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Research article

Global dynamics of a two-strain flu model with a single vaccination and general incidence rate

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Abstract: Influenza remains one of the major infectious diseases that target humankind, therefore, understand transmission mechanisms and control strategies can help us obtain more accurate predictions. There are many control strategies, one of them is vaccination. In this paper, our purpose is to extend the incidence rate of a two-strain flu model with a single vaccination, which includes a wide range of incidence rates among them, some cases are not monotonic nor concave, which may be used to reflect media education or psychological effect. Our main aim is to mathematically analyze the effect of the vaccine for strain 1, the general incidence rate of strain 1 and the general incidence rate of strain 2 on the dynamics of the model. Four equilibrium points were obtained and the global dynamics of the model are completely determined via suitable Lyapunov functions. We illustrate our results by some numerical simulations. Our results showed that the vaccination is always beneficial for controlling strain 1, its impact on strain 2 depends on the force of infection of strain 2. Also, the psychological effect is always beneficial for controlling the disease.

Keywords: general nonlinear incidence rate; mathematical model; basic reproduction number; Lyapunov functional; globally asymptotically stable; vaccination; influenza

1. Introduction

Seasonal influenza is an acute respiratory infection caused by influenza viruses. Worldwide, these annual epidemics are estimated to result in about 3 to 5 million cases of severe illness, and about 290,000 to 650,000 respiratory deaths [1]. This infection can have an endemic, epidemic or pandemic behavior.

There were, three major flu pandemics during the 20th century, the so-called Spanish flu (H1N1) in 1918 was the most devastating pandemic. It has been estimated that the Spanish flu claimed around 40–50 million deaths (as much as 3% of the total population), and it also infected 20–40% of the

whole population. In 1957–1958, the Asian flu or bird flu pandemic (H2N2) caused more than two million deaths [2]. Unlike the Spanish flu, this time the infection-causing virus was detected earlier due to the advancement of science and technology. A vaccine was made available but with limited supply. After a decade (in 1968), a flu pandemic (H3N2) that originated again in Hong Kong hit mankind. That flu pandemic also claimed one million lives. In 2009, the H1N1 swine flu is one of the more publicized pandemics that attracted the attention of all scientists and health professionals in the world and made them very much concerned. However, the pandemic did not result in great casualties like before. As of July 2010, only about 18,000 related deaths had been reported [2]. Besides the 4 influenza pandemics since 1918, annual seasonal influenza epidemics have spread among nations on smaller scales. There are many methods of preventing the spread of infectious disease, one of them is vaccination. Vaccination is the administration of agent-specific, but relatively harmless, antigenic components that in vaccinated individuals can induce protective immunity against the corresponding infectious agent [3].

Influenza causes serious public-health problems around the world, therefore, we need to understand transmission mechanisms and control strategies. Mathematical models also provided insight into the severity of past influenza epidemics. Some models were used to investigate the three most devastating historical pandemics of influenza in the 20th century [4–6]. There are a lot of pathogens with several circulating strains.

An important factor when analyzing the dynamics of a disease is the way in which it is transmitted from an infected individual to a healthy one. The incidence rate of a disease is defined as the number of susceptible individuals that become infected per unit of time. It measures the number of new cases of a disease in a period of time. There are different types of incidence functions that have been used in literature in order to model the force of infection of a disease. For example, Rahman and Zou [2] used the bilinear incidence rate $\beta S I$. However, there are more realistic incidence rates than the bilinear incidence rate, For instance, Capasso and his co-workers observed in the seventies [7] that the incidence rate may increase more slowly as I increases, so they proposed a saturated incidence rate $\frac{\beta IS}{1+\tau/I}$.

Baba and Hincal [8] studied an epidemic model consisting of three strains of influenza $(I_1, I_2, \text{ and } I_3)$ where we have vaccine for strain 1 (V_1) only, and force of infection $\frac{\beta S I}{1+\zeta S}$ for strain 2. Baba et al. [9] studied an epidemic model consisting of two strains of influenza $(I_1 \text{ and } I_2)$ where force of infection $\frac{\beta S I_2}{1+\zeta I_2^2}$ for strain 2. As models with more general incidence functions are considered, the dynamics of the system become more complicated. Models with incidence functions of the form g(I)h(S) have been studied, such as [10]. In the most general case, the transmission of the disease may be given by a non-factorable function of *S* and *I*.

In this paper, our purpose is to study model considered in [2] modifying the force of infection in the compartments I_1 and I_2 , by extending the incidence function to a more general form F(S, I), which is based on the incidence rate studied in [11]. Our main aim is to mathematically analyze the effect of the vaccine for strain 1, the general incidence rate of strain 1 ($F_1(S, I_1)$), and the general incidence rate of strain 2 ($F_2(S, I_2)$) on the dynamics of the model (2.2).

This paper is organized as follows. In section 2.1, we formulate the model. In section 3.1, we investigate the disease dynamics described by the model. In section 3.2, we calculate the basic reproduction number. In section 3.3, we establish the existence of equilibrium points. In section 3.4, we study the stability of the model. In section 3.5, provides some numeric simulations to illustrate our main theoretical results. The paper ends with some remarks.

2.1. The model

Rahman and Zou [2] proposed a two-strain model with a single vaccination, namely.

$$\dot{S} = \Lambda - (\beta_1 I_1 + \beta_2 I_2 + \lambda) S$$

$$\dot{V}_1 = rS - (\mu + kI_2)V_1$$

$$\dot{I}_1 = \beta_1 I_1 S - \alpha_1 I_1$$

$$\dot{I}_2 = \beta_2 I_2 S + kI_2 V_1 - \alpha_2 I_2$$

$$\dot{R} = \gamma_1 I_1 + \gamma_2 I_2 - \mu R.$$
(2.1)

where $\lambda = r + \mu$, $\alpha_1 = \gamma_1 + \nu_1 + \mu$, $\alpha_2 = \gamma_2 + \nu_2 + \mu$. The compartments are S(t), $V_1(t)$, $I_1(t)$, $I_2(t)$ and R(t) which denote the population of susceptible, vaccine of strain 1, infective with respect to strain 1, infective with respect to strain 2 and removed individuals at time t, respectively. We assume that all the parameters are positive constants that can be interpreted as follows:

- Λ is the birth rate.
- μ is the death rate.
- *r* is the rate of vaccination with strain 1.
- *k* is the transmission coefficient of vaccinated individuals to strain 2.
- β_1 is the transmission coefficient of susceptible individuals to strain 1.
- β_2 is the transmission coefficient of susceptible individuals to strain 2.
- $\frac{1}{-}$ is the average infection period of strain 1. • $\frac{\gamma_1}{-}$ is the average infection period of strain 2.
- $v_1^{\prime 2}$ is the infection-induced death rate of strain 1.
- v_2 is the infection-induced death rate of strain 2.

The modification of the model (2.1) is given by the following system:

$$\dot{S} = \Lambda - F_1(S, I_1) - F_2(S, I_2) - \lambda S$$

$$\dot{V}_1 = rS - (\mu + kI_2)V_1$$

$$\dot{I}_1 = F_1(S, I_1) - \alpha_1 I_1$$

$$\dot{I}_2 = F_2(S, I_2) + kI_2V_1 - \alpha_2 I_2$$

$$\dot{R} = \gamma_1 I_1 + \gamma_2 I_2 - \mu R.$$
(2.2)

Whose state space is $\mathbb{R}^{5}_{+} = \{(S, V_{1}, I_{1}, I_{2}, R) : S \ge 0, V_{1} \ge 0, I_{1} \ge 0, I_{2} \ge 0, R \ge 0\}$ and subject to the initial conditions $S(0) = S_0 \ge 0$, $V_1(0) = V_{10} \ge 0$, $I_1(0) = I_{10} \ge 0$, $I_2(0) = I_{20} \ge 0$ and $R(0) = R_0 \ge 0$.

We make the following hypotheses on F_i , i = 1, 2.:

H1)
$$F_i(S, I_i) = I_i f_i(S, I_i)$$
 with F_i , $f_i \in \mathbb{C}^2(\mathbb{R}^2_+ \to \mathbb{R}_+)$ and $F(0, I_i) = F(S, 0) = 0$ for all $S, I_i \ge 0$.
H2) $\frac{\partial f_i}{\partial S}(S, I_i) > 0$ and $\frac{\partial f_i}{\partial I_i}(S, I_i) \le 0$ for all $S, I_i \ge 0$.

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H3)
$$\lim_{I_i \to 0^+} \frac{F_i(S, I_i)}{I_i}$$
 exists and is positive for all $S > 0$.

The first of this hypotheses is a basic requirement for any biologically feasible incidence rate, since the disease cannot spread when the number of susceptible or infected individuals is zero.

As for (H2), the condition $\frac{\partial f_i}{\partial S}(S, I_i) > 0$ ensures the monotonicity of $f_i(S, I_i)$ on S, while $\frac{\partial f_i}{\partial I_i}(S, I) \le 0$ suggests that $\frac{F_i(S, I_i)}{I_i}$ is non-increasing with respect to I_i . In the case when f_i monotonically increases with respect to both variables and is concave with respect to I_i , the hypothesis (H2) naturally holds. Concave incidence functions have been used to represent the saturation effect in the transmission rate when the number of infected is very high and exposure to the disease is virtually certain.

(H3) is needed only to ensure that the basic reproduction number is well defined. Some examples of incidence functions studied in the literature that satisfy (H1)–(H3) are as follows:

- (C1) $F(S,I) = \beta S I [2].$
- (C2) $F(S,I) = \frac{\beta SI}{1+\zeta S}$, where $\zeta \ge 0$ describes the psychological effect of general public towards the infective [8].
- (C3) $F(S,I) = \frac{\beta SI}{1+\zeta I^2}$, where $\zeta \ge 0$ measures the psychological or inhibitory effect of the population [9].

A more thorough list can be found in [11]. It should be noted that model (2.2) extends as well as generalizes many special cases.

3. Results

3.1. Disease dynamics described by the model

Lemma 1. Under the initial value $(S_0, V_{10}, I_{10}, I_{20}, R_0) \in \mathbb{R}^5_+$ the system (2.2) has a unique positive and bounded solution in \mathbb{R}^5_+ for t > 0. All solutions ultimately enter and remain in the following bounded and positively invariant region

$$\Omega = \left\{ (S, V_1, I_1, I_2, R) \in \mathbb{R}^5_+ | N = S + V_1 + I_1 + I_2 + R \le \frac{\Lambda}{\mu} \right\}.$$

Proof. The right hand side of system (2.2) is continuous and satisfies the Lipschitz condition in \mathbb{R}^{5}_{+} . Then the system (2.2) has a unique solution ($S(t, V_1(t), I_1(t), I_2(t), R(t))$) in $[0, t_m)$ for some $t_m > 0$. Adding all equations in (2.2), the total population $N = S + V_1 + I_1 + I_2 + R$ satisfies:

$$\begin{split} \dot{N} &= \dot{S} + \dot{V_1} + \dot{I_1} + \dot{I_2} + \dot{R} \\ &= \Lambda - \mu S - \mu V_1 - \mu I_1 - \mu I_2 - \mu R - v_1 I_1 - v_2 I_2 \\ &\leq \Lambda - \mu (S + V_1 + I_1 + I_2 + R) \\ &= \Lambda - \mu N. \end{split}$$

The comparison theorem implies that $\lim_{t\to\infty} \sup N(t) \le \frac{\Lambda}{\mu}$. Hence N(t) is bounded and so are all components S(t), $V_1(t)$, $I_1(t)$, $I_2(t)$ and R(t). This in turn shows that the solution exists globally, i.e. for all $t \ge 0$. Consequently, the solutions S(t), $V_1(t)$, $I_1(t)$, $I_2(t)$, R(t) of (2.2) are ultimately bounded in the positively invariant region Ω .

Let $(S(t), V_1(t), I_1(t), I_2(t), R(t))$ be a solution of system (2.2) with positive initial conditions. Assume by contradiction that there exists t > 0 such that $S(t) \le 0$, $V_1(t) \le 0$, $I_1(t) \le 0$, $I_2(t) \le 0$ or $R(t) \le 0$. By continuity of solutions, this implies that there is a minimal $t_0 > 0$ such that $S(t_0), V_1(t_0)$, $I_1(t_0), I_2(t_0)$ or $R(t_0)$ is zero.

If $S(t_0) = 0$, then $\dot{S} = \Lambda > 0$ at t_0 , so S is increasing in a neighbourhood $(t_0 - \epsilon, t_0 + \epsilon)$ of t_0 . Thus $S(t_0 - \frac{\epsilon}{2}) < S(t_0) = 0$, and since S(0) > 0 and $S(t_0 - \frac{\epsilon}{2}) < 0$, there exists a $t_1 \in (0, t_0 - \frac{\epsilon}{2})$ with $S(t_1) = 0$. But $t_1 < t_0$, which contradicts the minimality of t_0 , then $S(t_0) > 0$.

If $V_1(t_0) = 0$, then $\dot{V_1} = rS(t_0) > 0$ at t_0 . So V_1 is increasing in a neighbourhood $(t_0 - \epsilon, t_0 + \epsilon)$ of t_0 . Thus $V_1(t_0 - \frac{\epsilon}{2}) < V_1(t_0) = 0$, and since $V_1(0) > 0$ and $V_1(t_0 - \frac{\epsilon}{2}) < 0$, there exists a $t_1 \in (0, t_0 - \epsilon/2)$ with $V_1(t_1) = 0$. But $t_1 < t_0$, which contradicts the minimality of t_0 , then $V_1(t_0) > 0$.

If $I_1(t_0) = 0$, then $\dot{I_1} = 0$ at t_0 . On the other hand, any solution with $I_1(0) = 0$ satisfies I(t) = 0 for all t > 0. Since $I_1(0) > 0$ and $I_1(t_0) = 0$, this contradicts the uniqueness of solutions. Similar contradictions are obtained if we assume that $I_2(t_0) = 0$ or $R(t_0) = 0$. Thus we conclude that the solutions of (2.2) are positive for all t > 0.

