density is renewed frequently by
relatively small increments, and M_r will fluctuate less through time. If reservoir groups are
140 large ($G_1 \rightarrow N_1$), less frequent visits, each with larger average increments, will produce
greater variance in the parasite densities. Below we show how this effect of reservoir group
142 size on parasite-density variance influences the expected waiting time until a host-jump
occurs.

144 Appendix A shows that the characteristic time scale of parasite mortality is $t_c = 2/\xi$. If
the rate at which reservoir-host groups arrive at patches is sufficiently large, we have:

$$p_{11}\lambda_1(N_1/G_1), \quad (1 - p_{11})\lambda_1(N_1/G_1) \gg t_c^{-1} = \xi/2 \quad (5)$$

146 If these expressions hold, the stationary distribution of each parasite density will be, by the
central limit theorem, approximately normal [Lowen and Teich 1990]. Since host-jumps are
148 far rarer than patch visits, we can assume that each M_r has a normal density with mean
and variance given by the stationary values in Eq. (4).

150 **3.2 Novel-host patch use**

We index the novel host (*i.e.*, initially parasite free) as Species 2. N_2 novel-host individuals
152 are distributed uniformly among groups of size G_2 , where $G_2 \in \{1, 2, \dots, N_2\}$. We take the
number of groups N_2/G_2 as an integer.

154 The rate at which novel hosts visit patches, and their patch preferences, will together
govern novel-host exposure to the parasite and, therefore, affect the probability of a
156 host-jump. In general, Species 1 and 2 may differ in population size, the overall rate at
which groups visit patches, and in the distribution of those visits across patches.

158 Each novel-host group enters the set of patches as an independent Poisson process. The
combined probabilistic rate is $\lambda_2 N_2/G_2$; λ_2 is a constant. p_{2r} is the probability that any
160 given visit by a novel-host group occurs at patch r . $X_{2r}(t)$ counts all novel-host visits to
patch r over t time units; $r = 1, 2$. The expected value of $X_{2r}(t)$ is $E[X_{2r}(t)] = p_{2r} \lambda_2 N_2 t / G_2$.
162 Larger group size G_2 implies fewer total groups, hence fewer patch-visits per unit time. The
 $X_{2r}(t)$ are independent Poisson random variables; see Appendix A.

164 **3.3 Host-jumping**

We take parasite densities as stationary random variables. $f(M_r)$ represents the probability
166 density of M_r , an approximately normal variate with mean $E[M_r]$ and variance $V[M_r]$, each
given in Eq. (4). The N_2/G_2 novel-host groups enter the system at time $\tau = 0$. Each visit
168 to patch r exposes the G_2 group members to an independent realization of the random
variable M_r .

170 We assume no immunological effects of past exposure [Breban et al. 2009]. Each
exposure to the parasite infects/fails to infect each group member independently. The

172 chance any novel host is infected upon exposure must increase monotonically in M_r .

Given exposure to the parasite, the probability of infection is a matter of dose-response
 174 analysis [Tenuis et al. 1996, Strachan et al. 2005], and we use a model favored by empirical
 studies. Consider the g – th group member of a novel host-group during a single visit to
 176 patch r ; $g = 1, 2, \dots, G_2$. The conditional probability of infection is

$\Pr[g \text{ infected} | M_r] = 1 - e^{-\gamma M_r}$; $\gamma > 0$. γ is the susceptibility parameter, an attribute of the
 178 parasite and novel-host combination. If γ is very small, susceptibility presents a
 between-host barrier [Woolhouse et al. 2005]. Some generalist endoparasites, however, jump
 180 hosts more readily, implying larger values of γ ; furthermore, γ may vary with phylogenetic
 distance between host species [Cooper et al. 2012].

