

On a discrete West Nile epidemic model

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Abstract. A West Nile epidemic model in discrete-time is proposed. The model consists of two interacting populations, the vector and the avian populations. The avian population is classified into susceptible, infective, and recovered classes while an individual vector is either susceptible or infective. The transmission of the disease is assumed only by mosquitoes bites and vertical transmission in the vector population. The model behavior depends on a lumped parameter R_0 . The disease-free equilibrium is locally asymptotically stable if $R_0 < 1$. The system is uniformly persistent and possesses a unique endemic equilibrium if $R_0 > 1$. Consequently, the disease can persist in the populations if $R_0 > 1$.

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1 Introduction

West Nile virus (WNV) is a kind of arthropod-borne virus that are maintained in nature through biological transmission between susceptible vertebrate hosts and blood-feeding arthropods such as mosquitoes. Vertebrates can become infected when an infected arthropod bites them to take a blood meal. The susceptible vectors then become infected once feed on an infected host.

WNV was first isolated from a woman in the West Nile District of Uganda in 1937 and has emerged in recent years in many regions of the United States and Canada. The disease presents a threat and challenge to public and animal health. West Nile virus has been detected in dead birds of at least 138

species. Although birds, particularly crows and jays, infected with the virus can die or become ill, most infected birds do survive. We refer the reader to www.cdc.gov/ncidod/dvbid/westnile for more information about the virus history and its ecology.

Since data collected for the West Nile virus are usually discrete, we develop a discrete-time West-Nile model to investigate evolution of the disease between mosquitoes and bird reservoir hosts. Discrete time West Nile models have been studied in [15, 17]. However, our modeling assumptions are different from that given in [15, 17]. In [15] the vector population is partitioned into larval, susceptible, exposed, and infective classes, and all the newborns are in the larval class, while in [17] the vector population also has an exposed compartment and there is no vertical transmission. Moreover, our incidence rate is different from that studied in [15, 17]. Our model derivation is based on a recent continuous-time model proposed by Cruz-Pacheco et al. [5]. Although other vertebrates such as horses and humans do become infected, these populations are not modeled here.

The resulting epidemic model is a four-dimensional system of difference equations. Sufficient conditions for which solutions remain nonnegative are derived. It is shown that the disease-free equilibrium always exists. Its stability depends on a threshold R_0 . The disease-free equilibrium is locally asymptotically stable if $R_0 < 1$ and unstable if $R_0 > 1$. There exists a unique endemic equilibrium and the system is uniformly persistent when $R_0 > 1$. Consequently, the WNV can persist if $R_0 > 1$. When there is no disease related death for the avian population, it can be shown that the disease-free equilibrium is globally asymptotically stable for $R_0 \leq 1$.

In the following section, model derivation will be presented. Section 3 provides stability analysis of the model. Numerical simulations and a brief summary are given in the last section.

2 Model derivation

Our model consists of two interacting populations: birds and mosquitoes. The transmission of the disease is only by mosquitoes bites and vertical transmission in the vector population. Let $N_a(t)$ and $N_v(t)$ denote the avian and vector pop-

ulations at time t , respectively for $t = 0, 1, \dots$. We assume that the mosquito population under the period of study is a constant, N_v , and the bird population has a constant recruitment rate Λ_a per unit time due to birth and immigration. However, the new arrival birds are all susceptible. The death rates of avian and vector populations are denoted by μ_a and μ_v , respectively.

For simplicity, the birth rate of the vector population is μ_v which is the same as its death rate. That is, μ_v is the number of births per individual per unit time for the mosquito population. It is also assumed that the bird population in the absence of the disease is governed by the difference equation $N_a(t+1) = \Lambda_a + (1 - \mu_a)N_a(t)$. As a result, the bird population in the absence of the disease will always stabilize at the level $(\Lambda_a)/(\mu_a)$.

Similar to the idea used by Kermack and McKendrick [12] for modeling epidemics, the avian population at time t is separated into three compartments: susceptible $S_a(t)$, the healthy susceptible individuals who can contract the disease, infectives $I_a(t)$, the individuals who are infected and are infectious, and recovered $R_a(t)$, who are cured. That is, $N_a(t) = S_a(t) + I_a(t) + R_a(t)$ for $t \geq 0$. Since mosquitos have short life span, the vector population at any given time t is only classified into susceptible, $S_v(t)$, and infectives, $I_v(t)$. There is no recovered class for the vector population and $S_v(t) + I_v(t) = N_v > 0$ for $t \geq 0$.

Let b be the average number of bites per mosquito per unit time. The transmission probability from vectors to birds and from birds to vectors are constants and denoted by β_a and β_v , respectively. Hence a bird receives on average $b \frac{N_v}{N_a}$ bites per unit time. Therefore the infection rate per susceptible bird is

$$b\beta_a \frac{N_v}{N_a} \frac{I_v}{N_v} = b\beta_a \frac{I_v}{N_a},$$

and the infection rate per susceptible mosquito is

$$b\beta_v \frac{I_a}{N_a}.$$

We assume that the infected birds are recovered at a constant rate γ_a , and let α_a be the disease related death rate for the avian population. From the data given in [13, 16] (cf. [5]) it is reasonable to assume that

$$(H1) \quad \alpha_a \leq \gamma_a.$$

We remark that assumption (H1) may not be satisfied for some species of birds such as American crow and blue jay. However, many other species of birds such as common grackle, house sparrow, European starling etc. do have small WNV mortality ([13]), and consequently (H1) will fit in to these particular species of birds.