Since the equation for \dot{R} is actually decoupled from the rest in Eq (2.2), we only need to consider dynamics of the following four-dimensional sub-system:

$$\dot{S} = \Lambda - F_1(S, I_1) - F_2(S, I_2) - \lambda S$$

$$\dot{V}_1 = rS - (\mu + kI_2)V_1$$

$$\dot{I}_1 = F_1(S, I_1) - \alpha_1 I_1$$

$$\dot{I}_2 = F_2(S, I_2) + kI_2V_1 - \alpha_2 I_2.$$
(3.1)

3.2. Basic reproduction number

The basic reproduction number is a dimensionless quantity denoted by \mathcal{R}_0 . It is defined as the expected number of secondary infection cases caused by a single typical infective case during its entire period of infectivity in a wholly susceptible population. Then, referring to the method of [12].

$$\mathcal{F} := \begin{pmatrix} F_1(S, I_1) \\ F_2(S, I_2) + kI_2V_1 \end{pmatrix}.$$
$$\mathcal{V} := \begin{pmatrix} \alpha_1 I_1 \\ \alpha_2 I_2 \end{pmatrix}.$$

Then

$$F' = \begin{pmatrix} \frac{\partial F_1(S,I_1)}{\partial I_1} & 0\\ 0 & \frac{\partial F_2(S,I_2)}{\partial I_2} + kV_1 \end{pmatrix} \Big|_{E_0} = \begin{pmatrix} \frac{\partial F_1(S_0,0)}{\partial I_1} & 0\\ 0 & \frac{\partial F_2(S_0,0)}{\partial I_2} + \frac{kr\Lambda}{\mu\lambda} \end{pmatrix}.$$

$$V' = \begin{pmatrix} \alpha_1 & 0 \\ 0 & \alpha_2 \end{pmatrix} \Big|_{E_0} = \begin{pmatrix} \alpha_1 & 0 \\ 0 & \alpha_2 \end{pmatrix}.$$

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where $E_0 = (S^0, V_1^0, 0, 0) = \left(\frac{\Lambda}{\lambda}, \frac{r\Lambda}{\mu\lambda}, 0, 0\right)$. The matrix F is non-negative and is responsible for new infections (transmission matrix), while the V is invertible and is referred to as the transition matrix for the model (3.1). It follows that,

$$F'V'^{-1} = \begin{pmatrix} \frac{\sigma_1}{\alpha_1} & 0\\ 0 & \frac{\sigma_2}{\alpha_2} + \frac{kr\Lambda}{\alpha_2\mu\lambda} \end{pmatrix}.$$

where $\sigma_i = \frac{\partial F_i(S^0, 0)}{\partial I_i}$, for i = 1, 2. Thus, the basic reproduction number can be calculated as

$$\mathcal{R}_0 = \rho(F'V'^{-1}) = \max\left\{\frac{\sigma_1}{\alpha_1}, \frac{\sigma_2}{\alpha_2} + \frac{kr\Lambda}{\alpha_2\mu\lambda}\right\}.$$

where $\rho(A)$ denotes the spectral radius of a matrix A. Let

$$\mathcal{R}_1 = \frac{\sigma_1}{\alpha_1}$$
 and $\mathcal{R}_2 = \frac{\sigma_2}{\alpha_2} + \frac{kr\Lambda}{\alpha_2\mu\lambda}$.

Then

$$\mathcal{R}_0 = \max\{\mathcal{R}_1, \mathcal{R}_2\}.$$

Therefore $\mathcal{R}_1, \mathcal{R}_2 \leq \mathcal{R}_0$.

3.3. Existence of equilibrium solutions

The four possible equilibrium points for the system (3.1) are: Disease-free equilibrium, single-strain (I_1)-infection, single-strain (I_2)-infection and endemic equilibrium. The system (3.1) has disease-free equilibrium $E_0 = \left(\frac{\Lambda}{\lambda}, \frac{r\Lambda}{\mu\lambda}, 0, 0\right)$ for all parameter values. We will now prove the existence of the other equilibrium points. First we will show some lemmas.

Lemma 2. *For i*=1,2.

$$\frac{\partial F_i(S, I_i)}{\partial I_i} = I \frac{\partial f_i(S, I_i)}{\partial I_i} + \frac{F_i(S, I_i)}{I_i}.$$

Also:

$$\frac{\partial F_i(S, I_i)}{\partial I_i} \le \frac{F_i(S, I_i)}{I_i}$$

Proof. By H1)

$$F_i(S, I_i) = I_i f_i(S, I_i)$$

Then

$$\frac{\partial F_i(S, I_i)}{\partial I_i} = I_i \frac{\partial f_i(S, I_i)}{\partial I_i} + f_i(S, I_i)$$

By H2) $\frac{\partial f_i(S, I_i)}{\partial I_i} \le 0$, then:

$$\frac{\partial F_i(S, I_i)}{\partial I_i} \le f_1(S, I_i) = \frac{F_i(S, I_i)}{I_i}.$$

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Lemma 3. For model (3.1), the closed set $\Omega_1 = \{(S, V_1, I_1, I_2) \in \Omega | S \leq S^0 \text{ and } V_1 \leq V_1^0\}$ is a positively invariant set.

Proof. As Ω is a positively invariant set for model (3.1), it will be enough to show that if $S = S^0$, then $\dot{S} \leq 0$ and if $S \leq S^0$ and $V_1 = V_1^0$, then $\dot{V}_1 \leq 0$. If $S = S^0$, then

$$\dot{S} = \Lambda - F_1(S^0, I_1) - F_2(S^0, I_2) - \lambda S^0$$

= $\lambda S^0 - F_1(S^0, I_1) - F_2(S^0, I_2) - \lambda S^0$
= $-F_1(S^0, I_1) - F_2(S^0, I_2) \le 0$

If $S \leq S^0$ and $V_1 = V_1^0$, Then

$$\dot{V}_1 \leq rS^0 - (\mu + kI_2)V_1^0 = rS^0 - \mu V_1^0 - kI_2V_1^0 = -kI_2V_1^0 \leq 0$$

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Lemma **4.** *For i*=1,2.

$$\frac{\partial F_i(S,I_i)}{\partial S} \geq 0$$

Proof. By (H1)

$$F_i(S, I_i) = I_i f_i(S, I_i)$$

Then

$$\frac{\partial F_i(S, I_i)}{\partial S} = I_i \frac{\partial f_i(S, I_i)}{\partial S} \ge 0 \text{ By H2}.$$

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Remark 1. By (H2) given a and b, if $S \leq a$ and $I_i \geq b$, then $f_i(S, I_i) \leq f_i(a, b)$, i = 1, 2.

- **Theorem 1.** (1) The model (3.1) admits a unique single-strain (I_1)-infection equilibrium $E_1 = (\bar{S}, \bar{V}_1, \bar{I}_1, 0)$ if and only if $\mathcal{R}_1 > 1$.
- (2) The model (3.1) admits a unique single-strain (I₂)-infection equilibrium $E_2 = (\tilde{S}, \tilde{V}_1, 0, \tilde{I}_2)$ if and only if $\mathcal{R}_2 > 1$.

Proof. (1) If $I_2 = 0$ and $\mathcal{R}_1 > 1$, we consider the system

$$\Lambda - F_1(\bar{S}, \bar{I}_1) - \lambda \bar{S} = 0 \tag{3.2}$$

$$r\bar{S} - \mu\bar{V}_1 = 0 \tag{3.3}$$

$$F_1(\bar{S}, \bar{I}_1) - \alpha_1 \bar{I}_1 = 0. \tag{3.4}$$

By (3.3) and (3.4)

$$\bar{V}_1 = \frac{rS}{\mu}, \ F_1(\bar{S}, \bar{I}_1) = \alpha_1 \bar{I}_1.$$

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Substituting in (3.2).

$$\begin{aligned} \Lambda - \alpha_1 \bar{I}_1 - \lambda \bar{S} &= 0\\ \bar{S} &= \frac{\Lambda - \alpha_1 \bar{I}_1}{\lambda}. \end{aligned}$$

Note that $\bar{S} \ge 0$ if and only if $\bar{I}_1 \le \frac{\Lambda}{\alpha_1}$. \bar{I}_1 being determined by the positive roots of the equation.

$$G(\bar{I}_1) \equiv F_1(\frac{\Lambda - \alpha_1 \bar{I}_1}{\lambda}, \bar{I}_1) - \alpha_1 \bar{I}_1.$$
(3.5)

Then

$$G'(\bar{I_1}) = \frac{-\alpha_1}{\lambda} \frac{\partial F_1(\frac{\Lambda - \alpha_1 \bar{I_1}}{\lambda}, \bar{I_1})}{\partial S} + \frac{\partial F_1(\frac{\Lambda - \alpha_1 \bar{I_1}}{\lambda}, \bar{I_1})}{\partial \bar{I_1}} - \alpha_1.$$

And

$$G(0) = F_1(\frac{\Lambda}{\lambda}, 0) = 0$$
 by H1.

$$G'(0) = \frac{-\alpha_1}{\lambda} \frac{\partial F_1(\frac{\Lambda}{\lambda}, 0)}{\partial S} + \frac{\partial F_1(\frac{\Lambda}{\lambda}, 0)}{\partial \bar{I}_1} - \alpha_1$$
$$= \frac{\partial F_1(S^0, 0)}{\partial \bar{I}_1} - \alpha_1 \text{ by } H1$$
$$= \alpha_1 \left(\frac{\sigma_1}{\alpha_1} - 1\right) = \alpha_1 \left(\mathcal{R}_1 - 1\right) > 0.$$

Therefore $G(\overline{I}_1) > 0$ by I_1 sufficiently small. Also

$$G(\frac{\Lambda}{\alpha_1}) = F_1(0, \bar{I}_1) - \Lambda = -\Lambda < 0.$$

Then Eq (3.5) has a positive root. Also if E_1 exists then

$$f_1(\bar{S},\bar{I}_1)-\alpha_1=0.$$

Note that $\bar{S} < S^0$. Then by Lemma 2 and remark 1

$$0 < f_1(S^0, 0) - \alpha_1$$

= $\frac{\partial F_1(S^0, 0)}{\partial I_1} - \alpha_1$
= $\alpha_1 (\mathcal{R}_1 - 1).$

Then $\mathcal{R}_1 > 1$.

Next, we shall show that \bar{I}_1 is unique. From (3.4), it follows that

$$\alpha_1 = f_1(\bar{S}, \bar{I}_1)$$

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Using (H2) and Lemma 2, we have that $\frac{-\alpha_1}{\lambda} \frac{\partial F_1(\bar{S}, \bar{I}_1)}{\partial S} < 0$ and $\bar{I}_1 \frac{\partial f_1(\bar{S}, \bar{I}_1)}{\partial I_1} \le 0$. Furthermore, it can be found that

$$\begin{split} G'(\bar{I_1}) &= \frac{-\alpha_1}{\lambda} \frac{\partial F_1(\frac{\Lambda - \alpha_1 \bar{I_1}}{\lambda}, \bar{I_1})}{\partial S} + \frac{\partial F_1(\frac{\Lambda - \alpha_1 \bar{I_1}}{\lambda}, \bar{I_1})}{\partial \bar{I_1}} - \alpha_1. \\ &= \frac{-\alpha_1}{\lambda} \frac{\partial F_1(\frac{\Lambda - \alpha_1 \bar{I_1}}{\lambda}, \bar{I_1})}{\partial S} + \bar{I_1} \frac{\partial f_1(\bar{S}, \bar{I_1})}{\partial \bar{I_1}} + f_1(\bar{S}, \bar{I_1}) - f_1(\bar{I_1}, \bar{I_1}) \\ &= \frac{-\alpha_1}{\lambda} \frac{\partial F_1(\frac{\Lambda - \alpha_1 \bar{I_1}}{\lambda}, \bar{I_1})}{\partial S} + \bar{I_1} \frac{\partial f_1(\bar{S}, \bar{I_1})}{\partial \bar{I_1}} < 0. \end{split}$$

Which implies that $G(\bar{I}_1)$ strictly decreases at any of the zero points of (3.5). Let us suppose that (3.5) has more than one positive root. Without loss of generality, we choose the one, denoted by $\bar{I_1}^*$, that is the nearest to $\bar{I_1}$. Because of the continuity of $G(\bar{I_1})$, we must have $G'(\bar{I_1}^*) \ge 0$, which results in a contradiction with the strictly decreasing property of $G(\overline{I}_1)$ at all the zero points. (2) If $I_1 = 0$ and $\mathcal{R}_2 > 1$, we consider the system

$$\Lambda - F_2(\tilde{S}, \tilde{I}_2) - \lambda \tilde{S} = 0 \tag{3.6}$$

$$r\tilde{S} - (\mu + k\tilde{I}_2)\tilde{V}_1 = 0$$
(3.7)

$$F_2(\tilde{S}, \tilde{I}_2) + k\tilde{I}_2\tilde{V}_1 - \alpha_2\tilde{I}_2 = 0.$$
(3.8)

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By (3.7) and (3.8)

$$\tilde{V}_1 = \frac{rS}{\mu + k\tilde{I}_2}, \ F_2(\tilde{S}, \tilde{I}_2) = -k\tilde{I}_2\tilde{V}_1 + \alpha_2\tilde{I}_2.$$

Substituting in (3.6).

$$\begin{split} \Lambda &- \alpha_2 \tilde{I}_2 + k \tilde{I}_2 \tilde{V}_1 - \lambda \tilde{S} = 0\\ & \left(\lambda - \frac{k r \tilde{I}_2}{\mu + k \tilde{I}_2}\right) \tilde{S} = \Lambda - \alpha_2 \tilde{I}_2.\\ & \left(\frac{\lambda (\mu + k \tilde{I}_2) - k r \tilde{I}_2}{\mu + k \tilde{I}_2}\right) \tilde{S} = \Lambda - \alpha_2 \tilde{I}_2\\ & \left(\frac{\lambda \mu + (\mu + r) k \tilde{I}_2 - k r \tilde{I}_2}{\mu + k \tilde{I}_2}\right) \tilde{S} = \Lambda - \alpha_2 \tilde{I}_2\\ & \tilde{S} = \left(\Lambda - \alpha_2 \tilde{I}_2\right) \left(\frac{\mu + k \tilde{I}_2}{\lambda \mu + \mu k \tilde{I}_2}\right). \end{split}$$

