Dynamics of SI models with both horizontal and vertical transmissions as well as Allee effects

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Abstract

A general SI (Susceptible-Infected) epidemic system of host-parasite interactions operating under Allee effects, horizontal and/or vertical transmission, and where infected individuals experience pathogeninduced reductions in reproductive ability, is introduced. The initial focus of this study is on the analyses of the dynamics of Density-Dependent and Frequency-Dependent effects on SI models (SI-DD and SI-FD). The analyses identify conditions involving horizontal and vertical transmitted reproductive numbers, namely those used to characterize and contrast SI-FD and SI-DD dynamics. Conditions that lead to disease-driven extinction, or disease-free dynamics, or susceptible-free dynamics, or endemic disease patterns are identified. The SI-DD system supports richer dynamics including limit cycles while the SI-FD model only supports equilibrium dynamics. SI models under "small" horizontal transmission rates may result in susceptible-free dynamics. SI models under with and inefficient reproductive infectious class may lead to disease-driven extinction scenarios. The SI-DD model supports stable periodic solutions that emerge from unstable equilibrium provided that either the Allee threshold and/or the disease transmission rate is large; or when the disease has limited influence on the infectives growth rate; and/or when disease-induced mortality is low. Host-parasite systems where diffusion or migration of local populations manage to destabilize them are examples of what is known as diffusive instability. The exploration of SI-dynamics in the presence of dispersal brings up the question of whether or not diffusive instability is a possible outcome. Here, we briefly look at such possibility within two-patch coupled SI-DD and SI-FD systems. It is shown that relative high levels of asymmetry, two modes of transmission, frequency dependance, and Allee effects are capable of supporting diffusive instability.

Keywords: Allee Effects, Horizontal Transmission, Vertical Transmission, Disease-Driven Extinction, Disease-Free Dynamics, Susceptibles-Free Dynamics, Diffusive Instability.

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1. Introduction

Parasitism contributes to the selection of future generations of hosts through their impact on factors that lead to reductions in fitness (Hudson *et al.* 2002) and as a result, wildlife managers must account for emerging and/or re-emerging diseases. Competition for space and resources (finding mates or food) also impact the reproductive ability and likelihood of survival of individuals, particularly those housing pathogens or parasites. Hosts' dynamics (survival in particular) often depends on the ability of a population to maintain a critical mass (Kang and Castillo-Chavez 2012). The impact of heterogenous transmission factors including multiple transmission modes by altering a population's dynamics may lessen the plausibility of conservation goals or the economic viability of selected management policies (Potapov *et al* 2012). Hence, it is not surprising that the pressure which parasites or pathogens place on their hosts and the relation of such interactions to community and/or ecosystem structure has been the subject of continuous empirical and theoretical studies. Some of the theoretical consequences associated to host-pathogen dynamics when factors like: i) multiple modes of disease transmission; (ii) host population density; and (iii) the presence or absence of critical host population thresholds, are addressed in this manuscript.

Modes of disease transmission, like horizontal and vertical, differentially facilitate the colonization of host populations by bacteria, fungi, or viruses. Colonization (horizontal transmission) is sometimes seen as the result of close interactions (contacts) between disease-free host and infected individuals. A contact process that implicitly assumes the sharing of a common, local habitat. The passage of a disease-causing agent from a mother to offspring during the "birth" process is also sometimes possible (vertical transmission). Feline leukemia (FeLV) and feline immunodeficiency (FIV) viruses are transmitted horizontally and vertically. Leishmaniasis, a disease caused by the protozoan parasite *Leishmania infantum*, is transmitted horizontally and vertically. Domesticated dog populations are presumed to be a reservoir for *Leishmania infantum*; a reservoir maintained by the differential contributions of multiple modes of transmission (Santaella *et al.* 2011). The deadly *septicaemia*, which manages to kill 80% of *septicaemia*-infected birds, gets lodged in the ovary of surviving birds; passed later to the birds eggs (vertical transmission); spreading horizontally within the hatcher and brooder.

Teasing out the roles of density- and frequency-dependent transmission (DDT versus FDT) on the dynamics of host-parasite systems is carried out for theoretical and policy reasons. FDT is the result of density-independent contact rates between susceptible and infected individuals. DDT assumes that *infection risks increase* with host density. Density-dependent transmission (DDT) may require a minimal number of available susceptible hosts, that is, a threshold density, for transmission to occur. Density-dependent parasitic disease transmission plays a role in regulating host population size (Anderson and May 1978 &1991) while frequency-dependent parasitic transmission does not require host density thresholds or regulatory host population constraints on the birth or death rates to "work" (Getz and Pickering 1983).

In population biology we often lack absolutes. And so, vector- and sexually-transmitted diseases have been seen to thrive in frequency-dependent transmission settings while density-dependent infections that lead to pathogens being shed by infected hosts into common environments may sometimes need a critical mass of susceptible individuals to thrive (Anderson and May 1991; Antonovics *et al.* 1995). Pathogens can be spread via "direct" contacts (kissing can spread herpes viruses), aerosol (sneezing can spread influenza viruses), or via indirect contacts (ingesting water contaminated with fecal material can cause result in cholera infections), or through vectors (ticks and mosquitoes often spread viruses and bacteria to their hosts), or via some combination of direct and indirect modes, sometimes mediated by a vector. Empirical work on mice, voles, lady bird beetles, frogs, and plants has shown that pathogen transmission often involve DD and FD transmission modes, with one predominant mode (Hudson *et al.* 2002). The negative impact of deliberate releases of pathogens via aerosol or in water systems tends to increase with host density. On the other hand, sexually transmitted pathogens seem to thrive equally well or bad in small or large population settings while some vector-borne diseases have been shown to support frequency-dependent transmission patterns (Anderson and May 1991; Antonovics and Alexander 1992; Ferrari *et al* 2011). Antonovics and Alexander (1992) manipulated the density and frequency of infected hosts *Silene latifolia* and in the process they found out that deposition of the anther smut fungus *Microbotryum violaceum* by pollinating insects managed to increase with the frequency of infection.

A pathogen may or may not be deleterious enough to regulate the dynamics of host populations and so it is not surprising that the impact of pathogens on hosts is tied in to virulence. Pathogen's levels of virulence differentially impacts host's fitness. Often, increases in virulence result in a reduced probability survival or a diminished ability of a host to reproduce successfully, or both (Anderson and May 1979; Hudson et al. 2001; Hilker et al. 2009). Pathogens whose transmission successes increases with host density seem to have managed to select for variants capable of regulating a host population. Dwyer et al (1990) studied a host-pathogen system where a detailed account of virus titer on infected hosts could be estimated. Their study focused on studying the ability of the Myxoma virus to control an exploding rabbit populations over a long window in time. Empirical evidence from systems involving conjunctivitis in house finches or parasitic nematodes in red grouse and feral Soay sheep provide an example of a system where disease regulates population size (Gulland 1992; Hudson et al. 1998; Hochachka and Dhondt 2000). Pathogen infections are contributors to the decline or the extinction of some species (Dwyer et al 1990; Daszak et al. 1999; Harvell et al. 2002; Smith et al. 2006; Thieme et al. 2009). The deleterious role of *chytridiomycosis* in amphibians, chestnut blight in American chestnuts, avian malaria in Hawaiian birds, devil facial tumour disease in Tasmanian devils, or sudden oak death in Californian trees provide classical examples of the role of disease in regulating a population. Theory suggests that density-dependent specialist pathogens (i.e., those infecting a single host) alone rarely drive their hosts' extinction but can lead to extinction of the pathogen while frequency-dependent transmission may be capable of supporting significant decreases, including the potential extinction of host and parasite populations in the presence of moderately lethal pathogens (de Castro and Bolker 2005; Ferrari et al 2011; Kilpatrick and Altizer 2012).

The impact of disease outbreaks can be devastating and their dynamics must be particularly monitored within populations near extinction; that is, those with population levels near established Allee effects thresholds (Allee 1938; Stephens and Sutherland 1999; Stephens *et al* 1999; Courchamp *et al* 2009; Kang and Lanchier 2011). The relevance of threshold effects has been identified within a wide array of taxa (Courchamp *et al.* 2008; Kramer *et al.* 2009). Populations under Allee effects or facing extinction or both must be effectively managed (Drake 2004; Hilker 2009). The fragility of these populations means that limiting the transmission of highly deleterious diseases is critical (Deredec and Courchamp 2007; Hilker 2009). Recurrent infectious disease outbreaks tend to enhance the deleterious role of Allee effects within diseases capable of inducing reductions in host fitness (Hilker *et al.* 2005; Deredec and Courchamp 2006; Yakubu 2007; Hilker *et al.* 2009; Thieme *et al.* 2009; Hilker 2010; Friedman and Yakubu 2011; Kang and Castillo-Chavez 2013). The results of this manuscript seem to be in sync with the overall conclusions reached the study of predator-prey systems (e.g., Cushing 1994; Emmert and Allen 2004; Drew *et al* 2006; Jang and Diamond 2007; Berezovskaya *et al.* 2010; Kang and Armbruster 2011; Kang and Castillo-Chavez 2012).

Parasites and hosts co-evolve in response to environmental clues and/or selective pressures (Kilpatrick and Altizer 2012). Mammals, birds, fish, and insects generate mobility patterns as they track resources and as it is well known movement and/or dispersal can impact disease dynamics (Altizer 2010). In short, mobility has been a key player in the evolution of host-parasite systems. Studies that in addition to disease and mobility (dispersal) also include the impact of Allee effects are not well understood (Rios-Soto *et al.* 2006; Hikler *et al.* 2007; Kang and Castillo-Chavez 2013b). Hilker *et al.* (2007) used a reaction-diffusion SI model within a frequency-dependent transmission framework in their explorations of the impact of disease and mobility on the spatiotemporal patterns of disease transmission. SI models that incorporate disease-reduced fertility have been explored by a number of researchers (Diekmann and Kretzshmar (1991) and Berezovskaya *et al.* (2004)). In Kang and Castillo-Chavez (2013b) a two-patch SI model with density-dependent transmission is used to show that the differential movement of susceptible and infected individuals can enhance or suppress the spread of a disease. A SI model that incorporates a horizontally and vertically transmitted disease; infectives giving birth to infectives; susceptibles giving birth to susceptibles; Allee effects within the net reproduction term; disease-induced death rate; and disease reduced reproductive ability, is used in this manuscript to begin to address questions that include: What is the role of multiple modes of transmission? Will density-dependent and frequency-dependent vertical transmission affect host-parasite dynamics differentially? Under what conditions would Allee effects alter disease-free dynamics or facilitate disease-driven extinction? Would Allee thresholds on reproductive fitness become altered (reduced) by disease? What is the role of DDT or FDT in support of diffusive instability?

In Section 2, a general SI (Susceptible-Infected) model with Allee effects built in the reproduction that incorporates horizontal and vertical transmission modes, is formulated. The basic dynamic properties of the model are characterized, in particular, sufficient conditions in support of disease-free and persistence of species results are identified (Theorem 2.1 and its corollary 2.1). In Section 3, the dynamics of SI models under frequency-dependent or density-dependent horizontal transmission are contrasted. Boundary dynamics are characterized (Proposition 3.1) and sufficient conditions for diseaseand susceptible-population persistence are provided in Theorem 3.1. A classification of interior dynamics comes in Theorem 3.2. In Section 4, disease-driven extinction, disease-free or susceptible-free dynamics, and endemic persistent dynamics are characterized. The nature of bifurcations supported by SI models is studied with the aid of the reproduction numbers linked to horizontal and vertical transmission modes. In Section 5, sufficient conditions leading to diffusive instability (Theorem 5.1) are identified. The nature of mechanisms potentially capable of supporting diffusive instability in SI-models and preypredator models, is briefly discussed. The implications of the results in this manuscript are discussed in the Conclusion.

2. An SI model with Allee effects and vertical transmission

The model outlined in this section deals with a population facing a disease that can be effectively captured within a SI framework under assumptions that include the possibility of multiple modes of transmission, that is, horizontal and vertical. It is therefore assumed that infected individuals can give birth to infected hosts; that Allee effects alter the net reproduction term (possibly due to mating limitations or predator saturation); the presence of increases in mortality due to disease-induced deaths; and the fact that infected individuals may experience reductions in reproductive ability.

We let S and I denote the susceptible and infective populations, respectively, with N = S + I denoting the total host population. The approach followed from Deredec and Courchamp (2006) leads to the following set of nonlinear system after the incorporation of the above assumptions:

$$\frac{dS}{dt} = \underbrace{rSf(N)}_{\text{Growth with Allee effects}} - \underbrace{\phi(N)\frac{I}{N}S}_{\text{Horizontal transmission}}$$
(1)
$$\frac{dI}{dt} = \phi(N)\frac{I}{N}S + \underbrace{\rho rIf(N)}_{\text{Vertical transmission}} - \underbrace{dI}_{\text{Additional death due to infections}}$$

where $r > 0, \rho \in [0, 1], d > 0$ are respectively the intrinsic growth rate, the reduction of growth rate due to disease, and the excess death rate from the disease. The horizontal transmission term $\phi(N)$ includes density-dependent transmission, $\phi(N) = \beta N$ (the law of mass action) or frequency-dependent transmission, $\phi(N) = \beta$ (proportionate mixing or standard incidence). In the absence of disease, the SI Model (1) reduces to the following single species growth model:

$$\frac{dN}{dt} = rNf(N) \tag{2}$$

where the per capita growth function rf(N) is subject to strong Allee effects, i.e., there exists an Allee threshold K^- and a carrying capacity K^+ such that

$$f(N) < 0 \text{ if } 0 < N < K^{-} \text{ or } N > K^{+}; f(N) > 0 \text{ if } K^{-} < N < K^{+}; f(K^{-}) = f(K^{+}) = 0.$$
(3)

Thus, the population model described by Equation (2) converges to 0 if initial conditions are below K^- or converges to its carrying capacity K^+ whenever the initial conditions are above K^- .

Note: If $\rho = 1$, Model (1) is reduced to Model (6), which was introduced by Deredec and Chourchamp (2006) while if $\rho = 0$, Model (1) can be seen as a special case of the models studied in Kang and Chavez-Castillo (2012). The formulation of this *SI* Model (1) is similar in approach to that found in Boukal and Berec (2002), Courchamp *et al* (2009), Hilker *et al* (2009), and Thieme *et al* (2009), particularly in the way we model the effects of Allee effects and disease. The literature on the use of phenomenological models like Model (2) is extensive (e.g., see Lewis and Kareiva 1993; Gruntfest *et al* 1997; Alvarez 1998; Padrón and Trevisan 2000; Shi and Shivaji 2006; Hilker 2010; Friedman and Yakubu 2012). Our models allow for infectives to give birth to infectives with the caveat that their reproductive ability may be reduced; a feat that being captured with the parameter ρ .

The need for biological consistency (well posed model) is addressed with the help of the statespace naturally associated with Model (1), namely, $X = \{(S, I) \in \mathbb{R}^2_+\}$ with its interior defined as $\mathring{X} = \{(S, I) \in \mathbb{R}^2_+ : SI > 0\}$. The state space when $\phi(N) = \beta$ is $X = \{(S, I) \in \mathbb{R}^3_+ : S + I > 0\}$. The assumption that f(N) is differentiable leads to the following theorem for Model (1):

Theorem 2.1 (Basic dynamical features of (1)). Assume that $r > 0, d > 0, \rho \in [0, 1]$ and both f, ϕ are continuous in X with f satisfying Condition (3), then System (1) is positively invariant and bounded in X with the following property

$$\limsup_{t \to \infty} N(t) = \limsup_{t \to \infty} S(t) + I(t) \le K^+.$$

In addition, we have the following:

- 1. If $N(0) \in (0, K^{-})$, then $\lim_{t \to \infty} N(t) = 0$.
- 2. If there exists a positive number $\alpha > K^-$ such that $r\rho f(\alpha) > d$, then

$$\liminf_{t \to \infty} N(t) = \liminf_{t \to \infty} S(t) + I(t) \ge \alpha \text{ for any } N(0) > \alpha.$$

3. If
$$\max_{K^- \le N \le K^+} \{ \frac{K^+ \phi(N)}{N} + r \rho f(N) \} < d$$
, then $\limsup_{t \to \infty} I(t) = 0$.