Notice the average infectious period for an infected bird is

$$\frac{1}{\gamma_a + \mu_a + \alpha_a}.$$

Furthermore, it is assumed that all these parameters $\Lambda_a, \mu_a, \mu_v, b, \beta_a, \beta_v, N_v, \gamma_a$ and α_a are positive. Since vertical transmission of the vector population has been found as an important mechanism in maintaining the virus in natural populations [3, 7, 10], we assume a constant fraction $p, 0 \leq p \leq 1$, of the offspring of the infectious vectors is infectious. Under these biological assumptions, the interaction between vector and avian populations are given below:

$$\left\{ \begin{array}{l} S_a(t+1) = \Lambda_a + (1 - \mu_a)S_a(t) - \frac{b\beta_a}{N_a(t)}I_v(t)S_a(t) \\ I_a(t+1) = \frac{b\beta_a}{N_a(t)}I_v(t)S_a(t) + (1 - \gamma_a - \mu_a - \alpha_a)I_a(t) \\ R_a(t+1) = (1 - \mu_a)R_a(t) + \gamma_a I_a(t) \\ N_a(t+1) = \Lambda_a + (1 - \mu_a)N_a(t) - \alpha_a I_a(t) \\ S_v(t+1) = S_v(t) + (1 - p)\mu_v I_v(t) - \frac{b\beta_v}{N_a(t)}I_a(t)S_v(t) \\ I_v(t+1) = (1 - \mu_v)I_v(t) + p\mu_v I_v(t) + \frac{b\beta_v}{N_a(t)}I_a(t)S_v(t) \\ S_a(0), I_a(0), R_a(0), S_v(0), I_v(0) \geq 0, N_a(0) > 0. \end{array} \right. \quad (2.1)$$

Notice as the birth and death rates of the vector population are the same and offsprings of susceptible mosquitoes are born susceptible, the equation for S_v has the above form.

Discrete time epidemic models have been studied by Allen [1, 2], Lewis et al. [15], Thomas and Urena [17], and more recently by Franke [9] on periodic epidemic models. In [1, 2], the models are expressed in terms of time unit Δt

and nonnegativity of the solutions are derived using the quantity Δt and model parameters. Since data for the West Nile epidemics given in the literature [3, 5, 7, 10, 16] are in terms of days, our time unit is taken to be a day and model (2.1) does not involve time unit Δt as in [1, 2]. We now impose the following conditions on the parameters so that solutions of (2.1) will remain nonnegative as shown in Proposition 2.1.

$$(H2) \quad b\beta_a N_v \leq \Lambda_a, \gamma_a + \mu_a + \alpha_a \leq 1, b\beta_v \leq 1, \text{ and } \mu_v \leq 1.$$

We remark that the first three conditions in (H2) imposed on the parameters are reasonable restrictions. For example, since the time unit is taken to be one day, then according to the data given in [5, 16], we have $\beta_a = 1$, $b = 0.75$, and the maximum values of β_v is 0.68, of γ_a is 0.36, of α_a is 0.19, of μ_a is 0.0004, and of μ_v is 0.06 for a variety of bird species such as blue jay, common grackle, American crow, house sparrow, American robin, rock dove etc. and different species of mosquitos. Therefore the first three conditions in (H2) are easily satisfied. However, we would need the total population of vector to be small or the new arrival of birds to be large in our study for the last inequality in (H2) to be true.

Proposition 2.1. *Solutions of system (2.1) remain nonnegative and are bounded.*

Proof. Let $(S_a(t), I_a(t), R_a(t), N_a(t), S_v(t), I_v(t))$ be a solution of (2.1) with $S_a(0), I_a(0), R_a(0), S_v(0), I_v(0) \geq 0$ and $N_a(0) > 0$. It is sufficient to prove nonnegativity for $t = 1$. Since $S_a(0) + I_a(0) + R_a(0) = N_a(0) > 0$ and $S_v(0) + I_v(0) = N_v > 0$, $S_a(1) \geq \Lambda_a + (1 - \mu_a)S_a(0) - b\beta_a I_v(0) \geq \Lambda_a + (1 - \mu_a)S_a(0) - b\beta_a N_v \geq (1 - \mu_a)S_a(0) \geq 0$ by (H2) and (H3). It is clear that $I_a(1), R_a(1) \geq 0$ by (H3). Moreover, $S_v(1) \geq S_v(0) + (1 - p)\mu_v I_v(0) - b\beta_v S_v(0) \geq 0$ by (H4). Similarly, $N_a(1) \geq \Lambda_a + (1 - \mu_a - \alpha_a)N_a(0) \geq \Lambda_a$ and $I_v(1) \geq 0$ by (H3) and (H5), respectively. Therefore, solutions of (2.1) remain nonnegative by induction.

Notice $N_a(t + 1) \leq \Lambda_a + (1 - \mu_a)N_a(t)$ for $t \geq 0$ implies

$$\limsup_{t \rightarrow \infty} N_a(t) \leq \frac{\Lambda_a}{\mu_a}.$$

As $S_a(t)$, $I_a(t)$, $R_a(t)$ are nonnegative and satisfy $S_a(t) + I_a(t) + R_a(t) = N_a(t)$ for $t \geq 0$, we have

$$\limsup_{t \rightarrow \infty} S_a(t) \leq \frac{\Lambda_a}{\mu_a}, \quad \limsup_{t \rightarrow \infty} I_a(t) \leq \frac{\Lambda_a}{\mu_a}, \quad \text{and} \quad \limsup_{t \rightarrow \infty} R_a(t) \leq \frac{\Lambda_a}{\mu_a}.$$

Moreover, since $S_v(t+1) + I_v(t+1) = N_v$ for $t \geq 0$ and solutions remain nonnegative, $S_v(t), I_v(t) \leq N_v$ for $t \geq 0$. Therefore, solutions of (2.1) are bounded.