Note that $\tilde{S} \ge 0$ if and only if $\tilde{I}_2 \le \frac{\Lambda}{a_2}$. \tilde{I}_2 being determined by the positive roots of the equation.

$$H(\tilde{I}_{2}) \equiv F_{2}\left(\frac{(\Lambda - \alpha_{2}\tilde{I}_{2})(\mu + k\tilde{I}_{2})}{\lambda\mu + k\mu\tilde{I}_{2}}, \bar{I}_{2}\right) + k\tilde{I}_{2}\tilde{V}_{1} - \alpha_{2}\tilde{I}_{2}$$
$$= F_{2}\left(\frac{\Lambda\mu + (\Lambda k - \alpha_{2}\mu)\tilde{I}_{2} - k\alpha_{2}\tilde{I}_{2}^{2}}{\lambda\mu + k\mu\tilde{I}_{2}}, \tilde{I}_{2}\right)$$
$$+ \left(\frac{\Lambda rk\tilde{I}_{2} - \alpha_{2}rk\tilde{I}_{2}^{2}}{\lambda\mu + k\mu\tilde{I}_{2}}\right) - \alpha_{2}\tilde{I}_{2}.$$
(3.9)

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Then

$$H'(\tilde{I}_{2}) = \frac{(k\mu)(-k\alpha_{2}\tilde{I}_{2}^{2} - \Lambda\mu) + \lambda\mu(\Lambda k - \alpha_{2}\mu - 2k\alpha_{2}\tilde{I}_{2})}{\left(\lambda\mu + \mu k\tilde{I}_{2}\right)^{2}}$$

$$\times \frac{\partial F_{2}\left(\frac{\Lambda\mu + (\Lambda k - \alpha_{2}\mu)\tilde{I}_{2} - k\alpha_{2}\tilde{I}_{2}^{2}}{\lambda\mu + \mu k\tilde{I}_{2}}, \tilde{I}_{2}\right)}{\partial S}$$

$$+ \frac{\partial F_{2}\left(\frac{\Lambda\mu + (\Lambda k - \alpha_{2}\mu)\tilde{I}_{2} - k\alpha_{2}\tilde{I}_{2}^{2}}{\lambda\mu + \mu k\tilde{I}_{2}}, \tilde{I}_{2}\right)}{\partial \tilde{I}_{2}}$$

$$+ \left(\frac{\lambda\mu(\Lambda rk - \alpha_{2}rk\tilde{I}_{2}) - (k\mu)\alpha_{2}rk\tilde{I}_{2}^{2}}{(\lambda\mu + \mu k\tilde{I}_{2})^{2}}\right) - \alpha_{2}.$$

And

$$H(0) = F_2\left(\frac{\Lambda}{\lambda}, 0\right) = 0$$
 by H1.

$$H'(0) = \frac{\partial F_2(S^0, 0)}{\partial I_2} + \frac{\Lambda rk}{\lambda \mu} - \alpha_2 \text{ by } H1$$
$$= \alpha_2 \left(\frac{\sigma_2}{\alpha_2} + \frac{\Lambda rk}{\alpha_2 \lambda \mu} - 1 \right) = \alpha_2 (\mathcal{R}_2 - 1) > 0.$$

Therefore $H(\tilde{I}_2) > 0$ by \tilde{I}_2 sufficiently small. Also

$$H\left(\frac{\Lambda}{\alpha_2}\right) = F_2\left(0, \frac{\Lambda}{\alpha_2}\right) - \Lambda = -\Lambda < 0.$$

Then Eq (3.9) has a positive root. Also if E_2 exists then

$$\Lambda - F_2(\tilde{S}, \tilde{I}_2) - \lambda \tilde{S} = 0$$

$$f_2(\tilde{S}, \tilde{I}_2) + k \tilde{V}_1 - \alpha_2 = 0.$$

Note that by H1 we have $-F_2(\tilde{S}, \tilde{I}_2) < 0$, then $\Lambda - \lambda \tilde{S} > 0$, therefore $\tilde{S} < S^0$ and $\tilde{V}_1 < V_1^0$. Then by Lemma 2 and remark 1

$$0 < f_2(S^0, 0) + kV_1^0 - \alpha_2$$

= $\frac{\partial F_2(S^0, 0)}{\partial I_2} + kV_1^0 - \alpha_2$
= $\alpha_2 (\mathcal{R}_2 - 1).$

Then $\mathcal{R}_2 > 1$.

Next, we shall show that \tilde{I}_2 is unique. From (3.8), it follows that

$$\alpha_2 - k\tilde{V}_1 = f_2(\tilde{S}, \tilde{I}_2).$$

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Furthermore, it can be found that

$$H'(\tilde{I}_{2}) = \frac{-\alpha_{2}r\mu - \alpha_{2}\mu^{2} - 2\alpha_{2}\mu k\tilde{I}_{2} - 2\alpha_{2}kr\tilde{I}_{2} - \alpha_{2}k^{2}\tilde{I}_{2}^{2} + k\Lambda r}{\mu(\lambda + k\tilde{I}_{2})^{2}}$$

$$\times \frac{\partial F_{2}(\tilde{S}, \tilde{I}_{2})}{\partial S} + \frac{\partial F_{2}(\tilde{S}, \tilde{I}_{2})}{\partial \tilde{I}_{2}} + k\tilde{V}_{1} - \frac{kr(\alpha_{2}\lambda + k\Lambda)\tilde{I}_{2}}{\mu(\lambda + k\tilde{I}_{2})^{2}} - \alpha_{2}$$

$$= \frac{-\alpha_{2}r\mu - \alpha_{2}\mu^{2} - 2\alpha_{2}\mu k\tilde{I}_{2} - 2\alpha_{2}kr\tilde{I}_{2} - \alpha_{2}k^{2}\tilde{I}_{2}^{2} + k\Lambda r}{\mu(\lambda + k\tilde{I}_{2})^{2}}$$

$$\times \frac{\partial F_{2}(\tilde{S}, \tilde{I}_{2})}{\partial S} + \tilde{I}_{2}\frac{\partial f_{2}(\tilde{S}, \tilde{I}_{2})}{\partial \tilde{I}_{2}} - \frac{kr(\alpha_{2}\lambda + k\Lambda)\tilde{I}_{2}}{\mu(\lambda + k\tilde{I}_{2})^{2}}.$$

If $-\alpha_2 r\mu - \alpha_2 \mu^2 + k\Lambda r \le 0$, then $H'(\tilde{I}_2) < 0$ which implies that $H(\tilde{I}_2)$ strictly decreases at any of the zero points of (3.9). Let us suppose that (3.9) has more than one positive root. Without loss of generality, we choose the one, denoted by \tilde{I}_2^* , that is the nearest to \tilde{I}_2 . Because of the continuity of $H(\tilde{I}_2)$, we must have $H'(\tilde{I}_2^*) \ge 0$, which results in a contradiction with the strictly decreasing property of $H(\tilde{I}_2)$ at all the zero points.

If
$$-\alpha_2 r\mu - \alpha_2 \mu^2 + k\Lambda r > 0$$
. Next, we show that $\tilde{I}_2 \notin \left[0, \frac{-r\alpha_2 - \alpha_2 \mu + \sqrt{r\alpha_2(r\alpha_2 + \alpha_2 \mu + k\Lambda)}}{\alpha_2 k}\right]$. Note that
 $\tilde{S}(\tilde{I}_2) = \left(\Lambda - \alpha_2 \tilde{I}_2\right) \left(\frac{\mu + k\tilde{I}_2}{\lambda \mu + \mu k\tilde{I}_2}\right)$.

Then

$$\tilde{S}'(\tilde{I}_2) = \frac{-\alpha_2 r \mu - \alpha_2 \mu^2 - 2\alpha_2 \mu k \tilde{I}_2 - 2\alpha_2 k r \tilde{I}_2 - \alpha_2 k^2 \tilde{I}_2^2 + k \Lambda r}{\mu (\lambda + k \tilde{I}_2)^2}.$$

If $\tilde{I}_2 \in \left[0, \frac{-r\alpha_2 - \alpha_2 \mu + \sqrt{r\alpha_2(r\alpha_2 + \alpha_2 \mu + k\Lambda)}}{\alpha_2 k}\right)$, then $\tilde{S}'(\tilde{I}_2) > 0$, therefore $\tilde{S} \geq S^0$, which results in a contradiction, since $\tilde{S} < S^0$.

Thus $\tilde{I}_2 \in \left[\frac{-r\alpha_2 - \alpha_2\mu + \sqrt{r\alpha_2(r\alpha_2 + \alpha_2\mu + k\Lambda)}}{\alpha_2 k}, \frac{\Lambda}{\alpha_2}\right]$, which implies that $H(\tilde{I}_2)$ strictly decreases at any of the zero points of (3.9). Let us suppose that (3.9) has more than one positive root in $\left[\frac{-r\alpha_2 - \alpha_2\mu + \sqrt{r\alpha_2(r\alpha_2 + \alpha_2\mu + k\Lambda)}}{\alpha_2 k}, \frac{\Lambda}{\alpha_2}\right]$. Without loss of generality, we choose the one, denoted by \tilde{I}_2^* , that is the nearest to \tilde{I}_2 . Note that $H'(\tilde{I}_2^*) < 0$ and $H'(\tilde{I}_2) < 0$. Because of the continuity of $H(\tilde{I}_2)$, we must have $H'(\tilde{I}_2^*) \ge 0$, which results in a contradiction.

The model (3.1) can have endemic infection equilibrium $E_3 = (S^*, V_1^*, I_1^*, I_2^*)$. To find E_3 , we consider the system

$$\Lambda - F_1(S^*, I_1^*) - F_2(S^*, I_2^*) - \lambda S^* = 0$$
(3.10)

$$rS^* - (\mu + kI_2^*)V_1^* = 0 \tag{3.11}$$

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$$F_1(S^*, I_1^*) - \alpha_1 I_1^* = 0 \tag{3.12}$$

$$F_2(S^*, I_2^*) + kI_2^*V_1^* - \alpha_2 I_2^* = 0.$$
(3.13)

By (3.11), (3.12) and (3.13)

$$V_1^* = \frac{rS^*}{\mu + kI_2^*}, \ F_1(S^*, I_1^*) = \alpha_1 I_1^*, \ F_2(S^*, I_2^*) = -kI_2^*V_1^* + \alpha_2 I_2^*.$$

Substituting in (3.10).

$$\begin{split} \Lambda &- \alpha_1 I_1^* - \alpha_2 I_2^* + k I_2^* V_1^* - \lambda S^* = 0 \\ \left(\lambda - \frac{k r I_2^*}{\mu + k I_2^*}\right) S^* &= \Lambda - \alpha_1 I_1^* - \alpha_2 I_2^* \\ \left(\frac{\lambda \mu + (\mu + r) k I_2^* - k r I_2^*}{\mu + k I_2^*}\right) S^* &= \Lambda - \alpha_1 I_1^* - \alpha_2 I_2^* \\ S^* &= (\Lambda - \alpha_1 I_1^* - \alpha_2 I_2^*) \left(\frac{\mu + k I_2^*}{\lambda \mu + \mu k I_2^*}\right). \end{split}$$

Note that $S^* \ge 0$ if and only if $I_1^* \le \frac{\Lambda - \alpha_2 I_2^*}{\alpha_1}$ and $I_2^* \le \frac{\Lambda - \alpha_1 I_1^*}{\alpha_2}$. \bar{I}_2 being determined by the positive roots of the equation.

$$G_2(I_2^*) \equiv f_2\left(\frac{(\Lambda - \alpha_1 I_1^* - \alpha_2 I_2^*)(\mu + kI_2^*)}{\lambda \mu + k \mu I_2^*}, I_2^*\right) + kV_1^* - \alpha_2.$$

 I_1^* being determined by the positive roots of the equation.

$$G_1(I_1^*) \equiv f_1\left(\frac{(\Lambda - \alpha_1 I_1^* - \alpha_2 I_2^*)(\mu + k I_2^*)}{\lambda \mu + k \mu I_2^*}, I_1^*\right) - \alpha_1.$$

3.4. Stability of equilibrium

In this section we will study the local and global stability of the equilibrium points.

Theorem 2. The disease-free equilibrium $E_0 = \left(\frac{\Lambda}{\lambda}, \frac{r\Lambda}{\mu\lambda}, 0, 0\right)$ is unstable if $\mathcal{R}_0 > 1$ while it is locally asymptotically stable if $\mathcal{R}_0 < 1$.

