

Exploring vector-borne infection ecology in multi-host communities: a case study of West Nile virus

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Abstract

In this study, we develop a model to investigate how ecological factors might affect the **dynamics** of a vector-borne pathogen in a population composed by different hosts which interact with each other. Specifically, we consider the case when different host species compete with each other, as they share the same habitat, and the vector might have different feeding preference, which can also be time dependent. As a prototypical example, we apply our model to study the invasion and spread, during a typical season, of West Nile virus in an ecosystem composed of two competent avian host species and possibly of dead-end host species. We found that competition and vector feeding preferences can profoundly influence pathogen invasion, influencing its probability to start an epidemic, and influencing transmission rates. Finally, when considering time-dependent feeding preferences, as observed in the field, we noted that the virus circulation could be amplified and that the timing of epidemic peaks could be changed. Our work highlights that ecological interactions between hosts can have a profound influence on the dynamics of the pathogen and that, when modeling vector-borne infections, vector feeding behavior should, for this reason, be carefully evaluated.

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1. Introduction

Ecological interactions within and between species, such as competition and consumer-resource relations, can be influenced by infection dynamics of pathogens and parasites, and vice versa. A recent attempt to more systematically address their mutual interactions (Roberts and Heesterbeek, 2013) has focused on invasion of infection into ecological communities. For this purpose, they computed the basic reproduction number \mathcal{R}_0 , defined as the average number of new cases of an infection caused by one typical infected individual, in a population consisting of susceptibles only (Diekmann et al., 2010).

Following Roberts and Heesterbeek (2013), we investigate the case of a vector-borne infection spreading in a population where different species of hosts compete with each other, for instance for food or habitat, and where the vector bites the hosts according to its feeding preferences. Several papers investigating an infection spreading into competing hosts have been published (Bowers and Turner, 1997; Han and Pugliese, 2009; Bokil and Manore, 2013), but to the best of our knowledge none analyzing the case of a vector-borne infection. In this type of infection, the pathogen is usually transmitted to and from the host when the latter is bitten by the vector to obtain a blood meal required for reproduction.

We focus on mosquito-borne infections and analyze a basic system where one vector species takes blood meals from two competent host species that compete ecologically. We show how to compute \mathcal{R}_0 allowing for different reservoir host competence (i.e. probability of transmitting the pathogen to the vector). As a prototypical example, we consider West Nile virus (WNV) in an ecosystem composed of two avian host species. However, the formula for \mathcal{R}_0 is easily generalizable to ecosystems of arbitrary numbers of host and non-host species that interact by competition and/or food web relations (Roberts and Heesterbeek,

28 2013).

29 WNV is a flavivirus first isolated in Uganda in 1937 (Smithburn et al., 1940)

30 and now present on every continent (Reisen, 2013). It is maintained in a bird-

31 mosquito transmission cycle primarily involving mosquitoes belonging to *Culex*

32 species, of which the *pipiens* sub-complex is thought to be one of the most im-

33 portant in Europe (Zeller and Schuffenecker, 2004) and North America (Reisen,

34 2013). Humans and other mammals (e.g. horses) are considered dead-end hosts,

35 i.e. they can not transmit the virus.

36 *Culex* mosquitoes and WNV have a broad host range, and mosquito feeding

37 preferences can change during the season. In fact mosquitoes seem to pref-

38 erentially bite certain hosts more than others, even if they are less available

39 (Kilpatrick et al., 2006a; Thiemann et al., 2011; Simpson et al., 2012; Tay-

40 lor et al., 2012; Rizzoli et al., 2015); moreover its preferences seem to change

41 during the breeding season (Kilpatrick et al., 2006b; Thiemann et al., 2011;

42 Burkett-Cadena et al., 2012).

43 Many models have been proposed to study West Nile virus dynamics among

44 different bird species (Cruz-Pacheco et al., 2005, 2012; Maidana and Yang,

45 2011; Simpson et al., 2012) but they do not explicitly investigate ecological

46 interactions between the hosts or the effects of changes in host preference over

47 the season. Our aim is to investigate how ecological interactions, such as com-

48 petition, and shifting mosquito feeding preferences can affect the invasion of

49 a pathogen and therefore change the outcome relatively to a baseline scenario

50 which does not include such features.

51 **2. The model**

52 We analyze the simplest case with only two competing species, both compe-

53 tent hosts for an infection transmitted by a vector with population size V . In

54 addition, we assume that hosts can not recover, but may die due to the infection.

55 To this aim, we develop a compartmental model similar to the one proposed by

56 Lord et al. (1996) with hosts and vectors classified according to whether they

57 are susceptible S or infected I . Although vector-borne infections are usually
 58 transmitted only by the vector, we consider also the possibility of host-to-host
 59 transmission, as this has been shown to be possible for West Nile virus among
 60 crows by Komar et al. (2003).

61 To model the competition among birds we assume, as in (Gamarra et al., 2005),
 62 that they both follow a Lotka-Volterra dynamics. In addition, the mosquito
 63 population dynamics is assumed to be density dependent; in particular, we as-
 64 sume that density can affect larval development and survival, as observed by
 65 Agnew et al. (2010).

66 The equations of the model are

$$\begin{cases}
 N'_1 &= r_1 \left(1 - \frac{N_1 + c_{12}N_2}{K_1}\right) N_1 - \alpha_1 I_1 \\
 N'_2 &= r_2 \left(1 - \frac{N_2 + c_{21}N_1}{K_2}\right) N_2 - \alpha_2 I_2 \\
 V' &= (n_E \sigma b_{max} (1 - \rho_V V) - \mu_V) V \\
 I'_1 &= \left[p_{V1} b_1 \frac{I_V}{N_1} + \beta_{11} p_{11} \frac{I_1}{N_1} + \beta_{12} p_{21} \frac{I_2}{N_2} \right] S_1 - \left(\alpha_1 + \mu_1 + r_1 \frac{a_{11}N_1 + a_{12}c_{12}N_2}{K_1} \right) I_1 \\
 I'_2 &= \left[p_{V2} b_2 \frac{I_V}{N_2} + \beta_{22} p_{22} \frac{I_2}{N_2} + \beta_{21} p_{12} \frac{I_1}{N_1} \right] S_2 - \left(\alpha_2 + \mu_2 + r_2 \frac{a_{22}N_2 + a_{21}c_{21}N_1}{K_2} \right) I_2 \\
 I'_V &= \left[p_{1V} b_1 \frac{I_1}{N_1} + p_{2V} b_2 \frac{I_2}{N_2} \right] S_V + (q_V n_E \sigma b_{max} (1 - \rho_V V) - \mu_V) I_V
 \end{cases} \quad (1)$$

67 where

- 68 • $N_i = S_i + I_i$ is the number of individuals of species i with $i \in \{1, 2\}$;
- 69 • $r_i = \eta_i - \mu_i > 0$ is the growth rate of species $i \in \{1, 2\}$, where η_i and
 70 μ_i are the birth and death rate respectively. They are assumed not to be
 71 influenced by the vector. Each species has a certain carrying capacity K_i ;
- 72 • c_{ij} represents the effect of competition of species j on species i with
 73 $i, j \in \{1, 2\}, i \neq j$;
- 74 • $a_{ij} \in [0, 1]$ is the proportion of competition from species j that affects the
 75 death rate of species i with $i, j \in \{1, 2\}$;
- 76 • n_E is the number of eggs laid by a gravid mosquito and σ is the probability
 77 that an egg becomes an adult;