It follows from the proof of Proposition 2.1 that $N_a(t) \geq \Lambda_a$ for $t \geq 1$ if $N_a(0) > 0$. Therefore system (2.1) is well-defined. Furthermore, since $I_a(t) + S_a(t) + R_a(t) = N_a(t)$ and $S_v(t) + I_v(t) = N_v$ for $t \geq 0$ from modeling assumptions, we are able to reduce the dimension of system (2.1) so that system (2.1) is equivalent to the following four-dimensional system of difference equations

$$\left\{ \begin{array}{l} S_a(t+1) = \Lambda_a + (1 - \mu_a)S_a(t) - \frac{b\beta_a}{N_a(t)}I_v(t)S_a(t) \\ I_a(t+1) = \frac{b\beta_a}{N_a(t)}I_v(t)S_a(t) + (1 - \gamma_a - \mu_a - \alpha_a)I_a(t) \\ N_a(t+1) = \Lambda_a + (1 - \mu_a)N_a(t) - \alpha_a I_a(t) \\ I_v(t+1) = (1 - \mu_v)I_v(t) + p\mu_v I_v(t) + \frac{b\beta_v}{N_a(t)}I_a(t)(N_v - I_v(t)) \\ S_a(0), I_a(0), I_v(0) \geq 0, N_a(0) > 0. \end{array} \right. \quad (2.2)$$

3 Mathematical analysis

We first study the existence of steady state solutions of (2.2). Clearly there always exists a trivial steady state $E_0 = \left(\frac{\Lambda_a}{\mu_a}, 0, \frac{\Lambda_a}{\mu_a}, 0\right)$, the disease-free equilibrium. The Jacobian matrix of (2.2) evaluated at E_0 has the following form

$$J(E_0) = \begin{pmatrix} 1 - \mu_a & 0 & 0 & -b\beta_a \\ 0 & 1 - \gamma_a - \mu_a - \alpha_a & 0 & b\beta_a \\ 0 & -\alpha_a & 1 - \mu_a & 0 \\ 0 & \mu_a b\beta_v N_v / \Lambda_a & 0 & 1 - (1 - p)\mu_v \end{pmatrix}. \quad (3.1)$$

Let J_1 be the lower 3×3 submatrix of $J(E_0)$. Then J_1 is similar to

$$\begin{pmatrix} 1 - \gamma_a - \mu_a - \alpha_a & b\beta_a & 0 \\ \mu_a b\beta_v N_v / \Lambda_a & 1 - (1 - p)\mu_v & 0 \\ -\alpha_a & 0 & 1 - \mu_a \end{pmatrix}. \quad (3.2)$$

Therefore eigenvalues of $J(E_0)$ are $1 - \mu_a$ of multiplicity 2 and eigenvalues of

$$J_2 = \begin{pmatrix} 1 - \gamma_a - \mu_a - \alpha_a & b\beta_a \\ \mu_a b\beta_v N_v / \Lambda_a & 1 - (1 - p)\mu_v \end{pmatrix}. \quad (3.3)$$

Notice

$$\text{tr } J_2 = 2 - \gamma_a - \mu_a - \alpha_a - (1 - p)\mu_v$$

and

$$\det J_2 = (1 - \gamma_a - \mu_a - \alpha_a)[1 - (1 - p)\mu_v] - \frac{b^2 \mu_a N_v \beta_v \beta_a}{\Lambda_a}.$$

Jury conditions imply that eigenvalues λ of J_2 satisfy $|\lambda| < 1$ if and only if $|\text{tr } J_2| < 1 + \det J_2 < 2$ [8]. It follows from (H2) that $\text{tr } J_2 > 0$ and thus we need to verify $\text{tr } J_2 < 1 + \det J_2 < 2$ for the local stability of E_0 .

Notice $\det J_2 < 1$ is trivially true and thus $1 + \det J_2 < 2$ holds. To verify $\text{tr } J_2 < 1 + \det J_2$ we shall separate our discussion into two cases: $0 \leq p < 1$ and $p = 1$. When $0 \leq p < 1$, a simple computation yields

$$\text{tr } J_2 < 1 + \det J_2 \text{ if and only if } \frac{b^2 \mu_a N_v \beta_a \beta_v}{\Lambda_a \mu_v (1 - p)(\gamma_a + \mu_a + \alpha_a)} < 1.$$

Let

$$R_0 = \frac{b^2 \mu_a N_v \beta_a \beta_v}{\Lambda_a \mu_v (1 - p)(\gamma_a + \mu_a + \alpha_a)}. \quad (3.4)$$

It follows that E_0 is locally asymptotically stable if $R_0 < 1$, and unstable if $R_0 > 1$. When $p = 1$,

$$\text{tr } J_2 = 2 - \gamma_a - \mu_a - \alpha_a \quad \text{and} \quad \det J_2 = 1 - \gamma_a - \mu_a - \alpha_a - \frac{b^2 \mu_a N_v \beta_v \beta_a}{\Lambda_a}.$$

Therefore $\text{tr } J_2 < 1 + \det J_2$ if and only if

$$2 - \gamma_a - \mu_a - \alpha_a < 2 - \gamma_a - \mu_a - \alpha_a - \frac{b^2 \mu_a N_v \beta_v \beta_a}{\Lambda_a}.$$

The above inequality is never valid. Therefore E_0 is always unstable when $p = 1$.

Observe that $\frac{b\mu_a N_v \beta_v}{\Lambda_a(\gamma_a + \mu_a + \alpha_a)}$ can be interpreted as the number of infections produced by a single infected bird during its infectious period in a susceptible mosquito population when the avian population is stabilized at the population level Λ_a/μ_a . Similarly, $\frac{b\beta_a}{(1-p)\mu_v}$ is the number of infections produced by a single infectious mosquito during its lifetime in a susceptible avian population. Therefore, $\sqrt{R_0}$, the geometric mean of these two quantities, may be regarded as the basic reproductive number of the disease.