Proof. The Jacobian matrix of the model we get is the following one

$$J := \begin{pmatrix} -\frac{\partial F_1}{\partial S} - \frac{\partial F_2}{\partial S} - \lambda & 0 & -\frac{\partial F_1}{\partial I_1} & -\frac{\partial F_2}{\partial I_2} \\ r & -\mu - kI_2 & 0 & -kV_1 \\ \frac{\partial F_1}{\partial S} & 0 & \frac{\partial F_1}{\partial I_1} - \alpha_1 & 0 \\ \frac{\partial F_2}{\partial S} & kI_2 & 0 & \frac{\partial F_2}{\partial I_2} + kV_1 - \alpha_2 \end{pmatrix}.$$
(3.14)

Then Eq (3.14) at the disease-free equilibrium E_0 is

$$J_{E_0} = \begin{pmatrix} -\frac{\partial F_1(S^0,0)}{\partial S} - \frac{\partial F_2(S^0,0)}{\partial S} - \lambda & 0 & -\frac{\partial F_1(S^0,0)}{\partial I_1} & -\frac{\partial F_2(S^0,0)}{\partial I_2} \\ r & -\mu & 0 & -kV_1^0 \\ \frac{\partial F_1(S^0,0)}{\partial S} & 0 & \frac{\partial F_1(S^0,0)}{\partial I_1} - \alpha_1 & 0 \\ \frac{\partial F_2(S^0,0)}{\partial S} & 0 & 0 & \frac{\partial F_2(S^0,0)}{\partial I_2} + kV_1^0 - \alpha_2 \end{pmatrix}$$

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$$= \begin{pmatrix} -\lambda & 0 & -\frac{\partial F_{1}(S^{0},0)}{\partial l_{1}} & -\frac{\partial F_{2}(S^{0},0)}{\partial l_{2}} \\ r & -\mu & 0 & -kV_{1}^{0} \\ 0 & 0 & \frac{\partial F_{1}(S^{0},0)}{\partial l_{1}} - \alpha_{1} & 0 \\ 0 & 0 & 0 & \frac{\partial F_{2}(S^{0},0)}{\partial l_{2}} + \frac{kr\Lambda}{\mu\lambda} - \alpha_{2} \end{pmatrix}$$

$$= \begin{pmatrix} -\lambda & 0 & -\sigma_{1} & -\sigma_{2} \\ r & -\mu & 0 & -kV_{1}^{0} \\ 0 & 0 & \alpha_{1}\left(\frac{\sigma_{1}}{\alpha_{1}} - 1\right) & 0 \\ 0 & 0 & 0 & \alpha_{2}\left(\frac{\sigma_{2}}{\alpha_{2}} + \frac{kr\Lambda}{\mu\lambda\alpha_{2}} - 1\right) \end{pmatrix}$$

$$= \begin{pmatrix} -\lambda & 0 & -\sigma_{1} & -\sigma_{2} \\ r & -\mu & 0 & -kV_{1}^{0} \\ 0 & 0 & \alpha_{1}\left(\mathcal{R}_{1} - 1\right) & 0 \\ 0 & 0 & 0 & \alpha_{2}\left(\mathcal{R}_{2} - 1\right) \end{pmatrix}. \tag{3.15}$$

Thus the eigenvalues of the above Eq (3.15) are

$$\lambda_1 = -\lambda, \ \lambda_2 = -\mu, \ \lambda_3 = \alpha_1(\mathcal{R}_1 - 1), \ \lambda_4 = \alpha_2(\mathcal{R}_2 - 1).$$
 (3.16)

From (3.16), if $\mathcal{R}_0 < 1$, then $\lambda_3, \lambda_4 < 0$ and we obtain that the disease-free equilibrium E_0 of Model (3.1) is locally asymptotically stable. If $\mathcal{R}_0 > 1$, then the disease-free equilibrium loses its stability. \Box

Theorem 3. Let $\bar{\mathcal{R}}_2 = \frac{1}{\alpha_2} \frac{\partial F_2(\bar{S},0)}{\partial l_2} + \frac{k\bar{V}_1}{\alpha_2}$. The equilibrium E_1 is unstable if $\bar{\mathcal{R}}_2 > 1$ while it is locally asymptotically stable if $\bar{\mathcal{R}}_2 < 1$.

Proof. Then Eq (3.14) at the equilibrium E_1 is

$$J_{E_1} = \begin{pmatrix} A_{11} & 0 & A_{13} & A_{14} \\ r & -\mu & 0 & A_{24} \\ A_{31} & 0 & A_{33} & 0 \\ 0 & 0 & 0 & A_{44} \end{pmatrix}.$$
 (3.17)

where

$$\begin{split} A_{11} &= -\frac{\partial F_1(\bar{S}, \bar{I}_1)}{\partial S} - \lambda < 0 \\ A_{13} &= -\frac{\partial F_1(\bar{S}, \bar{I}_1)}{\partial I_1} < 0 \\ A_{14} &= -\frac{\partial F_2(\bar{S}, 0)}{\partial I_2} \\ A_{24} &= -k\bar{V}_1 < 0 \\ A_{31} &= \frac{\partial F_1(\bar{S}, \bar{I}_1)}{\partial S} > 0 \\ A_{33} &= \frac{\partial F_1(\bar{S}, \bar{I}_1)}{\partial I_1} - \alpha_1 = \bar{I}_1 \frac{\partial f_1(\bar{S}, \bar{I}_1)}{\partial I_1} + f_1(\bar{S}, \bar{I}_1) - \alpha_1 = \bar{I}_1 \frac{\partial f_1(\bar{S}, \bar{I}_1)}{\partial I_1} \le 0 \\ A_{44} &= \frac{\partial F_2(\bar{S}, 0)}{\partial I_2} + k\bar{V}_1 - \alpha_2 = \alpha(\bar{R}_2 - 1). \end{split}$$

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The last equality regarding A_{33} , is because Eq (3.4) implies that $f_1(\bar{S}, \bar{I}_1) - \alpha_1 = 0$. The corresponding characteristic polynomial is

$$p(x) = -(A_{44} - x)(x^3 + a_2x^2 + a_1x + a_0).$$

Then an eigenvalue is A_{44} and the remaining ones satisfy

$$(x^3 + a_2x^2 + a_1x + a_0) = 0.$$

where

$$a_{2} = -(A_{11} - \mu + A_{33}) > 0$$

$$a_{1} = -\mu A_{11} - \mu A_{33} + A_{11}A_{33} - A_{13}A_{31}$$

$$a_{0} = \mu A_{11}A_{33} - \mu A_{13}A_{31}.$$

Note that

$$\begin{aligned} A_{11}A_{33} - A_{13}A_{31} &= \left(-\frac{\partial F_1(\bar{S},\bar{I_1})}{\partial S} - \lambda\right) \left(\frac{\partial F_1(\bar{S},\bar{I_1})}{\partial I_1} - \alpha_1\right) + \frac{\partial F_1(\bar{S},\bar{I_1})}{\partial I_1} \frac{\partial F_1(\bar{S},\bar{I_1})}{\partial S} \\ &= -\lambda \left(\frac{\partial F_1(\bar{S},\bar{I_1})}{\partial I_1} - \alpha_1\right) + \alpha_1 \frac{\partial F_1(\bar{S},\bar{I_1})}{\partial S} > 0. \end{aligned}$$

Then $a_1, a_0 > 0$ and

$$a_{2}a_{1} - a_{0} = -(A_{11} + A_{33})a_{1} + \mu(-\mu A_{11} - \mu A_{33}) + \mu(A_{11}A_{33} - A_{13}A_{31}) - a_{0}$$

= -(A_{11} + A_{33})a_{1} + \mu(-\mu A_{11} - \mu A_{33}) > 0.

Applying the Routh–Hurwitz criterion, we see that all roots of $x^3 + a_2x^2 + a_1x + a_0$ have negative real parts. If $\bar{R}_2 > 1$, then $A_{44} > 0$ therefore E_1 is unstable and if $\bar{R}_2 < 1$, then $A_{44} < 0$ therefore E_1 is stable.

Remark 2. $\bar{S} \leq S^0$ and $\bar{V}_1 \leq V_1^0$, then $\bar{\mathcal{R}}_2 \leq \mathcal{R}_2$, therefore if $\mathcal{R}_2 < 1$ then $\bar{\mathcal{R}}_2 < 1$.

Theorem 4. Let $\tilde{\mathcal{R}}_1 = \frac{1}{\alpha_1} \frac{\partial F_1(\tilde{S},0)}{\partial l_1}$. If $\frac{\partial F_2(\tilde{S},\tilde{I}_2)}{\partial l_2} \leq 0$ the equilibrium E_2 is unstable if $\tilde{\mathcal{R}}_1 > 1$ while it is locally asymptotically stable if $\tilde{\mathcal{R}}_1 < 1$.

Proof. Then Eq (3.14) at the equilibrium E_1 is

$$J_{E_2} = \begin{pmatrix} B_{11} & 0 & B_{13} & B_{14} \\ r & B_{22} & 0 & B_{24} \\ 0 & 0 & B_{33} & 0 \\ B_{41} & B_{42} & 0 & B_{44} \end{pmatrix}.$$
 (3.18)

Where

$$B_{11} = -\frac{\partial F_2(\tilde{S}, \tilde{I}_2)}{\partial S} - \lambda < 0$$

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$$B_{13} = -\frac{\partial F_1(\tilde{S}, 0)}{\partial I_1}$$

$$B_{14} = -\frac{\partial F_2(\tilde{S}, \tilde{I}_2)}{\partial I_2}$$

$$B_{22} = -\mu - k\tilde{I}_2 < 0$$

$$B_{24} = -k\tilde{V}_1 < 0$$

$$B_{33} = \frac{\partial F_1(\tilde{S}, 0)}{\partial I_1} - \alpha_1 = \alpha_1 \left(\tilde{\mathcal{R}}_1 - 1\right)$$

$$B_{41} = \frac{\partial F_2(\tilde{S}, \tilde{I}_2)}{\partial S} > 0$$

$$B_{42} = k\tilde{I}_2 > 0$$

$$B_{44} = \frac{\partial F_2(\tilde{S}, \tilde{I}_2)}{\partial I_2} + k\tilde{V}_1 - \alpha_2 = \tilde{I}_2 \frac{\partial f_2(\tilde{S}, \tilde{I}_2)}{\partial I_2} < 0.$$

The last equality regarding B_{44} , is because Eq (3.8) implies that $k\tilde{V}_1 - \alpha_2 = -f_2(\tilde{S}, \tilde{I}_2)$. The corresponding characteristic polynomial is

$$p(x) = -(B_{33} - x)(x^3 + b_2x^2 + b_1x + b_0)$$

Then (3.18) has an eigenvalue equal to B_{33} and the remaining ones satisfy

$$(x^3 + b_2 x^2 + b_1 x + b_0) = 0.$$

where

$$b_2 = -(B_{11} + B_{22} + B_{44}) > 0.$$

$$b_1 = B_{22}B_{11} + B_{22}B_{44} + B_{11}B_{44} - B_{14}B_{41} - B_{24}B_{42}$$

$$b_0 = -B_{22}B_{11}B_{44} - rB_{14}B_{42} + B_{14}B_{22}B_{41} + B_{11}B_{24}B_{42}.$$

Note that

$$B_{11}B_{44} - B_{14}B_{41} = -\lambda \left(\frac{\partial F_2(\bar{S}, \bar{I}_2)}{\partial I_2} + k\tilde{V}_1 - \alpha_2\right) + \left(-\frac{\partial F_2(\tilde{S}, \bar{I}_2)}{\partial S}\right) \left(k\tilde{V}_1 - \alpha_2\right) > 0.$$

And

$$\begin{split} -B_{22}B_{11}B_{44} - rB_{14}B_{42} + B_{14}B_{22}B_{41} &= \left(\frac{\partial F_2(\tilde{S},\tilde{I_2})}{\partial S} + \lambda\right) \left(k\tilde{V_1} - \alpha_2\right) \left(-\mu - k\tilde{I_2}\right) \\ &- \left(-\mu\right) \left(\frac{\partial F_2(\bar{S},\bar{I_2})}{\partial I_2}\right) \left(-\mu - k\tilde{I_2}\right) \\ &- \left(-r\right) \left(\frac{\partial F_2(\bar{S},\bar{I_2})}{\partial I_2}\right) (-\mu) > 0. \end{split}$$

Then b_1 , $b_0 > 0$. Also

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$$b_{2}b_{1} - b_{0} = -B_{44}b_{1} - B_{22} (B_{22}B_{11} + B_{22}B_{44} - B_{24}B_{42}) - B_{22} (B_{11}B_{44} - B_{14}B_{41}) -B_{11} (B_{22}B_{11} + B_{22}B_{44} + B_{11}B_{44} - B_{14}B_{41}) + B_{11}B_{24}B_{42} +B_{22}B_{11}B_{44} + rB_{14}B_{42} - B_{14}B_{22}B_{41} - B_{11}B_{24}B_{42} = -B_{44}b_{1} - B_{22} (B_{22}B_{11} + B_{22}B_{44} - B_{24}B_{42}) -B_{11} (B_{22}B_{11} + B_{22}B_{44} + B_{11}B_{44} - B_{14}B_{41}) + rB_{14}B_{42} > 0.$$

Applying the Routh–Hurwitz criterion, we see that all roots of $x^3 + b_2x^2 + b_1x + b_0$ have negative real parts. If $\tilde{\mathcal{R}}_1 > 1$, then $B_{33} > 0$ therefore E_2 is unstable and if $\tilde{\mathcal{R}}_1 < 1$, then $B_{33} < 0$ therefore E_2 is stable.

Remark 3. $\tilde{S} \leq S^0$, then $\tilde{\mathcal{R}}_1 \leq \mathcal{R}_1$, therefore if $\mathcal{R}_1 < 1$ then $\tilde{\mathcal{R}}_1 < 1$. **Remark 4.** If $\frac{\partial F_2(\tilde{S}, \tilde{I}_2)}{\partial I_2} > 0$, then $b_i > 0$ i = 0, 1, 2.

Remark 5. The Theorem 4 is valid for $\frac{\partial F_2(\tilde{S}, \tilde{I}_2)}{\partial I_2} > 0$ if $b_2 b_1 - b_0 > 0$.

Theorem 5. If $\overline{\mathcal{R}}_2 > 1$ and $\widetilde{\mathcal{R}}_1 > 1$ then system (3.1) is uniformly persistent.