- 78 • b_{max} is the vector biting rate, which can be thought of as the inverse
79 of the length of the gonotrophic cycle, i.e. the interval spanned between
80 the blood meal and the oviposition. Bites are divided between the two
81 host populations with b_1, b_2 , denoting the biting rates on species 1 and 2
82 respectively, and therefore $b_{max} = b_1 + b_2$;
- 83 • $\rho_V < 1$ is the density dependent factor on vector fecundity. We can then
84 define K_V as the vector carrying capacity, as follows:

$$K_V := \frac{1}{\rho_V} \left(1 - \frac{\mu_V}{n_E \sigma b_{max}} \right).$$

- 85 • μ_V is the vector death rate;
- 86 • α_i is the additional death rate for species i due to the infection;
- 87 • p_{ij} is the probability that an infected individual of type $i \in \{1, 2, V\}$
88 infects a susceptible individual of type $j \in \{1, 2, V\}$, given contact or bite;
- 89 • β_{ij} is the direct transmission rate between host species i and j ;
- 90 • q_V is the probability of vertical transmission, i.e. the probability that an
91 infected mosquito passes the virus to its offspring.

92 If there is no infection, the Jacobian of system (1) at the Infection-free Equilib-
93 rium is given by

$$J = \begin{pmatrix} C & D \\ 0 & H \end{pmatrix}$$

94 where

$$C = \begin{pmatrix} \frac{r_1}{K_1} (K_1 - c_{12}N_2 - 2N_1) & -\frac{c_{12}r_1N_1}{K_1} & 0 \\ -\frac{c_{21}r_2N_2}{K_2} & \frac{r_2}{K_2} (K_2 - c_{21}N_1 - 2N_2) & 0 \\ 0 & 0 & n_E \sigma b_{max} (1 - 2\rho_V V) - \mu_V \end{pmatrix}$$

95 represents the ecological community dynamics of the two host species and the
 96 vector. The lower 3×3 matrix H in the Jacobian represents the epidemiological
 97 dynamics of the two host species and the vector species:

$$H = \begin{pmatrix} p_{11}\beta_{11} - (\alpha_1 + \tilde{\mu}_1) & \frac{\beta_{12}p_{21}N_1}{N_2} & p_{V1}b_1 \\ \frac{p_{12}\beta_{21}N_2}{N_1} & \beta_{22}p_{22} - (\alpha_2 + \tilde{\mu}_2) & p_{V2}b_2 \\ \frac{p_{1V}b_1V}{N_1} & \frac{p_{2V}b_2V}{N_2} & q_V n_E \sigma b_{max} (1 - \rho_V V) - \mu_V \end{pmatrix}$$

98 where $\tilde{\mu}_i = \mu_i + r_i \frac{a_{ii}N_i + a_{ij}c_{ij}N_j}{K_i}$ with $i \in \{1, 2\}$.

99 The matrix D in the upper right corner is

$$D = \begin{pmatrix} -\alpha_1 & 0 & 0 \\ 0 & -\alpha_2 & 0 \\ 0 & 0 & 0 \end{pmatrix}.$$

100 The infection-free non trivial equilibrium is

$$N_1^* = \frac{K_1 - c_{12}K_2}{1 - c_{12}c_{21}}, \quad N_2^* = \frac{K_2 - c_{21}K_1}{1 - c_{12}c_{21}}, \quad V^* = K_V \quad (2)$$

101 which exists (i.e. $N_1^* \geq 0, N_2^* \geq 0, V^* \geq 0$) and is stable provided

$$K_i > c_{ij}K_j, \quad \mu_V < n_E \sigma b_{max}.$$

102 We observe that $N_i^*/K_i > N_j^*/K_j$ (i.e. population size of species i is depressed
 103 by competition less than species j) when $c_{ji} \frac{K_i}{K_j} > c_{ij} \frac{K_j}{K_i}$. In particular when
 104 $K_i = K_j$ (which can also be assumed through an appropriate scaling) we can
 105 equate size of competition coefficients with depression of population size.

106 As in (Roberts and Heesterbeek, 2013) we write $H = T + \Sigma$ where T is the
 107 epidemiological transmission matrix

$$T = \begin{pmatrix} p_{11}\beta_{11} & \frac{\beta_{12}p_{21}N_1^*}{N_2^*} & p_{V1}b_1 \\ \frac{p_{12}\beta_{21}N_2^*}{N_1^*} & \beta_{22}p_{22} & p_{V2}b_2 \\ \frac{p_{1V}b_1V^*}{N_1^*} & \frac{p_{2V}b_2V^*}{N_2^*} & q_V n_E \sigma b_{max} (1 - \rho_V V^*) \end{pmatrix}$$

and Σ is the epidemiological transition matrix

$$\Sigma = \begin{pmatrix} -(\alpha_1 + \tilde{\mu}_1) & 0 & 0 \\ 0 & -(\alpha_2 + \tilde{\mu}_2) & 0 \\ 0 & 0 & -\mu_V \end{pmatrix}$$

108 and therefore the so called next-generation matrix with large domain (Diekmann
109 et al., 2010) is

$$\mathcal{K} = -T\Sigma^{-1} = \begin{pmatrix} \frac{p_{11}\beta_{11}}{(\alpha_1 + \tilde{\mu}_1)} & \frac{\beta_{12}p_{21}N_1^*}{N_2^*(\alpha_2 + \tilde{\mu}_2)} & \frac{p_{V1}b_1}{\mu_V} \\ \frac{p_{12}\beta_{21}N_2^*}{N_1^*(\alpha_1 + \tilde{\mu}_1)} & \frac{\beta_{22}p_{22}}{(\alpha_2 + \tilde{\mu}_2)} & \frac{p_{V2}b_2}{\mu_V} \\ \frac{p_{1V}b_1V^*}{N_1^*(\alpha_1 + \tilde{\mu}_1)} & \frac{p_{2V}b_2V^*}{N_2^*(\alpha_2 + \tilde{\mu}_2)} & q_V \end{pmatrix}$$

110 and \mathcal{R}_0 is the dominant eigenvalue of \mathcal{K} .

111 2.1. Infections without horizontal transmission

112 This is probably the most common case, since, as explained above, most
113 vector-borne infections are transmitted only by the vector, so $p_{ij} = 0, i, j \in \{1, 2\}$.

114 In this case the next-generation matrix becomes

$$\mathcal{K} = \begin{pmatrix} 0 & 0 & \frac{p_{V1}b_1}{\mu_V} \\ 0 & 0 & \frac{p_{V2}b_2}{\mu_V} \\ \frac{p_{1V}b_1V^*}{N_1^*(\alpha_1 + \tilde{\mu}_1)} & \frac{p_{2V}b_2V^*}{N_2^*(\alpha_2 + \tilde{\mu}_2)} & q_V \end{pmatrix}$$

115 and the formula for \mathcal{R}_0 is

$$\mathcal{R}_0 = \frac{1}{2} \left(\sqrt{\frac{4p_{V1}p_{1V}b_1^2}{\mu_V(\alpha_1 + \tilde{\mu}_1)} \frac{V^*}{N_1^*} + \frac{4p_{V2}p_{2V}b_2^2}{\mu_V(\alpha_2 + \tilde{\mu}_2)} \frac{V^*}{N_2^*} + q_V^2} + q_V \right) \quad (3)$$

116 with N_1^*, N_2^*, V^* as in (2).