182 On any *single visit* to patch r , the conditional probability of a host jump, given M_r , is
 simply 1 minus the probability that no host is infected:

$$\Pr [\text{Host jump} | M_r] = 1 - \left(e^{-\gamma M_r} \right)^{G_2} \quad (6)$$

184 We assume that the same value of γ applies to all novel-host individuals. If susceptibility
 varies among individuals as a beta variate [Tenuis and Havelaar 2000], the results do not
 186 affect the model’s qualitative predictions. Proceeding, the unconditional probability of a
 host-jump during a single patch-visit is:

$$\begin{aligned} \Pr [\text{Host jump}] &= 1 - \int_0^\infty e^{-\gamma G_2 M_r} f(M_r) dM_r \\ &\approx 1 - \exp \left(-\gamma G_2 E[M_r] + \frac{(\gamma G_2)^2}{2} V[M_r] \right) \end{aligned} \quad (7)$$

188 by similarity to the normal distribution’s moment generating function. The chance that the

parasite jumps to the novel host during a group's single exposure must increase with mean
 190 parasite density. But the same probability decreases with the variance of the parasite
 density. Larger reservoir-host group size G_1 increases each $V[M_r]$, and so can decrease the
 192 probability of a host-jump.

Intuitively, a larger novel-host group size must increase the probability of host-jumping
 194 on a single visit, since more hosts are exposed per visit. However, increasing G_2 decreases
 the number of novel host groups (N_2/G_2), so that the expected number of patch-visits per
 196 unit time declines. To explore these effects in combination, consider the probability that the
 parasite has failed to invade the novel-host population after multiple visits.

198 Applying the unconditional probability of avoiding infection on a single visit, and
 conditioning on the number of novel-host visits to patch r by time τ , the probability the
 200 novel-host remains without infestation after χ_r visits is:

$$\Pr[\text{No jump} | X_{2r}(\tau) = \chi_r] = \left[\exp \left(-\gamma G_2 E[M_r] + \frac{(\gamma G_2)^2}{2} V[M_r] \right) \right]^{\chi_r} \quad (8)$$

Unconditionally, the probability that the novel host has not acquired the endoparasite
 202 within patch r by time τ is:

$$\Pr[\text{No jump by } \tau] = \sum_{X_{2r}(\tau)=0}^{\infty} \Pr[\text{No jump} | X_{2r}(\tau)] \Pr[X_{2r}(\tau)] \quad (9)$$

for both $r = 1$ and $r = 2$ independently. Recall that $\Pr[X_{2r}(\tau)]$ follows a Poisson probability
 204 function with mean $E[X_{2r}(\tau)] = p_{2r} \lambda_2 N_2 \tau / G_2$. Then Eq. (9) has the form of a Poisson
 probability generating function [Ross 1983]. If $G(z)$, where $0 < z < 1$, is the generating
 206 function for a Poisson random variable x , then $G(z) = \sum_{x=0}^{\infty} z^x Pr[x]$. If k is the mean of the

Poisson variable x , we have $G(z) = \exp(k[z - 1])$. We can substitute the exponential term
 208 in Eq. (8) for z , and $E[X_{2r}(\tau)]$ for k . Assembling the pieces, the probability that the
 reservoir host avoids parasite infection across all visits to patch r through time τ is:

$$\zeta_r(\tau) = \exp\left([p_{2r}\lambda_2 N_2 \tau / G_2] \left[\exp\left(-\gamma G_2 E[M_r] + \frac{(\gamma G_2)^2}{2} V[M_r]\right) - 1 \right]\right) \quad (10)$$

210 Of course, the chance that the novel host avoids infection at patch r declines as time τ
 increases. Since the random variables $X_{21}(\tau)$ and $X_{22}(\tau)$ are independent, the overall
 212 probability that the parasite has not invaded the novel host by time τ is $\zeta_1(\tau) \zeta_2(\tau)$.

Let the random variable T ($T > 0$) represent the first time a novel host is infected; T is
 214 then the waiting time until the parasite jumps to the novel host. We have

$\Pr[T > \tau] = \zeta_1(\tau) \zeta_2(\tau)$. Then the expected waiting time until the parasite jumps is:

$$\begin{aligned} E[T] &= \int_0^\infty \Pr[T > \tau] d\tau = \int_0^\infty \zeta_1(\tau) \zeta_2(\tau) d\tau \\ &= \frac{G_2}{\lambda_2 N_2} \left\{ \sum_{r=1}^2 p_{2r} \left[1 - \exp\left(-\gamma G_2 E[M_r] + \frac{(\gamma G_2)^2}{2} V[M_r]\right) \right] \right\}^{-1} \end{aligned} \quad (11)$$

216 $E[T]$ the product of the average time elapsing between consecutive patch-visits by the novel
 host, and the expected number of visits (to both patches) for occurrence of the first
 218 infection. The latter may not be obvious; see Appendix A.