Proof. The result follows from an application of Theorem 4.6 in [13], with $X_1 = int(\mathbb{R}^4_+)$ and $X_2 = bd(\mathbb{R}^4_+)$ this choice is in accordance with the conditions stated in the theorem. Now, note that by of Lemma 1 there exists a compact set Ω in which all solution of system (3.1) initiated in \mathbb{R}^4_+ ultimately enter and remain forever after. The condition $(C_{4,2})$ is easily verified for this set Ω_1 . On other hand, we denote the omega limit set of the solution $x(t, x_0)$ of system (3.1) starting in $x_0 \in \mathbb{R}^4_+$ by $w(x_0)$. Note that $w(x_0)$ is bounded (Lemma 1), we need to determine the following set:

$$\Omega_2 = \bigcup_{y \in Y_2} w(y), \text{ where } Y_2 = \{x_0 \in X_2 | x(t, x_0) \in X_2, \forall t > 0\}.$$

From the system equations (3.1) it follows that all solutions starting in $bd(\mathbb{R}^4_+)$ but not on the I_1 axis or on the I_2 axis leave $bd(\mathbb{R}^4_+)$. This implies that

$$Y_2 = \left\{ (S, V_1, I_1, I_2) \in bd(\mathbb{R}^4_+) | I_1 = 0 \text{ or } I_2 = 0 \right\}.$$

Furthermore, we see that $\Omega_2 = \{E_0, E_1, E_2\}$, then $\bigcup_{i=1}^3 \{E_i\}$ is a covering of Ω_2 , which is isolated (since E_i (i = 1, 2, 3) is a saddle point) and acyclic. Finally we need to prove that E_i (i = 1, 2, 3) is a weak repeller for X_1 to end the prove.

By definition E_i is a weak repeller for X_1 if for every solution $(S(t), V_1(t), I_1(t), I_2(t))$ starting in $(S_0, V_{10}, I_{10}, I_{20}) \in X_1$

$$\limsup_{t \to +\infty} \|(S(t), V_1(t), I_1(t), I_2(t)) - E_i\| > 0.$$

We will first show that E_0 is a weak repeller for X_1 . Since $\overline{\mathcal{R}}_2 > 1$ and $\overline{\mathcal{R}}_1 > 1$, then $\mathcal{R}_2 = \frac{1}{\alpha_2} \left(f_2(S^0, 0) + kV^0 \right) > 1$ and $\mathcal{R}_1 = \frac{1}{\alpha_1} \left(f_1(S^0, 0) \right) > 1$, therefore $f_2(S^0, 0) + kV^0 - \alpha_2 > 0$ and $f_1(S^0, 0) - \alpha_1 > 0$. Because of the continuity of $f_2(S, I_2) + kV_1 - \alpha_2$ and $f_1(S, I_1) - \alpha_1$, there exists a sufficiently small constant $\eta_2 > 0$, such that $f_1(S^0 - \eta_2, \eta_2) - \alpha_1 > 0$ and $f_2(S^0 - \eta_2, \eta_2) + k(V_1^0 - \eta_2) - \alpha_1 > 0$.

Now, we suppose that E_0 is not a weak repeller for X_1 , i.e., there exists a solution $(S(t), V_1(t), I_1(t), I_2(t))$ starting in $(S_0, V_{10}, I_{10}, I_{20}) \in X_1$ such that

$$\limsup_{t \to +\infty} \|(S(t), V_1(t), I_1(t), I_2(t)) - E_0\| = 0.$$

Then exists $T_1 > 0$ such that for every $\eta_1 > 0$

$$S^0 - \eta_1 < S(t), V_1^0 - \eta_1 < V_1(t), 0 < I_1(t) < \eta_1 \text{ and } 0 < I_2(t) < \eta_1 \ \forall t \ge T_1.$$

Let $\eta_1 = \eta_2$, then for $t \ge T_1$.

$$I_{1} = I_{1} (f_{1}(S, I_{1}) - \alpha_{1})$$

$$\geq I_{1} (f_{1}(S^{0} - \eta_{2}, \eta_{2}) - \alpha_{1}).$$

and

$$\begin{split} \dot{I}_2 &= I_2 \left(f_2(S, I_2) + kV_1 - \alpha_2 \right) \\ &\geq I_2 \left(f_2(S^0 - \eta_2, \eta_2) + k(V_1^0 - \eta_2) - \alpha_2 \right) \end{split}$$

By comparison principle, we have

$$I_1(t) \ge I_1(T_1)e^{(f_1(S^0 - \eta_2, \eta_2) - \alpha_1)(t - T_1)} \text{ and } I_2(t) \ge I_2(T_1)e^{(f_2(S^0 - \eta_2, \eta_2) + k(V_1^0 - \eta_2) - \alpha_2)(t - T_1)}, \forall t \ge T_1.$$

Note that $f_1(S^0 - \eta_2, \eta_2) - \alpha_1 > 0$, $f_2(S^0 - \eta_2, \eta_2) + k(V_1^0 - \eta_2) - \alpha_1 > 0$, $I_1(T_1) > 0$ and $I_2(T_1) > 0$, which implies that $\lim_{t \to \infty} I_1 = \lim_{t \to \infty} I_2 = \infty$, this gives a contradiction. Then E_0 is a weak repeller for X_1 .

Similarly it is shown that E_1 and E_2 are weak repeller for X_1 . Then we conclude that system (3.1) is uniformly persistent.

Further, it is proved in [14] uniform persistence implies the existence of an interior equilibrium point. Therefore, we have established the following.

Theorem 6. The model (3.1) admits a endemic equilibrium $E_3 = (S^*, V_1^*, I_1^*, I_2^*)$ if $\overline{\mathcal{R}}_2 > 1$ and $\widetilde{\mathcal{R}}_1 > 1$.

Theorem 7. If $c_1c_2 - c_3 > 0$ and $c_1c_2c_3 - c_3^2 - c_1^2c_4 > 0$, where

$$c_{1} = -C_{44} - C_{33} - C_{22} - C_{11}$$

$$c_{2} = -C_{41}C_{14} - C_{42}C_{24} + C_{44}C_{33} + C_{44}C_{22} + C_{44}C_{11} - C_{31}C_{13} + C_{33}C_{22}$$

$$+C_{33}C_{11} + C_{22}C_{11}$$

$$c_{3} = -rC_{42}C_{14} + C_{41}C_{14}C_{33} + C_{41}C_{14}C_{22} + C_{42}C_{24}C_{33} + C_{42}C_{24}C_{11} + C_{44}C_{31}C_{13}$$

$$-C_{44}C_{33}C_{22} - C_{44}C_{33}C_{11} - C_{44}C_{22}C_{11} + C_{31}C_{13}C_{22} - C_{33}C_{22}C_{11}$$

$$c_{4} = rC_{42}C_{14}C_{33} - C_{41}C_{14}C_{33}C_{22} + C_{42}C_{24}C_{31}C_{13} - C_{42}C_{24}C_{33}C_{11}$$

$$-C_{44}C_{31}C_{13}C_{22} + C_{44}C_{33}C_{22}C_{11}.$$

Then E_3 is locally asymptotically stable.

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Proof. Then Eq (3.14) at the equilibrium E_3 is

$$J_{E_3} = \begin{pmatrix} C_{11} & 0 & C_{13} & C_{14} \\ r & C_{22} & 0 & C_{24} \\ C_{31} & 0 & C_{33} & 0 \\ C_{41} & C_{42} & 0 & C_{44} \end{pmatrix}.$$

Where

$$\begin{aligned} C_{11} &= -\frac{\partial F_1(S^*, I_1^*)}{\partial S} - \frac{\partial F_2(S^*, I_2^*)}{\partial S} - \lambda < 0 \\ C_{13} &= -\frac{\partial F_1(S^*, I_1^*)}{\partial I_1} \\ C_{14} &= -\frac{\partial F_2(S^*, I_2^*)}{\partial I_2} \\ C_{22} &= -\mu - kI_2^* < 0 \\ C_{24} &= -kV_1^* < 0 \\ C_{31} &= \frac{\partial F_1(S, I_1^*)}{\partial S} > 0 \\ C_{33} &= \frac{\partial F_1(S^*, I_1^*)}{\partial I_1} - \alpha_1 = I_1^* \frac{\partial f_1(S^*, I_1^*)}{\partial I_1} + f_1(S^*, I_1^*) - \alpha_1 = I_1^* \frac{\partial f_1(S^*, I_1^*)}{\partial I_1} \le 0 \\ C_{41} &= \frac{\partial F_2(S^*, I_2^*)}{\partial S} > 0. \\ C_{42} &= kI_2^* > 0. \\ C_{44} &= \frac{\partial F_2(S^*, I_2^*)}{\partial I_2} + kV_1^* - \alpha_2 = I_2^* \frac{\partial f_2(S^*, I_2^*)}{\partial I_2} \le 0. \end{aligned}$$

The corresponding characteristic polynomial is

$$p(x) = x^4 + c_1 x^3 + c_2 x^2 + c_3 x + c_4.$$

Note that $c_1 > 0$,

$$\begin{aligned} -C_{41}C_{14} + C_{44}C_{11} &= C_{44}(C_{11} + C_{41}) - C_{41}\left(kV_1^* - \alpha_2\right) > 0\\ C_{33}C_{11} - C_{31}C_{13} &= C_{33}(C_{11} + C_{31}) - C_{31}\left(-\alpha_1\right) > 0. \end{aligned}$$

then $c_2 > 0$, If $C_{14} \ge 0$ then $c_3 > 0$ and $c_4 > 0$, while if $C_{14} < 0$ we have that

$$C_{41}C_{14}C_{33} + C_{44}C_{31}C_{13} - C_{44}C_{33}C_{11} = -C_{44}C_{33}(C_{11} + C_{41} + C_{31}) - C_{44}(\alpha_1)(C_{31}) -(kV_1^* - \alpha_2)C_{33}(-C_{41}) > 0 -rC_{42}C_{14} - C_{44}C_{22}C_{11} + C_{41}C_{14}C_{22} = -C_{44}C_{22}(C_{11} + C_{41} + C_{31} + r) - (C_{14})(\mu)(r) +(kV_1^* - \alpha_2)C_{22}(C_{41} + r) + C_{44}C_{22}C_{31} > 0.$$

and let

$$\star = rC_{42}C_{14}C_{33} + C_{44}C_{33}C_{22}C_{11} - C_{41}C_{14}C_{33}C_{22} - C_{44}C_{31}C_{13}C_{22}$$

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then

$$\star = -C_{33} \left(-C_{44}C_{22}(C_{11} + C_{41} + C_{31} + r) - (C_{14})(\mu)(r) + (kV_1^* - \alpha_2)C_{22}(C_{41} + r) \right. \\ \left. + C_{44}C_{22}C_{31} \right) - C_{44}C_{31}C_{13}C_{22} \\ = -C_{33} \left(-C_{44}C_{22}(C_{11} + C_{41} + C_{31} + r) - (C_{14})(\mu)(r) + (kV_1^* - \alpha_2)C_{22}(C_{41} + r) \right) \\ \left. + C_{44}C_{22}C_{31}(\alpha_1) > 0. \right.$$

Then $c_3 > 0$ and $c_4 > 0$. If $c_1c_2 - c_3 > 0$ and $c_1c_2c_3 - c_3^2 - c_1^2c_4 > 0$ by Routh–Hurwitz criterion, we see that all roots of $x^4 + c_1x^3 + c_2x^2 + c_3x + c_4$ have negative real parts, then E_3 is locally asymptotically stable.

3.4.1. Global stability of equilibria

In this section, we study the global properties of the equilibria. We use Lyapunov function to show the global stabilities. Such Lyapunov functions all take advantage of the properties of the function.

$$g(x) = x - 1 - \ln(x).$$

which is positive in \mathbb{R}_+ except at x = 1, where it vanishes.

Theorem 8. The DFE E_0 is globally asymptotically stable if,

$$\mathcal{R}_0 < 1.$$

Proof. Consider the Lyapunov function

$$V(S, V_1, I_1, I_2) = I_1 + I_2,$$

Since $I_1, I_2 > 0$, then $V(S, V_1, I_1, I_2) \ge 0$ and $V(S, V_1, I_1, I_2)$ attains zero at $I_1 = I_2 = 0$. Now, we need to show $\dot{V} \le 0$.

$$\begin{split} \dot{V} &= \dot{I}_1 + \dot{I}_2 \\ &= F_1(S, I_1) - \alpha_1 I_1 + F_2(S, I_2) + k I_2 V_1 - \alpha_2 I_2. \\ &= I_1(f_1(S, I_1) - \alpha_1) + I_2(f_2(S, I_2) + k V_1 - \alpha_2). \end{split}$$

For $S \leq S^0$ and $V_1 \leq V_1^0$

$$\begin{split} \dot{V} &\leq I_1(f_1(S^0, 0) - \alpha_1) + I_2(f_2(S^0, 0) + kV_1^0 - \alpha_2). \\ &= I_1\left(\frac{\partial F_1(S^0, 0)}{\partial I_1} - \alpha_1\right) + I_2\left(\frac{\partial F_2(S^0, 0)}{\partial I_2} + kV_1^0 - \alpha_2\right) \\ &= \alpha_1 I_1(\mathcal{R}_1 - 1) + \alpha_2 I_2(\mathcal{R}_2 - 1) \leq 0. \end{split}$$

Furthermore, $\frac{dV}{dt} = 0$ if and only if $I_1 = I_2 = 0$, so the largest invariant set contained in $\{(S, V_1, I_1, I_2) \in \Omega_1 | \frac{dV}{dt} = 0\}$ is the hyperplane $I_1 = I_2 = 0$, By LaSalle's invariant principle, this implies that all solution in Ω_1 approach the hyperplane $I_1 = I_2 = 0$ as $t \to \infty$. Also, All solution of (3.1) contained in such plane satisfy $\dot{S} = \Lambda - \lambda S$, $\dot{V}_1 = rS - \mu V_1$, which implies that $S \to \frac{\Lambda}{\lambda}$ and

 $V_1 \rightarrow \frac{r\Lambda}{\mu\lambda}$ as $t \rightarrow \infty$, that is, all of these solution approach E_0 . Therefore we conclude that E_0 is globally asymptotically stable in Ω_1 .