117 We note that the biting rates b_1, b_2 play a crucial role for \mathcal{R}_0 , which depends
118 also on the vector to host ratio which is in turn driven by the competition co-
119 efficients and the carrying capacities.

120 In order to make (3) more perspicuous, we can simplify it by assuming that an

121 infected vector passes the pathogen to any susceptible host with same probabil-
 122 ity $p_{V1} = p_{V2} = p_{VH}$ and that it can not transmit the virus to its offspring (so
 123 $q_V = 0$). We can further assume that mosquitoes bite hosts according to their
 124 density, so $b_i = b_{max} \frac{N_i}{N_1 + N_2}$ with $i \in \{1, 2\}$. In this case (3) reduces to

$$\mathcal{R}_0 = \frac{b_{max}}{N_1^* + N_2^*} \sqrt{\frac{p_{VH}V^*}{\mu_V}} \sqrt{\frac{p_{1V}N_1^*}{\alpha_1 + \tilde{\mu}_1} + \frac{p_{2V}N_2^*}{\alpha_2 + \tilde{\mu}_2}} \quad (4)$$

125 **with N_1^*, N_2^* as in (2).** From (4) we see that competition does not affect \mathcal{R}_0
 126 linearly. In fact, for fixed c_{21} , increasing c_{12} will decrease N_1^* while at the same
 127 time N_2^* and $\tilde{\mu}_1$ will increase. Thus one term inside the square root will increase
 128 with c_{12} while the other will decrease; as a consequence, the overall effect on
 129 \mathcal{R}_0 is not straightforward.

130 2.2. Horizontal transmission

131 In this case we assume, as observed for WNV (Komar et al., 2003), that
 132 horizontal transmission can happen only between individuals belonging to the
 133 same species, so $p_{ij} = 0, i \neq j$. In this case

$$\mathcal{K} = \begin{pmatrix} \frac{p_{11}\beta_{11}}{(\alpha_1 + \tilde{\mu}_1)} & 0 & \frac{p_{V1}b_1}{\mu_V} \\ 0 & \frac{p_{22}\beta_{22}}{(\alpha_2 + \tilde{\mu}_2)} & \frac{p_{V2}b_2}{\mu_V} \\ \frac{p_{1V}b_1V^*}{N_1^*(\alpha_1 + \tilde{\mu}_1)} & \frac{p_{2V}b_2V^*}{N_2^*(\alpha_2 + \tilde{\mu}_2)} & q_V \end{pmatrix}.$$

134 \mathcal{R}_0 is then the largest root of a 3-rd order equation. We will consider some
 135 numerical examples in the next section.

136 Finally, in order to investigate the combined effect of horizontal and vector
 137 transmission, we consider, for the sake of simplicity, that there is only one host
 138 species, say species 1. Then, with only horizontal transmission

$$\mathcal{R}_0^h = \frac{p_{11}\beta_{11}}{(\alpha_1 + \tilde{\mu}_1)},$$

139 while, with only vector transmission (and $q_V = 0$),

$$\mathcal{R}_0^V = \sqrt{\frac{p_{V1}p_{1V}b_1^2V^*}{\mu_V N_1^*(\alpha_1 + \tilde{\mu}_1)}}.$$

140 When both transmission routes operate, one obtains

$$\mathcal{R}_0 = \sqrt{\left(\frac{p_{11}\beta_{11}}{4(\alpha_1 + \tilde{\mu}_1)}\right)^2 + \frac{p_{V1}p_{1V}b_1^2V^*}{\mu_V N_1^*(\alpha_1 + \tilde{\mu}_1)}} + \frac{p_{11}\beta_{11}}{\alpha_1 + \tilde{\mu}_1} = \sqrt{\frac{(\mathcal{R}_0^h)^2}{4} + (\mathcal{R}_0^V)^2} + \frac{\mathcal{R}_0^h}{2}. \quad (5)$$

141 From (5) we can note that $\mathcal{R}_0 \leq (>)1 \iff \mathcal{R}_0^h + (\mathcal{R}_0^V)^2 \leq (>)1$.

142 3. Numerical example

143 Here, we present a numerical example to explore the influence of vector and
 144 host ecology, in our setting, on invasion of the infectious agent. In particular, we
 145 study the invasion of WNV with two bird species. We selected their respective
 146 parameters among the most competent species, that are American crow (*Corvus*
 147 *brachyrhynchos*, species 1) and House finch (*Haemorhous mexicanus*, species 2),
 148 as found in (Komar et al., 2003). We assume horizontal transmission only in
 149 species 1 (American crow), since Komar et al. (2003) found its occurrence in
 150 this species only, and assume $p_{22} = 0$.

151 Finally, we also assume that the vector has a fixed daily biting rate $b_{max} = b_1 + b_2$.
 152 The baseline parameters, with their description, are reported in Table 1.

153

154 3.1. Effect of vector and host ecology on \mathcal{R}_0

155 We assume that the vector bites its hosts according to their density with a
 156 fixed daily rate $b_{max} = 0.2$, and $a_{11} = a_{22} = a_{12} = a_{21} = 0.5$. In Figure 1 the
 157 effect of competition on \mathcal{R}_0 is shown. In Figure 1a (left panel) we can see that
 158 for a fixed value of c_{ij} , increasing c_{ji} will increase \mathcal{R}_0 and the highest values are
 159 reached when c_{12} is particularly large, so when species 1 (which has a higher
 160 probability of transmitting the virus to the vector) is much less abundant than
 161 the other. The lowest values are expected when the competition is not very high.
 162 We remark that \mathcal{R}_0 is always greater than 1 and, as expected from formula (5),
 163 it is also greater than the one computed without vector transmission. In fact in
 164 this latter case $\mathcal{R}_0^h \sim 0.73$ (see Supplementary Material), thus mosquitoes are
 165 crucial for the pathogen invasion and transmission.

Parameter	Description	Value	Source
p_{V1}, p_{V2}	Transmission probability mosquito to bird	0.88	(Turell et al., 2001)
p_{1V}	Transmission probability species 1 to mosquito	0.5	(Komar et al., 2003)
p_{2V}	Transmission probability species 2 to mosquito	0.28	(Komar et al., 2003)
$p_{11}\beta_{11}$	Contact transmission rate in crows	0.33	(Hartemink et al., 2007)
μ_V	Death rate in mosquitoes (/day)	0.08	(Hartemink et al., 2007)
q_V	Transovarial transmission rate	0.004	(Hartemink et al., 2007)
α_1	Species 1 WNV-related mortality rate (/day)	0.2	(Komar et al., 2003)
α_2	Species 2 WNV-related mortality rate (/day)	0.11	(Komar et al., 2003)
n_E	Number of mosquito eggs in one batch	200	(Hartemink et al., 2007)
σ	Survival probability egg to female mosquito	0.1	(Hartemink et al., 2007)
μ_1, μ_2	Bird death rate	0.001	(Bowman et al., 2005)
r_1, r_2	Bird growth rate	0.5	(Bowman et al., 2005)
$\frac{K_V}{(K_1+K_2)}$	Mosquito to bird ratio	5	(Cruz-Pacheco et al., 2005)
K_1, K_2	Carrying capacities for birds	1000	Assumption
$a_{11}, a_{12}, a_{21}, a_{22}$	Proportion of competition affecting the death rate	Varying	

Table 1: parameters.