3.4 Predictions

220 Inspection of the solution for $E[T]$ yields some intuitive predictions. The mean waiting time
 for the first infection must decrease as $\lambda_2 N_2$ increases. Given group size G_2 , increasing λ_2 or
 222 N_2 increases exposure to the parasite since the rate at which the novel host visits patches is

increased.

224 Recall that increasing reservoir group size G_1 increases each variance $V[M_r]$, but does not
affect mean parasite densities $E[M_r]$. As a consequence, larger group size G_1 increases the
226 mean waiting time $E[T]$. Increasing G_1 reduces the rate at which reservoir hosts visit
patches, but increases the level of parasite shedding when a visit occurs. This combination
228 leaves stationary parasite densities more variable which, in turn, increases the mean time
elapsing before the parasite first infects a novel host.

230 Increasing the variance of parasite shedding by an individual reservoir-host (*i.e.*,
increasing σ^2 with μ fixed) increases parasite-density variances $V[M_r]$, and so increases the
232 expected waiting time $E[T]$. Increasing the mean individual-level parasite shedding μ
increases both the mean and variance of the stationary parasite densities. The former effect
234 can prove stronger, assuming $\gamma G_2 < 1$, so that increasing μ should decrease $E[T]$. In some
cases, σ^2 may depend on μ , and predictions would be revised accordingly.

236 A primary focus concerns the effect of novel-host group size G_2 on the waiting time until
the parasite jumps to the novel host. Differentiating $E[T]$ with respect to G_2 does not yield
238 a simple expression, but allows us to state two necessary, though not sufficient, conditions
such that the waiting time for the first infection should increase with novel-host group size.
240 The first condition depends on the the ζ_r , after Eq. (10), and the second depends on that
expression's derivative. So, $\partial E[T]/\partial G_2 > 0$ requires:

$$E[M_r] > (\gamma V[M_r]/2) G_2 \quad \text{and} \quad E[M_r] > \gamma V[M_r] G_2 - \frac{1}{\gamma G_2}; \quad \text{for } r = 1, 2 \quad (12)$$

242 Each condition should hold when the novel host's susceptibility to the parasite is not too
large. If host-jumping involves crossing a "species barrier," novel-host susceptibility will, by

244 definition, be quite small. These conditions assume that parasite-density means and
variances make Eq. (7) a proper probability; conditions (12) predict reasonably that the
246 waiting time for the parasite to invade the novel host population will increase with group
size G_2 .

248 The increase in the waiting time $E[T]$ with G_2 has a simple, intuitive explanation.
Following our example above, doubling novel-host group size halves the expected number of
250 patch visits per unit time. But the probability that at least one member of the larger group
is infected during a single visit is less than doubled, by the non-linearity in Eq. (6). Hence
252 larger reservoir-host groups expect more time to elapse before the first infection.

Figure 1 shows how the waiting time $E[T]$ depends on the respective host-species group
254 sizes. For any novel-host group size G_2 , waiting time for a host-jump increases with G_1 . For
any reservoir-host group size G_1 , waiting time increases with G_2 . The former effect is
256 reduced for $G_2 = 1, 2$, but $E[T]$ increases strongly with G_1 once $G_2 > 5$. Figure 1 assumes
the reservoir prefers patch 1; $p_{11} = 0.9$. Intuitively, as the novel host's use of patch 1
258 declines, $E[T]$ increases.

Figure 2 displays the waiting time $E[T]$ as a function of the reservoir host's proportional
260 use of patch 1, p_{11} . Note that $E[T]$ is independent of p_{21} when the reservoir host acts as an
ideal generalist ($p_{11} = p_{12} = 0.5$). Comparing plots 2A and 2B shows, for given parameter
262 values, that as the novel-host changes from visiting patches as solitaries to moving in groups
of 15, the expected time until their first infection increases approximately 2 orders of
264 magnitude.