We proceed to examine the existence of an interior steady state. An interior steady state $(\bar{S}_a, \bar{I}_a, \bar{N}_a, \bar{I}_v)$ must satisfy

$$\begin{cases} \Lambda_a - \mu_a \bar{S}_a - \frac{b\beta_a}{\bar{N}_a} \bar{I}_v \bar{S}_a = 0 \\ \frac{b\beta_a}{\bar{N}_a} \bar{I}_v \bar{S}_a - (\gamma_a + \mu_a + \alpha_a) \bar{I}_a = 0 \\ \Lambda_a - \mu_a \bar{N}_a - \alpha_a \bar{I}_a = 0 \\ -\mu_v \bar{I}_v + p\mu_v \bar{I}_v + \frac{b\beta_v}{\bar{N}_a} \bar{I}_a (N_v - \bar{I}_v) = 0. \end{cases} \quad (3.5)$$

Adding the first two equations of (3.5) resulting

$$\Lambda_a - \mu_a \bar{S}_a - (\gamma_a + \mu_a + \alpha_a) \bar{I}_a = 0.$$

Let

$$A = \gamma_a + \mu_a + \alpha_a.$$

Then

$$\bar{S}_a = \frac{\Lambda_a - A\bar{I}_a}{\mu_a},$$

and $\bar{S}_a > 0$ if and only if $\bar{I}_a < \frac{\Lambda_a}{A}$. The third and fourth equilibrium equations imply

$$\bar{N}_a = \frac{\Lambda_a - \alpha_a \bar{I}_a}{\mu_a},$$

and

$$\bar{I}_v = \frac{b\beta_v \mu_a N_v \bar{I}_a}{(\Lambda_a - \alpha_a \bar{I}_a)(1-p)\mu_v + \mu_a b\beta_v \bar{I}_a}.$$

Substituting these into the second equilibrium equation we have for $0 \leq p < 1$, $\bar{I}_a > 0$ must satisfy

$$\hat{A}x^2 + \hat{B}x + \hat{C} = 0, \quad (3.6)$$

where

$$\begin{aligned} \hat{A} &= \alpha_a [b\mu_a\beta_v - (1-p)\alpha_a\mu_v], \\ \hat{B} &= 2\Lambda_a(1-p)\mu_v\alpha_a - b\Lambda_a\mu_a\beta_v - (1-p)A\Lambda_a\mu_vR_0, \quad \text{and} \\ \hat{C} &= \Lambda_a^2\mu_v(1-p)(R_0 - 1). \end{aligned}$$

Let

$$f(x) = \hat{A}x^2 + \hat{B}x + \hat{C}.$$

Notice

$$f(0) = \hat{C}$$

and since $A = \alpha_a + \gamma_a + \mu_a$,

$$f\left(\frac{\Lambda_a}{A}\right) = \frac{1}{A^2} [b\Lambda_a^2\mu_a\beta_v(\alpha_a - A) - (1-p)\Lambda_a^2\mu_v(A - \alpha_a)^2] < 0.$$

When $R_0 < 1$, $\hat{C} < 0$ and hence for $f(x)$ to have at least one positive root in $(0, \Lambda_a/A)$ it is necessary that

$$\hat{A} < 0 < \frac{-\hat{B}}{2\hat{A}} < \frac{\Lambda_a}{A} \quad \text{and} \quad \hat{B}^2 - 4\hat{A}\hat{C} > 0.$$

Notice that the last inequality is equivalent to $\frac{-\hat{B}}{2\hat{A}} > \frac{-2\hat{C}}{\hat{B}}$, and

$$\frac{-2\hat{C}}{\hat{B}} > \frac{\Lambda_a}{A}$$

if and only if

$$b\mu_a\beta_v - A\mu_v(1-p)R_0 + 2(\mu_a + \gamma_a)\mu_v(1-p) > 0.$$

Substituting R_0 by the expression (3.4) and using (H2), one can see that the above inequality is trivially true. Therefore there exists no feasible solution \bar{I}_a

for (3.6) in $(0, \Lambda_a/A)$ if $0 \leq p < 1$ and $R_0 < 1$. Consequently, system (2.2) has no interior steady state if $0 \leq p < 1$ and $R_0 < 1$.

On the other hand if $R_0 > 1$ then since $\hat{C} > 0$ and $f(\Lambda_a/A) < 0$, it is clear that (3.6) has a unique solution $\bar{I}_a \in (0, \Lambda_a/A)$. As a result, system (2.2) has a unique endemic equilibrium $E_1 = (\bar{S}_a, \bar{I}_a, \bar{N}_a, \bar{I}_v)$ if $0 \leq p < 1$ and $R_0 > 1$. If $0 \leq p < 1$ and $R_0 = 1$, then $\hat{C} = 0$ and $f(x) = 0$ has solutions 0 and $-\hat{B}/\hat{A}$. Notice $\hat{B} = \Lambda_a(1-p)\mu_v\alpha_a - (1-p)(\mu_a + \gamma_a)\Lambda_a\mu_v - b\Lambda_a\mu_a\beta_v < 0$ by (H1). If $\hat{A} > 0$ then it is straightforward to show that $-\hat{B}/\hat{A} > \Lambda_a/A$, and if $\hat{A} < 0$ then it is trivial that $-\hat{B}/\hat{A} < 0$. Therefore if $0 \leq p < 1$ and $R_0 = 1$, there is no feasible \bar{I}_a . We conclude that system (2.2) has no interior steady state if $0 \leq p < 1$ and $R_0 \leq 1$, and has a unique endemic equilibrium if $R_0 > 1$.