166 Figure 1b (right panel) shows how \mathcal{R}_0 is influenced by competition and its
167 contribution to the death rate, represented by a_{12}, a_{21} . \mathcal{R}_0 is greater when host
168 death rates are less affected by competition (as this increases expected life of
169 infected individuals) and, in all three cases, it increases linearly with $c_{12} = c_{21}$.
170 Since the three cases ($a_{12} = a_{21} = 0.1, 0.5, 0.9$) do not differ substantially, from
171 now on we consider only the case with $a_{12} = a_{21} = a_{22} = a_{11} = 0.5$.

172 As observed in many field studies, *Culex pipiens* may show different feeding
173 preferences for different avian species (Kilpatrick et al., 2006a; Simpson et al.,
174 2012; Thiemann et al., 2011). We say that a species, say i , is preferred if
175 $\frac{b_i}{b_{max}} > \frac{N_i}{N_i + N_j}$, i.e. its fraction of bites is higher than its frequency. Hence we
176 can model the mosquito biting preference introducing, as in (Simpson et al.,
177 2012), the feeding preference index $\delta_V, \delta_V \geq 1$. According to this the biting
178 rates become

$$b_i = b_{max} \frac{\delta_V N_i}{\delta_V N_i + N_j}, \quad b_j = b_{max} \frac{N_j}{\delta_V N_i + N_j}.$$

179 We study the cases $b_{max} = 0.1, 0.2, 0.3$; as the biting rate can be interpreted as
180 the reciprocal of the duration of the gonotrophic cycle, we are assuming that it

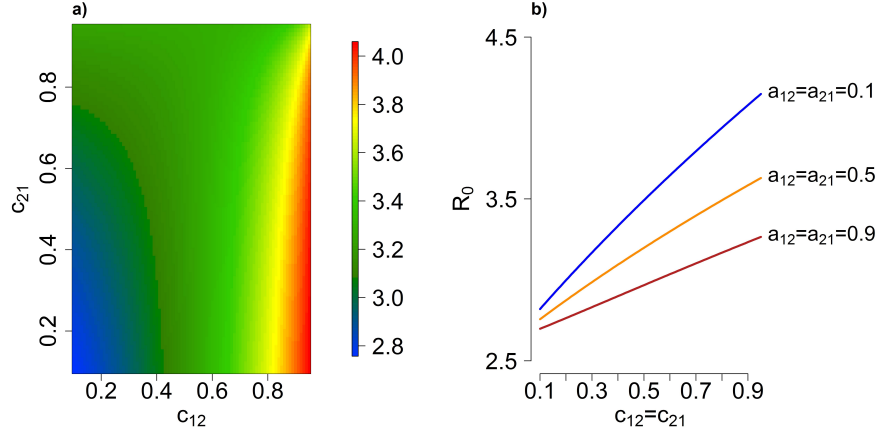


Figure 1: Panel a) \mathcal{R}_0 according to the competition coefficients with $a_{12} = a_{21} = 0.5$. Panel b) \mathcal{R}_0 according to the competition coefficients (only for the case $c_{21} = c_{12}$) and three different values for $a_{12} = a_{21}$. In both panels $b_{max} = 0.2$.

181 varies between 3 and 10 days, that seems to be a realistic estimate (Faraj et al.,
 182 2006; Jones et al., 2012).

183 Figure 2 shows how the value of \mathcal{R}_0 depends on the values of the different
 184 ecological ingredients (competition coefficients c_{12}, c_{21} and vector feeding pref-
 185 erence δ_V). Continuous lines represent the case when species 1 is preferred,
 186 while dashed lines when the vector prefers species 2. On the x -axis δ_V , the
 187 feeding preference index, ranges from 1 to 10. Different panels refer to different
 188 values of (c_{12}, c_{21}) that assume respectively the values of (0.1, 0.5, 0.9).

189 In every case we observe that $\mathcal{R}_0 > 1$, so the infection-free equilibrium will
 190 always be unstable.

191 The interplay of both competition and feeding preference is rather complex;
 192 however we see that they both affect significantly \mathcal{R}_0 . Higher values of \mathcal{R}_0 can
 193 be observed when the vector prefers to feed on the less abundant host. For
 194 example, if species 1 is preferred and $c_{12} = 0.9$, $c_{21} = 0.1$ (i.e. species 1 is
 195 less abundant than species 2), we can observe that \mathcal{R}_0 reaches its maximum
 196 values. Conversely, if the most abundant species is preferred, \mathcal{R}_0 does not seem

197 to increase significantly if δ_V increases. Actually, it may slightly decrease: for
 198 instance if $c_{12} = 0.5$, $c_{21} = 0.1$, $b_{max} = 0.1$ and species 2, the less infectious
 199 one, is preferred, then \mathcal{R}_0 is 1.49, 1.46, 1.49, 1.5 for $\delta_V = 1, 2, 3, 4$ respectively
 200 (see blue dashed line in upper central panel). Eventually, we can also note that
 201 if no species is ecologically advantaged (i.e. $c_{12} = c_{21}$, panels on the diagonal),
 202 then the patterns are quite similar but the values are higher when competition
 203 is strong ($c_{12} = c_{21} = 0.9$) and if species 1 is preferred.

204 Figure 2 shows the case with both horizontal and vector transmission. The
 205 results for the model considering only vector transmission are very similar and
 presented in **Figure B1** in the Supplementary Material.

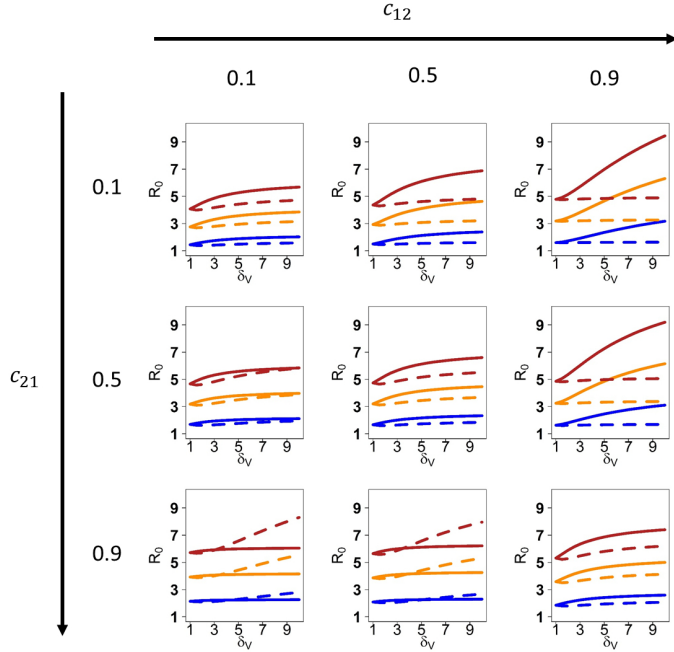


Figure 2: \mathcal{R}_0 with $b_{max} = 0.1$ (blue), 0.2 (orange) and 0.3 (red) as function of the competition coefficients $c_{12}, c_{21} \in \{0.1, 0.5, 0.9\} \times \{0.1, 0.5, 0.9\}$ and the feeding preference index δ_V ranging from 1 to 10. Continuous (dashed) lines regard the case when species 1 (2) is preferred.