Comparing plots 2B and 2C shows that increasing the absolute number of reservoir-host
266 patch visits, for any (p_{11}, p_{21}) -combination, decreases $E[T]$. Comparing plots 2B and 2D
shows that decreasing the variance in the number of parasites shed per reservoir host, σ^2 ,

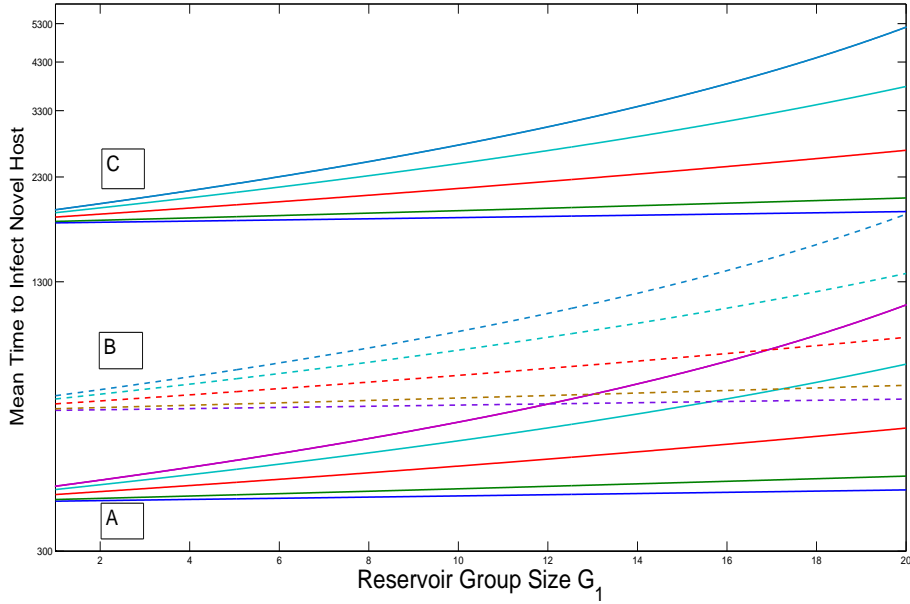


Figure 1: Mean time to jump to novel host: reservoir specializes. Abscissa: reservoir group size G_1 . Ordinate: waiting time $E[T]$, logarithmic scale. Reservoir prefers patch 1; $p_{11} = 0.9$. A. Novel host identical to reservoir; $p_{21} = 0.9$. B. Novel host generalizes; $p_{21} = 0.5$ (broken lines). C. Novel host specializes on patch 2; $p_{21} = 0.1$. A, B, C: As lines ascend in each set, $G_2 = 1, 2, 5, 8, 10$. Longest waiting times occur for different specialists. Parameters: $\lambda_1 N_1 = \lambda_2 N_2 = 0.5$, $\mu = 5$, $\sigma^2 = 25$, $\xi = 1$, $\gamma = 2.5 \times 10^{-3}$.

268 strongly decreases the expected time until the parasite jumps to the novel host. A decrease
in σ^2 decreases the stationary variance of the parasite densities, from Eq. (4). Any decrease
270 in the parasite-density variances $V[M_r]$ decreases $E[T]$, from Eq. (11).

3.4.1 Spatial overlap and infection hazard

272 The reservoir host's use of space influences parasite abundances. Given parasite densities,
the novel host's use of space will govern exposure to infection, and so affect the waiting time
274 until the parasite can jump to the new host. So, $E[T]$ depends directly on overlap between
novel host and parasite, and indirectly on overlap between reservoir and novel hosts. To
276 clarify the role of patch preferences, we show how the hazard of infection varies with the p_{jr} .