Now we will show that every solution $(S(t), V_1(t), I_1(t), I_2(t)) \in \mathbb{R}^4_+$, where $t \to \infty$ $(S(t), V_1(t), I_1(t), I_2(t)) \in \Omega_1$, let $(S(t), V_1(t), I_1(t), I_2(t)) \in \mathbb{R}^4_+$. Then

$$\dot{S} \leq \Lambda - \lambda S$$

By the comparison principle $\lim_{t\to\infty} \sup S(t) \le \frac{\Lambda}{\lambda} = S^0$. Then $S(t) \le S^0$ for t sufficiently large. Also if $S(t) \le S^0$.

$$\dot{V}_1 \leq rS^0 - (\mu + kI_2)V_1 \leq rS^0 - \mu V_1$$

By the comparison principle $\lim_{t\to\infty} \sup V(t) \le \frac{rS^0}{\mu} = V_1^0$. Therefore E_0 is globally asymptotically stable.

From now on, we assume that

H4) For i = 1, 2. $f_i(S, I_i) = Sg_i(S, I_i)$.

Lemma 5. Let a > 0 be a constant, for i = 1, 2 if $\frac{\partial F_i(S,I_i)}{\partial I_i} \ge 0$ for all I_i , then

$$\left(\frac{I_i}{a} - \frac{F_i(S, I_i)}{F_i(S, a)}\right) \left(\frac{F_i(S, a)}{F_i(S, I_i)} - 1\right) \le 0$$

Proof. Note that

$$\left(\frac{I_i}{a} - \frac{F_i(S, I_i)}{F_i(S, a)}\right) \left(\frac{F_i(S, a)}{F_i(S, I_i)} - 1\right) = \frac{I_i}{a} \left(1 - \frac{f_i(S, I_i)}{f_i(S, a)}\right) \left(\frac{F_i(S, a)}{F_i(S, I_i)} - 1\right)$$

If $a \ge I_i$, then

$$\frac{f_i(S, I_i)}{f_i(S, a)} \ge 1 \text{ and } \frac{F_i(S, a)}{F_i(S, I_i)} \ge 1.$$

If $a \leq I_i$, then

$$\frac{f_i(S, I_i)}{f_i(S, a)} \le 1 \text{ and } \frac{F_i(S, a)}{F_i(S, I_i)} \le 1.$$

Therefore

$$\left(\frac{I_i}{a} - \frac{F_i(S, I_i)}{F_i(S, a)}\right) \left(\frac{F_i(S, a)}{F_i(S, I_i)} - 1\right) \le 0.$$

Theorem 9. Suppose that $\frac{\partial F_1(S,I_1)}{\partial I_1} \ge 0$ for all I_1 , then E_1 is globally asymptotically stable if,

$$\mathcal{R}_2 < 1.$$

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Proof. Consider the Lyapunov function

$$V(S, V_1, I_1, I_2) = I_2,$$

Since $I_2 > 0$, then $V(S, V_1, I_1, I_2) \ge 0$ and $V(S, V_1, I_1, I_2)$ attains zero at $I_2 = 0$. Now, we need to show $\dot{V} \le 0$.

$$\dot{V} = \dot{I}_2 = F_2(S, I_2) + kI_2V_1 - \alpha_2I_2. = I_2(f_2(S, I_2) + kV_1 - \alpha_2).$$

For $S \leq S^0$ and $V_1 \leq V_1^0$

$$\dot{V} \leq I_2(f_2(S^0, 0) + kV_1^0 - \alpha_2).$$

$$= I_2\left(\frac{\partial F_2(S^0, 0)}{\partial I_2} + kV_1^0 - \alpha_2\right)$$

$$= \alpha_2 I_2(\mathcal{R}_2 - 1) \leq 0.$$

Furthermore, $\frac{dV}{dt} = 0$ if and only if $I_2 = 0$. Suppose that $(S(t), V_1(t), I_1(t), I_2(t))$ is a solution of (3.1) contained entirely in the set $M = \{(S(t), V_1(t), I_1(t), I_2(t)) \in \Omega_1 | \dot{V} = 0\}$. Then, $\dot{I}_2 = 0$ and, from the above inequalities, we have $I_2 = 0$. Thus, the largest positively invariant set contained in M is the plane $I_2 = 0$. By LaSalle's invariance principle, this implies that all solutions in approach the plane $I_2 = 0$ as $t \to \infty$. On the other hand, solutions of (3.1) contained in such plane satisfy

$$\dot{S} = \Lambda - F_1(S, I_1) - \lambda S$$

 $\dot{V}_1 = rS - (\mu)V_1$
 $\dot{I}_1 = F_1(S, I_1) - \alpha_1 I_1.$

Now we will show that $S(t) \to \overline{S}$, $V_1(t) \to \overline{V_1}$ and $I_1(t) \to \overline{I_1}$ Consider the Lyapunov function

$$V(S, V_1, I_1) = \int_{\bar{S}}^{S} \left(1 - \frac{F_1(\bar{S}, \bar{I}_1)}{F_1(\chi, \bar{I}_1)} \right) d\chi + \bar{I}_1 g\left(\frac{I_1}{\bar{I}_1}\right).$$

Note that $1 - \frac{F_1(\bar{S},\bar{I}_1)}{F_1(\chi,\bar{I}_1)} = \frac{I_1(f_1(\chi,\bar{I}_1) - f_1(\bar{S},\bar{I}_1))}{F_1(\chi,\bar{I}_1)}$, by H2) $f_1(S,\bar{I}_1) - f_1(\bar{S},\bar{I}_1) \ge 0$ if $S \ge \bar{S}$ and $f_1(S,\bar{I}_1) - f_1(\bar{S},\bar{I}_1) \le 0$ if $S \le \bar{S}$, then $\int_{\bar{S}}^{S} \left(1 - \frac{F_1(\bar{S},\bar{I}_1)}{F_1(\chi,\bar{I}_1)}\right) d\chi \ge 0$ for all S. Therefore, $V(S, V_1, I_1) \ge 0$ and $V(S, V_1, I_1)$ attains zero at $S(t) = \bar{S}$, and $I_1(t) = \bar{I}_1$.

Now, we need to show $\dot{V} \leq 0$.

$$\begin{split} \dot{V} &= \left(1 - \frac{F_1(\bar{S}, \bar{I}_1)}{F_1(S, \bar{I}_1)}\right) \dot{S} + \left(1 - \frac{\bar{I}_1}{I_1}\right) \dot{I}_1 \\ &= \left(1 - \frac{F_1(\bar{S}, \bar{I}_1)}{F_1(S, \bar{I}_1)}\right) (\Lambda - F_1(S, I_1) - \lambda S) + \left(1 - \frac{\bar{I}_1}{I_1}\right) (F_1(S, I_1) - \alpha_1 I_1) \\ &= \left(1 - \frac{F_1(\bar{S}, \bar{I}_1)}{F_1(S, \bar{I}_1)}\right) \left(\lambda \bar{S} + F_1(\bar{S}, \bar{I}_1) - F_1(S, I_1) - \lambda S\right) \end{split}$$

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$$\begin{split} &+F_1(S,I_1) - \alpha_1 I_1 - \bar{I}_1 f_1(S,I_1) + \alpha_1 \bar{I}_1 \\ &= \lambda(\bar{S} - S) \left(1 - \frac{F_1(\bar{S},\bar{I}_1)}{F_1(S,\bar{I}_1)} \right) + \left(1 - \frac{F_1(\bar{S},\bar{I}_1)}{F_1(S,\bar{I}_1)} \right) F_1(\bar{S},\bar{I}_1) - F_1(S,I_1) \\ &+ \frac{F_1(\bar{S},\bar{I}_1)}{F_1(S,\bar{I}_1)} F_1(S,I_1) + F_1(S,I_1) - \frac{I_1F_1(\bar{S},\bar{I}_1)}{\bar{I}_1} - \bar{I}_1 f_1(S,I_1) + F_1(\bar{S},\bar{I}_1) \\ &= \left(2 - \frac{F_1(\bar{S},\bar{I}_1)}{F_1(S,\bar{I}_1)} + \frac{F_1(S,I_1)}{F_1(S,\bar{I}_1)} - \frac{I_1}{\bar{I}_1} - \frac{\bar{I}_1 f_1(S,I_1)}{F_1(\bar{S},\bar{I}_1)} \right) F_1(\bar{S},\bar{I}_1) \\ &+ \lambda(\bar{S} - S) \left(1 - \frac{F_1(\bar{S},\bar{I}_1)}{F_1(S,\bar{I}_1)} \right). \end{split}$$

Note that

$$\lambda(\bar{S} - S) \left(1 - \frac{F_1(\bar{S}, \bar{I}_1)}{F_1(S, \bar{I}_1)} \right) = \lambda(\bar{S} - S) \left(1 - \frac{f_1(\bar{S}, \bar{I}_1)}{f_1(S, \bar{I}_1)} \right) \le 0.$$

and

$$2 - \frac{F_{1}(\bar{S},\bar{I}_{1})}{F_{1}(S,\bar{I}_{1})} + \frac{F_{1}(S,I_{1})}{F_{1}(S,\bar{I}_{1})} - \frac{I_{1}}{\bar{I}_{1}} - \frac{\bar{I}_{1}f_{1}(S,I_{1})}{F_{1}(\bar{S},\bar{I}_{1})} = 2 - \frac{F_{1}(\bar{S},\bar{I}_{1})}{F_{1}(S,\bar{I}_{1})} + \frac{F_{1}(S,I_{1})}{F_{1}(S,\bar{I}_{1})} - \frac{I_{1}}{\bar{I}_{1}} - \frac{I_{1}}{\bar{I}_{1}} - \frac{\bar{I}_{1}F_{1}(S,I_{1})}{I_{1}F_{1}(\bar{S},\bar{I}_{1})} + 1 - \frac{F_{1}(S,I_{1})F_{1}(S,\bar{I}_{1})}{F_{1}(S,I_{1})F_{1}(S,\bar{I}_{1})} + \frac{IF_{1}(S,I_{1})}{\bar{I}_{1}F_{1}(S,I_{1})} - \frac{IF_{1}(S,I_{1})F_{1}(S,\bar{I}_{1})}{\bar{I}_{1}F_{1}(S,I_{1})} - \frac{IF_{1}(S,I_{1})}{\bar{I}_{1}F_{1}(S,I_{1})} + \frac{IF_{1}(S,I_{1})}{F_{1}(S,\bar{I}_{1})} - \frac{IF_{1}(S,I_{1})}{\bar{I}_{1}F_{1}(S,I_{1})} - \frac{IF_{1}(S,I_{1})}{\bar{I}_{1}F_{1}(S,I_{1})} + \frac{(I_{1}}{\bar{I}_{1}} - \frac{F_{1}(S,I_{1})}{F_{1}(S,\bar{I}_{1})} - \frac{IF_{1}(S,I_{1})}{\bar{I}_{1}F_{1}(S,I_{1})} - \frac{IF_{1}(S,I_{1})}{\bar{I}$$

The last inequality is due to the Lemma 5 and the relation of the geometric and arithmetic means, then $\dot{V} \leq 0$. Furthermore, $\frac{dV}{dt} = 0$ if and only if $S = \bar{S}$ and $I_1 = \bar{I}_1$, which implies that $S \to \bar{S}$, $I_1 \to \bar{I}_1$ and $I_2 \to 0$ as $t \to \infty$. By LaSalle's invariant principle, this implies that all solutions in Ω_1 approach the plane $S = \bar{S}$, $I_1 = \bar{I}_1$ and $I_2 = 0$ as $t \to \infty$. Also, All solutions of (3.1) contained in such plane satisfy $\dot{V}_1 = r\bar{S} - \mu V_1$, which implies that $V_1 \to \frac{r\bar{S}}{\mu} = \bar{V}_1$ as $t \to \infty$, that is, all of these solution approach E_1 . Therefore we conclude that E_1 is globally asymptotically stable in Ω_1 .

Corollary 1. If $2 - \frac{F_1(\bar{S},\bar{I_1})}{F_1(\bar{S},\bar{I_1})} + \frac{F_1(\bar{S},I_1)}{F_1(\bar{S},\bar{I_1})} - \frac{I_1}{\bar{I_1}} - \frac{\bar{I_1}f_1(\bar{S},I_1)}{F_1(\bar{S},\bar{I_1})} \le 0$ and $\mathcal{R}_2 < 1$ then E_1 is globally asymptotically stable.