Proof. Since both S = 0 and I = 0 are invariant manifold for System (1), then according to the continuity of the system, we can easily show that (1) is positively invariant in X. In addition, System (1) gives the following equation:

$$\frac{dN}{dt} = r(S + \rho I)f(N) - dI$$
(4)

where N = S + I. Thus, if $K^- \leq N \leq K^+$, we have the following

$$r\rho Nf(N) - dN = N(r\rho f(N) - d) \le \frac{dN}{dt} = r(S + \rho I)f(N) - dI \le rNf(N)$$
(5)

which indicates that

$$\lim_{t \to \infty} N(t) = \lim_{t \to \infty} S(t) + I(t) = 0 \text{ if } N(0) < K^{-1}$$

since f(N) satisfies Condition (3), i.e.,

$$f(N) > 0$$
 when $K^- \le N \le K^+$.

If $N < K^-$ or $N > K^+$, then we have the following

$$rNf(N) - dN = N(rf(N) - d) \le \frac{dN}{dt} = r(S + \rho I)f(N) - dI \le r\rho Nf(N)$$
(6)

which indicates that

$$\limsup_{t \to \infty} N(t) = \limsup_{t \to \infty} S(t) + I(t) \le K^+ \text{ if } N(0) > K^-$$

since f(N) satisfies Condition (3), i.e.,

$$f(N) < 0$$
 when $N < K^-$ or $N > K^+$.

If there exists a positive number $\alpha > K^-$ such that $r\rho f(\alpha) > d$, then according to (6), we have

$$\frac{dN}{Ndt}\Big|_{N=\alpha} \ge r\rho f(N) - d\Big|_{N=\alpha} > 0.$$

Therefore, $\liminf_{t\to\infty} N(t) = \liminf_{t\to\infty} S(t) + I(t) \ge \alpha$ for any $N(0) > \alpha$.

If $\max_{K^- < N < K^+} \{ \frac{K^+ \phi(N)}{N} + r \rho f(N) \} < d$, then from (1) and the fact that $\limsup_{t \to \infty} N(t) \le K^+$, we have

$$\frac{dI}{dt} = \phi(N)\frac{I}{N}S + \rho r If(N) - dI = I\left(\frac{S\phi(N)}{N} + r\rho f(N) - d\right) < I\left(\frac{K^+\phi(N)}{N} + r\rho f(N) - d\right) < 0$$

which indicates that $\limsup_{t\to\infty} I(t) = 0$.

Notes: Some of the consequences that follow from Theorem 2.1 are:

- 1. The size of the initial population is extremely important for persistence regardless of the disease due to Allee effects.
- 2. The total population population will not be above its carrying capacity K^+ in the long run.
- 3. Species persistence requires that the initial population α is above the Allee threshold K^- , and excess deaths $\frac{d}{r\rho}$ are small enough, smaller than the per capita growth function evaluated at the total population α , i.e., $f(\alpha) > \frac{d}{r\rho}$.
- 4. In the absence of vertical transmission, disease free dynamics requires that

$$\max_{K^- \le N \le K^+} \left\{ \frac{\phi(N)}{N} \right\} < \frac{d}{K^+}$$

While in the presence of vertical transmission, disease free dynamics requires that

$$\max_{K^- \le N \le K^+} \{\frac{\phi(N)}{N} + \frac{r\rho f(N)}{K^+}\} < \frac{d}{K^+}.$$

For convenience, we can consider that f(N) has a generic form of $(N-K^-)(K^+-N)$ and $\phi(N) = \beta$ (i.e., frequency-dependent) or βN (i.e., the law of mass action) then scaling and setting

$$S \to \frac{S}{K^+}, \ I \to \frac{I}{K^+}, \ K^- \to \frac{K^-}{K^+}, \ t \to rt, \ \rho \to \frac{\rho}{r}, \ \beta \to \frac{\beta}{r}, \ d \to \frac{d}{r}$$

leads to the following two SI models with both horizontal and vertical transmission and Allee effects:

$$\frac{dS}{dt} = S(N-\theta)(1-N) - \frac{\beta SI}{N}$$
(7)

$$\frac{dI}{dt} = \frac{\beta SI}{N} - dI + \rho I(N - \theta) (1 - N)$$
(8)

and

$$\frac{dS}{dt} = S(N-\theta)(1-N) - \beta SI$$
(9)
$$\frac{dI}{dI} = \beta SI - dI + \rho I(N-\theta)(1-N)$$
(10)

$$\frac{dI}{dt} = \beta SI - dI + \rho I (N - \theta) (1 - N)$$
(10)

where $f(N) = r(N-\theta)(1-N)$ is the per capita growth in the absence of disease; the parameter $\theta = \frac{K^{-1}}{K^{+1}}$ represents the Allee threshold; $\rho \in [0, 1]$ represents the reduce reproductive ability due to the disease; β represents the disease transmission rate while d denotes the additional death rate coming from infections. The direct application of Theorem 2.1 to System (7)-(8) and (9)-(10) gives the following corollary:

Corollary 2.1 (Basic dynamic features of (7)-(8) and (9)-(10)). System (7)-(8) and System (9)-(10)are positively invariant and bounded in their state space X with the following property

$$\limsup_{t \to \infty} N(t) = \limsup_{t \to \infty} S(t) + I(t) \le 1$$

In addition, we have the following:

- 1. If $N(0) \in (0, \theta)$, then $\lim_{t \to \infty} N(t) = 0$.
- 2. If there exists a positive number $\alpha > \theta$ such that $\rho f(\alpha) > d$, then

$$\liminf_{t \to \infty} N(t) = \liminf_{t \to \infty} S(t) + I(t) \ge \alpha \text{ for any } N(0) > \alpha.$$

3. If $\beta + \frac{\rho(1-\theta)^2}{4} < d$, we have $\limsup_{t\to\infty} I(t) = 0$.

Proof. The application of Theorem 2.1 is direct. We only show the item 3. Since $N = S + I \ge S$ and $\limsup_{t\to\infty} N(t) = \limsup_{t\to\infty} S(t) + I(t) \le 1$, therefore, for System (7)-(8), we have

$$\frac{S\phi(N)}{N} + r\rho f(N) = \frac{\beta S}{N} + \rho(N-\theta)(1-\theta) \le \beta + \rho(N-\theta)(1-\theta) \le \beta + \frac{\rho(1-\theta)^2}{4}.$$

For System (9)-(10), we also have

$$\frac{S\phi(N)}{N} + r\rho f(N) = \beta S + \rho(N-\theta)(1-\theta) \le \beta + \rho(N-\theta)(1-\theta) \le \beta + \frac{\rho(1-\theta)^2}{4}.$$

Therefore, if $\beta + \frac{\rho(1-\theta)^2}{4} < d$, we have $\limsup_{t\to\infty} I(t) = 0$ for System (7)-(8) and System (9)-(10).

Notes: The traditional basic reproduction number for SI- Allee effects free- and vertical transmission free-models, namely $R_0 = \frac{d}{\beta}$ is naturally no longer relevant. The remainder of this article focuses on the dynamics of System (7)-(8) and System (9)-(10).

3. Mathematical analysis

Notice that System (7)-(8) is not defined at (S, I) = (0, 0) but from Corollary 2.1, we know that

$$\lim_{t\to\infty} (S(t), I(t)) = (0, 0) \text{ whenever } S(0) + I(0) < \theta$$

Thus, we artificially define (0,0) as the extinction equilibrium. Hence, System (7)-(8) and System (9)-(10) have the same boundary dynamics since both of them can be reduced to the system given by

$$\frac{dS}{dt} = S(S - \theta) (1 - S) \text{ if } I = 0$$

and

$$\frac{dI}{dt} = \rho I (I - \theta) (1 - I) - dI \text{ if } S = 0$$

Therefore, System (7)-(8) and System (9)-(10) have the following three boundary equilibria

$$E_{0,0} = (0,0), E_{\theta,0} = (\theta,0), E_{1,0} = (1,0).$$

If, in addition, $(1 - \theta)^2 > 4d/\rho$ holds, then systems (7)-(8) and (9)-(10) support the following additional boundary equilibria on the *I*-axis:

$$E_{0,\theta} = \left(0, \frac{1+\theta}{2} - \frac{\sqrt{(1-\theta)^2 - 4d/\rho}}{2}\right), \ E_{0,1} = \left(0, \frac{1+\theta}{2} + \frac{\sqrt{(1-\theta)^2 - 4d/\rho}}{2}\right)$$

We have arrived at the following proposition regarding the boundary equilibria of System (7)-(8) and System (9)-(10):

Proposition 3.1 (Boundary equilibria of System (7)-(8) and System (9)-(10)). System (7)-(8) and System (9)-(10) always have boundary equilibria $E_{0,0} = (0,0)$, $E_{\theta,0} = (\theta,0)$, $E_{1,0} = (1,0)$. If in addition Condition $(1-\theta)^2 > 4d/\rho$ holds then both systems will support two additional boundary equilibria $E_{0,\theta} = (0,\theta_2)$ and $E_{0,1} = (0, K_2)$ where

$$\theta < \theta_2 = \frac{1+\theta}{2} - \frac{\sqrt{(1-\theta)^2 - 4d/\rho}}{2} < K_2 = \frac{1+\theta}{2} + \frac{\sqrt{(1-\theta)^2 - 4d/\rho}}{2} < 1$$

The nature of the stability of these boundary equilibria is listed in Table 1.

Proof. If S = 0, System(7)-(8) and System (9)-(10) reduced to the following equation:

$$\frac{dI}{dt} = I(\rho(I-\theta)(1-I) - d) = 0 \Rightarrow \rho(I-\theta)(1-I) - d = 0.$$

Therefore, if $(1 - \theta)^2 - 4d/\rho$, we have

$$E_{0,\theta} = (0,\theta_2) = \left(0, \frac{1+\theta}{2} - \frac{\sqrt{(1-\theta)^2 - 4d/\rho}}{2}\right), \ E_{0,1} = (0,K_2) = \left(0, \frac{1+\theta}{2} + \frac{\sqrt{(1-\theta)^2 - 4d/\rho}}{2}\right).$$

Notice that $\sqrt{(1-\theta)^2 - 4d/\rho} < \frac{1-\theta}{2}$, therefore,

$$\theta_2 = \frac{1+\theta}{2} - \frac{\sqrt{(1-\theta)^2 - 4d/\rho}}{2} > \frac{1+\theta}{2} - \frac{1-\theta}{2} > \theta$$

and

$$K_2 = \frac{1+\theta}{2} + \frac{\sqrt{(1-\theta)^2 - 4d/\rho}}{2} < \frac{1+\theta}{2} - \frac{1-\theta}{2} = 1.$$

Boundary Equilibria	Stability Condition		
$E_{0,0}$	Always locally asymptotically stable		
F	For Model (7)-(8)- Saddle if $\frac{d}{\beta} > 1$; Source if $\frac{d}{\beta} < 1$.		
$E_{ heta,0}$	For Model (9)-(10)-Saddle if $\frac{d}{\beta} > \theta$; Source if $\frac{d}{\beta} < \theta$		
$E_{1,0}$	Saddle if $\frac{d}{\beta} < 1$; Locally asymptotically stable if $\frac{d}{\beta} \ge 1$		
	For Model (7)-(8)- Saddle if $\frac{d}{\rho} < \beta$; Source if $\frac{d}{\rho} > \beta$		
$E_{0, heta}$	For Model (9)-(10)-Saddle if $\frac{1+\theta-\sqrt{(1-\theta)^2-4d/\rho}}{2} > \frac{d}{\rho\beta}$ (i.e., $\theta_2 > \frac{d}{\rho\beta}$);		
	Source if $\frac{1+\theta-\sqrt{(1-\theta)^2-4d/\rho}}{2} < \frac{d}{\rho\beta}$ (i.e., $\theta_2 < \frac{d}{\rho\beta}$)		
_	For Model (7)-(8)- Saddle if $\frac{d}{\rho} > \beta$; Locally asymptotically stable if $\frac{d}{\rho} < \beta$		
$E_{0,1}$	For Model (9)-(10)-Saddle if $\frac{1+\theta+\sqrt{(1-\theta)^2-4d/\rho}}{2} < \frac{d}{\rho\beta}$ (i.e., $K_2 < \frac{d}{\rho\beta}$);		
	Locally asymptotically stable if $\frac{1+\theta+\sqrt{(1-\theta)^2-4d/\rho}}{2} > \frac{d}{\rho\beta}$ (i.e., $K_2 > \frac{d}{\rho\beta}$)		

Table 1: The local stability of boundary equilibria for System (7)-(8) and System (9)-(10)

Thus, we have

$$\theta < \theta_2 = \frac{1+\theta}{2} - \frac{\sqrt{(1-\theta)^2 - 4d/\rho}}{2} < \frac{1+\theta}{2} < K_2 = \frac{1+\theta}{2} + \frac{\sqrt{(1-\theta)^2 - 4d/\rho}}{2} < 1.$$

The stability of the boundary equilibria is obtained from the signs of eigenvalues of the corresponding Jacobian matrices. We omit the details but collect the results in Table 1.

Notes: The results in Proposition 3.1 are used to determine the global dynamics in the absence of interior equilibrium (see the proof of Theorem 3.2 for details).

Theorem 3.1 (Persistence of disease or susceptibles). Assume that $\frac{(1-\theta)^2}{4} > \frac{d}{\rho}$ and the initial condition $N(0) \in (\theta_2, K_2)$ then the following statement follows:

- 1. For System (7)-(8) and System (9)-(10), a sufficient condition for the persistence of disease is $\frac{d}{\beta} < 1$.
- 2. For System (7)-(8), a sufficient condition for the persistence of susceptibles is $\frac{d}{\beta\rho} > 1$ while for System (9)-(10) is $\frac{d}{\beta\rho} > K_2$.

The persistence of disease (or susceptibles) means that there exists a positive number ϵ such that

$$\liminf_{t \to \infty} I(t)[or \ S(t)] \ge \epsilon \text{ for any } N(0) \in (\theta_2, K_2).$$

Proof. The condition $\frac{(1-\theta)^2}{4} > \frac{d}{\rho}$ leads to the equalities

$$h(N) = \rho(N - \theta)(1 - N) - d = \frac{\rho}{d}(N - \theta_2)(K_2 - N),$$

which indicates that $h(\alpha) > 0$ for any $\alpha \in (\theta_2, K_2)$. Since $h(N) = \rho f(N) - d$ then Theorem 2.1 and (its) Corollary 2.1 imply that

$$\liminf_{t \to \infty} N(t) = \liminf_{t \to \infty} S(t) + I(t) \ge \alpha \text{ for any } N(0) = \alpha \in (\theta_2, K_2).$$

The use of Theorem 2.1 and (its) Corollary 2.1 again allows to conclude that System (7)-(8) and System (9)-(10) attract to the compact set $0 \le N \le 1$ and are positively invariant within $\alpha \le N \le 1$ for any $\alpha \in (\theta_2, K_2)$.

Letting $B_S = \{(S, I) \in X : \alpha \leq S + I \leq 1\} \cap \{I = 0\}$ and $B_I = \{(S, I) \in X : \alpha \leq S + I \leq 1\} \cap \{S = 0\}$ leads to the facts that (i) B_S and B_I are positively invariant and that (ii) the omega limit set of B_S is $E_{1,0}$ while the omega limit set of B_I is $E_{0,1}$.

The results (Theorem 2.5 and its corollary) in Hutson (1984) guarantee that the persistence of disease is determined by the sign of $\frac{dI}{Idt}\Big|_{B_S} = \frac{dI}{Idt}\Big|_{E_{1,0}}$ while the persistence of susceptibles is determined by the

sign of $\frac{dS}{Sdt}\Big|_{B_I} = \frac{dI}{Idt}\Big|_{E_{0,1}}$.