206

207 *3.2. Effect of competition and shifting mosquito feeding preference on infection*
 208 *seasonal dynamics*

209 To model a typical season, we add another host type M , representing a
 210 mammal species, individuals of which are bitten at rate b_M . They do not have
 211 any interaction with other hosts and we assume they are a closed population.
 212 Moreover, we assume that they are dead-end hosts, so they do not infect the
 213 vector, and that they can recover and become immune for life.
 214 Hence, we add to system (1) the three following equations

$$\begin{cases} N'_M &= 0 \\ I'_M &= p_{VM}b_M I_V \frac{S_M}{N_M} - \alpha_M I_M \\ R'_M &= R_M + \alpha_M I_M \end{cases}$$

215 The description of the new parameters and their values are reported in Table
 216 2. As far as we know, there are no empirical estimates for the probability of
 217 transmission to any mammal species. Hence in our simulations we consider two
 218 values for p_{VM} . In the first case we assume $p_{VM} = p_{V1} = p_{V2} = 0.88$ as in
 219 (Bowman et al., 2005), in the other case we assume p_{VM} value and order of
 220 magnitude less than p_{V1}, p_{V2} , i.e. $p_{VM} = 0.088$.

221 An important ecological aspect that affects mosquito seasonal dynamics is the
 222 diapause (Denlinger and Armbuster, 2014). It is a common mechanism adopted
 223 by mosquitoes to survive winter; in the case of *Culex pipiens*, only adult females
 224 undergo diapause, i.e. they do not lay new eggs until the following spring.
 225 Daylight duration plays a key role in its activation (Spielman and Wong, 1973;
 226 Denlinger and Armbuster, 2014). To take into account this feature, we introduce
 227 a new variable γ , which is a function of the daylight duration **as published**
 228 **in (Marini et al., 2016)**. It ranges from 0 to 1 and it is shown in Figure 3
 229 (dotted line in panel d). The equation for V in (1) is replaced by

$$V' = [n_E \sigma \gamma b_{max} (1 - \rho_V V) - \mu_V] V.$$

230 The simulations start on June 1 in a given year with an infected bird belonging
 231 to species 1 and lasts 6 months. γ is modeled according to the daylight duration

232 recorded at 46°N latitude (see Supplementary Material for more details).
 233 Instead of simulating the deterministic system (1), we consider a Markov chain
 234 whose transition probabilities corresponds to the rates of the differential equa-
 235 tions in (1) (see **section A in the** Supplementary Material for more details).
 236 We decided to follow a stochastic approach to be able to account for demo-
 237 graphic stochasticity, relevant for instance at the invasion stage.

Parameter	Description	Value	Source
N_M	Number of mammals	1000	Assumption
α_M	Recovery rate from WNV for dead end hosts (/day)	0.07	(Bowman et al., 2005)
p_{VM}	Transmission probability mosquito to mammal	Varying	

Table 2: Mammal parameters.

238

239 3.2.1. Baseline case

240 Here we present the outcome of the model when inter-species competition is
 241 absent, i.e. $c_{12} = c_{21} = 0$, the vector does not have a preferred avian species (i.e.
 242 $\delta_V = 1$) and its biting rate is $b_{max} = 0.2$. So for each host type $i \in \{1, 2, M\}$
 243 the biting rate is

$$b_i = b_{max} \frac{N_i}{N_1 + N_2 + N_M}.$$

244 Figure 3 shows the prevalence, i.e. the number of infected divided by the total
 245 number of individuals, for each host type i and the vector during the season.
 246 Avian species 1 experiences a higher infection than the other two host popu-
 247 lations. Highest prevalence in the vector is recorded very late in the season;
 248 this may be due to our assumption that towards the end of the season almost
 249 no mosquitoes reproduce, as γ is very close to zero, and therefore the influx of
 250 susceptible vectors is very low in that period. Maximal prevalence in the host
 251 species is expected much earlier. For mammal, this occurs two months after the
 252 beginning of the season (10th of August), when $p_{VM} = p_{V1} = p_{V2}$ (black line
 253 in panel c), or around three months after the beginning (middle of September),

254 when $p_{VM} = p_{V1} \cdot 10^{-1}$ (red line in panel c), **similar** to what happens in the
 255 avian populations.

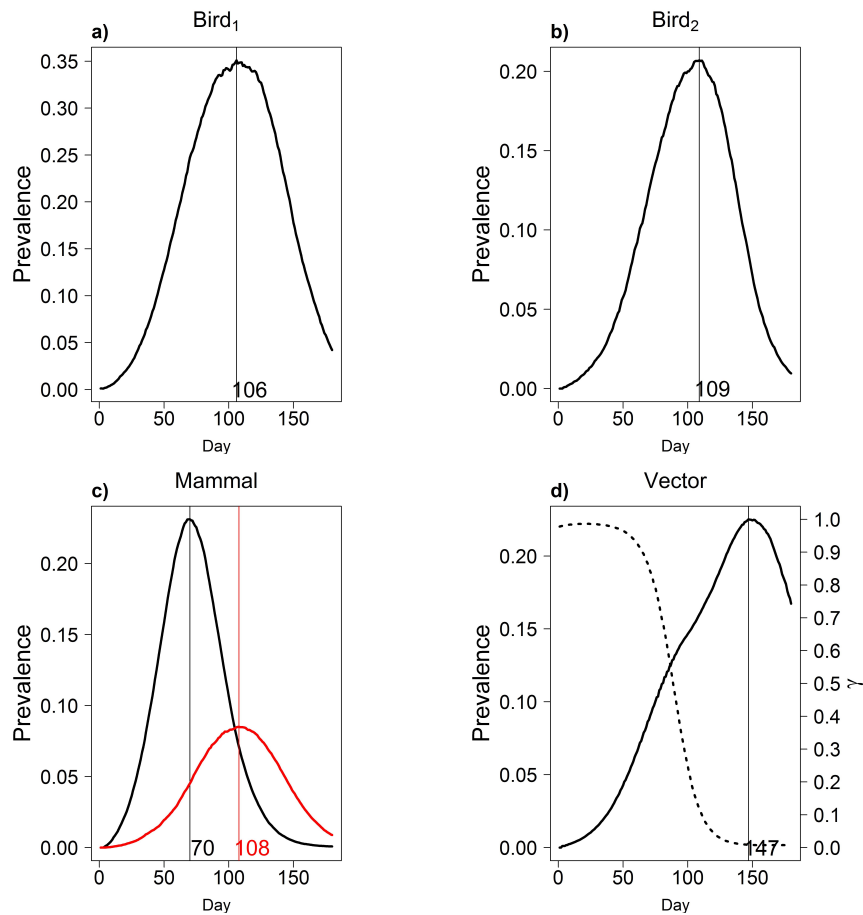


Figure 3: **Prevalence** for the three host types (panels a-c) and the vector (panel d). Black (red) line in panel c represent the outcome with $p_{VM} = 0.88$ (0.088). The vertical lines show the day at which the maximal prevalence is reached. Parameter γ is shown in panel d (dotted line).