Theorem 10. Suppose that $\frac{\partial F_2(S,I_2)}{\partial I_2} \ge 0$ for all I_2 , then E_2 is globally asymptotically stable if,

$$\mathcal{R}_1 < 1 \text{ and } 2 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)} + \frac{SF_2(\tilde{S}, \tilde{I}_2)}{\tilde{S}F_2(S, \tilde{I}_2)} - \frac{V_1}{\tilde{V}_1} - \frac{S\tilde{V}_1}{\tilde{S}V_1} \le 0.$$

Proof. Consider the Lyapunov function

$$V(S, V_1, I_1, I_2) = I_1.$$

Since $I_1 > 0$, then $V(S, V_1, I_1, I_2) \ge 0$ and $V(S, V_1, I_1, I_2)$ attains zero at $I_1 = 0$. Now, we need to show $\dot{V} \le 0$.

$$\dot{V} = \dot{I}_1$$

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$$= F_1(S, I_1) - \alpha_1 I_1 = I_1(f_1(S, I_1) - \alpha_1)$$

For $S \leq S^0$

$$\dot{V} \leq I_1(f_1(S^0, 0) - \alpha_1) \\ = I_1\left(\frac{\partial F_1(S^0, 0)}{\partial I_1} - \alpha_1\right) = \alpha_1 I_1(\mathcal{R}_1 - 1) \leq 0.$$

Furthermore, $\frac{dV}{dt} = 0$ if and only if $I_1 = 0$. Suppose that $(S(t), V_1(t), I_1(t), I_2(t))$ is a solution of (3.1) contained entirely in the set $M = \{(S(t), V_1(t), I_1(t), I_2(t)) \in \Omega_1 | \dot{V} = 0\}$. Then, $\dot{I_1} = 0$ and, from the above inequalities, we have $I_1 = 0$. Thus, the largest positively invariant set contained in M is the plane $I_1 = 0$. By LaSalle's invariance principle, this implies that all solutions in approach the plane $I_1 = 0$ as $t \to \infty$. On the other hand, solutions of (3.1) contained in such plane satisfy.

$$\dot{S} = \Lambda - F_2(S, I_2) - \lambda S$$

$$\dot{V}_1 = rS - (\mu + kI_2)V_1$$

$$\dot{I}_2 = F_2(S, I_2) + kV_1I_2 - \alpha_2I_2$$

Now we will show that $S(t) \to \tilde{S}$, $V_1(t) \to \tilde{V}_1$ and $I_1(t) \to \tilde{I}_1$ Consider the Lyapunov function

$$V(S, V_1, I_2) = \int_{\tilde{S}}^{S} \left(1 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(\chi, \tilde{I}_2)} \right) \mathrm{d}\chi + \tilde{V}_1 g\left(\frac{V_1}{\tilde{V}_1}\right) + \tilde{I}_2 g\left(\frac{I_2}{\tilde{I}_2}\right).$$

Now, we need to show $\dot{V} \leq 0$.

$$\begin{split} \dot{V} &= \left(1 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)}\right) \dot{S} + \left(1 - \frac{\tilde{V}_1}{V_1}\right) \dot{V}_1 + \left(1 - \frac{\tilde{I}_2}{I_2}\right) \dot{I}_2 \\ &= \left(1 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)}\right) (\Lambda - F_2(S, I_2) - \lambda S) + \left(1 - \frac{\tilde{V}_1}{V_1}\right) (rS - (\mu + kI_2)V_1) \\ &+ \left(1 - \frac{\tilde{I}_2}{I_2}\right) (F_2(S, I_2) + kI_2V_1 - \alpha_2I_2) \\ &= \left(1 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)}\right) \left(\lambda \bar{S} + F_2(\tilde{S}, \tilde{I}_2) - F_2(S, I_2) - \lambda S\right) + rS - (\mu + kI_2)V_1 \\ &- r\frac{S\tilde{V}_1}{V_1} + (\mu + kI_2)\tilde{V}_1 + F_2(S, I_2) + kI_2V_1 - \alpha_2I_2 - \tilde{I}_2f_2(S, I_2) - k\tilde{I}_2V_1 + \alpha_2\tilde{I}_2 \\ &= \mu(\tilde{S} - S) \left(1 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)}\right) + r\left(\tilde{S} - \tilde{S}\frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)} - S + S\frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)}\right) \\ &+ \left(1 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)}\right) F_2(\tilde{S}, \tilde{I}_2) + \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)} F_2(S, I_2) + rS \\ &- \frac{r\tilde{S}}{\tilde{V}_1}V_1 - r\frac{S\tilde{V}_1}{V_1} + r\tilde{S} - \frac{I_2F_2(\tilde{S}, \tilde{I}_2)}{\tilde{I}_2} - \tilde{I}_2f_2(S, I_2) + F_2(\tilde{S}, \tilde{I}_2) \\ &= \mu(\tilde{S} - S) \left(1 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)}\right) + r\tilde{S} \left(2 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)} + \frac{SF_2(\tilde{S}, \tilde{I}_2)}{\tilde{S}F_2(S, \tilde{I}_2)} - \frac{V_1}{\tilde{V}_1} - \frac{S\tilde{V}_1}{\tilde{S}V_1}\right) \end{split}$$

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$$+\left(2-\frac{F_2(\tilde{S},\tilde{I}_2)}{F_2(S,\tilde{I}_2)}+\frac{F_2(S,I_2)}{F_2(S,\tilde{I}_2)}-\frac{I_2}{\tilde{I}_2}-\frac{\tilde{I}_2f_2(S,I_2)}{F_2(\tilde{S},\tilde{I}_2)}\right)F_2(\tilde{S},\tilde{I}_2).$$

Note that

$$\mu(\tilde{S}-S)\left(1-\frac{F_2(\tilde{S},\tilde{I}_2)}{F_2(S,\tilde{I}_2)}\right) \le 0.$$

$$2 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)} + \frac{F_2(S, I_2)}{F_2(S, \tilde{I}_2)} - \frac{I_2}{\tilde{I}_2} - \frac{\tilde{I}_2 f_2(S, I_2)}{F_2(\tilde{S}, \tilde{I}_2)} \le 0$$

The last inequality is due to the Lemma 5 and the relation of the geometric and arithmetic means, then $\dot{V} \leq 0$. Furthermore, $\dot{V} = 0$ if and only if $S = \tilde{S}$, $I_2 = \tilde{I}_2$ and $V_1 = \tilde{V}_1$. Therefore E_2 is globally asymptotically stable.

Remark 6. Note that if $\frac{\partial g_2(S,I_2)}{\partial S} \ge 0$, then

$$2 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)} + \frac{SF_2(\tilde{S}, \tilde{I}_2)}{\tilde{S}F_2(S, \tilde{I}_2)} - \frac{V_1}{\tilde{V}_1} - \frac{S\tilde{V}_1}{\tilde{S}V_1} = 3 - \frac{V_1}{\tilde{V}_1} - \frac{S\tilde{V}_1}{\tilde{S}V_1} - \frac{\tilde{S}}{\tilde{S}} + \left(-1 + \frac{\tilde{S}}{\tilde{S}}\right) \left(1 - \frac{g_2(\tilde{S}, \tilde{I}_2)}{g_2(S, \tilde{I}_2)}\right) \le 0.$$

Corollary 2. If $2 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(\tilde{S}, \tilde{I}_2)} + \frac{F_2(S, I_2)}{F_2(\tilde{S}, \tilde{I}_2)} - \frac{I_2}{\tilde{I}_2} - \frac{\tilde{I}_2 f_2(S, I_2)}{F_2(\tilde{S}, \tilde{I}_2)} \le 0$, $\mathcal{R}_1 < 1$ and $2 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)} + \frac{SF_2(\tilde{S}, \tilde{I}_2)}{\tilde{S}F_2(S, \tilde{I}_2)} - \frac{V_1}{\tilde{V}_1} - \frac{S\tilde{V}_1}{\tilde{S}V_1} \le 0$ then E_2 is globally asymptotically stable.

Theorem 11. E₃ is globally asymptotically stable if

$$F_{1}(S^{*}, I_{1}^{*})\left(2 - \frac{S^{*}}{S} - \frac{Sg_{1}(S, I_{1})}{S^{*}g_{1}(S^{*}, I_{1}^{*})}\right) + F_{2}(S^{*}, I_{2}^{*})\left(2 - \frac{S^{*}}{S} - \frac{Sg_{2}(S, I_{2})}{S^{*}g_{2}(S^{*}, I_{2}^{*})}\right) + rS^{*}\left(3 - \frac{S^{*}}{S} - \frac{V_{1}}{V_{1}^{*}} - \frac{SV_{1}^{*}}{S^{*}V_{1}}\right) + \mu S^{*}\left(2 - \frac{S^{*}}{S} - \frac{S}{S^{*}}\right) + I_{1}\left(S^{*}g_{1}(S, I_{1}) - \alpha_{1}\right) + I_{2}\left(S^{*}g_{2}(S, I_{2}) + kV_{1}^{*} - \alpha_{2}\right) < 0.$$

Proof. Assume E_3 exists. Consider the Lyapunov function

$$V(S, V_1, I_1, I_2) = S^* g\left(\frac{S}{S^*}\right) + V_1^* g\left(\frac{V_1}{V_1^*}\right) + I_1^* g\left(\frac{I_1}{I_1^*}\right) + I_2^* g\left(\frac{I_2}{I_2^*}\right).$$

Where g(x) = x - 1 - ln(x). Then $V(S, V_1, I_1, I_2) \ge 0$ and $V(S, V_1, I_1, I_2)$ attains zero at E_3 . Now, we need to show $\dot{V} \le 0$.

$$\begin{split} \dot{V} &= \left(1 - \frac{S^*}{S}\right) \dot{S} + \left(1 - \frac{V_1^*}{V_1}\right) \dot{V}_1 + \left(1 - \frac{I_1^*}{I_1}\right) \dot{I}_1 + \left(1 - \frac{I_2^*}{I_2}\right) \dot{I}_2 \\ &= \left(1 - \frac{S^*}{S}\right) (\Lambda - F_1(S, I_1) - F_2(S, I_2) - \lambda S) + \left(1 - \frac{V_1^*}{V_1}\right) (rS - (\mu + kI_2)V_1) \\ &+ \left(1 - \frac{I_1^*}{I_1}\right) (F_1(S, I_1) - \alpha_1 I_1) + \left(1 - \frac{I_2^*}{I_2}\right) (F_2(S, I_2) + kI_2V_1 - \alpha_2 I_2) \\ &= \Lambda - F_1(S, I_1) - F_2(S, I_1) - \lambda S - \Lambda \frac{S^*}{S} + I_1 S^* g_1(S, I_1) + I_2 S^* g_2(S, I_2) + \lambda S \end{split}$$

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$$\begin{split} &+rS - \mu V_1 - kI_2 V_1 - rS \frac{V_1^*}{V_1} + \mu V_1^* + kI_2 V_1^* + F_1(S, I_1) - \alpha_1 I_1 - I_1^* f_1(S, I_1) \\ &+ \alpha_1 I_1^* + F_2(S, I_2) + kI_2 V_1 - \alpha_2 I_2 - I_2^* f_2(S, I_2) - kI_2^* V_1 + \alpha_2 I_2^* \\ &= (F_1(S^*, I_1^*) + F_2(S^*, I_2^*) + \lambda S^*) - \lambda S - (F_1(S^*, I_1^*) + F_2(S^*, I_2^*) + \lambda S^*) \frac{S^*}{S} \\ &+ I_1 S^* g_1(S, I_1) + I_2 S^* g_2(S, I_2) + \lambda S^* + rS - \mu V_1 - rS \frac{V_1^*}{V_1} + \mu V_1^* + kI_2 V_1^* \\ &- \alpha_1 I_1 - I_1^* f_1(S, I_1) + F_1(S^*, I_1^*) - \alpha_2 I_2 - I_2^* f_2(S, I_2) - kI_2^* V_1 + F_2(S^*, I_2^*) \\ &+ kI_2^* V_1^* \\ &= \left(2F_1(S^*, I_1^*) - F_1(S^*, I_1^*) \frac{S^*}{S} - I_1^* f_1(S, I_1)\right) + \left(2F_2(S^*, I_2^*) - F_2(S^*, I_2^*) \frac{S^*}{S}\right) \\ &- I_2^* f_2(S, I_2) + \left(2\lambda S^* - \lambda S^* \frac{S^*}{S} - \lambda S + rS - rS \frac{V_1^*}{V_1} + rS^* - rS^* \frac{V_1}{V_1^*}\right) \\ &+ (I_1 S^* g_1(S, I_1) - \alpha_1 I_1) + (I_2 S^* g_2(S, I_2) + kI_2 V_1^* - \alpha_2 I_2) \\ &= F_1(S^*, I_1^*) \left(2 - \frac{S^*}{S} - \frac{S g_1(S, I_1)}{S^* g_1(S^*, I_1^*)}\right) + F_2(S^*, I_2^*) \left(2 - \frac{S^*}{S} - \frac{S g_2(S, I_2)}{S^* g_2(S^*, I_2^*)}\right) \\ &+ rS^* \left(3 - \frac{S^*}{S} - \frac{V_1}{V_1^*} - \frac{S V_1^*}{S^*V_1}\right) + \mu S^* \left(2 - \frac{S^*}{S} - \frac{S}{S^*}\right) \\ &+ I_1 (S^* g_1(S, I_1) - \alpha_1) + I_2 (S^* g_2(S, I_2) + kV_1^* - \alpha_2). \end{split}$$

By the relation of geometric and arithmetic means, we conclude $\dot{V} \leq 0$, with equality holding only at the equilibrium E_3 . Therefore E_3 is globally asymptotically stable.

3.5. Numerical simulations

In this section, we present some numerical simulations of the solutions for system (3.1) to verify the results obtained in section 3.3 and give examples to illustrate theorems in section 3.4. In system (3.1), we set:

$$F_1(S, I_1) = \frac{\beta_1 S I_1}{1 + \zeta_1 I_1^2}, F_2(S, I_2) = \frac{\beta_2 S I_2}{1 + \zeta_2 S}, \Lambda = 200, \gamma_1 = 0.07, \gamma_2 = 0.09, \mu = 0.02, \nu_1 = 0.1, \nu_2 = 0.1$$

and $k = 0.00002$.

In this case

$$g_1(S, I_1) = \frac{\beta_1}{1 + \zeta_1 I_1^2}, g_2(S, I_1) = \frac{\beta_2}{1 + \zeta_2 S}, \mathcal{R}_1 = \frac{\beta_1 \Lambda}{\alpha_1 \lambda} \text{ and } \mathcal{R}_2 = \frac{\beta_2 \Lambda}{\alpha_2 (\lambda + \zeta \Lambda)} + \frac{kr \Lambda}{\alpha_2 \mu \lambda}.$$

Parameters and units are arbitrary and have been used for illustration purposes only. Anyway, when considering a realistic scenario such values could be derived from statistical data.

• Example 6.1. In system (3.1), we set $\beta_1 = 0.00003$, r = 0.1, $\beta_2 = 0.0002$, $\zeta_1 = 0.7$ and $\zeta_2 = 0.9$. Then $S^0 \approx 1667$, $V^0 \approx 8333$, $\mathcal{R}_1 \approx 0.2632$, $\mathcal{R}_2 \approx 0.7947$. By Theorem 8, we see that the disease-free equilibrium E_0 is globally asymptotically stable. Numerical simulation illustrates our result (see Figure 1).



Figure 1. Numerical simulation of (3.1) indicates that E_0 is globally asymptotically stable.



Figure 2. Numerical simulation of (3.1) indicates that E_1 is globally asymptotically stable.