Letting $\phi(N) = \beta$ or βN means that the dynamics of *I*-class are governed by

$$\frac{dI}{dt} = \phi(N)\frac{I}{N}S + \rho If(N) - dI = I\left(\frac{S\phi(N)}{N} + \rho f(N) - d\right),$$

which gives

$$\frac{dI}{Idt}\Big|_{B_S} = \left(\beta + \rho f(N) - d\right)\Big|_{B_S} = \left(\beta + \rho f(S) - d\right)\Big|_{E_{1,0}} = \beta - d > 0 \text{ if } \phi(N) = \beta \text{ and } \frac{d}{\beta} < 1$$

and

$$\frac{dI}{Idt}\Big|_{B_S} = \left(\beta S + \rho f(N) - d\right)\Big|_{B_S} = \left(\beta S + \rho f(S) - d\right)\Big|_{E_{1,0}} = \beta - d > 0 \text{ if } \phi(N) = \beta N \text{ and } \frac{d}{\beta} < 1.$$

The results (Theorem 2.5 and its corollary) in Hutson (1984) guarantee that the persistence of disease for System (7)-(8) and Systeme (9)-(10) as long as $d < \min\{\frac{\rho(1-\theta)^2}{4}, \beta\}$ and the initial condition $N(0) \in (\theta_2, K_2)$.

The dynamics of the S-class are governed by

$$\frac{dS}{dt} = Sf(N) - \phi(N)\frac{I}{N}S = S\left(f(N) - \frac{I\phi(N)}{N}\right)$$

which gives

$$\frac{dS}{Sdt}\Big|_{B_{I}} = f(N) - \frac{I\phi(N)}{N}\Big|_{B_{I}} = f(I) - \beta\Big|_{E_{0,1}} = f(K_{2}) - \beta = \frac{d}{\rho} - \beta > 0 \text{ if } \phi(N) = \beta \text{ and } \frac{d}{\beta\rho} > 1$$

and

$$\frac{dS}{Sdt}\Big|_{B_{I}} = f(N) - \frac{I\phi(N)}{N}\Big|_{B_{I}} = f(I) - \beta I\Big|_{E_{0,1}} = \frac{d}{\rho} - \beta K_{2} > 0 \text{ if } \phi(N) = \beta N \text{ and } \frac{d}{\beta\rho} > K_{2}.$$

Therefore, applications of the results in Hutson ((Theorem 2.5 and its corollary, 1984) allows us to conclude that:

- 1. A sufficient condition for the persistence of susceptibles in System (7)-(8) is that $\frac{(1-\theta)^2}{4} > \frac{d}{\rho} > \beta$ as long initial condition are such that $N(0) \in (\theta_2, K_2)$.
- 2. A sufficient condition for the persistence of susceptibles in System (9)-(10) is that $\frac{(1-\theta)^2}{4} > \frac{d}{\rho} > \beta K_2$ as long initial condition are such that $N(0) \in (\theta_2, K_2)$.

Note: The System (7)-(8) or System (9)-(10) satisfy the definition of permanence provided that there exists a positive number ϵ such that for any $N(0) \in (\theta_2, K_2)$

$$\liminf_{t \to \infty} \min\{I(t), S(t)\} \ge \epsilon$$

An application of Theorem 3.1 leads to the following permanency results:

1. A sufficient condition for the permanence of System (7)-(8) is that the initial condition $N(0) \in (\theta_2, K_2)$ and

$$\rho < \frac{d}{\beta} < \min\{1, \frac{\rho(1-\theta)^2}{4\beta}\}.$$

2. A sufficient condition for the permanence of System (9)-(10) is that the initial condition $N(0) \in (\theta_2, K_2)$ and

$$\rho K_2 < \frac{d}{\beta} < \min\{1, \frac{\rho(1-\theta)^2}{4\beta}\}.$$

We postulate (throughout the rest of this manuscript) that System (7)-(8) or System (9)-(10) have **disease-free dynamics** if its attractor is $E_{0,0} \cup E_{1,0}$; or System (7)-(8) or (9)-(10) has **susceptibles-free dynamics** if its attractor is $E_{0,0} \cup E_{0,1}$; or System (7)-(8) or (9)-(10) has **disease-driven extinction** if its attractor is $E_{0,0} \cup E_{0,1}$; or System (7)-(8) or (9)-(10) has **disease-driven extinction** if its attractor is $E_{0,0} \cup E_{0,1}$; or System (7)-(8) or (9)-(10) has **disease-driven extinction** if its attractor is $E_{0,0} \cup E_{0,1}$; or System (7)-(8) or (9)-(10) has **disease-driven extinction** if its attractor is $E_{0,0} \cup E_{0,1}$; or System (7)-(8) or (9)-(10) has **disease-driven extinction** if its attractor is $E_{0,0} \cup E_{0,1}$; or System (7)-(8) or (9)-(10) has **disease-driven extinction** if its attractor is $E_{0,0}$.

3.1. Interior equilibrium

Notice that the equilibria of System (7)-(8) satisfy the following equations:

$$S' = S\left[(N-\theta) (1-N) - \frac{\beta I}{N} \right] = 0 \Rightarrow S = 0 \text{ or } I = \frac{N(N-\theta)(1-N)}{\beta},$$

$$I' = I\left[\frac{\beta S}{N} + \rho(N-\theta) (1-N) - d \right] = 0 \Rightarrow I = 0 \text{ or } S = \frac{N[d-\rho(N-\theta)(1-N)]}{\beta}.$$

while the equilibria of System (9)-(10) satisfy the following equations

$$\begin{array}{rcl} S' &=& S\left[(N-\theta) \left(1-N \right) - \beta I \right] = 0 \Rightarrow S = 0 \text{ or } I = \frac{(N-\theta)(1-N)}{\beta}, \\ I' &=& I\left[\beta S + \rho(N-\theta) \left(1-N \right) - d \right] = 0 \Rightarrow I = 0 \text{ or } S - \frac{d}{\beta} = -\frac{\rho(N-\theta)(1-N)}{\beta} \end{array}.$$

If we let (S^*, I^*) be an interior equilibrium of System (7)-(8) or System (9)-(10) then we have that:

1. The following equation

$$N^* = S^* + I^* = \frac{N^* [d + (1 - \rho)(N^* - \theta)(1 - N^*)]}{\beta} \Rightarrow \frac{\beta - d}{1 - \rho} = (N^* - \theta)(1 - N^*)$$
(11)

for System (7)-(8) must be satisfied, and so, we see that System (7)-(8) has no interior equilibrium if $d \ge \beta$ and

$$N^{*} = \frac{1+\theta}{2} \pm \frac{\sqrt{(1-\theta)^{2} - \frac{4(\beta-d)}{1-\rho}}}{2} \text{ if } \frac{(1-\theta)^{2}}{4} > \frac{(\beta-d)}{1-\rho} > 0$$

$$I^{*} = \frac{N^{*}(N^{*}-\theta)(1-N^{*})}{\beta},$$

$$S^{*} = N^{*} \frac{d-\rho(N^{*}-\theta)(1-N^{*})}{\beta} \left[= \rho N^{*} \frac{(N^{*}-\theta_{2})(N^{*}-K_{2})}{\beta} \text{ if } \frac{(1-\theta)^{2}}{4} > \frac{d}{\rho} \right]$$

System (7)-(8) may have the following two interior equilibria $N_i^*, i = 1, 2, \text{ i.e.},$

$$\theta < S_1^* + I_1^* = N_1^* = \frac{1+\theta}{2} - \frac{\sqrt{(1-\theta)^2 - \frac{4(\beta-d)}{1-\rho}}}{2} < \frac{1+\theta}{2}$$

and

$$\frac{1+\theta}{2} < S_2^* + I_2^* = N_2^* = \frac{1+\theta}{2} + \frac{\sqrt{(1-\theta)^2 - \frac{4(\beta-d)}{1-\rho}}}{2} < 1$$

The Jacobian matrix of Model (7)-(8) evaluated at the interior equilibrium (S^*, I^*) can be represented as follows

$$J_{(S^*,I^*)} = \begin{pmatrix} S^* \left[1 + \theta - 2N^* + \frac{\beta I^*}{(N^*)^2} \right] & S^* \left[1 + \theta - 2N^* - \frac{\beta S^*}{(N^*)^2} \right] \\ I^* \left[\beta + \rho(1+\theta) - 2\rho N^* + \frac{\beta I^*}{(N^*)^2} \right] & I^* \left[\rho(1+\theta - 2N^*) - \frac{\beta S^*}{(N^*)^2} \right] \end{pmatrix}$$
(12)

where $N^* = S^* + I^*$. Its two eigenvalues $\lambda_i, i = 1, 2$ satisfy the following equalities:

$$\lambda_1 + \lambda_2 = (S^* + \rho I^*)(1 + \theta - 2N^*) \text{ and } \lambda_1 \lambda_2 = \frac{\beta S^* I^* (1 - \rho)(2N^* - 1 - \theta)}{N^*}.$$
 (13)

Thus, if an interior equilibrium (S^*, I^*) exists, it would be locally asymptotically stable provided that $N^* > \frac{1+\theta}{2}$, or a saddle if $N^* < \frac{1+\theta}{2}$. We conclude that System (7)-(8) has either no interior or two interior equilibria $N_i^*, i = 1, 2$ and if we happen to have two interior equilibria then we must have that N_1^* is always a saddle and N_2^* is always a source.

2. The following equation

$$N^* = S^* + I^* = \frac{d}{\beta} + (1-\rho)\frac{(N^* - \theta)(1 - N^*)}{\beta} \Rightarrow \frac{\beta}{1-\rho}(N^* - \frac{d}{\beta}) = (N^* - \theta)(1 - N^*)$$
(14)

for System (9)-(10). According to Corollary 2.1, we have $N^* < 1$, thus (14) implies that System (9)-(10) has no interior equilibrium if $\frac{d}{\beta} \geq 1$ and

$$N^{*} = \frac{1+\theta}{2} - \frac{\beta}{2(1-\rho)} \pm \frac{\sqrt{\left(1+\theta-\frac{\beta}{1-\rho}\right)^{2}+4\left(\frac{d}{1-\rho}-\theta\right)}}{2} \text{ if } \frac{\left(1+\theta-\frac{\beta}{1-\rho}\right)^{2}}{4} > \theta - \frac{d}{1-\rho}$$
$$= \frac{1+\theta}{2} - \frac{\beta}{2(1-\rho)} \pm \frac{\sqrt{\left(1-\theta-\frac{\beta}{1-\rho}\right)^{2}-4\frac{d-\theta\beta}{1-\rho}}}{2} \text{ if } \frac{\left(1+\theta-\frac{\beta}{1-\rho}\right)^{2}}{4} > \theta - \frac{d}{1-\rho}$$
$$I^{*} = \frac{(N^{*}-\theta)(1-N^{*})}{\beta},$$
$$S^{*} = \frac{d-\rho(N^{*}-\theta)(1-N^{*})}{\beta} \left[= \rho \frac{(N^{*}-\theta_{2})(N^{*}-K_{2})}{\beta} \text{ if } \frac{(1-\theta)^{2}}{4} > \frac{d}{\rho} \right].$$

System (9)-(10) may have the following two interior equilibria:

$$S_1^* + I_1^* = N_1^* = \frac{1+\theta}{2} - \frac{\beta}{2(1-\rho)} - \frac{\sqrt{\left(1+\theta - \frac{\beta}{1-\rho}\right)^2 + 4\left(\frac{d}{1-\rho} - \theta\right)}}{2} < \frac{1+\theta}{2} - \frac{\beta}{2(1-\rho)}$$

and

$$\frac{1+\theta}{2} - \frac{\beta}{2(1-\rho)} < S_2^* + I_2^* = N_2^* = \frac{1+\theta}{2} - \frac{\beta}{2(1-\rho)} + \frac{\sqrt{\left(1+\theta - \frac{\beta}{1-\rho}\right)^2 + 4\left(\frac{d}{1-\rho} - \theta\right)}}{2} < 1+\theta - \frac{\beta}{(1-\rho)} < 0$$

if

$$\frac{1+\theta}{2} > \frac{\beta}{2(1-\rho)}, \left(1+\theta-\frac{\beta}{1-\rho}\right)^2 > 4(\theta-\frac{d}{1-\rho}) \text{ and } \theta > \frac{d}{1-\rho}$$

On the other hand if $\theta < \frac{d}{1-\rho}$ then System (9)-(10) may have at most one interior equilibrium, namely,

$$S_2^* + I_2^* = N_2^* = \frac{1+\theta}{2} - \frac{\beta}{2(1-\rho)} + \frac{\sqrt{\left(1+\theta - \frac{\beta}{1-\rho}\right)^2 + 4\left(\frac{d}{1-\rho} - \theta\right)}}{2}$$

The Jacobian matrix of System (9)-(10) evaluated at the interior equilibrium (S^*, I^*) is represented as

$$J_{(S^*,I^*)} = \begin{pmatrix} S^*(1+\theta-2N^*) & S^*(1+\theta-\beta-2N^*) \\ I^*(\beta+\rho(1+\theta)-2\rho N^*) & \rho I^*(1+\theta-2N^*) \end{pmatrix}$$
(15)

where $N^* = S^* + I^*$. Its two eigenvalues $\lambda_i, i = 1, 2$ satisfy the following equalities:

$$\lambda_1 + \lambda_2 = (S^* + \rho I^*)(1 + \theta - 2N^*) \text{ and } \lambda_1 \lambda_2 = \beta S^* I^* \left[(1 - \rho)(2N^* - 1 - \theta) + \beta \right].$$
(16)

Thus, when the interior (S^*, I^*) exists, it is locally asymptotically stable as long as $N^* > \frac{1+\theta}{2}$ while a saddle whenever $N^* < \frac{1+\theta}{2} - \frac{\beta}{2(1-\rho)}$. It is a source if

$$\frac{1+\theta}{2} - \frac{\beta}{2(1-\rho)} < N^* < \frac{1+\theta}{2}.$$

If System (9)-(10) has two interior equilibria N_i^* , i = 1, 2 then using their expressions and the criteria for interior stability allow us to conclude that N_1^* is always a saddle while N_2^* can be a sink or source. If, on the other hand, System (9)-(10) has only one interior equilibrium, namely N_2^* , then we see that it can be a sink or source depending on parameter values.

The above discussion can be summarized in the following theorem:

Theorem 3.2 (Interior equilibria of Models). Let $E_{i1} = (S_1^*, I_1^*)$ and $E_{i2} = (S_2^*, I_2^*)$ then existence and stability conditions for the interior equilibria of System (7)-(8) are listed in Table 2.

Interior Equilibrium	Condition for existence	Condition for local asymptotically stable
E_{i1}	$0 < \frac{\beta - d}{1 - \rho} < \frac{(1 - \theta)^2}{4} < \frac{d}{\rho}$ or	
	$0 < \frac{\beta - d}{1 - \rho} < \frac{d}{\rho} < \frac{(1 - \theta)^2}{4}$	Always a saddle
E_{i2}	$0 < \frac{\beta - d}{1 - \rho} < \frac{(1 - \theta)^2}{4} < \frac{d}{\rho}$ or	
	$0 < \frac{\beta - d}{1 - \rho} < \frac{d}{\rho} < \frac{(1 - \theta)^2}{4}$	Always locally asymptotically stable.

Table 2: The local stability of interior equilibrium for System (7)-(8)

Existence and stability conditions for the interior equilibria of System (9)-(10) are listed in Table 3. Sufficient conditions leading to no interior equilibrium and the related global dynamics of System (7)-(8) and System (9)-(10) are listed in Table 4.

Proof. The above discussion shows that a necessary condition for System (7)-(8) and System (9)-(10) to a have interior equilibrium is that $\frac{d}{\beta} < 1$ while the existence of interior equilibrium for System (7)-(8) and System (9)-(10) can be classified with the conditions $\frac{d}{\rho} > \frac{(1-\theta)^2}{4}$ and $\frac{d}{\rho} < \frac{(1-\theta)^2}{4}$.

If $\frac{d}{\rho} > \frac{(1-\theta)^2}{4}$ then System (7)-(8) and System (9)-(10) have no boundary equilibria $E_{0,\theta}$ and $E_{0,1}$ on *I*-axis and therefore

$$h(N) = \rho(N - \theta)(1 - N) - d < 0$$
 for all $N > 0$.

