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# Global dynamics of humoral immunity Chikungunya virus with two routes of infection and Holling type-II



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# Abstract

In this work, we analyze the global dynamics of within-host Chikungunya virus (CHIKV) infection model with humoral immune response. We incorporate two modes of infections, attaching a CHIKV to a host monocyte, and contacting an infected monocyte with an uninfected monocyte. The infection incident rate is given by Holling type-II. The basic reproduction number  $\mathcal{R}_0$  is used to prove that the CHIKV-free equilibrium  $E_0$  is globally asymptotically stable when  $\mathcal{R}_0 \leq 1$  and the infected equilibrium  $E_1$  is globally asymptotically stable when  $\mathcal{R}_0 > 1$ . Numerical simulations have been performed to confirm the theoretical results.

**Keywords:** Chikungunya virus, holling type-II, global stability, Lyapunov function, viral and cellular infections. **2010 MSC:** 34D23, 93D20, 93D05, 93C55.

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# 1. Introduction

Mosquito is one of the dangerous insect throughout the world. It can carry and spread viruses to humans and animals causes many of deaths every year. A great efforts has been paid to develop and analyze mathematical models that describe the population dynamics of mosquito-borne diseases such as Zika [2, 4, 7], dengue [1, 27, 43, 48], malaria [3, 5, 6, 36], yellow fever [40] and chikungunya [8–10, 34, 37–39, 46]. Chikungunya virus (CHIKV) is transmitted to humans by infected Aedes albopictus and Aedes agypti mosquito. CHIKV causes severe joint and muscle pain, fever, rash, headache, nausea and fatigue. Wang and Liu [45] have proposed and studied a within-host CHIKV dynamics model which contains four compartments, uninfected-monocytes (s), infected monocytes (y), free CHIKV particles (p) and antibodies (x). The model has been extended in [13, 14] by considering general CHIKV-monocyte incidence rate. In [13, 14, 45] it has been assumed that the uninfected monocyte becomes infected by contacting with CHIKV (CHIKV-to-monocyte transmission). Long and Heise [35] have reported that the CHIKV can also spread by infected-to-monocyte transmission. Mathematical models of different viruses with both cellular and

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viral infections have been studied in several works [24, 25, 31–33, 41, 44, 47]. In a very recent work, Elaiw et al. [15] have studied the dynamics of CHIKV model with two routes of infection, however, they did not consider the holling-II. The aim of the present paper is to propose and analyze a CHIKV dynamics model where the infection rate is given by Holling type-II incidence. The proposed model is given as:

$$\dot{s}(t) = \beta - \delta s(t) - \frac{\eta_1 s(t) p(t)}{1 + \omega s(t)} - \frac{\eta_2 s(t) y(t)}{1 + \omega s(t)},$$
(1.1)

$$\dot{y}(t) = \frac{\eta_1 s(t) p(t)}{1 + \omega s(t)} + \frac{\eta_2 s(t) y(t)}{1 + \omega s(t)} - \epsilon y(t), \tag{1.2}$$

$$\dot{p}(t) = \pi y(t) - cp(t) - rx(t)p(t),$$
 (1.3)

$$\dot{\mathbf{x}}(t) = \lambda + \rho \mathbf{x}(t)\mathbf{p}(t) - \mathbf{m}\mathbf{x}(t). \tag{1.4}$$

The uninfected monocytes are generated monocytes by rate  $\beta$ , die with rate  $\delta s(t)$  and be infected by CHIKV and infected monocytes with rate  $\frac{\eta_1 s(t) p(t)}{1+\omega s(t)} + \frac{\eta_2 s(t) y(t)}{1+\omega s(t)}$ , where  $\omega$  is the uninfected monocyte Holling type-II constant, and  $\eta_1$ , and  $\eta_2$  are the incidence rate constants. Constants  $\epsilon$ , c, and m represent, respectively, the death rate constants of the infected monocytes, CHIKV, and antibodies. Constant  $\pi$  is the production rate constant of the CHIKV from infected monocytes. Antibodies attack the CHIKV at rate rx(t)p(t). Once antigen is encountered, the antibodies expand at a constant rate  $\lambda$  and proliferate at rate  $\rho x(t)p(t)$ . All the parameters of the model are positive.