- Example 6.2. In system (3.1), we set $\beta_1 = 0.0002$, r = 0.1, $\beta_2 = 0.0002$, $\zeta_1 = 0$ and $\zeta_2 = 0.9$. Then $\overline{S} \approx 950$, $\overline{V_1} \approx 4750$, $\overline{I_1} \approx 453$, $\mathcal{R}_1 \approx 1.7544$, $\mathcal{R}_2 \approx 0.7947$. By Theorem 9, we see that the E_1 is globally asymptotically stable. Numerical simulation illustrates our result (see Figure 2).
- Example 6.3. In system (3.1), we set $\beta_1 = 0.00003$, r = 0.1, $\beta_2 = 0.0002$, $\zeta_1 = 0.7$ and $\zeta_2 = 0.001$. Then $\tilde{S} \approx 1317$, $\tilde{V}_1 \approx 4814$, $\tilde{I}_2 \approx 368$, $\mathcal{R}_1 \approx 0.2632$, $\mathcal{R}_2 \approx 1.3889$ and $2 \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)} + \frac{SF_2(\tilde{S}, \tilde{I}_2)}{\tilde{S}F_2(S, \tilde{I}_2)} \frac{V_1}{\tilde{V}_1} \frac{S\tilde{V}_1}{\tilde{S}V_1} \leq 0$ (see Figure 3). By corollary 2, we see that the E_2 is globally asymptotically stable. Numerical simulation illustrates our result (see Figure 4).
- **Example 6.4.** In system (3.1), we set $\beta_1 = 0.0002$, r = 0.01, $\beta_2 = 0.0002$, $\zeta_1 = 0.0001$ and $\zeta_2 = 0.0001$. Then $\mathcal{R}_1 \approx 7.0175$, $\mathcal{R}_2 \approx 4.1270$, $\tilde{S} \approx 1134$, $\bar{S} \approx 5310$, $\bar{V}_1 \approx 2655$, $\bar{R}_2 \approx 3.555$ and $\tilde{R}_1 \approx 1.194$. Then by Theorem 6, $E_3 = (S^*, V_1^*, I_1^*, I_2^*)$ exists ($S^* \approx 1133$, $V_1^* \approx 320$, $I_1^* \approx 44$, $I_2^* \approx 774$), Also $c_1 \approx 0.2501 c_2 \approx 0.0171 c_3 \approx 3.4759 \times 10^{-04} c_4 \approx 3.4759 \times 3.924210^{-06}$, $c_1c_2-c_3^2 \approx 0.0043$ and $c_1c_2c_3 c_3^2 c_1^2c_4 \approx 1.1218e \times 10^{-06}$ by Theorem 7, E_3 is locally asymptotically stable. Also E_3 satisfies $F_1(S^*, I_1^*) \left(2 \frac{S^*}{S} \frac{Sg_1(S, I_1)}{S^*g_1(S^*, I_1^*)}\right) + F_2(S^*, I_2^*) \left(2 \frac{S^*}{S} \frac{Sg_2(S, I_2)}{S^*g_2(S^*, I_2^*)}\right) + \mu S^* \left(2 \frac{S^*}{S} \frac{S}{S^*}\right) + I_1(S^*g_1(S, I_1) \alpha_1) + I_2(S^*g_2(S, I_2) + kV_1^* \alpha_2) < 0$. By Theorem 11, we see that the E_3 is globally asymptotically stable. Numerical simulation illustrates our result (see Figure 5).



Figure 3. Graph of $2 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)} + \frac{SF_2(\tilde{S}, \tilde{I}_2)}{\tilde{S}F_2(S, \tilde{I}_2)} - \frac{V_1}{\tilde{V}_1} - \frac{S\tilde{V}_1}{\tilde{S}V_1}$.



Figure 4. Numerical simulation of (3.1) indicates that E_2 is globally asymptotically stable.



Figure 5. Numerical simulation of (3.1) indicates that E_3 is globally asymptotically stable.

4. Conclusions and discussions

In this paper, we studied a system of ordinary differential equations to model the disease dynamics of two strains of influenza with only one vaccination for strain 1 being implemented, and general incidence rate for strain 1 and strain 2. We obtained four equilibrium points:

- E_0 disease-free equilibrium, I_1 and I_2 are both zero.
- E_1 single-strain-infection equilibria, I_2 are zero.
- E_2 single-strain-infection-equilibria, I_1 are zero.
- E_3 double-strain-infection equilibrium, I_1 and I_2 are both positive.

We have investigated the topics of existence and non-existence of equilibrium points and their stabilities. We also used the next-generation matrix method to obtain two threshold quantities \mathcal{R}_1 and \mathcal{R}_2 , called the basic reproduction ratios for strain 1 and 2 respectively. It was shown that the global stability of each of the equilibrium points depends on the magnitude of these threshold quantities. More precisely, we have proved the following:

- If $\mathcal{R}_0 < 1$ the disease free equilibrium E_0 is globally asymptotically stable and if $\mathcal{R}_0 > 1$, then E_0 is unstable.
- If $\mathcal{R}_1 > 1$ the model (3.1) admits a unique single-strain-infection-equilibria E_1 . Also if $\mathcal{R}_2 < 1$ then E_1 is globally asymptotically stable and if $\overline{\mathcal{R}}_2 > 1$, then E_1 is unstable.
- If $\mathcal{R}_2 > 1$ the model (3.1) admits a unique single-strain-infection equilibria E_2 . Also if $\mathcal{R}_1 < 1$ and $2 - \frac{F_2(\tilde{S}, \tilde{I}_2)}{F_2(S, \tilde{I}_2)} + \frac{SF_2(\tilde{S}, \tilde{I}_2)}{\tilde{S}F_2(S, \tilde{I}_2)} - \frac{V_1}{\tilde{V}_1} - \frac{S\tilde{V}_1}{\tilde{S}V_1} < 0$, then E_2 is globally asymptotically stable and if $\tilde{\mathcal{R}}_1 > 1$, then E_2 is unstable.
- If $\bar{\mathcal{R}}_2 > 1$ and $\tilde{\mathcal{R}}_1 > 1$ the model (3.1) admits a double strain infection equilibrium E_3 . Also if $F_1(S^*, I_1^*) \left(2 \frac{S^*}{S} \frac{S_{g_1}(S, I_1)}{S^* g_1(S^*, I_1^*)}\right) + F_2(S^*, I_2^*) \left(2 \frac{S^*}{S} \frac{S_{g_2}(S, I_2)}{S^* g_2(S^*, I_2^*)}\right) + rS^* \left(3 \frac{S^*}{S} \frac{V_1}{V_1^*} \frac{SV_1^*}{S^* V_1}\right) + \mu S^* \left(2 \frac{S^*}{S} \frac{S}{S^*}\right) + I_1(S^* g_1(S, I_1) \alpha_1) + I_2(S^* g_2(S, I_2) + kV_1^* \alpha_2) < 0$. Then E_3 is globally asymptotically stable.

In order to discuss the meaning of our mathematical results, let us rewrite the two key indirect parameters \mathcal{R}_1 and \mathcal{R}_2 in terms of the rate of vaccination (*r*), the incidence rate of strain 1 ($F_1(S, I_1)$) and the incidence rate of strain 2 ($F_2(S, I_2)$) as shown below:

$$\mathcal{R}_1 = \frac{f_1\left(\frac{\Lambda}{r+\mu}, 0\right)}{\alpha_1}, \quad \mathcal{R}_2 = \frac{f_2\left(\frac{\Lambda}{r+\mu}, 0\right)}{\alpha_2} + \frac{kr\Lambda}{\alpha_2\mu(r+\mu)}$$

Also, the derivative of \mathcal{R}_2 with respect to *r* is,

$$\frac{\Lambda}{\alpha_2(r+\mu)^2} \left(-\frac{\partial f_2\left(\frac{\Lambda}{r+\mu},0\right)}{\partial S} + k \right)$$

Note that $\mathcal{R}_1(r)$ is decreasing and $\mathcal{R}_2(r)$ depends on $\frac{\partial f_2(\frac{\Lambda}{\mu}, 0)}{\partial S}$. Now we will analyse some cases of incidence rate.

(C1)
$$F_i(S, I) = \beta_i S I_i$$
, then $\frac{\partial f_2(\frac{\Lambda}{\mu}, 0)}{\partial S} = \beta_i$.
(C2) $F_i(S, I) = \frac{\beta_i S I_i}{1 + \zeta_i S}$, then $\frac{\partial f_2(\frac{\Lambda}{\mu}, 0)}{\partial S} = \frac{\beta_i}{1 + \zeta_i (\frac{\Lambda}{r + \mu})}$

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(C3) $F_i(S, I) = \frac{\beta_i S I_i}{1 + \zeta_i I_i^2}$, then $\frac{\partial f_2(\frac{\Lambda}{\mu}, 0)}{\partial S} = \beta_i$.

Note that for (C1) and (C3), $\mathcal{R}_2(r)$ is increasing if $\beta_i < k$, $\mathcal{R}_2(r)$ is decreasing if $\beta_i > k$ and $\mathcal{R}_2(r)$ is constant if $\beta_i = k$. For (C2), $\mathcal{R}_2(r)$ is increasing if $\beta_i \leq k$ ($\zeta \neq 0$). If $\beta_i > k \mathcal{R}_2(r)$ is increasing if $\frac{\zeta_i k \Lambda}{\beta_i - k} - \mu < r$ and decreasing if $\frac{\zeta_i k \Lambda}{\beta_i - k} - \mu > r$.

Furthermore, if the force of infection of strain 1 is (C2), then $\mathcal{R}_1 = \frac{\beta_1}{\alpha_1(1+\zeta_1 S^0)}$, note that \mathcal{R}_1 is decreasing in ζ_1 . If the force of infection of strain 2 is (C2), then $\mathcal{R}_2 = \frac{\beta_2 \Lambda}{\alpha_2(\lambda+\zeta_2 \Lambda)} + \frac{kr\Lambda}{\alpha_2\mu\lambda}$, note that \mathcal{R}_2 is decreasing in ζ_2 .

With the above information and the results in section 3.4, we conclude that the vaccination is always beneficial for controlling strain 1, its impact on strain 2 depends on the force of infection of strain 2. For example, if the force of infection of strain 2 is (C2), the impact of vaccination depends on values of β_2 , k and ζ_2 . If $\zeta_2 = 0$ and $\beta_2 > k$ it plays a positive role and if $\zeta_2 = 0$ and $\beta_2 < k$, it has a negative impact in controlling strain 2. This is reasonable because larger k (than β_2) means that vaccinated individuals are more likely to be infected by strain 2 than those who are not vaccinated, and thus, is helpful to strain 2. Smaller k (than β_2) implies the opposite. If $\zeta_2 \neq 0$ and $\beta_2 \leq k$, it plays a negative role and if $\zeta_2 \neq 0$ and $\beta_2 > k$, not necessarily has a positivity impact in controlling strain 2. This is reasonable because larger k (than β_2) means that vaccinated individuals are more likely to be infected by strain 2 than those who are not vaccinated, but if ζ_2 is large it means that the population is taking precautions to avoid the infection of strain 1 and ζ_2 (of the force of infection (C2)) is always beneficial for controlling strain 2, it means that it is very important that people are taking precautions not to get infected.

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Conflict of interest

No conflict of interest.

References

- 1. World Health Organization. Influenza (Seasonal), 6 November 2018. Available from: https://www.who.int/es/news-room/fact-sheets/detail/influenza-(seasonal).
- A. Rahman, X. Zou, Flu epidemics: a two strain flu model with a single vaccination, *J. Biol. Dyn.*, 5 (2011), 376–390.
- 3. M. Medina, E. Vintiñi, J. Villena, R. Raya, S. Alvarez, Lactococcus lactisas an adjuvant and delivery vehicle of antigens against pneumococcal respiratory infections, *Bioeng. Bugs*, **1** (2010), 313–325.
- G. Chowell, C. Ammon, N. Hengartner, J. Hyman, Transmission dynamics of the great influenza pandemic of 1918 in Geneva, Switzerland: Assessing the effects of hypothetical interventions, *J. Theor. Biol.*, **241** (2006), 193–204.

- 5. C. Mills, J. Robins, M. Lipsitch, Transmissibility of 1918 pandemic influenza, *Nature*, **432** (2004), 904–906.
- 6. S. Cauchemez, A. Valleron, P. Boëlle, A. Flahault, N. Ferguson, Estimating the impact of school closure on influenza transmission from Sentinel data, *Nature*, **452** (2008), 750–754.
- 7. V. Capasso, G. Serio, A Generalization of the Kermack-Mckendrick deterministic epidemic model, *Math. Biosci.*, **42** (1978), 43–61.
- 8. I. Baba, E. Hincal, A model for influenza with vaccination and awareness, *Chaos, Solitons, Fractals*, **106** (2018), 49–55.
- 9. I. Baba, E. Hincal, Global stability analysis of two-strain epidemic model with bilinear and non-monotone incidence rates, *Eur. Phys. J. Plus*, **132** (2017), 208.
- A. Korobeinikov, P. K. Maini, Non-linear incidence and stability of infectious disease models, *Math. Med. Biol.*, 22 (2005), 113–128.
- 11. L. Wang, X. Zhang, Z. Liu, An SEIR Epidemic Model with Relapse and General Nonlinear Incidence Rate with Application to Media Impact, *Qual. Theory Dyn. Syst.*, **17** (2017), 309–329.
- 12. P. Van den Driessche, J. Watmough, Reproduction numbers and sub-threshold endemic equilibria for compartmental models of disease transmission, *Math. Biosci.*, **180** (2002), 29–48.
- 13. H. R. Thieme, Persistence under relaxed point-dissipativity (with application to an endemic model), *SIAM J. Math. Anal.*, **24** (1993), 407–435.
- 14. G. J. Butler, H. I. Freedman, P. Waltman, Uniformly persistent systems, *Proc. Am. Math. Soc.*, **96** (1989), 425–430.



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