When $p = 1$, the threshold R_0 in (3.4) is not defined and the I_a -component, $\bar{I}_a > 0$, of an interior steady state must satisfy (3.6) with

$$\hat{A} = \alpha_a b \mu_a \beta_v, \hat{C} = \frac{b^2 \beta_a \beta_v \mu_a N_v \Lambda_a}{A}$$

and

$$\hat{B} = -b\Lambda_a\mu_a\beta_v - b^2\beta_a\beta_v\mu_aN_v.$$

Since $\hat{A} > 0$, $\hat{C} > 0$ and $f(\Lambda_a/A) < 0$, $f(x) = 0$ has a unique solution \bar{I}_a in $(0, \Lambda_a/A)$. Consequently, (2.2) has a unique endemic equilibrium $E_1 = (\bar{S}_a, \bar{I}_a, \bar{N}_a, \bar{I}_v)$ when $p = 1$. Recall in this case that the disease-free equilibrium $E_0 = (\Lambda_a/\mu_a, 0, \Lambda_a/\mu_a, 0)$ is unstable. The above discussion is summarized below.

Proposition 3.1. *If $0 \leq p < 1$ and $R_0 \leq 1$, then $E_0 = (\Lambda_a/\mu_a, 0, \Lambda_a/\mu_a, 0)$ is the only equilibrium and E_0 is locally asymptotically stable if $R_0 < 1$. If $0 \leq p < 1$ and $R_0 > 1$, then E_0 is unstable and system (2.2) has a unique interior steady state $E_1 = (\bar{S}_a, \bar{I}_a, \bar{N}_a, \bar{I}_v)$. If $p = 1$, then E_0 is unstable and E_1 exists for (2.2).*

Our next goal is to determine local stability of the steady state E_1 . When $p = 1$, it follows from (3.5) that $\bar{I}_v = N_v$. Therefore the Jacobian matrix of

system (2.2) evaluated at E_1 has the following form

$$J(E_1) = \begin{pmatrix} 1 - \mu_a - b\beta_a N_v / \bar{N}_a & 0 & b\beta_a N_v \bar{S}_a / \bar{N}_a^2 & -b\beta_a \bar{S}_a / \bar{N}_a \\ b\beta_a N_v / \bar{N}_a & 1 - A & -b\beta_a N_v \bar{S}_a / \bar{N}_a^2 & b\beta_a \bar{S}_a / \bar{N}_a \\ 0 & -\alpha_a & 1 - \mu_a & 0 \\ 0 & 0 & 0 & 1 - b\beta_v \bar{I}_a / \bar{N}_a \end{pmatrix}.$$

Clearly $1 - b\beta_v \bar{I}_a / \bar{N}_a$ is an eigenvalue of $J(E_1)$ which is less than 1 but greater than zero by (H4). The upper 3×3 submatrix of $J(E_1) - \lambda I$ is similar to the following matrix

$$\hat{J} = \begin{pmatrix} 1 - \mu_a - \lambda & 1 - A - \lambda & 0 \\ b\beta_a N_v / \bar{N}_a & 1 - A - \lambda & -b\beta_a N_v \bar{S}_a / \bar{N}_a^2 \\ 0 & -\alpha_a & 1 - \mu_a - \lambda \end{pmatrix}.$$

Using the third row expansion we see that $1 - \mu_a$ is another eigenvalue and the rest of the two eigenvalues satisfy

$$\lambda^2 + \text{tr } \bar{J} \lambda + \det \bar{J} = 0,$$

where

$$\text{tr } \bar{J} = b\beta_a N_v / \bar{N}_a - 2 + \mu_a + A$$

and

$$\det \bar{J} = (1 - A)(1 - \mu_a) - b\alpha_a \beta_a N_v \bar{S}_a / \bar{N}_a^2 - (1 - A)b\beta_a N_v / \bar{N}_a.$$

Since $\det \hat{J} < 1$, applying the Jury conditions, we need to verify that $-1 - \det \bar{J} < \text{tr } \bar{J} < 1 + \det \bar{J}$. A straightforward calculation yields $\text{tr } \bar{J} < 1 + \det \bar{J}$ if and only if

$$A\mu_a + \frac{Ab\beta_a N_v}{\bar{N}_a} - \frac{b\alpha_a \beta_a \bar{S}_a N_v}{\bar{N}_a^2} > 0.$$

Since $\bar{S}_a < \bar{N}_a$ and $A > \alpha_a$, the above inequality is clearly true. Moreover, $-1 - \det \bar{J} < \text{tr } \bar{J}$ if and only if

$$-\mu_a A + (1 - A) \frac{b\beta_a N_v}{\bar{N}_a} + \frac{b\alpha_a \beta_a N_v \bar{S}_a}{\bar{N}_a^2} < \frac{b\beta_a N_v}{\bar{N}_a}.$$

This inequality also holds as $A > \alpha_a$ and $\bar{S}_a < \bar{N}_a$. We now summarize our discussion in the following proposition.

Proposition 3.2. *System (2.2) has steady states $E_0 = (\Lambda_a/\mu_a, 0, \Lambda_a/\mu_a, 0)$ and $E_1 = (\bar{S}_a, \bar{I}_a, \bar{N}_a, \bar{I}_v)$ when $p = 1$, where E_0 is unstable and E_1 is locally asymptotically stable.*

It is not easy to verify whether the endemic-equilibrium E_1 is locally asymptotically stable when $0 \leq p < 1$ and $R_0 > 1$. We show that the disease can persist by showing that the system is uniformly persistent. We first briefly discuss terminology used in Hofbauer and So [11] which will be adopted for our analysis. Let (X, d) be a metric space and $h : X \rightarrow X$ be continuous with a closed subspace Y such that $X \setminus Y$ is forward invariant under h . It is assumed that X has a global attractor \mathcal{A} . Let M be the maximal compact invariant set in Y . Then h is uniformly persistent (with respect to Y) i.e., there exists $m > 0$ such that $\liminf_{t \rightarrow \infty} d(h^t(x), Y) > m$ for all $x \in X \setminus Y$ if and only if M is isolated in \mathcal{A} and $W^s(M) = \{x \in X : h^t(x) \rightarrow M \text{ as } t \rightarrow \infty\} \subset Y$ [11, Theorem 4.1].