Interior Equilibrium	Condition for existence	Condition for stability
E_{i1}	$\min\{\frac{d}{1-\rho}, \frac{d}{\beta}, \frac{\beta+\rho-1}{1-\rho}\} < \theta \text{ and (a)}$ $\frac{d}{\rho} > \frac{(1-\theta)^2}{4} \text{ or (b) } \frac{d}{\rho} < \frac{(1-\theta)^2}{4}, N_1 < \theta_2$	Always a saddle
E_{i2}	Case (1):min $\{\frac{d}{1-\rho}, \frac{d}{\beta}, \frac{\beta+\rho-1}{1-\rho}\} < \theta$ and (a) $\frac{d}{\rho} > \frac{(1-\theta)^2}{4}$ or (b) $\frac{d}{\rho} < \frac{(1-\theta)^2}{4}, N_2 < \theta_2$ or (c) $\frac{d}{\rho} < \frac{(1-\theta)^2}{4}, N_2 > K_2;$ Case (2):1 > $\frac{d}{\beta} > \theta$ and (a) $\frac{d}{\rho} > \frac{(1-\theta)^2}{4}$ or (b) $\frac{d}{\rho} < \frac{(1-\theta)^2}{4}, N_2 < \theta_2$ or (c) $\frac{d}{\rho} < \frac{(1-\theta)^2}{4}, N_2 > K_2$	Locally asymptotically stable if $N_2^* > \frac{1+\theta}{2}$ while it's a source if $\frac{1+\theta}{2} - \frac{\beta}{2(1-\rho)} < N_2^* < \frac{1+\theta}{2}$

Table 3: The local stability of interior equilibrium for System (9)-(10)

Necessary conditions for the existence of interior equilibrium for System (7)-(8) and System (9)-(10) are tied in to the existence of solutions of the following equations:

$$\frac{\beta - d}{1 - \rho} = (N - \theta) (1 - N), 0 < \frac{\beta - d}{1 - \rho} < \frac{(1 - \theta)^2}{4} < \frac{d}{\rho}$$
(17)

and

$$\frac{\beta}{1-\rho}(N-\frac{d}{\beta}) = (N-\theta)(1-N), \frac{d}{\beta} > \theta, \text{ or}
\frac{\beta}{1-\rho}(N-\frac{d}{\beta}) = (N-\theta)(1-N), \frac{1+\theta}{2} > \frac{\beta}{2(1-\rho)}, \left(1-\theta-\frac{\beta}{1-\rho}\right)^2 > \frac{4(\beta\theta-d)}{1-\rho} > 0.$$
(18)

Therefore, if $\frac{d}{\rho} > \frac{(1-\theta)^2}{4}$ then System (7)-(8) and System (9)-(10) have no interior if $\frac{d}{\beta} \ge 1$ or if the following conditions

$$\frac{\beta - d}{1 - \rho} > \frac{(1 - \theta)^2}{4} \text{ for System (7)} - (8)$$

and

$$\left(1-\theta-\frac{\beta}{1-\rho}\right)^2 < \frac{4(\beta\theta-d)}{1-\rho} < 0 \text{ for System (9)} - (10)$$

for each model hold.

If $\frac{d}{\rho} < \frac{(1-\theta)^2}{4}$, then System (7)-(8) and System (9)-(10) have boundary equilibria $E_{0,\theta}$ and $E_{0,1}$ on the *I*-axis. Additional conditions are needed to guarantee the existence of interior equilibrium for System (7)-(8) and System (9)-(10). System (7)-(8) and System (9)-(10) are discussed separately.

1. For System (7)-(8), if $\frac{\beta-d}{1-\rho} > \frac{(1-\theta)^2}{4}$ nEquation (17) should have solutions in the interval $(0, \theta_2)$ or $(K_2, 1)$ since

$$S^* = N^* \frac{d - \rho(N^* - \theta) \left(1 - N^*\right)}{\beta} = \rho N^* \frac{\left(N^* - \theta_2\right) \left(N^* - K_2\right)}{\beta} > 0.$$

The schematic nullclines for System (7)-(8) when $\frac{\beta-d}{1-\rho} < \frac{(1-\theta)^2}{4}$ are illustrated in Figure 1. Two interior equilibria occur whenever $\frac{\beta-d}{1-\rho} < \frac{d}{\rho} < \frac{(1-\theta)^2}{4}$ with one interior a saddle (i.e., the horizontal

Cases	System (7)-(8)	System (9)-(10)	
No interior equilibrium	Case (1): $\frac{d}{\beta} \ge 1$; Case (2): $\frac{\beta-d}{1-\rho} > \frac{(1-\theta)^2}{4}$; Case (3): $\frac{d}{\rho} < \frac{\beta-d}{1-\rho} < \frac{(1-\theta)^2}{4}$.	$\begin{array}{c} \text{Case (1): } \frac{d}{\beta} \geq 1; \text{ Case (2):} \\ \frac{\beta\theta-d}{1-\rho} > \frac{(1-\theta-\frac{1}{1-\rho})^2}{4}; \text{ Case (3):} \\ \min\{\frac{d}{1-\rho}, \frac{d}{\beta}, \frac{\beta+\rho-1}{1-\rho}\} < \theta, \frac{d}{\rho} < \\ \frac{(1-\theta)^2}{4}, \theta_2 < N_i^* < K_2, i = 1, 2; \\ \text{ Case (4): } \frac{d}{\beta} > \theta, \frac{d}{\rho} < \\ \frac{(1-\theta)^2}{4}, \theta_2 < N_2 < K_2 \end{array}$	
Disease-free dynamics	$\frac{d}{eta} \ge 1$	$\frac{d}{\beta} \ge 1$	
Susceptible-free dynamics	$\min\{\beta, \frac{(1-\theta)^2}{4}\} > \frac{d}{\rho}.$	$\frac{d}{\rho} < \min\{\beta K_2, \frac{(1-\theta)^2}{4}\} \text{ and}$ Conditions of Case (2) or Case (3) or Case (4)	
Disease-driven extinction	$\frac{(1-\theta)^2}{4} < \min\{\frac{\beta-d}{1-\rho}, \frac{d}{\rho}\}.$	$\frac{\frac{d}{\rho} < \frac{(1-\theta)^2}{4} \text{ and}}{\frac{\beta\theta-d}{1-\rho} > \frac{(1-\theta-\frac{\beta}{1-\rho})^2}{4}.$	
Permanence	$N(0) \in (\theta_2, K_2) \text{ and} \beta < \frac{d}{\rho} < \min\{\frac{\beta}{\rho}, \frac{(1-\theta)^2}{4}\}.$	$N(0) \in (\theta_2, K_2) \text{ and} \beta K_2 < \frac{d}{\rho} < \min\{\frac{\beta}{\rho}, \frac{(1-\theta)^2}{4}\}.$	

Table 4: No interior equilibrium and the related global dynamics for System (7)-(8) and System (9)-(10)

line intercepts in the green region of $(N - \theta)(1 - N)$ and the other a sink (i.e., the horizontal line intercepts in the blue region of $(N - \theta)(1 - N)$). There is no interior equilibrium when the horizontal line intercepts (crosses) the black region of $(N - \theta)(1 - N)$, i.e., $\frac{d}{\rho} < \frac{\beta - d}{1 - \rho} < \frac{(1 - \theta)^2}{4}$.

2. For System (9)-(10), whenever $\frac{\beta\theta-d}{1-\rho} < \frac{(1-\theta-\frac{\beta}{1-\rho})^2}{4}$ Equation (18) should have solutions in the interval $(0,\theta_2)$ or $(K_2,1)$ since

$$S^* = \frac{d - \rho(N^* - \theta) (1 - N^*)}{\beta} = \rho \frac{(N^* - \theta_2) (N^* - K_2)}{\beta} > 0.$$

The schematic nullclines for System (9)-(10) when $\frac{\beta\theta-d}{1-\rho} < \frac{(1-\theta-\frac{\beta}{1-\rho})^2}{4}$ are found in Figure 2. There are two cases depending on the sign of $\frac{d}{\beta} - \theta$: (1) If $\frac{d}{\beta} > \theta$ (see Figure 2(a)), two interior equilibria occur whenever

$$\frac{1+\theta}{2} > \frac{\beta}{2(1-\rho)}, \left(1-\theta-\frac{\beta}{1-\rho}\right)^2 > \frac{4(\beta\theta-d)}{1-\rho} > 0$$

and

$$N_1^* < \theta_2, N_2^* < \theta_2 \text{ or } N_2^* < K_2$$

where N_1^* is always a saddle (i.e., the line $\frac{\beta}{1-\rho}(N-\frac{d}{\beta})$ intercepts the green region of $(N-\theta)(1-N)$) and N_2^* can be sink (i.e., the line intercepts the red region of $(N-\theta)(1-N)$) or source (i.e., the line intercepts the blue region of $(N-\theta)(1-N)$). If Condition $N_2^* < \theta_2$ or $N_2^* < K_2$ does not hold, i.e., $\theta_2 < N_2^* < K_2$ then System (9)-(10) has only one interior equilibrium N_1^* , a saddle. (2) If $\frac{d}{\beta} < \theta$ (see Figure 2(b)) then only one interior equilibria occurs whenever

$$\theta < \frac{d}{\beta} < 1, \, \theta < \frac{d}{1-\rho} \text{ and } [N_2^* < \theta_2 \text{ or } N_2^* < K_2]$$



Figure 1: Schematic nullclines for System (7)-(8) when $\frac{\beta-d}{1-\rho} < \frac{(1-\theta)^2}{4}$. Potential (two) interior equilibria are the intercepts of the horizontal line $\frac{\beta}{1-\rho}$ and the curve $(N-\theta)(1-N)$ with their stability determined by the location of the intercept. The green region of $(N-\theta)(1-N)$ is a saddle and the blue region is a sink. Two interior equilibria (see the purple horizontal line) occur whenever $\frac{\beta-d}{1-\rho} < \frac{d}{\rho} < \frac{(1-\theta)^2}{4}$ where one interior is saddle and the other one a sink.

d

and this interior equilibrium N_2^* can be a sink or a source, depending on parameters' values. There is no interior if the line intercepts the black region of $(N - \theta)(1 - N)$ when

$$\mathbf{or}$$

$$\frac{\alpha}{\beta} < \theta, \, \theta_2 < N_2^* < K_2$$
$$1 - \theta - \frac{\beta}{1 - \epsilon} \Big)^2 < \frac{4(\beta \theta - d)}{1 - \epsilon}$$

In short, sufficient conditions for the existence of interior equilibria and their stability have been identified and listed in Table 2 for System (7)-(8) and in Table 3 for System (9)-(10).

The above analysis has identified conditions (sufficient) that guarantee the absence of interior equilibria for System (7)-(8) and System (9)-(10); listed inTable 4. In the absence of interior equilibrium, we can conclude thanks to the Poincaré-Bendixson Theorem (Guckenheimer & Holmes 1983), that a trajectory starting with arbitrary initial conditions in X converge to its locally asymptotically stable boundary equilibria since System (7)-(8) and System (9)-(10) support each a global compact attractor $\{(S,I) \in X : 0 \leq S + I \leq 1\}$ in X. The fact that $E_{0,0}$ is always an attractor results according to Proposition 3.1 in the following three cases:

1. Disease-free dynamics that corresponds to the case where $E_{0,0}$ and $E_{1,0}$ are the only locally asymptotically stable boundary equilibria while other existing boundary equilibria are unstable.



(a) Schematic nullclines for System (9)-(10) when $\frac{d}{\beta} < \theta$. (b) Schematic nullclines for System (9)-(10) when $\frac{d}{\beta} > \theta$.

Figure 2: Schematic nullclines for System (9)-(10) when $\frac{\beta\theta-d}{1-\rho} < \frac{(1-\theta-\frac{\beta}{1-\rho})^2}{4}$. Potential interior equilibria are intercepts of the line $\frac{\beta}{1-\rho}(N-\frac{d}{\beta})$ and the curve $(N-\theta)(1-N)$ with their stability determined by the location of the intercept. The green region of $(N-\theta)(1-N)$ is a saddle; the red region is a source and the blue region is a sink. The left graph corresponds to the case when $\frac{d}{\beta} < \theta$: one (see the black line $\frac{\beta}{1-\rho}(N-\frac{d}{\beta})$) or two interior equilibria (see the purple or dark green line $\frac{\beta}{1-\rho}(N-\frac{d}{\beta})$) are possible. The right graph corresponds to the case when $1 > \frac{d}{\beta} > \theta$, potentially one interior equilibria (see the dark green or purple line $\frac{\beta}{1-\rho}(N-\frac{d}{\beta})$).

This implies that $\beta \ge d$ is a sufficient condition in support of disease-free dynamics within System (7)-(8) and System (9)-(10).

- 2. Susceptible-free dynamics that corresponds to the case where $E_{0,0}$ and $E_{0,1}$ are the only locally asymptotically stable boundary equilibria while other existing boundary equilibria (including those on the *I*-axis) are unstable. This implies that $\frac{d}{\beta} < \rho K_2$ and $\frac{d}{\rho} < (1-\theta)^2/4$ for System (7)-(8) and System (9)-(10), in addition to the conditions of non-existence of interior equilibrium.
- 3. Disease-driven extinction that corresponds to the case where $E_{0,0}$ is the only locally asymptotically stable boundary equilibria provided that there is no boundary equilibria on the *I*-axis a result based on Theorem 3.1. This implies that $\rho K_2 < \frac{d}{\beta} < 1$ and $\frac{d}{\rho} > \frac{(1-\theta)^2}{4}$ for System (7)-(8) and System (9)-(10) in conjunction to the conditions that there are no interior equilibrium.

Detailed conditions on the three cases discussed above are listed in Table 4.

Notes: Theorem 3.2 implies the following:

- 1. Small values of ρ make System (7)-(8) and System (9)-(10) prone to disease-driven extinction since one necessary condition for **disease-driven extinction** requires that $\frac{d}{\rho} > \frac{(1-\theta)^2}{4}$ according to Theorem 3.1. This also suggests that vertical transmission may save a species from extinction provided that the reproductive ability of infectives is large enough (some additional conditions must be met).
- System (7)-(8) has simpler dynamics than System (9)-(10). In fact, System (7)-(8) has no interior equilibria or two interior equilibria (a saddle and a sink) while System (9)-(10) may have one or two interior equilibria.

4. Classifications on dynamics and related bifurcation diagrams

This section focuses on the classification of the dynamics and related bifurcations of System (7)-(8) and System (9)-(10). We define $R_0^h = \frac{\beta}{d}$ to be the horizontal transmission reproduction number and $R_0^v = \frac{\rho}{d}$ as the vertical transmission reproduction number.

4.1. The SI model with frequency-dependent horizontal transmission

For System (7)-(8), notice that

$$\frac{\beta-d}{1-\rho} < \frac{d}{\rho} \Rightarrow R_0^v = \frac{\rho}{d} > \frac{1}{\beta}$$

Thus, if System (7)-(8) has boundary equilibria on the *I*-axis, i.e., $\frac{d}{\rho} < \frac{(1-\theta)^2}{4}$, then the no interior equilibria condition $\frac{\beta-d}{1-\rho} > \frac{(1-\theta)^2}{4}$ says that the boundary equilibrium $E_{0,\theta}$ is a source while $E_{0,1}$ is locally asymptotically stable according to Proposition 3.1 and the fact that

$$\frac{\beta - d}{1 - \rho} > \frac{d}{\rho} \Rightarrow R_0^v = \frac{\rho}{d} < \frac{1}{\beta}.$$

Therefore, System (7)-(8) have two interior equilibria provided that

$$0 < \frac{\beta - d}{1 - \rho} < \frac{(1 - \theta)^2}{4} < \frac{d}{\rho} \Rightarrow R_0^v = \frac{\rho}{d} < \min\{\frac{1}{\beta}, \frac{4}{(1 - \theta)^2}\}$$
$$0 < \frac{\beta - d}{4} < \frac{d}{2} < \frac{(1 - \theta)^2}{2} \Rightarrow \frac{4}{2} < R_0^v = \frac{\rho}{2} < \frac{1}{2},$$

or

$$0 < \frac{\beta - d}{1 - \rho} < \frac{d}{\rho} < \frac{(1 - \theta)^2}{4} \Rightarrow \frac{4}{(1 - \theta)^2} < R_0^v = \frac{\rho}{d} < \frac{1}{\beta}.$$

Corollary 2.1, Proposition 3.1, and Theorem 3.1-3.2 lead to the study of three cases for System (7)-(8):

1. The disease-driven extinction occurs in the situation depicted in Figure 3. First, no interior equilibrium, which requires $\frac{\beta-d}{1-\rho} > \frac{(1-\theta)^2}{4}$. Within Figure 3, we see that the existence and the stability of boundary equilibria requires $R_0^v = \frac{\rho}{d} < \frac{4}{(1-\theta)^2}$ and $\frac{d}{\beta} < 1$ (i.e., $R_0^h > 1$). Thus, a sufficient condition that makes Figure 3 possible is

$$\max\{\frac{\rho}{d}, \frac{1-\rho}{\beta-d}\} < \frac{4}{(1-\theta)^2} \text{ and } R_0^h = \frac{\beta}{d} > 1.$$

System (7)-(8) may also support disease-driven extinction whenever it supports an interior equilibrium. In such a case, disease-driven extinction occurs as a result of catastrophic events, that is, when a stable limit cycles merges with the adjacent saddle, leading to the annihilation of the susceptible and infected sub-populations.