#### 1.1. Basic properties

The following lemma establishes the nonnegativity and boundedness of the solutions of system (1.1)-(1.4).

**Lemma 1.1.** There exist  $M_1, M_2, M_3 > 0$ , such that the following compact set is positively invariant for system (1.1)-(1.4);

$$\Gamma = \{(s, y, p, x) \in \mathbb{R}^4_{\geq 0} : 0 \leqslant s, y \leqslant M_1, 0 \leqslant p \leqslant M_2, 0 \leqslant x \leqslant M_3\}.$$

Proof. We have

$$\begin{split} \dot{s} \mid_{s=0} &= \beta > 0, \\ \dot{y} \mid_{y=0} &= \frac{\eta_1 s p}{1 + \omega s} \ge 0 \text{ for all } s, p \ge 0, \\ \dot{p} \mid_{p=0} &= \pi y \ge 0 \text{ for all } y \ge 0, \\ \dot{x} \mid_{x=0} &= \lambda > 0. \end{split}$$

This shows that  $(s(t), y(t), p(t), x(t)) \in \mathbb{R}^4_{\geq 0}$  with  $(s(0), y(0), p(0), x(0)) \in \mathbb{R}^4_{\geq 0}$ . Let us define

$$H_1(t) = s(t) + y(t),$$
  $H_2(t) = p(t) + \frac{r}{\rho}x(t).$ 

Then from Eqs. (1.1)-(1.4) we get

$$\dot{H}_{1}(t) = \beta - \delta s(t) - \varepsilon y(t) \leqslant \beta - \sigma_{1}(s(t) + y(t)) = \beta - \sigma_{1}H_{1}(t),$$

where,  $\sigma_1 = \min\{\delta, \epsilon\}$ . Hence  $H_1(t) \leq M_1$ , if  $H_1(0) \leq M_1$ , where  $M_1 = \frac{\beta}{\sigma_1}$ . It follows that  $0 \leq s(t), y(t) \leq M_1$  if  $0 \leq s(0) + y(0) \leq M_1$ . Moreover, we have

$$\dot{H}_{2}(t) = \pi y(t) - cp(t) + \frac{r}{\rho}\lambda - \frac{mr}{\rho}x(t) \leqslant \pi M_{1} + \frac{r}{\rho}\lambda - \sigma_{2}\left(p(t) + \frac{r}{\rho}x(t)\right) = \pi M_{1} + \frac{r}{\rho}\lambda - \sigma_{2}H_{2}(t),$$

where,  $\sigma_2 = \min\{c, m\}$ . Hence  $H_2(t) \leq M_2$ , if  $H_2(0) \leq M_2$ , where  $M_2 = \frac{\pi M_1 + \frac{\tau}{\rho}\lambda}{\sigma_2}$ . Since p(t) and x(t) are all non-negative, then  $0 \leq p(t) \leq M_2$  and  $0 < x(t) \leq M_3$  if  $0 < p(0) + \frac{\tau}{\rho}x(0) \leq M_2$ , where  $M_3 = \frac{\rho M_2}{r}$ .  $\Box$ 

#### 1.2. Equilibria

We define the basic reproduction number as:

$$\Re_0 = \frac{(\eta_1 \pi \mathfrak{m} + \eta_2 \mathfrak{c} \mathfrak{m} + \eta_2 r \lambda)\beta}{\varepsilon(\mathfrak{c} \mathfrak{m} + r \lambda)(\delta + \beta \omega)}.$$

# **Lemma 1.2.** *Consider system* (1.1)-(1.4)*, then*

- if  $\mathfrak{R}_0\leqslant 1,$  then there exists only one equilibrium  $\mathsf{E}_0\in \Gamma,$  and
- *if*  $\Re_0 > 1$ , then there exist two equilibria  $E_0 \in \Gamma$  and  $E_1 \in \overset{\circ}{\Gamma}$ , where  $\overset{\circ}{\Gamma}$  is the interior of  $\Gamma$ .