Eq. (6) gives us the conditional probability of a host-jump on a single visit to patch r ,
278 given parasite density M_r , as $1 - e^{-\gamma M_r G_2}$. We associate a hazard function with this

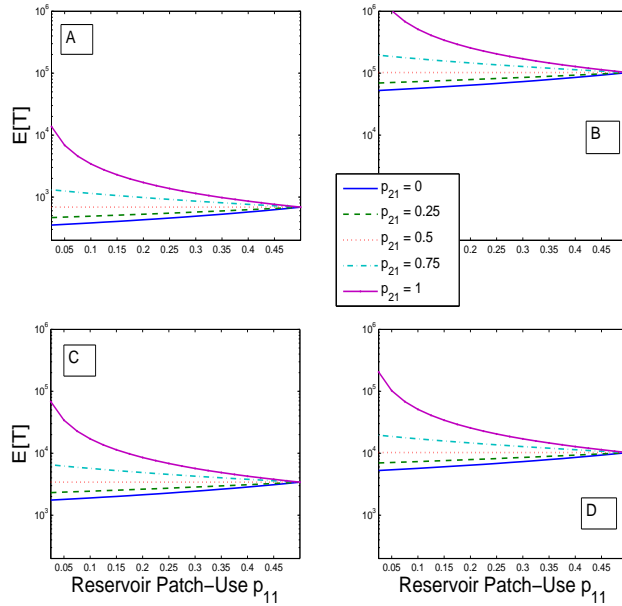


Figure 2: Mean time to jump to novel host as patch-use varies. Abscissa: reservoir preference for patch 1, p_{11} . Ordinate: waiting time $E[T]$, logarithmic scale. A. $G_2 = 1$, $\lambda_2 N_2 = 0.5$, $\sigma^2 = 60$. B. $G_2 = 15$, $\lambda_2 N_2 = 0.5$, $\sigma^2 = 60$. C. $G_2 = 15$, $\lambda_2 N_2 = 15$, $\sigma^2 = 60$. D. $G_2 = 15$, $\lambda_2 N_2 = 0.5$, $\sigma^2 = 0$. Legend shows that in each plot $E[T]$ increases as p_{21} increases, but results converge across levels of p_{21} as reservoir host changes from patch-2 specialist to generalist. Fixed parameters: $G_1 = 10$, $\lambda_1 N_1 = 0.5$, $\mu = 10$, $\xi = 2$, $\gamma = 2.5 \times 10^{-3}$.

infection probability: $h_r(G_2) = \gamma M_r G_2$; we recall that a larger novel-host group increases

280 the hazard of parasite invasion on a single visit. If we average the single-visit hazard

between patches, we have $\langle h_r(G_2) \rangle = E[\gamma G_2 (p_{21} M_1 + [1 - p_{21}] M_2)]$.

282 Eq. (4) provides the mean parasite densities $E[M_r]$, so that the mean hazard of infection becomes:

$$\langle h_r(G_2) \rangle = \frac{\gamma G_2 \lambda_1 N_1 \mu}{\xi} (p_{21} [2p_{11} - 1] + 1 - p_{11}) \quad (13)$$

284 Intuitively, the infection hazard has the minimal feasible value at $p_{21} = 1$ when $p_{11} < 1/2$,

and at $p_{21} = 0$ when $p_{11} > 1/2$. Any degree of bias in the reservoir host's patch (i.e.,

286 $p_{11} \neq 1/2$) implies that strong specialization by the novel host will increase $E[T]$. These

considerations might prove important if hosts avoid locations where a parasite's infectious

288 stage is concentrated

[Freeland 1976, Hausfater and Meade 1982, Hutchings et al. 2001, Turner et al. 2014]. If

290 $p_{11} = 1/2$, $f(M_1) = f(M_2)$, and any p_{21} generates the same mean infection hazard. The
greatest hazard rates occur when both species exhibit low patch-use diversity and prefer the
292 same patch. The lowest hazard rates occur when spatial overlap is minimal; each host
prefers a different patch. Both infection-hazard extremes occur when each species has low
294 resource-use diversity [Patil and Taillie 1979]; the difference between low and high hazards
then depends on spatial niche overlap.

296 4 Discussion

Our results offer two important insights regarding a parasite's jump between host species.
298 First, the time expected to elapse before an environmentally transmitted parasite first
infects a novel host increases both with group size in the endemically infected reservoir
300 population, and with group size in the novel host. Second, overlap in the two species' use of
space will not always predict the likelihood of novel-host infection; its predictive utility
302 varies with the degree of reservoir-host patch specialization.