256

257 *3.2.2. Including feeding preference shift only*

258 For the sake of simplicity we assume, in this Subsection, that there is no
 259 competition, i.e. $c_{12} = c_{21} = 0$, and that there is no preference between the
 260 two avian species, i.e. $\delta_V = 1$, but there is preference between birds and mam-
 261 mals. In particular this preference, which shifts through the season, is modeled
 262 according to the functions presented in (Kilpatrick et al., 2006b). More specif-
 263 ically, we assume that at time t the vector bites a host of type $i \in \{bird, M\}$
 264 with probability $f_i(t)$ with $f_{bird}(t) + f_M(t) = 1$. Therefore the biting rates are

$$b_i(t) = f_i(t)b_{max} \quad i \in \{bird, M\}$$

265 with $f_i(t)$ as shown in Figure 4 (panels a-c) and

$$b_i = b_{bird} \frac{N_i}{N_i + N_j} \quad i, j \in \{1, 2\}.$$

266 The inclusion of shifts in feeding preference significantly affects virus spread.
 267 As shown in Figure 4, **the highest prevalences for all populations oc-**
 268 **cur earlier compared to the scenario in which the** biting rates are time
 269 independent (Figure 3). More precisely, they are expected about two months
 270 (beginning of August) after the introduction of the first infected bird. The two
 271 avian species exhibit a similar pattern, but we can note that, compared to the
 272 baseline case, the prevalences have a much higher maximum (about 60%), and
 273 they decrease to zero more quickly.

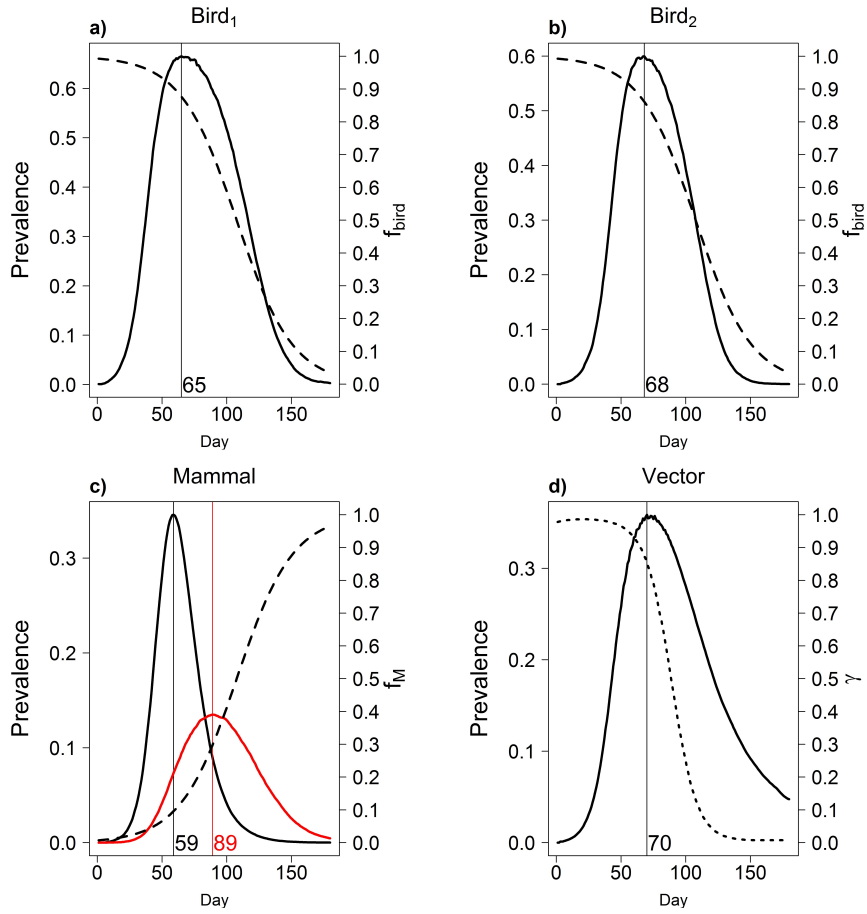


Figure 4: **Prevalence** for the three host types (panels a-c) and the vector (panel d). Black (red) line in panel c represent the outcome with $p_{VM} = 0.88$ (0.088). The shifting mosquito feeding preferences are represented in panel a, b ($f_{bird}(t)$) and c ($f_M(t)$) with dashed lines. The vertical lines show the times when the maximal prevalence is reached. The function γ (diapause rate) is shown in panel d (dotted line).

274 *3.2.3. The complete model without feeding preference shift*

275 In this case we allow c_{12} and c_{21} to be different from 0, while $b_{max} = 0.2$ as
276 previously and $p_{VM} = 0.88$ as in (Bowman et al., 2005). We explore a range of
277 c_{12}, c_{21} combinations, with $(c_{12}, c_{21}) \in (0.05, 1) \times (0.05, 1)$. Moreover, we study
278 three cases of vector preference for birds: no preference, preference for species
279 1 or for species 2, in which cases $\delta_V = 5$.

280 The inclusion of competition produces rather different outcomes, depending on
281 vector preference, as shown in Figure 5, where the central column presents the
282 no-preference case while the first and third column show the cases of prefer-
283 ence for species 1 and 2 respectively. When there is no preference between the
284 two bird species, all four populations present smaller maximal prevalences if
285 the species with higher infectiousness (1) is less abundant, i.e. if it is severely
286 affected by the competition with the other ($c_{12} \gg c_{21}$). The same observation
287 can be made for the vector in the case it prefers the more infectious bird species
288 (1), and when this latter has a strong ecological disadvantage ($c_{12} > 0.6$). For
289 avian species, the maximal prevalence is higher for the preferred bird popu-
290 lation. Moreover, for the preferred avian species (say i), its highest maximal
291 prevalence values are reached when $c_{ij} \gg c_{ji}$, i.e. when it has a strong ecological
292 disadvantage. This corresponds to what found for the value of \mathcal{R}_0 in Section
293 3.1.

294 If species 1 (with higher probability of infecting the vector) is preferred, then
295 the maximal prevalence for both host and vector population is recorded much
296 earlier in the season **compared to when the vector prefers species 2 or**
297 **there is no preference between them** (see Figure 6, blue boxplots). More-
298 over, as shown in Figure C1 in the Supplementary Material, if this host type
299 is also extremely affected by competition with the other species, then the vec-
300 tor maximal prevalence is expected less than two months after the beginning
301 of the epidemics, so much earlier with respect to cases with low competition
302 coefficients values.