2. An endemic situation occurs whenever System (7)-(8) supports the interior equilibria shown in Figure 4. A necessary condition is that $\frac{\beta-d}{1-\rho} < \frac{(1-\theta)^2}{4}$ and thus we can conclude that the sufficient condition leading to Figure 4(a) is

$$R_0^h = \frac{\rho}{d} < \frac{4}{(1-\theta)^2} < \frac{1-\rho}{\beta-d} \text{ and } \rho < \frac{d}{\beta} < 1 \text{ (or } 1 < R_0^h < \frac{1}{\rho})$$

while the sufficient condition leading to Figure 4(b) is

$$\frac{4}{(1-\theta)^2} < \frac{1-\rho}{\beta-d} < R_0^h = \frac{\rho}{d} \text{ and } \rho < \frac{d}{\beta} < 1 \text{ (or } 1 < R_0^h < \frac{1}{\rho}).$$



Figure 3: Schematic phase plane for System (7)-(8) when it experiences the possibility of disease-driven extinction.

3. Disease-free or susceptible-free dynamics occur when System (7)-(8) has no interior equilibrium with either $E_{1,0}$ or $E_{0,1}$ locally asymptotically stable, as shown in Figure 5. Figure 5(a) highlights a disease-free situation for which the condition

$$R_0^h \ge 1 \text{ and } \frac{4}{(1-\theta)^2} < R_0^v = \frac{\rho}{d}$$

is sufficient.

Figure 5(b) highlights a susceptible-free state for which sufficient the condition is

$$\frac{1-\rho}{\beta-d} < \frac{4}{(1-\theta)^2} < R_0^v = \frac{\rho}{d} \text{ and } \frac{d}{\beta} < \rho \text{ (or } R_0^h > \frac{1}{\rho} \text{)}.$$

The vertical transmission reproduction number, $R_0^v = \frac{\rho}{d}$, and the horizontal transmission reproduction number, $R_0^h = \frac{\beta}{d}$ help, using the above discussions and the analytical results in previous sections, understand the effects of parameters ρ, d, β, θ on the dynamics of System (7)-(8). The results can briefly summarized as follows:

- 1. A horizontal transmission reproduction number R_0^h less than 1 supports disease-free dynamics for System (7)-(8) (see Theorem 3.2 when combined with the relevant results in Table 4).
- 2. Both initial condition N(0) = S(0) + I(0) and the value of the vertical transmission reproduction number, R_0^v , are important in determining global dynamics (see Corollary 2.1 and Theorem 3.2). We can conclude that large values of R_0^v tend to lead to susceptible-free dynamics; while intermediate values of R_0^v tends lead to the coexistence of susceptibles and infectives; and small values of R_0^v tends to lead to disease-driven extinction.
- 3. The SI model with frequency-dependent transmissions or System (7)-(8) supports relatively simple equilibrium dynamics. It can support no interior or two interior equilibria, with one of the interior equilibrium always stable (see Theorem 3.2, Table 3-4 and Figure 1).



(a) Schematic phase plane

(b) Schematic phase plane

Figure 4: Schematic phase plane for System (7)-(8) when it has the endemic occurs.



Figure 5: Schematic phase plane for System (7)-(8) when it has no coexistence of susceptibles and infectives.

4.2. SI model with density-dependent horizontal transmissions

In this subsection, the dynamics and potential bifurcations of the SI model, with density-dependent horizontal transmission, given by System (9)-(10), are classified. The classification of stability of boundary equilibria for System (9)-(10) on the *I*-axis $E_{0,\theta}$ and $E_{0,1}$ when $\frac{(1-\theta)^2}{4} > \frac{d}{\rho}$ can be determined from the signs of $\theta_2 - \frac{d}{\rho\beta}$ and $K_2 - \frac{d}{\rho\beta}$. Since

$$\theta < \theta_2 = \frac{1+\theta}{2} - \frac{\sqrt{(1-\theta)^2 - 4d/\rho}}{2} < K_2 = \frac{1+\theta}{2} + \frac{\sqrt{(1-\theta)^2 - 4d/\rho}}{2} < 1$$

Hence, the signs can be determined by solving $R_0^v = \frac{\rho}{d}$ from the equations $\theta_2 = \frac{d}{\rho\beta}$ and $K_2 = \frac{d}{\rho\beta}$, i.e.,

$$\frac{1+\theta}{2} - \frac{\sqrt{(1-\theta)^2 - 4d/\rho}}{2} = \frac{d}{\rho\beta} \Rightarrow R_0^v = \frac{\rho}{d} = \frac{1+\theta - 4\beta \pm \sqrt{(4\beta - 1)^2 + (1-\theta)^2 - 1 - 8\beta\theta}}{2\beta\theta}$$

and

$$\frac{1+\theta}{2} + \frac{\sqrt{(1-\theta)^2 - 4d/\rho}}{2} = \frac{d}{\rho\beta} \Rightarrow \frac{\rho}{d} = \frac{1+\theta - 4\beta \pm \sqrt{(4\beta - 1)^2 + (1-\theta)^2 - 1 - 8\beta\theta}}{2\beta\theta}$$

Letting $c_1 = \frac{1+\theta-4\beta-\sqrt{(4\beta-1)^2+(1-\theta)^2-1-8\beta\theta}}{2\beta\theta}$ and $c_2 = \frac{1+\theta-4\beta+\sqrt{(4\beta-1)^2+(1-\theta)^2-1-8\beta\theta}}{2\beta\theta}$ leads, making use of Proposition 3.1, to the following (a two dimensional bifurcation diagram example is shown in Figure 6 when $\beta = 0.1$ and $\theta = 0.15$) results:

1. Black area in Figure 6: $E_{0,\theta}$ is a saddle and $E_{0,1}$ is locally asymptotically stable if

$$R_0^v = \frac{\rho}{d} > \max\left\{\frac{4}{(1-\theta)^2}, c_2\right\}.$$

2. Cyan area in Figure 6: $E_{0,\theta}$ is a source and $E_{0,1}$ is locally asymptotically stable if

$$\max\left\{\frac{4}{(1-\theta)^2}, c_1\right\} < R_0^v = \frac{\rho}{d} < c_2.$$

3. Green area in Figure 6: $E_{0,\theta}$ is a source and $E_{0,1}$ is a saddle if

$$\frac{4}{(1-\theta)^2} < R_0^v = \frac{\rho}{d} < c_1$$

4. White area in Figure 6: there is no boundary equilibrium on I-axis, i.e.,

$$R_0^v = \frac{\rho}{d} < \frac{4}{(1-\theta)^2}.$$

We therefore identify only four cases for System (9)-(10):

- 1. Case one: There is no boundary equilibrium $E_{0,\theta}$ and $E_{0,1}$ when the reproduction number of vertical transmission $R_0^v = \frac{\rho}{d}$ is small enough, i.e., $R_0^v < \frac{4}{(1-\theta)^2}$. For a certain range of parameter values, System (9)-(10) can have a unique interior attractor, which can be an interior equilibrium (see Figure 7; where within the the sub-figure (a) corresponds to the white area with blue dots, i.e., $R_0^h = \frac{\beta}{d} > \frac{1}{\theta}$, and (b) corresponds to the white area with red dots, i.e., $1 < R_0^h < \frac{1}{\theta}$, below the green area of Figure 6) or a stable limit cycle through Hopf-Bifurcation. This is the case when System (9)-(10) can support disease-driven extinction as it was the case for System (7)-(8).
- 2. Case two: There are boundary equilibria $E_{0,\theta}$ and $E_{0,1}$ when the reproduction number of vertical transmission $R_0^v = \frac{\rho}{d}$ is large enough, i.e., $\frac{4}{(1-\theta)^2} < R_0^v$ and the reproduction number of horizontal transmission has a large value, i.e., $R_0^h > \frac{1}{\theta}$. An example of this case is shown in the black area [whose dynamics is corresponding to the sub-figure (a) of Figure 8] and the green area [whose dynamics is corresponding to the sub-figure (b) of Figure 8] on the right of the purple vertical line d = 0.015 of Figure 6.
- 3. Case three: There are boundary equilibria $E_{0,\theta}$ and $E_{0,1}$ and the reproduction number of horizontal transmission has intermediate values, i.e., $1 < R_0^h < \frac{1}{\theta}$. An example of this case is shown in the green area [whose dynamics corresponds to the sub-figure (a) of Figure 9] and the black area [whose dynamics is corresponding to the sub-figure (b) of Figure 9] on the left of the purple vertical line d = 0.015 of Figure 6.
- 4. Case four: There are boundary equilibria $E_{0,\theta}$ and $E_{0,1}$ and the reproduction number of vertical transmission has intermediate values, i.e., $\max\left\{\frac{4}{(1-\theta)^2}, c_1\right\} < R_0^h < c_2$. An example of this case is shown in the cyan area of Figure 6 whose dynamics on the right side of the purple vertical line d = 0.015, i.e., the reproduction number of horizontal transmission has large values, i.e., $R_0^h > \frac{1}{\theta}$. Dynamics represented by the sub-figure (a) of Figure 10 and the dynamics on the left side of the purple vertical line d = 0.015, i.e., $1 < R_0^h < \frac{1}{\theta}$, represented by the sub-figure (b) of Figure 10.

The above discussion and the associated analytical results, including Proposition 3.1, Theorem 2.1, 3.1, 3.2 lead us to conclude that the effects of parameters ρ, d, β, θ on the dynamics of System (9)-(10) can be summarized as follows:

- 1. The values of the reproduction number of horizontal transmission $R_0^h = \frac{\beta}{d}$ and the reproduction number of vertical transmission $R_0^v = \frac{\rho}{d}$ determine the dynamics of System (9)-(10):
 - If $1 < R_0^h < \frac{1}{\beta}$ and $R_0^v < \frac{4}{(1-\theta)^2}$, then System (9)-(10) has no boundary equilibrium on the *I*-axis and it may have the disease-driven extinction in certain range of parameter values.
 - If System (9)-(10) has the intermediate values of R_0^v , i.e., $\max\left\{\frac{4}{(1-\theta)^2}, c_1\right\} < R_0^v < c_2$, then the system tends to have susceptible-free dynamics.
 - The values of R_0^h determines whether System (9)-(10) can have unique interior equilibrium $(R_0^h > \frac{1}{\beta})$ or two interior equilibria $(1 < R_0^h < \frac{1}{\beta})$.
- 2. The large values of ρ, θ, β and the small values of d can destablize System (9)-(10) (see Figure 11).
- 3. System (9)-(10) can have a stable limit cycle; an example is included in Figure 12.

5. Diffusive instability

The dynamics and evolution of host-pathogen or host-parasite systems is of theoretically challenging for factors that include the impact of recurrent disease invasion, the ability of a parasite or pathogen to modify a host's mobility-tied fitness, or reducing life span, or limiting/eliminating reproductive ability. Dispersal is capable of shaping the boundaries of habitats through increases or reductions on the size of the sphere-of-influence of infectious hosts, a cumulative process possibly altering infection rates (reductions or increases in effective contact rates), or its ability to generate clusters, or disease-driven selection of particular behavioral types (Altizer 2010; Diaz 2010; Levin 1974; Murray 2003).

Diffusive instability arises when diffusion or migration destabilizes stable situations (Segel and Jackson 1972; Levin 1974; Segel and Levin 1976). It may emerge as a result of the dynamics of metapopulation systems when coupled by dispersal or reaction-diffusion (Diaz 2010; Levin 1974; Murray 2003). The possible emergence of diffusive instability from two-patch systems, coupled by dispersal, when the local dynamics are governed by variants of the general *SI*-FD or *SI*-DD systems, is briefly address in this section.

Segel and Jackson (1972) using a simple predator-prey model studied the possibility of diffusive instability in predator-prey systems. Hence, first, some classical results addressing the emergence of diffusive instability in predator-prey systems, are revisited. Segel and Jackson (1972) showed that the addition of random dispersal was enough to generate instability from an otherwise initially stable uniform steady-state distribution. *Diffusive instability*, as shown by Levin (1974), also arises from the effects of dispersal on predator-prey interactions under the pressure of Allee effects. Segel and Levin (1976) used approximate methods and a multiple-time scale theoretical approach in their development of a small amplitude nonlinear theory of prey-predator interactions under random dispersal; a process modeled via diffusion-like terms in discrete and continuous settings. Segel and Levin (1976) showed that dispersal can destabilize spatially uniform states; diffusive instability moving the system to new nonuniform steady states. Levin and Segel (1976 & 1985) noted that the emergence of diffusive instabilities may explain some of the spatial irregularities observed in nature. The possibility of *diffusive instability* in general *SI* models is briefly discussed since identifying conditions that lead to *diffusive instability* on systems where disease and dispersal play a non-independent role are explored. The discussion below, we hope, will instigate further research on the study of diffusive instability in the settings introduced in this manuscript.

A general SI-model can be represented abstractly via the following set of equations

$$\frac{dS}{dt} = f(S, I)
\frac{dI}{dt} = g(S, I),$$
(19)

operating under the assumption that System (19) has a local asymptotically stable interior equilibrium (S^*, I^*) , an assumption formulated using the inequalities

$$f_S + g_I < 0 \text{ and } f_S g_I - f_I g_S > 0$$
 (20)

where $f_S = \frac{\partial f}{\partial S}(S^*, I^*), f_I = \frac{\partial f}{\partial I}(S^*, I^*), g_S = \frac{\partial g}{\partial S}(S^*, I^*), g_I = \frac{\partial g}{\partial I}(S^*, I^*).$ The inclusion of dispersal leads, for example, to the study of symmetric two-patch models. An

The inclusion of dispersal leads, for example, to the study of symmetric two-patch models. An example of such a system is given by the following set of equations:

$$\frac{dS_1}{dt} = f(S_1, I_1) + l_S(S_2 - S_1)$$

$$\frac{dI_1}{dt} = g(S_1, I_1) + l_I(I_2 - I_1)$$

$$\frac{dS_2}{dt} = f(S_2, I_2) + l_S(S_1 - S_2)$$

$$\frac{dI_2}{dt} = g(S_2, I_2) + l_I(I_1 - I_2)$$
(21)

where l_S is the dispersal rate of the S-class and l_I is the dispersal rate of I-class. A typical pseudo diffusion model analog, involving constant diffusion coefficients, is given by the following system:

$$\frac{\partial S}{dt} = D_S \Delta S + f(S, I)$$

$$\frac{\partial I}{dt} = D_I \Delta I + g(S, I)$$
(22)

where Δ is the Laplacian; D_S , D_I are the constant diffusion coefficients for susceptibles and infectives, respectively. We say the SI Model (19) supports diffusive instability (or Turing Effects) if (S^*, I^*) is a locally asymptotically stable interior equilibrium of System (19) but (S^*, I^*, S^*, I^*) becomes unstable when embedded in the symmetric two-patch model given by System (21) for certain values of l_I , l_S . We can achieve similar results as long as the (S^*, I^*) equilibrium is unstable for the Diffusion System (22) at least for some values of D_S , I_S . The following theorem provides conditions that support the diffusive instability of System (7)-(8) and System(9)-(10):

Theorem 5.1 (Diffusive instability). The general SI model (19) can have diffusive instability only if $f_S g_I < 0$. In particular, System (7)-(8) can support diffusive instability provided that the following inequalities hold

$$\frac{\beta - d}{1 - \rho} < \min\left\{\frac{d}{\rho}, \frac{(1 - \theta)^2}{4}\right\} \text{ and } \frac{1 + \theta}{2} + \frac{\sqrt{(1 - \theta)^2 - \frac{4(\beta - d)}{1 - \rho}}}{2} < \frac{(1 + \theta) + \sqrt{(1 + \theta)^2 - 3\theta}}{3}.$$

System (9)-(10) does not support diffusive instability.