*Proof.* Let E(s, y, p, x) be any equilibrium satisfying

$$0 = \beta - \delta s - \frac{\eta_1 s p}{1 + \omega s} - \frac{\eta_2 s y}{1 + \omega s'}$$
(1.5)

$$0 = \frac{\eta_1 s p}{1 + \omega s} + \frac{\eta_2 s y}{1 + \omega s} - \epsilon y, \tag{1.6}$$

$$0 = \pi y - cp - rxp, \tag{1.7}$$

$$0 = \lambda + \rho x p - m x. \tag{1.8}$$

By solving Eqs. (1.5)-(1.8) we get two equilibria a CHIKV-free equilibrium  $E_0 = (s_0, 0, 0, x_0)$ , where  $s_0 = \frac{\beta}{\delta}$  and  $x_0 = \frac{\lambda}{m}$ . Moreover, we have

$$\frac{C_1 p^3 + C_2 p^2 + C_3 p + C_4}{\bar{C}_1 p + \bar{C}_2} = 0,$$

where

$$\begin{split} C_1 &= c \varepsilon \rho^2 (-\pi \eta_1 - c \eta_2 + c \varepsilon \omega), \\ C_2 &= C_{21} + C_{22} + C_{23} + C_{24} + C_{25}, \\ C_3 &= C_{31} + C_{32} + C_{33} + C_{34} + C_{35} + C_{36}, \\ C_4 &= m \pi (m \pi \beta \eta_1 - c m (\delta \varepsilon - \beta \eta_2 + \beta \ \varepsilon \omega) - r \lambda (\delta \varepsilon - \beta \eta_2 + \beta \ \varepsilon \omega)), \\ \bar{C_1} &= \bar{C_{11}} + \bar{C_{12}}, \\ \bar{C_2} &= \bar{C_{21}} + \bar{C_{22}}, \end{split}$$

and

$$\begin{array}{ll} C_{21}=\rho\pi\eta_{1}(r\varepsilon\lambda+\pi\beta\rho), & C_{22}=2\rho c^{2}m\varepsilon(\eta_{2}-\varepsilon\omega), & C_{23}=2\rho cm\pi\varepsilon\eta_{1}, \\ C_{24}=2\rho cr\varepsilon\lambda(\eta_{2}-\varepsilon\omega), & C_{25}=-\pi\rho^{2}c(\delta\varepsilon-\beta\eta_{2}+\beta\varepsilon\omega), & C_{31}=-m\pi\eta_{1}(r\varepsilon\lambda+2\pi\beta\rho), \\ C_{32}=c^{2}m^{2}\varepsilon(-\eta_{2}+\varepsilon\omega), & C_{33}=r^{2}\lambda^{2}\varepsilon(-\eta_{2}+\varepsilon\omega), & C_{34}=r\lambda\pi\rho(\delta\varepsilon-\beta\eta_{2}+\beta\varepsilon\omega), \\ C_{35}=cm(-m\pi\varepsilon\eta_{1}+2r\varepsilon\lambda(-\eta_{2}+\varepsilon\omega)), & C_{36}=2cm\pi\rho(\delta\varepsilon-\beta\eta_{2}+\beta\varepsilon\omega), & \bar{C_{11}}=\pi\rho r\lambda(\eta_{2}-\varepsilon\omega), \\ \bar{C_{12}}=2\pi\rho m(\pi\eta_{1}+c(\eta_{2}-\varepsilon\omega)), & \bar{C_{21}}=-m\pi(m\pi\eta_{1}+cm(\eta_{2}-\varepsilon\omega)), & \bar{C_{22}}=-m\pi r\lambda(\eta_{2}-\varepsilon\omega). \end{array}$$