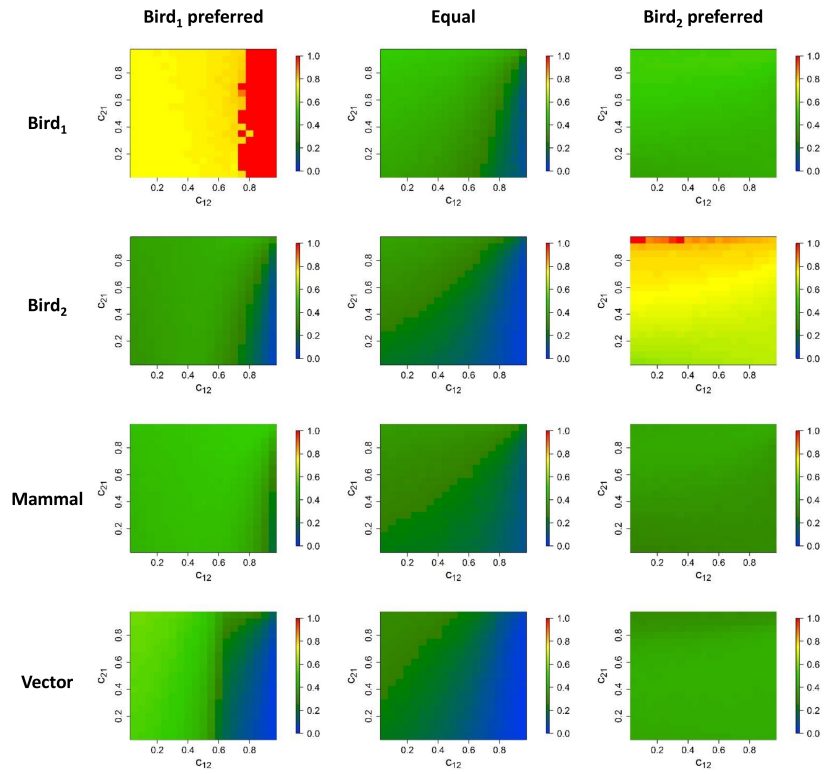


Figure 5: Maximal prevalence for the three host types and the vector for the model without feeding preference shift (Subsection 3.2.3), where species 1 (left) or species 2 (right) is preferred ($\delta_V = 5$), or there is no preference ($\delta_V = 1$, center). Values range from 0 (dark blue) to 1 (bright red). Avian species 1 is assumed to be more infectious than species 2 ($p_{1V} > p_{2V}$).

303 *3.2.4. The complete model with feeding preference shift*

304 In this Subsection, we study how the virus circulation is affected by com-
305 petition, as in Subsection 3.2.3, but with the time-dependent biting rates as
306 presented in Subsection 3.2.2.

307 As shown in Figure 7, if the vector changes its feeding preference during the
308 season, then the expected maximal prevalence in vector, hosts and mammals
309 increases in comparison with the case with constant preference. This is consis-
310 tent with what observed in Subsection 3.2.2 where we investigated the case with
311 time-dependent biting rates but without competition.

312 If there is no preference between the two bird populations (central column), it
313 can be seen that the lowest maximal prevalences are expected when the compe-
314 tition is not particularly high, similarly to what we observed for \mathcal{R}_0 in Figure
315 1. On the other hand, if there is a preference for species i , its maximal preva-
316 lence is much higher than that of the other avian population. Furthermore, if
317 the preferred host species is strongly affected by competition (large c_{ij}), both
318 avian maximal prevalences are larger, consistent with the computation of \mathcal{R}_0 in
319 Section 3.1 and with the simulations presented in Subsection 3.2.3, where we
320 studied the same scenario with time-independent biting rates.

321 In this case, infection prevalence in mammals and vector is not significantly
322 affected by bird competition and its value is around 50% for every (c_{12}, c_{21})
323 combination.

324 As found previously in the case without preference between birds and no com-
325 petition (Subsection 3.2.2), **maximal prevalence occurs earlier than when**
326 **there is no shift**. In fact, as shown in Figure 6, avian and vector infection
327 prevalence peaks are expected to occur from two to three months earlier re-
328 spectively. Moreover, the maximal prevalence is recorded earlier in the season
329 when mosquitoes prefer to feed on the more infectious avian species (see Figure
330 6, orange boxplots), while there does not seem to be a significant difference
331 between the cases $\delta_V = 1$ and $\delta_V = 5$ when species 2 is preferred.

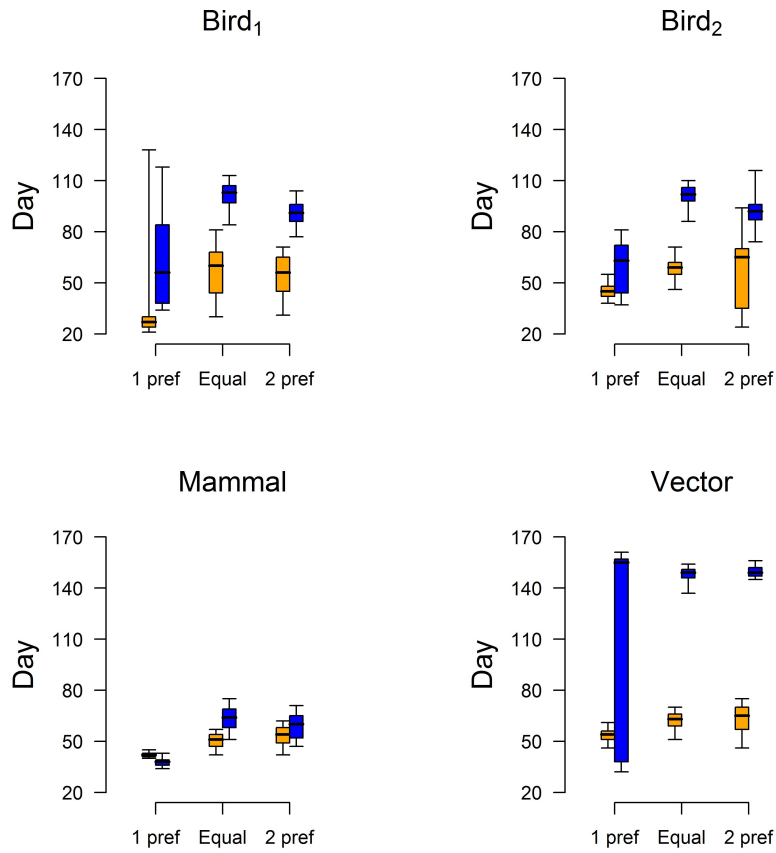


Figure 6: Boxplot of the maximal prevalence recording time (number of days after the introduction of the first infected host) in the cases with competition coefficients $(c_{12}, c_{21}) \in (0.05, 1) \times (0.05, 1)$, and where species 1 or species 2 is preferred ($\delta_V = 5$), and when there is no preference ($\delta_V = 1$) with (without) the assumption of shifting vector feeding preference in orange (in blue). **Whiskers: 2.5% and 97.5% quantiles; box: 25% and 75% quantiles; thick line: median.**