Theorem 3.3. *System (2.2) is uniformly persistent if either $0 \leq p < 1$ and $R_0 > 1$ or if $p = 1$.*

Proof. Let $X = \mathbb{R}_+^4$ and $Y = \partial\mathbb{R}_+^4$, the boundary of X . Let H denote the map induced by system (2.2). It follows from the proof of Proposition 2.1 that $S_a(t), I_a(t), N_a(t), I_v(t) > 0$ for $t \geq 1$ if the initial condition is positive. Therefore $X \setminus Y$ is positively invariant for system (2.2). Clearly system (2.2) has a global attractor and the only invariant set in Y is $\{E_0\}$, which is moreover isolated in $\{(S_a, I_a, N_a, I_v) \in \mathbb{R}_+^4 : S_a + I_a \leq \Lambda_a/\mu_a, N_a \leq \Lambda_a/\mu_a, I_v \leq N_v\}$.

To show $W^s(\{E_0\}) \subset Y$, suppose on the contrary that there exists a solution $(S_a(t), I_a(t), N_a(t), I_v(t))$ with $S_a(0) > 0, I_a(0) > 0, N_a(0) > 0$, and $I_v(0) > 0$ such that $\lim_{t \rightarrow \infty} S_a(t) = \lim_{t \rightarrow \infty} N_a(t) = \Lambda_a/\mu_a$ and $\lim_{t \rightarrow \infty} I_a(t) = \lim_{t \rightarrow \infty} I_v(t) = 0$. Then for any $\epsilon > 0$ there exists $t_0 > 0$ such that

$$1 - \epsilon < \frac{S_a(t)}{N_a(t)} < 1 + \epsilon, 0 < I_a(t), I_v(t) < \epsilon, \text{ and } N_a(t) < \Lambda_a/\mu_a + \epsilon$$

for $t \geq t_0$. We first consider the case when $0 \leq p < 1$ and $R_0 > 1$. Since $R_0 > 1$, we can choose $\epsilon > 0$ such that

$$\frac{b^2 \mu_a \beta_a \beta_v (N_v - \epsilon)(1 - \epsilon)}{(\Lambda_a + \epsilon \mu_a) \mu_v (1 - p)(\gamma_a + \mu_a + \alpha_a)} > 1. \quad (3.7)$$

We have by system (2.2) that

$$I_a(t+1) \geq b\beta_a(1-\epsilon)I_v(t) + (1-\gamma_a-\mu_a-\alpha_a)I_a(t)$$

$$I_v(t+1) \geq \frac{b\beta_v}{\Lambda_a/\mu_a + \epsilon}(N_v - \epsilon)I_a(t) + [1 - (1-p)\mu_v]I_v(t)$$

for $t \geq t_0$. Consider the following linear system

$$\begin{cases} x(t+1) = (1-\gamma_a-\mu_a-\alpha_a)x(t) + b\beta_a(1-\epsilon)y(t) \\ y(t+1) = \frac{b\beta_v\mu_a}{\Lambda_a + \epsilon\mu_a}(N_v - \epsilon)x(t) + [1 - (1-p)\mu_v]y(t) \\ x(t_0) = I_a(t_0), y(t_0) = I_v(t_0). \end{cases} \quad (3.8)$$

Let \mathbf{A} denote the map induced by system (3.8). Notice each entry of \mathbf{A} is positive and it follows from (3.7) that the spectral radius of \mathbf{A} is larger than unity. Since $x(t_0) = I_a(t_0) > 0$ and $y(t_0) = I_v(t_0) > 0$, solutions of (3.8) are unbounded. As a result, $I_a(t)$ and $I_v(t)$ also become unbounded large as $t \rightarrow \infty$. We obtain a contradiction and conclude that $W^s(\{E_0\}) \subset Y$. Therefore, system (2.2) is uniformly persistent with respect to Y by [11, Theorem 4.1], i.e., there exists $m > 0$ such that $\liminf_{t \rightarrow \infty} S_a(t) \geq m$, $\liminf_{t \rightarrow \infty} I_a(t) \geq m$, $\liminf_{t \rightarrow \infty} N_a(t) \geq m$ and $\liminf_{t \rightarrow \infty} I_v(t) \geq m$ for any solution $(S_a(t), I_a(t), N_a(t), I_v(t))$ with positive initial condition. The case when $p = 1$ can be shown similarly using instability of E_0 .

Although it is known that the crow family of birds have very high WNV mortality rate, the mortality rate of some other species of birds such as house barrow and common grackle are usually very small. In particular, European starling, rock dove, American robin, and several other species of birds have zero WNV mortality rate as shown in an experimental study by Komar [13]. Therefore, it is reasonable to consider the special case when there is no WNV related mortality for the avian population. In this situation $\lim_{t \rightarrow \infty} N_a(t) = \Lambda_a/\mu_a$ and (2.2) has the following three-dimensional limiting system

$$\begin{cases} S_a(t+1) = \Lambda_a + (1-\mu_a)S_a(t) - \frac{b\beta_a\mu_a}{\Lambda_a}I_v(t)S_a(t) \\ I_a(t+1) = \frac{b\beta_a\mu_a}{\Lambda_a}I_v(t)S_a(t) + (1-\gamma_a-\mu_a)I_a(t) \\ I_v(t+1) = (1-\mu_v)I_v(t) + p\mu_v I_v(t) + \frac{b\beta_v\mu_a}{\Lambda_a}I_a(t)(N_v - I_v(t)) \\ S_a(0), I_a(0), I_v(0) \geq 0. \end{cases} \quad (3.9)$$

Notice R_0 becomes

$$R_0 = \frac{b^2 \mu_a N_v \beta_a \beta_v}{\Lambda_a \mu_v (1-p)(\gamma_a + \mu_a)}.$$

We show that the disease-free equilibrium $(\Lambda_a/\mu_a, 0, \Lambda_a/\mu_a, 0)$ is globally asymptotically stable for (2.2) if $0 \leq p < 1$ and $R_0 \leq 1$.