Proof. Recall that the general SI model (19) has locally asymptotically stable interior equilibrium (S^*, I^*) if

$$f_S + g_I < 0$$
 and $f_S g_I - f_I g_S > 0$.

A simple calculation shows that (S^*, I^*, S^*, I^*) is an interior equilibrium of its two-patch model (21) with its stability being determined by the sign of

$$\Lambda = f_S g_I - f_I g_S - 2(f_S l_I + g_I l_S) + 4l_S l_I.$$

Thus, diffusive instability can occur only if $\Lambda < 0$ which indicates that $f_S g_I < 0$, that is, if either

 f_S or g_I , is positive then $(f_S l_I + g_I l_S) > 0$ can be made positive and large enough with the right combination of l_S, l_I ; in other words, we conclude that for these parameter values, we have that $\Lambda < 0$. For example, if $f_S > 0$ then we can select the dispersal rate of the I-class l_I large enough and the dispersal rate of S-class l_S small so that the condition $\Lambda < 0$ is met. Now, under $g_I > 0$ diffusive instability may be possible as long as l_S is large and l_I is small.

Relying on the discussion in Section 8.9 of Brauer and Castillo-Chavez (2012), we conclude that (S^*, I^*) is a steady state of its reaction-diffusion model, namely Model (22), where the necessary and sufficient conditions for *diffusive instability* are given by

$$f_S + g_I < 0, \ f_S g_I - f_I g_S > 0 \text{ and } f_S D_I + g_I D_S > \sqrt{D_S D_I} (f_S g_I - f_I g_S)$$

which also implies that $f_S g_I < 0$.

From Theorem 3.2, we know that if an interior equilibrium (S^*, I^*) is locally asymptotically stable then for System (7)-(8) or System (9)-(10).

$$N^* = S^* + I^* > \frac{1+\theta}{2}.$$

Thus, for System (7)-(8), its Jacobian matrix (12) evaluated at the interior equilibrium (S^*, I^*) gives

$$f_S = S^* \left[1 + \theta - 2N^* + \frac{\beta I^*}{(N^*)^2} \right] \text{ and } g_I = I^* \left[\rho (1 + \theta - 2N^*) - \frac{\beta S^*}{(N^*)^2} \right]$$

which implies that $g_I < 0$ since $N^* > \frac{1+\theta}{2}$. Therefore the possibility of *diffusive instability* in System (7)-(8) requires that $f_S > 0$. Since

$$I^* = \frac{N^*(N^* - \theta) \left(1 - N^*\right)}{\beta} \text{ and } N^* = \frac{1 + \theta}{2} + \frac{\sqrt{(1 - \theta)^2 - \frac{4(\beta - d)}{1 - \rho}}}{2} > \frac{1 + \theta}{2},$$

we have

$$f_S > 0 \Rightarrow 1 + \theta - 2N^* + \frac{\beta I^*}{(N^*)^2} = 1 + \theta - 2N^* + \frac{(N^* - \theta)(1 - N^*)}{N^*} = \frac{-3(N^*)^2 + 2(1 + \theta)N^* - \theta}{N^*} > 0.$$

Since

$$-3(N^*)^2 + 2(1+\theta)N^* - \theta = 0 \Rightarrow N^* = \frac{(1+\theta) \pm \sqrt{(1+\theta)^2 - 3\theta}}{3}$$

therefore $f_S > 0$ requires that $\frac{\beta-d}{1-\rho} < \min\left\{\frac{d}{\rho}, \frac{(1-\theta)^2}{4}\right\}$ for the existence of N^* based on Theorem 3.2 and

$$N^* = \frac{1+\theta}{2} + \frac{\sqrt{(1-\theta)^2 - \frac{4(\beta-d)}{1-\rho}}}{2} < \frac{(1+\theta) + \sqrt{(1+\theta)^2 - 3\theta}}{3}$$

For System (9)-(10), its Jacobian matrix (15) evaluated at (S^*, I^*) gives

$$f_S = S^*(1 + \theta - 2N^*)$$
 and $g_I = \rho I^*(1 + \theta - 2N^*)$

which implies that $f_S < 0$ and $g_I < 0$ since $N^* > \frac{1+\theta}{2}$. Therefore, we conclude that System (9)-(10) does not support diffusive instability.

Remark: A direct application of the proof for Theorem 5.1 leads to the following statements:

- 1. If $f_S > 0$ and $g_I < 0$, then *diffusive instability* for the patchy Model (21) requires that $\frac{l_I}{l_S}$ is large enough and $l_S < \frac{f_S}{2}$ while diffusive instability for the reaction-diffusion Model (22) requires that $\frac{D_I}{D_S}$ is large enough.
- 2. If $f_S < 0$ and $g_I > 0$, then diffusive instability for the patchy Model (21) requires $\frac{l_S}{l_I}$ to be large enough and $l_I < \frac{g_I}{2}$ while diffusive instability for the reaction-diffusion Model (22) requires $\frac{D_S}{D_I}$ to be large enough.

In addition, Theorem 5.1 indicates that the SI System (7)-(8) with frequency-dependent horizontal transmission can support *diffusive instability* under certain conditions. For example, when System (7)-(8) has $\beta = 1, d = 0.85, \rho = 0.05, \theta = 0.2$, then it has two interior equilibria

$$E_{i1} = (S_1^*, I_1^*) = (0.4666247185, 0.08749213472)$$
 and $E_{i2} = (S_2^*, I_2^*) = (0.5439015973, 0.1019815495)$

where E_{i1} is a saddle and E_{i2} is locally asymptotically stable with

$$f_S = 0.0830521552 > 0, f_I = -0.7590531079, g_S = 0.02446282446 \text{ and } g_I = -0.1334319123 < 0.02446282446$$

Thus if we choose $\frac{l_I}{l_S}$ (or $\frac{D_I}{D_S}$) large enough and $l_S < \frac{f_S}{2}$ then *diffusive instability* occurs. These results agree with the study of predator-prey systems by Timm and Okubo (1992), which suggest that the existence of *diffusive instability* in such systems may require density effects on intraspecific coefficients and on a predator's diffusivity that must be sufficiently larger when compared to the prey's. There is a critical value involving the ratio of the prey/predator diffusivities that must be crossed before diffusive instability sets in. An alternative SI Model (23)-(24) with Allee effects and horizontal and vertical transmission disease that can also support *diffusive instability* is given by the system

$$\frac{dS}{dt} = S(S-\theta)\left(1-S-I\right) - \beta SI$$
(23)

$$\frac{dI}{dt} = \beta SI - dI + \rho I (I - \theta) (1 - S - I)$$
(24)

with a locally asymptotically stable interior equilibrium (S^*, I^*) given by

$$f_{S} = S^{*}(1 - N^{*} - S^{*} + \theta), \qquad f_{I} = -S^{*}(S^{*} + \beta - \theta), g_{S} = I^{*}(-\rho I^{*} + \beta + \rho \theta), \qquad g_{I} = \rho I^{*}(1 - N^{*} - I^{*} + \theta)$$
(25)

For example, when $\beta = .1, \theta = .2, d = 0.095, \rho = 0.001$, System (26)-(27) has a unique locally asymptotically stable interior equilibrium $(S^*, I^*) = (0.95, 0.044)$ with

$$f_S = -0.70680 < 0, f_I = -0.8075, g_S = 0.004406864$$
 and $g_I = 0.000006864 > 0.$

Thus, if we choose $\frac{l_S}{l_I}$ (or $\frac{D_S}{D_I}$) large enough and $l_I < \frac{g_I}{2}$, then diffusive instability occurs. This suggests that the existence of diffusive instability in (23)-(24) requires that susceptible's diffusivity is sufficiently larger than that of infectives with a critical value involving the ratio of the susceptible/infectives diffusivilies moving beyond a threshold after which diffusive instability sets in.

The SI System (26)-(27) with Allee effects and disease modified fitness studied by Kang and Castillo-Chavez (2013a) is given by

$$\frac{dS}{dt} = f(S,I) = \begin{cases} (S+\rho I)(S+\alpha_1 I-\theta)(1-S-\alpha_2 I)-\beta SI\\ 0, & \text{if } S=0 \text{ and } (\alpha_1 I-\theta)(1-\alpha_2 I) \le 0 \end{cases},$$
(26)

$$\frac{dI}{dt} = g(S,I) = \beta SI - dI, \qquad (27)$$

where the assumptions of the model and the detailed biological meaning of parameters can be found in Kang and Castillo-Chavez (2013a), cannot support *diffusive instability*. The model can have a locally asymptotically stable interior equilibrium (S^*, I^*) with

$$f_{S} = [2S^{*} + (\alpha_{1} + \rho)I^{*} - \theta] (1 - S^{*} - \alpha_{2}I^{*}) - (S^{*} + \rho I^{*})(S^{*} + \alpha_{1}I^{*} - \theta) - \beta I^{*},$$

$$f_{I} = [(\rho + \alpha_{1})S^{*} + 2\rho\alpha_{1}I^{*} - \theta] (1 - S^{*} - \alpha_{2}I^{*}) - \alpha_{2}(S^{*} + \rho I^{*})(S^{*} + \alpha_{1}I^{*} - \theta) - \beta S^{*}, \qquad (28)$$

$$g_{S} = \beta I^{*}, g_{I} = 0.$$

However, if we replace density-dependent transmission with frequency-dependent transmission in the SI System (26)-(27) then we obtain the following SI System (29)-(30) by letting $\rho = 0, \alpha_1 = \alpha_2 = 1$:

$$\frac{dS}{dt} = f(S,I) = S(N-\theta)(1-N) - \frac{\beta SI}{N}$$
(29)

$$\frac{dI}{dt} = g(S,I) = \frac{\beta SI}{N} - dI, \qquad (30)$$

who supports the unique locally asymptotically stable interior equilibrium

$$(S^*, I^*) = \left(\frac{dN^*}{\beta}, \frac{N^*(N^* - \theta)(1 - N)}{\beta}\right) \text{ where } \frac{d}{\beta} < 1, N^* = \frac{1 + \theta + \sqrt{(1 - \theta)^2 - 4(\beta - d)}}{2}$$

and

$$f_{S} = \frac{\partial f}{\partial S}(S^{*}, I^{*}) = S^{*} \left(1 + \theta - 2N^{*} + \frac{\beta I^{*}}{(N^{*})^{2}}\right), \qquad f_{I} = \frac{\partial f}{\partial I}(S^{*}, I^{*}) = S^{*} \left(1 + \theta - 2N^{*} - \frac{\beta S^{*}}{(N^{*})^{2}}\right) < 0$$

$$g_{S} = \frac{\partial g}{\partial S}(S^{*}, I^{*}) = \frac{\beta (I^{*})^{2}}{(N^{*})^{2}} > 0, \qquad g_{I} = \frac{\partial g}{\partial I}(S^{*}, I^{*}) = -\frac{\beta S^{*} I^{*}}{(N^{*})^{2}} < 0$$

$$(31)$$

Thus, System (29)-(30) can have *diffusive instability* if

$$f_S > 0 \Rightarrow 1 + \theta - 2N^* + \frac{\beta I^*}{(N^*)^2} = 1 + \theta - 2N^* + \frac{\beta I^*}{(N^*)^2} = \frac{-3(N^*)^2 + 2(1+\theta)N^* - \theta}{N} > 0.$$

Therefore, according to the proof of Theorem 5.1 and the discussion on System (7)-(8), we can conclude that sufficient conditions leading to *diffusive instability* are that $\frac{l_I}{l_S}$ (or $\frac{D_I}{D_S}$) is large enough and the following inequalities hold

$$l_S < \frac{f_S}{2}, \, \frac{d}{\beta} < 1, \, \frac{1 + \theta + \sqrt{(1 - \theta)^2 - 4(\beta - d)}}{2} < \frac{(1 + \theta) + \sqrt{(1 + \theta)^2 - 3\theta}}{3}.$$

Since SI-disease and prey-predator interaction models share structural similarities, we first look at the following two patch prey-predator model (32) with differential migration coefficients μ, ν introduced by Levin (1974):

$$\frac{dx_i}{dt} = x_i(K - ax_i - by_i) + \mu(x_j - x_i) = f(x_i, y_i) + \mu(x_j - x_i)$$

$$\frac{dy_i}{dt} = y_i(-L + cx_i + dy_i) + \nu(y_j - y_i) = g(x_i, y_i) + \nu(y_j - y_i)$$
(32)

which supports the equilibrium:

$$x^* = \bar{x}_1 = \bar{x}_2 = \frac{Lb - Kd}{bc - ad}, \ y^* = \bar{y}_1 = \bar{y}_2 = \frac{Kc - La}{bc - ad}$$

and

$$f_{S} = \frac{\partial f}{\partial x_{i}}(x^{*}, y^{*}) = -ax^{*} < 0, \qquad f_{I} = \frac{\partial f}{\partial y_{i}}(x^{*}, y^{*}) = -bx^{*} < 0$$

$$g_{S} = \frac{\partial g}{\partial x_{i}}(x^{*}, y^{*}) = cy^{*} > 0, \qquad g_{I} = \frac{\partial g}{\partial y_{i}}(x^{*}, y^{*}) = dy^{*} > 0$$
(33)

According to Theorem 5.1, we conclude that diffusive instability arises if $\nu < \frac{g_I}{2}$, $\frac{\mu}{\nu}$ is large enough and the following equalities hold

$$\frac{a}{c} < \frac{K}{L} < \min\left\{\frac{b}{d}, \frac{a(b+d)}{d(a+c)}\right\}, \ bc > ad.$$

Notice that the positivity of g_I comes from the assumption that y_i is able to survive without x_i , i.e., $y_i \to \infty$ if $y_i(0) > L/d$. If there is no Allee effects, i.e., d = 0, then the prey-predator Model (32) does not have *diffusive instability*. However, if we replace a Holling-Type I functional response with a Beddington-DeAngelis type functional response in the Prey-predator Model (32) with d = 0, then we obtain the following two-patch prey predator model that can have *diffusive instability*:

$$\frac{dx_i}{dt} = x_i \left(K - ax_i - \frac{by_i}{1 + h_1 x_i + h_2 y_i} \right) + \mu(x_j - x_i) = f(x_i, y_i) + \mu(x_j - x_i)
\frac{dy_i}{dt} = y_i \left(-L + \frac{cx_i}{1 + h_1 x_i + h_2 y_i} \right) + \nu(y_j - y_i) = g(x_i, y_i) + \nu(y_j - y_i)$$
(34)

which supports a unique locally asymptotically stable interior equilibrium (x^*, y^*) whenever

$$ch_2 > bh_1, \frac{cK}{a+h_1K} > L, \mu = \nu = 0$$

and

$$f_{S} = \frac{\partial f}{\partial x_{i}}(x^{*}, y^{*}) = x^{*} \left(-a + \frac{bh_{1}y^{*}}{(1+h_{1}x^{*}+h_{2}y^{*})^{2}} \right), \qquad f_{I} = \frac{\partial f}{\partial y_{i}}(x^{*}, y^{*}) = -\frac{bx^{*}(1+h_{1}x^{*})}{(1+h_{1}x^{*}+h_{2}y^{*})^{2}} < 0$$

$$g_{S} = \frac{\partial g}{\partial x_{i}}(x^{*}, y^{*}) = \frac{cy^{*}(1+h_{2}y^{*})}{(1+h_{1}x^{*}+h_{2}y^{*})^{2}} > 0, \qquad g_{I} = \frac{\partial g}{\partial y_{i}}(x^{*}, y^{*}) = -\frac{ch_{2}x^{*}y^{*}(1+h_{1}x^{*})}{(1+h_{1}x^{*}+h_{2}y^{*})^{2}} < 0$$

$$(35)$$

For example, if $K = .5, L = 0.01, c = 2.5, a = .1, h_1 = 1.5, h_2 = 1., b = 1, \mu = \nu = 0$, then Preypredator Model (32) has a unique locally asymptotically stable interior equilibrium $(x^*, y^*, x^*, y^*) = (0.008084, 1.008, 0.008084, 1.008)$ with $f_S = 0.0022 > 0, g_I = -0.005$.