The mechanisms generating the increase in the waiting time for the host-jump differ
304 between the two host species, but both results involve interaction between host group size
and the parasite-density variances. Our model's novel host uses patches independently of
306 both reservoir behavior and parasite densities. As a consequence of the parasite-density
variances interacting with a non-linear probability of infection, larger novel-host group sizes
308 extend the expected time elapsing until the first infection occurs. Once a host-jump has
occurred, however, the parasite's initial spread might be faster in a larger group
310 [Cross et al. 2005].

Reservoir patch preferences govern the degree of spatial heterogeneity. Reservoir group

312 size and random among-individual variation in parasite contamination influence temporal
heterogeneity of each within-patch parasite density. When the reservoir host specializes in
314 its patch use (plots A and C in figure 2) parasite densities become spatially heterogeneous,
and the waiting time for host-jumping increases as between-host species similarity in patch
316 use declines. However, when the reservoir host generalizes in its patch use (plots B and D in
figure 2) spatial heterogeneity of parasite densities disappears, and the novel host's patch
318 preferences have no effect on the waiting time.

For simplicity we have treated space implicitly. In particular applications, the number of
320 resource clusters will be large, and groups may visit specific locations in an approximately
periodic manner [Hausfater and Meade 1982]. Some gregarious species may actively avoid
322 food patches to reduce exposure to parasitism [Hutchings et al. 2001]. Other species may
seek the most productive food patches and, in doing so, increase the chance of infection
324 [Turner et al. 2014]. We also assumed that the probability of infecting a given host
increased with parasite density in a strictly concave manner. The minimal parasite exposure
326 required to infect a host, often termed the infective dose, varies substantially across
host-parasite interactions [Leggett et al. 2012]. As long as environmental densities of the
328 parasite (hence, exposure) characteristically exceed the infective dose, our model should
apply. However, more complex dose-response relationships, and sufficiently small $E[M_r]$,
330 could alter predictions.

Relationships between reservoir patch use and the distribution of parasite densities can,
332 of course, exhibit more complexity than we assume. Some microparasite populations grow
not only within hosts, but also in the environment [Bani-Yaghoub et al. 2012]. Abiotic
334 variation and biotic processes can affect patch-scale parasite densities, or modulate the
infectiousness of a given parasite density, as well as the susceptibility of a given host

336 [Caraco and Wang 2008, Cizauskas et al. 2014b].

As climate change and habitat destruction increasingly alter a host species' use of
338 resources, social behavior, and spatio-temporal overlap with other species, we anticipate
that parasite host-jumping will occur more frequently. Our results demonstrate how host
340 species' social organization and collective use of a spatially heterogeneous resource
environment can influence parasites switching hosts or changing from host specialization to
342 generalization, and suggest directions for empirical research.

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346 A Online appendix: Some model details

The expected parasite densities are:

$$\mathbb{E}[M_1(t)] = \frac{p_{11}}{\xi} \lambda_1 N_1 \mu (1 - e^{-\xi t}) \quad \mathbb{E}[M_2(t)] = \frac{1 - p_{11}}{\xi} \lambda_1 N_1 \mu (1 - e^{-\xi t}) \quad (\text{A1})$$

348 It seems useful to establish probabilistic independence of the parasite densities $M_r(t)$. First,
we assure that the number of patch visits are independent; the demonstration is standard
350 [Ross 1983]. A host species enters the resource set (patches 1 and 2) at constant
probabilistic rate $\lambda N/G$; given entry, p is the probability of visiting patch 1. At time t
352 ($t > 0$) patch 1 has been visited $X_1(t)$ times, and patch 2 has been visited $X_2(t)$ times; the
respective means are $p\lambda(N/G)$ and $(1 - p)\lambda(N/G)$.