Theorem 3.4. *The disease-free equilibrium $E_0 = (\Lambda_a/\mu_a, 0, \Lambda_a/\mu_a, 0)$ is the only equilibrium which is moreover globally asymptotically stable for system (2.2) if $\alpha_a = 0$, $0 \leq p < 1$, and $R_0 \leq 1$.*

Proof. It is clear that (2.2) has only the disease-free equilibrium. Since $S_a(t) + I_a(t) \leq \Lambda_a/\mu_a$ and $I_v(t) \leq N_v$ for $t \geq 0$, we let

$$\Delta = \left\{ (x, y, z) \in \mathbb{R}_+^3 : x + y \leq \frac{\Lambda_a}{\mu_a}, z \leq N_v \right\}.$$

We construct a Liapunov function V

$$V : \Delta \rightarrow \mathbb{R}_+ \text{ by } V(S_a, I_a, I_v) = \tilde{A} \left(\frac{\Lambda_a}{\mu_a} - S_a \right) + \tilde{B} I_a + \tilde{C} I_v,$$

where nonnegative \tilde{A} , \tilde{B} and \tilde{C} will be determined later. Let G denote the map induced by system (3.9). Then $V \geq 0$ on Δ and

$$\begin{aligned} V(G(S_a, I_a, I_v)) &\leq \tilde{A}(1 - \mu_a) \frac{\Lambda_a}{\mu_a} + \left[\tilde{B}(1 - \gamma_a - \mu_a) + \tilde{C} \frac{b\beta_v \mu_a}{\Lambda_a} N_v \right] I_a \\ &\quad + [\tilde{A}b\beta_a + \tilde{B}b\beta_a + \tilde{C}(1 - \mu_v) + \tilde{C}p\mu_v] I_v. \end{aligned}$$

We choose $\tilde{A} = 0$. Then \tilde{B} and \tilde{C} must satisfy

$$\tilde{B}(1 - \gamma_a - \mu_a) + \tilde{C} \frac{b\beta_v \mu_a}{\Lambda_a} N_v \leq \tilde{B} \quad \text{and} \quad \tilde{B}b\beta_a + \tilde{C}(1 - \mu_v + p\mu_v) \leq \tilde{C}.$$

We now let

$$\tilde{C} = 1 \quad \text{and} \quad \tilde{B} = \frac{b\beta_v \mu_a N_v}{\Lambda_a(\gamma_a + \mu_a)}.$$

Then \tilde{C} and \tilde{B} clearly satisfy the above inequalities as $R_0 \leq 1$. Hence $V(G(S_a, I_a, I_v)) \leq V(S_a, I_a, I_v)$ and V is a Liapunov function on Δ .

Let $\mathcal{M} = \{(S_a, I_a, I_v) \in \Delta : V(G(S_a, I_a, I_v)) = V(S_a, I_a, I_v)\}$. Then $\mathcal{M} = \{(S_a, I_a, I_v) \in \Delta : I_a = I_v = 0\}$ and the only invariant set in \mathcal{M} is $(\Lambda_a/\mu_a, 0, 0)$. Therefore, $(\Lambda_a/\mu_a, 0, 0)$ is globally asymptotically stable for system (3.9) by the LaSalle's invariance principle [8, 14]. Since the limiting system (3.9) has only one equilibrium which is moreover globally asymptotically stable when $R_0 \leq 1$, applying [6], we conclude that the disease-free equilibrium is globally asymptotically stable for system (2.2) when $\alpha_a = 0$, $0 \leq p < 1$, and $R_0 \leq 1$.

4 Discussion

It is showed in the previous section that the West Nile virus can be wiped out when $R_0 \leq 1$ and $\mu_a = 0$, and the disease can persist within the populations when $R_0 > 1$. Although it is proved that the disease-free equilibrium is globally asymptotically stable if $\mu_a = 0$, $0 \leq p < 1$ and $R_0 \leq 1$, it is suspected that the disease-free equilibrium is globally asymptotically when $R_0 < 1$, $\mu_a > 0$, and $0 \leq p < 1$. Since stability analysis does not provide any information about the transient behavior of the model which may be very important in terms of eradication and management plans, we next use simple numerical methods to study (2.2).

To simulate model (2.2), we adopt the following parameter values: $\Lambda_a = 140$, $\gamma_a = 0.1$, $\alpha_a = 0.1$, $\mu_a = 0.02$, $b = 0.7$, $\beta_a = 1.0$, $\beta_v = 0.38$, $\mu_v = 0.06$, $N_v = 200$ and $p = 0.2$. Initial conditions are chosen to be $S_a(0) = 1000$, $I_a(0) = 0$, $N_a(0) = 1000$ and $I_v(0) = 100$ for all simulations presented. Notice in this case that $R_0 = 0.5038 < 1$ and system (2.2) has only the disease-free equilibrium. Simulations for this set of parameter values are plotted in Figure 1(a). Both infected populations go to a peak at approximately the same time before they are diminished. Therefore there is a surge of the disease for a short period of time even when $R_0 < 1$.