Let define a function X(p) as:

$$X(p) = \frac{C_1 p^3 + C_2 p^2 + C_3 p + C_4}{\bar{C}_1 p + \bar{C}_2} = 0$$

we obtain

$$X(0) = \frac{\beta(\mathsf{cm}\varepsilon + \mathsf{r}\varepsilon\lambda)(\delta + \beta\omega)(\mathfrak{R}_0 - 1)}{\varepsilon(\mathsf{cm} + \mathsf{r}\lambda)(\delta + \beta\omega)(\mathfrak{R}_0 - 1) + \varepsilon\delta(\mathsf{cm} + \mathsf{r}\lambda)}, \qquad \qquad \lim_{p \to \left(\frac{m}{\alpha}\right)^-} X(p) = -\frac{\mathsf{m}\mathsf{r}\varepsilon\lambda}{\rho^2} < 0.$$

Therefore, if  $\Re_0 > 1$  then X(0) > 0 and there exists  $p_1 \in (0, \frac{m}{\rho})$  such that  $X(p_1) = 0$ . It follows from Eqs. (1.6)-(1.8) that

$$\begin{split} x_1 &= \frac{\lambda}{m - \rho p_1} > 0, \quad y_1 = \frac{p_1(c + rx_1)}{\pi} > 0, \\ s_1 &= \frac{-\delta - p_1 \eta_1 - y_1 \eta_2 + \beta \omega + \sqrt{4\beta \delta \omega + (-\delta - p_1 \eta_1 - y_1 \eta_2 + \beta \omega)^2}}{2\delta \omega} > 0 \end{split}$$

Therefore, if  $\Re_0 > 1$ , then the system has an infected equilibrium  $E_1 = (s_1, y_1, p_1, x_1)$ . Now we show that  $E_0 \in \Gamma$  and  $E_1 \in \overset{\circ}{\Gamma}$ . Clearly,  $E_0 \in \Gamma$ . From the equilibria conditions of  $E_1$  we have

$$\beta = \delta s_1 + \frac{\eta_1 s_1 p_1}{1 + \omega s_1} + \frac{\eta_2 s_1 y_1}{1 + \omega s_1} \Rightarrow \delta s_1 + \epsilon y_1 = \beta \Rightarrow 0 < s_1 < \frac{\beta}{\delta} \leq M_1, 0 < y_1 < \frac{\beta}{\epsilon} \leq M_1.$$

Moreover, from Eqs. (1.7) and (1.8) we have

$$cp_{1} = \pi y_{1} + \frac{r}{\rho}\lambda - \frac{mr}{\rho}x_{1} \Rightarrow cp_{1} + \frac{mr}{\rho}x_{1} = \pi y_{1} + \frac{r}{\rho}\lambda < \pi M_{1} + \frac{r}{\rho}\lambda,$$

$$p_{1} < \frac{\pi M_{1} + \frac{r}{\rho}\lambda}{c} \leqslant M_{2}, \quad x_{1} < \frac{\rho}{r}\frac{\pi M_{1} + \frac{r}{\rho}\lambda}{m} \leqslant \frac{\rho M_{2}}{r} = M_{3}.$$

It follows that,  $E_1 \in \Gamma$ .

# 2. Global properties

To investigate the global stability of the equilibria we construct Lyapunov functions using the method presented [30] and followed by [11, 12, 15–24, 26, 28, 29, 42]. Define  $F(v) = v - 1 - \ln v$ .

**Theorem 2.1.** For system (1.1)-(1.4), if  $\mathcal{R}_0 \leq 1$ , then  $\mathsf{E}_0$  is globally asymptotically stable in  $\Gamma$ .