354 Let $W(t) = X_1(t) + X_2(t)$, the total number of visits by time t . We want the joint distribution of the $X_r(t)$, and begin by conditioning on the realization of $W(T)$:

$$Pr[X_1(t) = y, X_2(t) = z] = \sum_{w=0}^{\infty} Pr[X_1(t) = y, X_2(t) = z|W(t) = w]Pr[W(t) = w] \quad (\text{A2})$$

356 Given the definition of $W(t)$, the right side of Eq. (A2) must be:

$$Pr[X_1(t) = y, X_2(t) = z|W(t) = y + z]Pr[W(t) = y + z]$$

Since p is a constant, $Pr[X_1(t) = y, X_2(t) = z|W(t) = y + z]$ is simply a binomial random variable:

$$Pr[X_1(t) = y, X_2(t) = z|W(t) = y + z] = \binom{y+z}{y} p^y (1-p)^z \quad (\text{A3})$$

Then, unconditionally:

$$\begin{aligned} Pr[X_1(t) = y, X_2(t) = z] &= \frac{(y+z)!}{y!z!} p^y (1-p)^z e^{-\lambda(N/G)} \frac{(\lambda(N/G))^{y+z}}{(y+z)!} \\ &= e^{-p\lambda(N/G)} \frac{(p\lambda(N/G))^y}{y!} e^{-(1-p)\lambda(N/G)} \frac{((1-p)\lambda(N/G))^z}{z!} \end{aligned} \quad (\text{A4})$$

360 Applying Eq. (A4), we infer that $X_{11}(t)$ and $X_{12}(t)$ are independent Poisson random variables. Our model assumes that each m , the increment to the local parasite density, is an independent, identically distributed realization of the same random variable. Then we conclude that the shot-noise processes, the parasite densities, $M_r(t)$ and their stationary distributions are probabilistically independent.

Here we specify the characteristic time of the parasite mortality process. The characteristic time t_c is usually taken as the square of the survival integral divided by the

integral of the square [Lowen and Teich 1990]. Since we assume simple exponential decay,

368 we have:

$$t_c = \left(\int_0^\infty e^{-\xi t} dt \right)^2 / \int_0^\infty e^{-2\xi t} dt \quad (\text{A5})$$

Then $t_c = (\xi)^{-2}/(2\xi)^{-1} = 2/\xi$, twice the mean longevity of a parasite.

370 The expected waiting time for the first infection, Eq. (11), can be approached informally,

as follows. Novel-host groups collectively enter the set of patches R at probabilistic rate

372 $\lambda N_2/G_2$; the mean time between consecutive visits is the inverse of this rate. Suppose that

the host-jump were to occur at the $w - th$ visit; that is, when $W(\tau) = w$. Conditioned on

374 no infection during the first $(w - 1)$ visits, the *a priori* probability of infection at $W = w$ is

$p_{21}\theta_1 + (1 - p_{21})\theta_2$. θ_r is the probability of a host jump during a single visit to patch r , and

376 does not depend on w . From Eq. (7), we have for $r = 1, 2$:

$$\theta_r = 1 - \exp\left(-\gamma G_2 E[M_r] + \frac{(\gamma G_2)^2}{2} V[M_r]\right) \quad (\text{A6})$$

If the host-jump occurs at visit w , the distribution between patches of the previous $(w - 1)$

378 visits, none of which led to infection, must be binomial. Then the probability that the first

infection occurs at visit w is:

$$\Omega(w) = [p_{21}\theta_1 + (1 - p_{21})\theta_2] \sum_{x=1}^{w-1} \binom{w-1}{x} p_{21}^x (1 - p_{21})^{w-1-x} (1 - \theta_1)^x (1 - \theta_2)^{w-1-x} \quad (\text{A7})$$

380 Then:

$$\begin{aligned} \Omega(w) &= [p_{21}\theta_1 + (1 - p_{21})\theta_2] \sum_{x=1}^{w-1} \frac{(w-1)!}{x! (w-1-x)!} [p_{21}(1 - \theta_1)]^x [(1 - p_{21})(1 - \theta_2)]^{w-1-x} \\ &= [p_{21}\theta_1 + (1 - p_{21})\theta_2] [p_{21}(1 - \theta_1) + (1 - p_{21})(1 - \theta_2)]^{w-1} \end{aligned} \quad (\text{A8})$$

$\Omega(w)$, for $w = 1, 2, \dots$ follows a geometric probability function. The mean number of visits
382 until the host-jump occurs is then $\left[\sum_{r=1}^2 p_{2r}\theta_r\right]^{-1}$. Multiplying the mean time between
novel-host patch-visits and the mean number of visits until first infection yields the text's
384 expression for $E[T]$.

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