In this section, we have discussed *diffusive instability* in the context of five different SI-models and three different Prey-predator models. So, what are the **criterions** and mechanisms leading to *diffusive instabilities*? Comparisons between models supporting *diffusive instability* [SI-Models (7)-(8), (23)-(24), (29)-(30); the prey-predator models (32), (34)] and models not supporting *diffusive instability* [SI-models (9)-(10), (26)-(27); the prey-predator model (32) with d = 0] are summarized in Table 5.

6. Conclusion

Parasites and pathogens are sometimes effective "regulators of natural populations" (Anderson and May 1979; Dwyer *et al* 1990). Hence, it is of theoretical and empirical interest to study when multiple transmission modes are preferred; or whether pathogen/disease transmission depends on either host population density or its frequency; or the role of small populations (Allee effects) on populations living under the selection pressures placed by pathogens or parasites. Answers to such questions are needed to assess the role and impact of selection on populations, communities and ecosystems. In this manuscript, we explore the contributions of some of these factors on the dynamics of host-parasite interactions within a controlled setting, namely a general SI model that includes: (a) Horizontally and vertically-transmitted disease modes, (b) Net reproduction terms that account for the limitations posed by Allee effects, (c) Disease induced death rates, and (d) Disease-driven reductions in reproduction ability. The analyses carried out in the prior sections leads to the following conclusions and observations:

- Density- versus frequency-dependent horizontal transmission: From Theorem 3.2, we know that System (7)-(8) can have two interior equilibria, one a saddle and one a locally asymptotically stable equilibrium. System (9)-(10) can support stable limit cycles that emerge via Hopf-Bifurcation (see Figure 6 and 12). In other words, the *SI* model with density-dependent horizontal transmission turns out to support more complicated outcomes than its frequency-dependent counterpart.
- Effects of ρ, β, d and θ : $R_0^h = \frac{\beta}{d}$ is identified as the horizontal-transmission reproduction number and $R_0^v = \frac{\rho}{d}$ as the vertical-transmission reproduction number.
 - 1. Theorem 2.1 and its Corollary 2.1 assert that for the SI System (7)-(8) and the SI System (9)-(10), sufficiently large initial conditions (N(0) = S(0) + I(0)) and R_0^v can prevent extinction.
 - 2. Proposition 3.1, Theorem 3.1 and Theorem 3.2 imply that whenever $R_0^h \leq 1$ we can expect disease-free dynamics in System (7)-(8) and System (9)-(10).
 - 3. The SI-FD given by System (7)-(8) tends to support susceptible-free dynamics under large values of R_0^v ; coexistence of susceptibles and infectives under intermediate values of R_0^v ; and disease-driven extinction for small values of R_0^v .
 - 4. The SI-DD System (9)-(10) supports the following outcomes: (i)No boundary equilibrium on the *I*-axis and possibly disease-driven extinction for a range of parameter values whenever $1 < R_0^h < \frac{1}{\beta}$ and $R_0^v < \frac{4}{(1-\theta)^2}$. (ii) Susceptible-free dynamics for R_0^v -intermediate values; values that satisfy the inequality max $\left\{\frac{4}{(1-\theta)^2}, c_1\right\} < R_0^v < c_2$. (iii) An unique interior equilibrium whenever $R_0^h > \frac{1}{\beta}$ and possibly two interior equilibria if $1 < R_0^h < \frac{1}{\beta}$. (iv) Large values of ρ, θ, β and small values of d can destabilize the system (bifurcation diagrams; Figure
- Horizontal versus vertical modes of transmission in *SI* Systems: Small values of the horizontal transmission rate can lead to the susceptible-free dynamics (in System (7)-(8) and System (9)-(10)) with low reproductive rates for infectives leading, under certain conditions, to diseasedriven extinction.

System (7)-(8) and System (9)-(10) have similar dynamics to those of the SI-model

$$\frac{dS}{dt} = S(N-\theta)(1-N)$$

$$\frac{dI}{dt} = \rho I(N-\theta)(1-N) - dI$$
(36)

whenever the horizontal-transmission reproductive number is not greater than 1 $(R_0^h \leq 1)$ and the vertical-transmission reproduction number is dominant. In this case, System (36) has $E_{0,0} \cup E_{1,0}$ as its global attractor (susceptible-free dynamics).

If horizontal transmission is dominant, that is, the vertical transmission rate is small due to the low reproductive ability (ρ) of those infected ($R_0^v \ge \frac{4}{(1-\theta)^2}$) then System (7)-(8) and System (9)-(10) have similar dynamics to those supported by the *SI*-Models (37) and (38), respectively:

$$\frac{dS}{dt} = S(N-\theta)(1-N) - \frac{\beta SI}{N}$$

$$\frac{dI}{dt} = \frac{\beta SI}{N} - dI$$
(37)

and

11).

$$\frac{dS}{dt} = S(N-\theta)(1-N) - \beta SI$$

$$\frac{dI}{dt} = \beta SI - dI.$$
(38)

Model (37) and (38) can in fact have the disease-driven extinction, under some conditions.

- **Diffusive instability:** Sufficient conditions leading to *diffusive instability* (Theorem 5.1 in Section 5) require that *SI*/Prey-Predator models support a locally asymptotically stable interior equilibrium with the product of the diagonal entries of the Jacobian matrix (evaluated at this interior equilibrium) being negative. In this manuscript, we have investigated possible ways for *diffusive instability* to emerge in five different *SI*-models while contrasting their behavior with those of three different prey-predator models. The results of these comparisons have been summarized (Table 5). From Table 5 we conclude that:
 - 1. In the context of our *SI* models, asymmetricity and nonlinearity that emerge as a result of frequency-dependent horizontal transmission or some forms of vertical transmission in populations under Allee effects, can generate *diffusive instability*.
 - 2. In the context of Prey-Predator models, asymmetricity and nonlinearity arising from certain forms of functional responses such as Beddington-DeAngelis type functional response or Allee effects in the predator population, can generate *diffusive instability*.

In conclusion, the presence of asymmetricity and nonlinearity such as nonlinear density dependent factors including Allee effects, could be critical for the generation of *diffusive instabilities* in both SI models and Prey-Predator models.

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7. References

- [1] W. C. Allee, 1938. The social life of animals. Norton, New York.
- [2] L. H. R. Alvarez, 1998. Optimal harvesting under stochastic fluctuations and critical depensation, Mathematical Biosciences, 152, 63-85.
- [3] R. M. Anderson and R. M. May, 1978. Regulation and stability of host-parasite population interactions I: Regulatory processes; II: Destabilizing processes. J. Anita. Ecol. 47, 219-247; 249-267.
- [4] E. Angulo, G. W. Roemer, L. Berec, J. Gascoigne, and F. Courchamp. 2007. Double Allee effects and extinction in the island fox. *Conservation Biology*, 21,1082-1091.
- [5] J. Antonovics and H. M. Alexander, 1992. Epidemiology of Anther-Smut infection of Silene-alba (= S-Latifolia) caused by Ustilago-Violacea-patterns of spore deposition in experimental populations, *Proceedings of the Royal Society of London B*, 250, 157-163.
- [6] L. Berec, D. S. Boukal and M. Berec, 2001. Linking the Allee effect, sexual reproduction, and temperature-dependent sex determination via spatial dynamics. *The American Naturalist*, 157, 217-230.
- [7] D. S. Boukal and L. Berec, 2002. Single-species models of the Allee effect: extinction boundaries, sex ratios and mate encounters, *Journal of Theoretical Biology*, 218, 375-394.

- [8] F.S. Berezovskaya, G. Karev, B. Song and C. Castillo-Chavez, 2004. A simple epidemic model with surprising dynamics, *Mathematical Biosciences and Engineering*, 1, 1-20.
- [9] F. S. Berezovskaya, B. Song and C. Castillo-Chavez, 2010. Role of Prey dispersal and refuges on predator-prey dynamics, SIAM J. APPL. MATH., 70, 1821-1839.
- [10] F. Brauer, 2004. Backward bifurcations in simple vaccination models, J. Math. Anal. Appl., 298, 418-431.
- [11] F. Brauer and C. Castillo-Chavez, 2012. Mathematical Models in Population Biology and Epidemiology, 2nd Edition, Texts in Applied Mathematics, 40. Springer-Verlag, 530 pages.
- [12] R. Burrows, H. Hofer, and M. L. East, 1995. Population dynamics, intervention and survival in African wild dogs (Lycaon pictus). Proceedings of the Royal Society B: Biological Sciences, 262, 235-245.
- [13] C. Castillo-Chavez, C., K., Cooke, W., Huang, W. and S. A. Levin, 1989. Results on the Dynamics for Models for the Sexual Transmission of the Human Immunodeficiency Virus, *Applied Math. Letters*, 2, 327-331.
- [14] C. Castillo-Chavez and A.A. Yakubu, 2001. Dispersal, disease and life history evolution, Math. Biosc., 173: 35–53.
- [15] C. Castillo-Chavez, Z. Feng and W. Huang, 2002. On the computation of r_0 and its role on global stability. In: Mathematical approaches for emerging and reemerging infectious diseases: an introduction. IMA, **125**, 229-250. Springer, New York.
- [16] C. Castillo-Chavez and B. Song, 2003. Models for the transmission dynamics of fanatic behaviors. In (Banks T. and C. Castillo-Chavez, eds.) *Bioterrorism: Mathematical Modeling Applications to Homeland Security*, SIAM Series Frontiers in Applied Mathematics, 28, 240.
- [17] C. Castillo-Chavez and B. Song, 2004. Dynamical Models of Tuberculosis and applications, Journal of Mathematical Biosciences and Engineering, 1, 361-404.
- [18] F. de Castro and B. Bolker, 2005. Mechanisms of disease-induced extinction, *Ecol. Lett.*, 8,117-126.
- [19] A. Cintron-Arias, C. Castillo-Chavez, L. M. Bettencourt, A. L. Lloyd and H.T. Banks, 2009. Estimation of the effective reproductive number from disease outbreak data, *Math. Biosc. & Eng.* 6, 261-282.
- [20] B.R. Clark and S.H. Faeth, 1997. The consequences of larval aggregation in the butterfly *Chlosyne lacinia*. Ecological Entomology, 22 408–415.
- [21] D. L. Clifford, J. A. K. Mazet, E. J. Dubovi, D. K. Garcelon, T. J. Coonan, P. A. Conrad and L. Munson, 2006. Pathogen exposure in endangered island fox (*Urocyon littoralis*) populations: implications for conservation management. *Biological Conservation*, **131**, 230-243.
- [22] F. Courchamp, T. Clutton-Brock, B. Grenfell, 2000. Multipack dynamics and the Allee effect in the African wild dog, Lycaon pictus. Animal Conservation, 3, 277-285.
- [23] A. Deredec and F. Courchamp, 2006. Combined impacts of Allee effects and parasitism, OIKOS, 112, 667-679.
- [24] F. Courchamp, L. Berec and J. Gascoigne, 2009. Allee effects in ecology and conservation. Oxford University Press.

- [25] J.M. Cushing, 1994. Oscillations in age-structured population models with an Allee effect. Oscillations in nonlinear systems: applications and numerical aspects, J. Comput. Appl. Math. 52, 71-80.
- [26] J. Cushing and J. Hudson, 2012. Evolutionary dynamics and strong Allee effects, Journal of Biological Dynamics, 6, 941-958.
- [27] P. Daszak, L. Berger, A.A. Cunningham, A.D. Hyatt, D.E. Green and R. Speare, 1999. Emerging infectious diseases and amphibian population declines. *Emerging Infectious Diseases*, 5, 735-748.
- [28] S. Del Valle, H.W. Hethcote, J.M. Hyman and C. Castillo-Chavez, 2005. Effects of behavioral changes in a smallpox attack model, *Mathematical Biosciences*, 195, 228-251.
- [29] A. Deredec and F. Courchamp 2006. Combined impacts of Allee effects and parasitism, OIKOS, 112, 667-679.
- [30] A. Drew, E.J. Allen, and L.J.S. Allen, 2006. Analysis of climate and geographic factors affecting the presence of chytridiomycosis in Australia, *Dis. Aquat. Org.* 68, 245-250.
- [31] O. Diekmann and M. Kretzshmar, 1991. Patterns in the effects of infectious diseases on population growth, *Journal of Mathematical Biology*, 29, 539-570.
- [32] J. Dushoff, W. Huang and C. Castillo-Chavez, 1998. Backwards bifurcations and catastrophe in simple models of fatal diseases. J. Math. Biol., 36, 227-248.
- [33] G. Dwyer, S. A. Levin, and L. Buttel, 1990. A simulation model of the population dynamics and evolution of myxomatosis, *Ecological Monographs*, 60, 423-447.
- [34] K.E. Emmert and L.J.S. Allen, 2004. Population persistence and extinction in a discrete-time stagestructured epidemic model, J. Differ. Eqn Appl., 10, 1177-1199.
- [35] S. H. Faeth, K. P. Hadeler and H. R. Thieme, 2007. An apparent paradox of horizontal and vertical disease transmission, J. Biol. Dynamics, 1, 45-62.
- [36] W. F. Fagan, M. A. Lewis, M. G. Neubert and P. Van Den Driessche, 2002. Invasion theory and biological control, *Ecology Letters*, 5, 148-157.
- [37] M. J. Ferrari, S. E. Perkins, L. W. Pomeroy and O. N. Bjrnstad, 2011. Pathogens, social networks, and the paradox of transmission scaling, *Interdisciplinary Perspectives on Infectious Diseases*, 10, 10 pages. DOI:10.11552011267049
- [38] E.P. Fenichel, C. Castillo-Chavez, M.G. Ceddiac, G. Chowell, P. Gonzalez, G. J. Hickling, G. Holloway, R. Horan, B. Morin, C. Perrings, M. Springborn, L. Velazquez and C. Villalobos, 2011. Adaptive human behavior in epidemiological models, *Proc. Natl. Acad. Sci.*, 108, 6306-6311.
- [39] Z. Feng, C. Castillo-Chavez and A. Capurro, 2000. A model for tb with exogenous re-infection. Journal of Theoretical Population Biology, 57, 235-247.
- [40] A. Friedman and A-A. Yakubu, 2011. Fatal disease and demographic Allee effect: population persistence and extinction, *Journal of Biological Dynamics*, DOI:10.108017513758.2011.630489.
- [41] J. C. Gascoigne and R. N. Lipcius, 2004. Allee effects driven by predation. Journal of Applied Ecology, 41, 801-810.
- [42] B. Gonzalez, E. Huerta-Sanchez, A. Ortiz-Nieves, T. Vazquez-Alvarez and C. Kribs-Zaleta, 2003. Am I too fat? Bulimia as an epidemic, *Journal of Mathematical Psychology*, 47, 515-526.
- [43] D. Greenhalgh and M. Griffiths, 2009. Dynamic phenomena arising from an extended core group model, *Mathematical Biosciences*, 221, 136-149.