We next keep the same parameter values but change β_v from 0.38 to 0.78. Then $R_0 = 1.0341 > 1$ and system (2.2) has a unique endemic equilibrium according to Theorem 3.3. The time evolution of the infected populations are plotted in Figure 1(b). It can be seen that both infected populations also increase before they decrease to the equilibrium levels for initial conditions with $I_v(0) \geq 10$. When $0 < I_v(0) < 10$, then both infected populations increase

to equilibrium levels with increasing time. It is known that vertical transmission of virus in the vector population is an important factor for contributing the spread of the disease [3, 7, 10]. We shall investigate this factor using our built model. We vary the parameter value p with the above fixed parameter values so that $R_0 > 1$. When $p = 0.2$, it is calculated $R_0 = 1.0341$, when $p = 0.3$, $R_0 = 1.1818$, and $R_0 = 1.3788$ when $p = 0.4$. The resulting time series of the infected mosquitoes and birds are plotted in Figure 1 (c) and (d) respectively. We see from these two plots that increasing the vertical transmission rate p increases the equilibrium levels and hence increases severity of the epidemics as the peaks of infectives increase with increasing p . However, the time that these peaks occurred is approximately independent of p .

In this manuscript, a simple West Nile epidemic model in discrete-time is proposed and analyzed. Our modeling assumptions are based on a continuous-time model developed by Cruz-Pacheco et al. [5]. In particular, the avian population in the absence of the disease is stabilized in a constant population level and the transmission of the virus is either through infected mosquito bites or natural birth of infected vectors. The dynamics of the epidemics depend on a lumped parameter R_0 . The disease-free equilibrium E_0 is the only equilibrium and is locally asymptotically stable if $R_0 < 1$. It is proved that E_0 is globally asymptotically stable when there is no disease related mortality for the avian population and $R_0 \leq 1$. As a result, the disease can be wiped out in this special situation. However, the epidemic can persist if $R_0 > 1$. From the data given in [5, 16], it is very often that $R_0 > 1$ for many species of birds along with vertical transmission of the vector population. Therefore, very likely that the West Nile epidemic can persist in natural populations as it has been observed in recent years in the U.S.

It is demonstrated numerically via simulations that both infected populations increase initially even when $R_0 < 1$ and the transient behavior of the model depends on initial conditions when $R_0 > 1$. If the initial infected vector population is small, then both infected populations will increase until they reach the equilibrium levels. However, if the initial infected vector population is large, then both infected vector and host populations will reach a maximum number which is much larger than the equilibrium value in a short period of time before they decrease to the equilibrium levels as shown in Figure 1. Therefore in

this situation there will be a severe outbreak of the disease in the beginning of the epidemic.

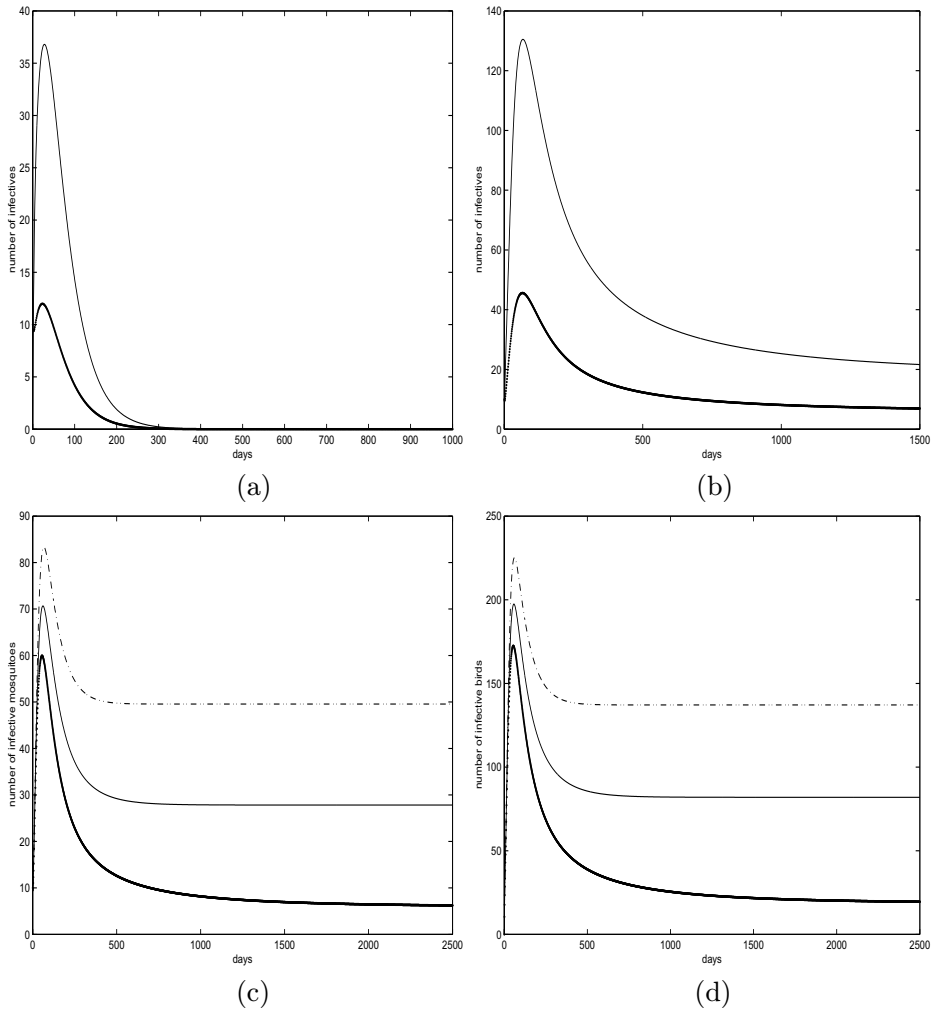


Figure 1 – (a) and (b) plot infective populations when $R_0 < 1$ and $R_0 > 1$, respectively. The solid lines are for the mosquito population and the dotted lines are for the bird population. (c) and (d) plot number of infective mosquitoes and birds versus time for different values of p , respectively. Solid lines are for $p = 0.2$, dotted lines are for $p = 0.3$, and dash-dotted lines are for $p = 0.4$.

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