*Proof.* Let  $\Re_0 \leq 1$  and construct a Lyapunov function  $U_0(s, y, p, x)$  as:

$$U_{0}(s, y, p, x) = s - s_{0} - \int_{s_{0}}^{s} \frac{s_{0}(1 + \omega\theta)}{\theta(1 + \omega s_{0})} d\theta + y + \frac{\eta_{1}s_{0}}{(c + rx_{0})(1 + \omega s_{0})} p + \frac{r\eta_{1}s_{0}}{\rho(c + rx_{0})(1 + \omega s_{0})} x_{0}F\left(\frac{x}{x_{0}}\right).$$

Clearly,  $U_0(s, y, p, x) > 0$  for all s, y, p, x > 0 and  $U_0(s_0, 0, 0, x_0) = 0$ . Calculating  $\frac{dU_0}{dt}$  along system (1.1)-(1.4) we obtain

$$\begin{split} \frac{d U_0}{dt} &= \left(1 - \frac{s_0(1+\omega s)}{s(1+\omega s_0)}\right) \left(\beta - \delta s - \frac{\eta_1 s p}{1+\omega s} - \frac{\eta_2 s y}{1+\omega s}\right) + \frac{\eta_1 s p}{1+\omega s} + \frac{\eta_2 s y}{1+\omega s} - \varepsilon y \\ &+ \frac{\eta_1 s_0}{(c+rx_0)(1+\omega s_0)} \left(\pi y - c p - r x p\right) + \frac{r \eta_1 s_0}{\rho(c+rx_0)(1+\omega s_0)} \left(1 - \frac{x_0}{x}\right) \left(\lambda + \rho x p - m x\right) \\ &= \left(1 - \frac{s_0(1+\omega s)}{s(1+\omega s_0)}\right) \left(\beta - \delta s\right) + \frac{\eta_2 s_0 y}{1+\omega s_0} - \varepsilon y + \frac{\eta_1 s_0}{(c+rx_0)(1+\omega s_0)} \pi y \\ &+ \frac{m_1 s_0}{\rho(c+rx_0)(1+\omega s_0)} \left(1 - \frac{x_0}{x}\right) \left(\lambda - m x\right). \end{split}$$

Substituting  $\beta = \delta s_0$  and  $\lambda = m x_0$  we get

$$\begin{split} \frac{\mathrm{d}\mathbf{U}_0}{\mathrm{d}\mathbf{t}} &= -\delta \frac{(\mathbf{s} - \mathbf{s}_0)^2}{\mathbf{s}(1 + \omega \mathbf{s}_0)} + \varepsilon \left( \frac{\eta_2 \mathbf{s}_0}{\varepsilon (1 + \omega \mathbf{s}_0)} + \frac{\eta_1 \mathbf{s}_0 \pi}{\varepsilon (\mathbf{c} + \mathbf{r} \mathbf{x}_0)(1 + \omega \mathbf{s}_0)} - 1 \right) \mathbf{y} - \frac{\eta_1 \mathbf{s}_0 m}{\rho (\mathbf{c} + \mathbf{r} \mathbf{x}_0)(1 + \omega \mathbf{s}_0)} \frac{(\mathbf{x} - \mathbf{x}_0)^2}{\mathbf{x}} \\ &= -\delta \frac{(\mathbf{s} - \mathbf{s}_0)^2}{\mathbf{s}(1 + \omega \mathbf{s}_0)} - \frac{\eta_1 \mathbf{s}_0 m}{\rho (\mathbf{c} + \mathbf{r} \mathbf{x}_0)(1 + \omega \mathbf{s}_0)} \frac{(\mathbf{x} - \mathbf{x}_0)^2}{\mathbf{x}} + \varepsilon (\mathcal{R}_0 - 1) \mathbf{y}. \end{split}$$

If  $\Re_0 \leq 1$ , then  $\frac{dU_0}{dt} \leq 0$  for all s, y, p, x > 0 and  $\frac{dU_0}{dt} = 0$  when  $s = s_0, x = x_0$  and y = 0. It can be easily shown that  $\frac{dU_0}{dt} = 0$  at  $E_0$ . Applying LaSalle's invariance principle, we get  $E_0$  is globally asymptotically stable when  $\Re_0 \leq 1$ .