- [44] Y. Gruntfest Y., R. Arditi and Y. Dombrovsky, 1997. A fragmented population in a varying environment, *Journal of Theoretical Biology*, 185, 539-547.
- [45] J. Guckenheimer and P. Holmes, 1983. Nonlinear Oscillations, Dynamical Systems, and Bifurcations of Vector Fields, Springer-Verlag.
- [46] S. Gupta, R. M. Anderson, and R. M. May, 1991. Potential of community-wide chemotherapy or immunotherapy to control the spread of HIV. *Nature*, 350, 356-359.
- [47] K.P. Hadeler and K. Dietz, 1983. Nonlinear hyperbolic partial differential equations for the dynamics of parasite populations. *Comput. Math. Appl.*, 9, 415-430.
- [48] K.P. Hadeler and J. Müfiller, 1992. The effects of vaccination on sexually transmitted disease in heterosexual populations. In *Mathematical Population Dynamics* O. Arino, D. Axelrod, M. Kimmel, and M. Langlois eds., 3d Int. Conf., 1, Wuerz, Winnipeg, 251-278.
- [49] K. P. Hadeler and C. Castillo-Chavez, 1995. A core group model for disease transmission, *Math. Biosci.*, 128, 41-55.
- [50] K. P. Hadeler and P. van den Driessche, 1997. Backward Bifurcation in Epidemic Control, Mathematical Biosciences, 146, 15-35.
- [51] C.D. Harvell, C.E. Mitchell, J.R. Ward, S. Altizer, A. P. Dobson, R. S. Ostfeld and M. D. Samuel, 2002. Climate warming and disease risks for terrestrial and marine biota, *Science*, **296**, 2158-2162.
- [52] H. Hethcote and J. Yorke, 1984. Gonorrhea: Transmission Dynamics and Control. Lecture Notes in Biomathematics, 56, Springer-Verlag, Berlin.
- [53] H. W. Hethcote and J. W. van Ark, 1987. Epidemiological models for heterogeneous populations: Proportionate mixing, parameter estimation, and immunization programs, *Math. Biosci.*, 84, 85-118.
- [54] F.M. Hilker, M.A. Lewis, H. Seno, M. Langlais and H. Malchow, 2005. Pathogens can slow down or reverse invasion fronts of their hosts, *Biol. Invasions*, 7, 817-832.
- [55] F. M. Hilker, M. Langlais, S. V. Petrovskii and H. Malchow, 2007. A diffusive SI model with Allee effect and application to FIV, *Mathematical Biosciences*, 206, 61-80.
- [56] F. M. Hilker, M.I Langlais and H. Malchow, 2009. The Allee Effect and Infectious Diseases: Extinction, Multistability, and the (Dis-)Appearance of Oscillations, *The American Naturalist*, 173, 72-88.
- [57] F. M. Hilker, 2010. Population collapse to extinction: the catastrophic combination of parasitism and Allee effect, *Journal of Biological Dynamics*, 4, 86-101.
- [58] K. R. Hopper and R. T. Roush, 1993. Mate finding, dispersal, number released, and the success of biological control introductions. *Ecological Entomology*, 18, 321-331.
- [59] V. Hutson, 1984. A theorem on average Liapunov functions. Monatshefte f
 ür Mathematik, 98, 267-275.
- [60] W. Huang, K. L. Cooke and C. Castillo-Chavez, 1992. Stability and bifurcation for a multiple group model for the dynamics of HIV/AIDS transmission, SIAM J. Appl. Math., 52, 835-854.
- [61] W. Huang and C. Castillo-Chavez, 2002. Age-structured Core Groups and their impact on HIV dynamics. In: Mathematical Approaches for Emerging and Reemerging Infectious Diseases: Models, Methods and Theory, IMA 126, 261-273, Springer-Verlag, Berlin-Heidelberg-New York. Edited by Carlos Castillo-Chavez with Pauline van den Driessche, Denise Kirschner and Abdul-Aziz Yakubu.

- [62] S.R.-J. Jang and S.L. Diamond, 2007. A host-parasitoid interaction with Allee effects on the host, Comp. Math. Appl., 53, 89-103.
- [63] Y. Kang and D. Armbruster, 2011. Dispersal effects on a two-patch discrete model for plantherbivore interactions, *Journal of Theoretical Biology*, 268, 84-97.
- [64] Y. Kang and N. Lanchier, 2011. Expansion or extinction: deterministic and stochastic two-patch models with Allee effects, *Journal of Mathematical Biology*, 62, 925-973.
- [65] Y. Kang and C. Castillo-Chavez, 2012. Multiscale analysis of compartment models with dispersal, Journal of Biological Dynamics, 6(2), 50-79.
- [66] Y. Kang and C. Castillo-Chavez, 2013a: A simple epidemiological model for populations in the wild with Allee effects and disease-modified fitness. Submitted to the Journal of Discrete and Continuous Dynamical Systems-B. Under review.
- [67] Y. Kang and C. Castillo-Chavez, 2013b: A simple two-patch epidemiological model with Allee effects and disease-modified fitness. Submitted to the *special AMS Contemporary Math Series* in honor of Ronald Mickens' 70th birthday. Under review.
- [68] W. O. Kermack and A. G. McKendrick, 1927. A contribution to the mathematical theory of epidemics, Proc. Roy. Soc. A, 115, 700-721.
- [69] A. M. Kilpatrick and S. Altizer, 2012. Disease Ecology, Nature Education Knowledge, 3(10), 55.
- [70] C. M. Kribs-Zaleta and J. X. Velasco-Hernandez, 2000. A simple vaccination model with multiple endemic states, *Mathematical Biosciences*, 164, 183-201.
- [71] C.M. Kribs-Zaleta, 2001. Center manifolds and normal forms in epidemic models, in *Mathematical Approaches for Emerging and Re-emerging Infectious Diseases: An Introduction*. Edited by C. Castillo-Chavez, S. Blower, D. Kirschner, P. van den Driessche, and A. A.Yakubu, Springer-Verlag, NewYork, 269-286.
- [72] A. Lajmanovich and J.A. Yorke, 1976. A deterministic model for gonorrhea in a nonhomogeneous population, *Math. Biosci.*, 28, 221-236.
- [73] R. Lande, 1998. Anthropogenic, ecological and genetic factors in extinction and conservation. Researches on Population Ecology, 40, 259-269.
- [74] M. A. Lewis and P. Kareiva, 1993. Allee dynamics and the spread of invading organisms, *Theoretical Population Biology*, 43, 141-158.
- [75] M. Lipsitch, M. A. Nowak, D. Ebert and R. M. May, 1995. The Population Dynamics of Vertically and Horizontally Transmitted Parasites, Proc. R. Soc. Lond. B, 260(1359), 321-327.
- [76] M. Lipsitch, S. Siller and M. A. Nowak, 1996. The evolution of virulence in pathogens with vertical and horizontal transmission, *Evolution*, 50, 1729-1741.
- [77] V. Padrón and M. C. Trevisan, 2000. Effect of aggregating behavior on population recovery on a set of habitat islands, *Mathematical Biosciences*, **165**, 63-78.
- [78] D. Pauly, V. Christensen, S. Guenette, T. J. Pitcher, U. R. Sumaila, C. J. Walters and D. Zeller, 2002. Towards sustainability in world fisheries, *Nature*, 418, 689-695.
- [79] A. Potapov, E. Merrill and M. Lewis, 2012. Wildlife disease elimination and density dependence, Proc. R. Soc. B, Published online May 16, 2012. DOI: 10.1098/rspb.2012.0520

- [80] L.J. Rachowicz, J.-M. Hero, R.A. Alford, J.W. Taylor, J.A.T. Morgan, V.T. Vredenburg, J.P. Collins and C.J. Briggs, 2005. The novel and endemic pathogen hypotheses: competing explanations for the origin of emerging infectious diseases of wildlife, *Conserv. Biol.*, **19**, 1441-1448.
- [81] L.J. Rachowicz, R.A. Knapp, J.A.T. Morgan, M.J. Stice, V.T. Vredenburg, J.M. Parker and C.J. Briggs, 2006. Emerging infectious disease as a proximate cause of amphibian mass mortality, *Ecology*, 87, 1671-1683.
- [82] K. R. Rios-Soto, C. Castillo-Chavez, M. Neubert, E. S. Titi and A-A Yakubu, 2006. *Epidemic Spread in Populations at Demographic Equilibrium* In: Mathematical Studies on Human Disease Dynamics: Emerging Paradigms and Challenges. Edited by Gumel A., Castillo-Chavez, C., Clemence, D.P. and R.E. Mickens, American Mathematical Society, **410**, 297-310.
- [83] F. Sanchez, X. Wang, C. Castillo-Chavez, P. Gruenewald and D. Gorman, 2007. Drinking as an epidemic, a simple mathematical model with recovery and relapse, In: *Therapist's Guide to Evidence Based Relapse Prevention*, Edited by Katie Witkiewitz and G. Alan Marlatt, 353-368.
- [84] J.n Santaella, C. B. Ocampo, N.G. Saravia, F. Mndez, R. Gngora, M. A. Gomez, L. E. Munstermann and R.J. Quinnell, 2011. Leishmania (Viannia) Infection in the Domestic Dog in Chaparral, Colombia, Am J Trop Med Hyg. 84(5), 674-680.
- [85] P. Scalia-Tomba, 1991. The effects of structural behavior change on the spread of HIV in one sex populations. *Math. Biosci.* 91, 547-555.
- [86] K. Sherman and A. M. Duda, 1999. Large Marine Ecosystems: An Emerging Paradigm for Fishery Sustainability, *Fisheries*, 24, 15-26.
- [87] J. Shi and R. Shivaji, 2006. Persistence in reaction diffusion models with weak Allee effect, Journal of Mathematical Biology, 52, 807-829.
- [88] L.F. Skerrat, L. Berger, R. Speare, S. Cashins, K.R. McDonald, A.D. Phillott, H.B. Hines and N. Kenyon, 2007. Spread of chytridiomycosis has caused the rapid global decline and extinction of frogs, *EcoHealth*, 4, 125-134.
- [89] K. F. Smith, D. F. Sax, and K. D. Lafferty, 2006. Evidence for the role of infectious disease in species extinction and endangerment, *Conservation Biology*, 20,1349-1357.
- [90] B. Song, 2002. Dynamical epidemical models and their applications, Phd Dissertation, Cornell University, Ithaca, NY.
- [91] B. Song, M. Garsow-Castillo, K. Rios-Soto, M. Mejran, L. Henso and C. Castillo-Chavez, 2006. Raves Clubs, and Ecstasy: The Impact of Peer Pressure, *Journal of Mathematical Biosciences and Engineering*, 3, 1-18.
- [92] P. A. Stephens and W. J. Sutherland, 1999. Consequences of the Allee effect for behaviour, ecology and conservation. *Trends in Ecology & Evolution* 14 401–405.
- [93] P. A. Stephens, W. J. Sutherland and R.P. Freckleton, 1999. What is the Allee effect?, Oikos, 87, 185-190.
- [94] H. R. Thieme and T. Dhirasakdanon, 2009. Persistence of vertically transmitted parasite strains which protect against more virulent horizontally transmitted strains, *Modeling and Dynamics of Infectious Diseases* (Z. Ma, Y. Zhou, J. Wu, eds.), 187-215, World Scientific, Singapore.
- [95] H. R. Thieme, T. Dhirasakdanon, Z. Han and R. Trevino, 2009. Species decline and extinction: synergy of infectious disease and Allee effect? *Journal of Biological Dynamics*, 3, 305-323.

- [96] P. van den Driessche and J. Watmough, 2000. A simple SIS epidemic model with a backward bifurcation, J. Math. Biol., 40, 525-540.
- [97] X. Wang, 2005. Backward bifurcation in a mathematical model for Tuberculosis with loss of immunity, Ph.D. Thesis, Purdue University.
- [98] J. Wang, J. Shi and J. Wei, 2011. Predator-prey system with strong Allee effect in prey, Journal of Mathematical Biology, 49, 291-331.
- [99] A-A. Yakubu, 2007. Allee effects in a discrete-time SIS epidemic model with infected newborns, Journal of Difference Equations and Applications, 13, 341-356.



Two dimensional Bifurcation Diagram: d-p

Figure 6: An example of two dimensional bifurcation diagram $(d-\rho)$ of System (9)-(10) when $\beta = 0.1$ and $\theta = 0.15$. The black area indicates that the reproduction number of vertical transmission $R_0^v = \frac{\rho}{d}$ is large, i.e., $R_0^v > \max\{\frac{4}{(1-\theta)^2}, c_2\}$; the cyan area indicates that R_0^v has intermediate values, i.e., $\max\{\frac{4}{(1-\theta)^2}, c_1\} < R_0^v < c_2$; the green area indicates that R_0^v has small values, i.e., $\frac{4}{(1-\theta)^2} < R_0^v < c_1$ and the white area indicates that $R_0^v < \frac{4}{(1-\theta)^2}$. The green dots indicate that System (9)-(10) has only one interior equilibrium which can be a source, saddle or sink while the red dots indicate that System (9)-(10) has two interior equilibria where one is a saddle and the other one can be a sink or source.



Figure 7: Schematic phase plane of System (9)-(10) when I-class is not able to persist in the absence of S-class, i.e., $\frac{(1-\theta)^2}{4} < \frac{d}{\rho}$.



Figure 8: Schematic phase plane of System (9)-(10) when I-class is able to persist in the absence of S-class, i.e., $\frac{\rho(1-\theta)^2}{4\beta} > \frac{d}{\beta}$.



Figure 9: Schematic phase plane of System (9)-(10) when I-class is able to persist in the absence of S-class, i.e., $\frac{\rho(1-\theta)^2}{4\beta} > \frac{d}{\beta}$.



Figure 10: Schematic phase plane of System (9)-(10) when I-class is able to persist in the absence of S-class, i.e., $\frac{\rho(1-\theta)^2}{4\beta} > \frac{d}{\beta}$.



(a) Bifurcation Diagram of ρ when $\beta = 1, \theta = 0.15, d = 0.45$ (b) Bifurcation Diagram of β when $\rho = 0.35, \theta = 0.15, d = 0.45$



(c) Bifurcation Diagram of d when $\beta = 1, \theta = 0.15, \rho = 0.35$ (d) Bifurcation Diagram of θ when $\beta = 1, d = 0.45, \rho = 0.35$

Figure 11: Bifurcation Diagrams of ρ , β , d, θ for System (9)-(10) where the blue dots in the left figure means the interior equilibrium is locally asymptotically stable while the red dots means source.



Figure 12: An example of a stable limit cycle when $\beta = 1, \theta = 0.15, d = 0.45, \rho = 0.35, S(0) = 0.26, I(0) = 0.2$ where the blue dots in the right figure means the population of I-class while the red dots means the population of S-class.

Models	f_S	f_I	Diffusive Instability	Potential Mechanisms
SI-model (7)-(8)	$f_S = f_S = S^* \left[1 + \theta - 2N^* + \frac{\beta I^*}{(N^*)^2} \right] > 0$	$g_I = g_I = I^* \left[\rho(1 + \theta - 2N^*) - \frac{\beta S^*}{(N^*)^2} \right] < 0$	Yes	Asymmetricity and nonlinearity arise from frequency-dependent horizontal transmission
SI-System (9)-(10)	$f_S = S^* (1 \! + \! \theta \! - \! 2N^*) < 0$	$g_I = \rho I^* (1 + \theta - 2N^*) < 0$	No	Symmetricity arises from certain forms of vertical transmission with Allee effects; Linearity arises from density-dependent vertical transmission
SI-System (23)-(24)	$f_S = \\ S^*(1 - N^* - S^* + \theta) < 0$	$g_I = \rho I^* (1 - N^* - I^* + \theta) > 0$	Yes	Asymmetricity and nonlinearity of arise from certain forms of vertical transmission with Allee effects
SI-System (26)-(27)	$f_{S} = [2S^{*} + (\alpha_{1} + \rho)I^{*} - \theta] (1 - S^{*} - \alpha_{2}I^{*}) - (S^{*} + \rho I^{*})(S^{*} + \alpha_{1}I^{*} - \theta) -\beta I^{*} < 0$	$g_I = 0$	No	Linearity arises from density-dependent horizontal transmission
SI-System (29)-(30)	$\begin{aligned} f_S &= \\ S^* \left(1 + \theta - 2N^* + \frac{\beta I^*}{(N^*)^2} \right) \\ &> 0 \end{aligned}$	$g_I = -\frac{\beta S^* I^*}{(N^*)^2} < 0$	Yes	Asymmetricity and nonlinearity arise from frequency-dependent horizontal transmission
PP-Model (32)	$f_S = -ax^* < 0$	$g_I = dy^* > 0$	Yes	Nonlinearity arises from Allee effects in predator
$\begin{array}{c} \text{PP-Model} \\ (32) \text{ with} \\ d = 0 \end{array}$	$f_S = -ax^* < 0$	$g_I = 0$	No	Linearity arises from a Holling Type I functional response
PP-Model (34)	$\begin{cases} f_S = \\ x^* \left[-a + \frac{bh_1 y^*}{(1+h_1 x^* + h_2 y^*)^2} \right] \\ > 0 \end{cases}$	$g_I = -\frac{ch_2 x^* y^* (1+h_1 x^*)}{(1+h_1 x^*+h_2 y^*)^2} < 0$	Yes	Asymmetricity and nonlinearity arise from a Beddington-DeAngelis type functional response

Table 5: Summary of *diffusive instability* for SI-models and Prey-Predator models