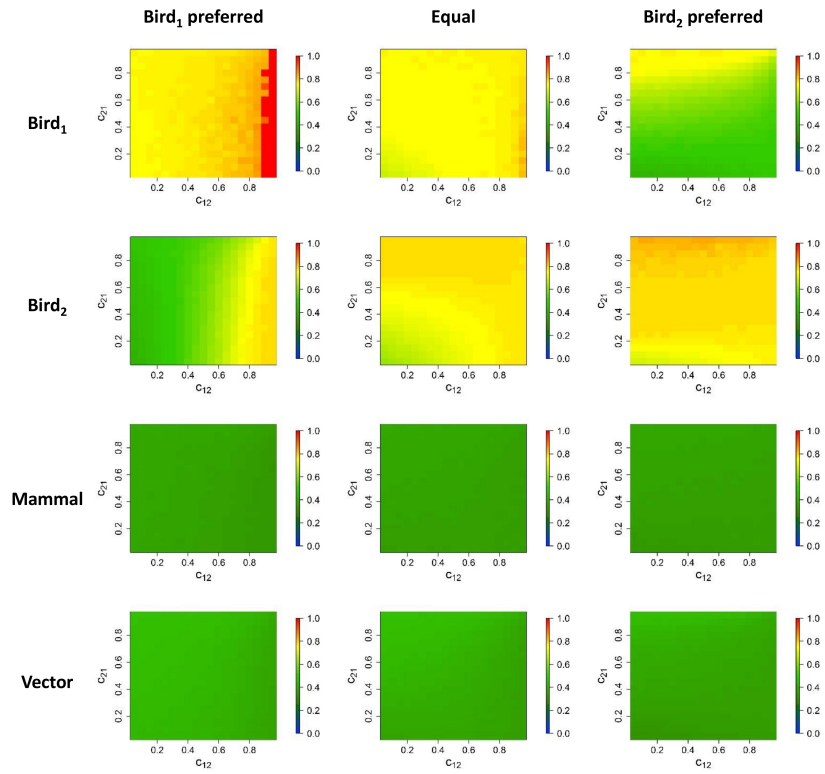


Figure 7: Maximal prevalence for the three host types and the vector in the cases with competition and time-dependent biting rates (Subsection 3.2.4), where species 1 (left) or species 2 (right) is preferred ($\delta_V = 5$), or there is no preference ($\delta_V = 1$, center). Values range from 0 (dark blue) to 1 (bright red).

332 4. Conclusions

333 In this paper, we presented a mathematical framework to investigate a
334 vector-borne infection spreading in a multi-host community where individuals
335 can interact with each other epidemiologically as well as ecologically (in particu-
336 lar by competition), following the study presented in (Roberts and Heesterbeek,
337 2013) and the model proposed in (Lord et al., 1996).

338 We observed that competition may increase \mathcal{R}_0 by decreasing host population
339 sizes (and thus increasing vector/host ratios), but that at the same time it might
340 decrease host life expectancy and in this way decrease \mathcal{R}_0 . A general pattern of
341 the effect of competition on \mathcal{R}_0 is therefore difficult to establish, as the influence
342 on infection dynamics very much depends on the ecological particularities of the
343 system one studies. \mathcal{R}_0 is also strongly influenced by the vectors' biting rate,
344 but also by vector feeding preferences, which may cause a large increase of \mathcal{R}_0
345 if the less abundant host is the preferred one. On the other hand, \mathcal{R}_0 might be
346 smaller if the vector tends to feed on the less competent host.

347 In order to be able to obtain more precise conclusions, we **focused on** a particu-
348 lar case, the **spread** of West Nile Virus within an avian population composed
349 by two different species that share the same habitat and compete for resources.
350 We explored a wide range of values for the ecological ingredients, such as ecolog-
351 ical interactions and vector feeding preference, using epidemiological parameter
352 values that have been estimated for WNV. We found that \mathcal{R}_0 can be strongly
353 influenced by competition and feeding preferences (see Figures 1 and 2).

354 We also used the model, parameterized for WNV, to simulate seasonal epi-
355 demics, and thus studying the effect of competition and vector preference on
356 transient dynamics. This model included also dead-end hosts, typically mam-
357 mals for WNV, and allowed for a shifting preference of vectors, from birds in
358 the first part of the season to mammals in the second part, as shown to occur
359 in natural systems by Kilpatrick et al. (2006b), Thiemann et al. (2011) and
360 Burkett-Cadena et al. (2012). One effect of the presence of dead-end hosts is
361 a dilution effect (Keesing et al., 2006), as they decrease the circulation of the

362 virus by wasting, from the pathogen transmission point of view, a proportion
363 of the vector bites. This effect is no longer observed when assuming time-
364 dependent vector feeding preference; in fact, in this case mosquitoes bite only
365 competent hosts at the beginning of the season, enhancing the increase of in-
366 fection prevalence; indeed, the virus would circulate among mosquitoes with a
367 higher incidence than in the case when mosquitoes are assumed to feed also on
368 mammals, which are assumed to be dead-end hosts, as studied in Subsection
369 3.2.1. From the simulations, it also appears that, with shifting vector prefer-
370 ences, infection prevalence in dead-end hosts and vectors is not influenced by
371 bird competition (compare Figures 5 and 7), which in this case affects infection
372 spread only among avian populations.

373 Shifting feeding preference during a season has another important consequence:
374 the times of highest prevalence in a season are recorded around the same pe-
375 riod for both vectors and birds, i.e. about two months after the start of the
376 epidemics. This result agrees with actual observations. For instance, Bell et al.
377 (2005), Lukacik et al. (2006) and Reisen et al. (2010) recorded the highest WNV
378 prevalence in mosquitoes in August in different parts of the US, while Nemeth
379 et al. (2007) and Kwan et al. (2010) noted that the highest records of WNV
380 avian cases are during summer (June-July). On the other hand, if it is assumed
381 that vector feeding preferences are fixed throughout the season (Subsections
382 3.2.1 and 3.2.3), one can see that the prevalence peaks later in the season and
383 in vectors later than in birds. We argue that the assumption of changing feed-
384 ing preferences is important when studying the seasonal pattern of infections in
385 vector-borne pathogen models.

386 The model considered here does not attempt to be realistic for any specific in-
387 fection, even though some assumptions and parameter values have been tailored
388 for WNV. In reality, *Culex* mosquitoes bite a large number of bird and mammal
389 species, some of which will be dead-end hosts, others will be of different compe-
390 tence for the transmission of WNV (Komar et al., 2003). The model we studied
391 considered only two avian species, both highly competent. Possibly, the rather
392 high prevalence of WNV in the simulations, as well as the high values of \mathcal{R}_0 , are

393 an artifact arising from this simplified situation. Another questionable assump-
394 tion is that birds are not allowed to recover, though antibody-positive birds are
395 not difficult to find in endemic areas (Jozan et al., 2003; Mckee et al., 2015).
396 Including a compartment of recovered birds would not change the values of \mathcal{R}_0
397 but would certainly decrease infection prevalence. Despite these limitations, we
398 believe that this study of a simplified situation gave important insights on the
399 importance of ecological interactions and vector feeding preferences in shaping
400 infection dynamics in a multi-host-vector system.

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