**Theorem 2.2.** For system (1.1)-(1.4), if  $\Re_0 > 1$ , then  $E_1$  is globally asymptotically stable in  $\overset{\circ}{\Gamma}$ .

*Proof.* Let a function  $U_1(s, y, p, x)$  be defined as:

$$U_{1}(s, y, p, x) = s - s_{1} - \int_{s_{1}}^{s} \frac{s_{1}(1 + \omega\theta)}{\theta(1 + \omega s_{1})} d\theta + y_{1}F\left(\frac{y}{y_{1}}\right) + \frac{\eta_{1}s_{1}p_{1}}{\pi y_{1}(1 + \omega s_{1})}p_{1}F\left(\frac{p}{p_{1}}\right) + \frac{r\eta_{1}s_{1}p_{1}}{\rho\pi y_{1}(1 + \omega s_{1})}x_{1}F\left(\frac{x}{x_{1}}\right).$$

Clearly,  $U_1(s, y, p, x) > 0$  for all s, y, p, x > 0 and  $U_1(s_1, y_1, p_1, x_1) = 0$ . Calculating  $\frac{dU_1}{dt}$  along the trajectories of (1.1)-(1.4) we obtain

$$\begin{split} \frac{\mathrm{d}U_1}{\mathrm{d}t} &= \left(1 - \frac{s_1(1+\omega s_1)}{s(1+\omega s_1)}\right) \left(\beta - \delta s - \frac{\eta_1 sp}{1+\omega s} - \frac{\eta_2 sy}{1+\omega s}\right) + \left(1 - \frac{y_1}{y}\right) \left(\frac{\eta_1 sp}{1+\omega s} + \frac{\eta_2 sy}{1+\omega s} - \epsilon y\right) \\ &+ \frac{\eta_1 s_1 p_1}{\pi y_1(1+\omega s_1)} \left(1 - \frac{p_1}{p}\right) \left(\pi y - cp - rxp\right) + \frac{m_1 s_1 p_1}{\rho \pi y_1(1+\omega s_1)} \left(1 - \frac{x_1}{x}\right) \left(\lambda + \rho xp - mx\right) \\ &= \left(1 - \frac{s_1(1+\omega s)}{s(1+\omega s_1)}\right) \left(\beta - \delta s\right) + \frac{\eta_1 s_1 p}{1+\omega s_1} + \frac{\eta_2 s_1 y}{1+\omega s_1} - \frac{\eta_1 sp}{1+\omega s} \frac{y_1}{y} - \frac{\eta_2 sy}{1+\omega s} \frac{y_1}{y} - \epsilon y + \epsilon y_1 \\ &+ \frac{\eta_1 s_1 p_1}{1+\omega s_1} \frac{y}{y_1} - \frac{\eta_1 s_1 p_1}{1+\omega s_1} \frac{p_1 y}{p_1} - \frac{\eta_1 s_1 p_1}{\pi y_1(1+\omega s_1)} cp + \frac{\eta_1 s_1 p_1}{\pi y_1(1+\omega s_1)} cp_1 + \frac{\eta_1 s_1 p_1}{\pi y_1(1+\omega s_1)} rxp_1 \\ &- \frac{m_1 s_1 p_1}{\pi y_1(1+\omega s_1)} x_1 p + \frac{m_1 s_1 p_1}{\rho \pi y_1(1+\omega s_1)} \left(1 - \frac{x_1}{x}\right) \left(\lambda - mx\right). \end{split}$$

Applying the equilibrium conditions for  $\mathsf{E}_1$ 

 $\beta = \delta s_1 + \frac{\eta_1 s_1 p_1}{1 + \omega s_1} + \frac{\eta_2 s_1 y_1}{1 + \omega s_1}, \quad \varepsilon y_1 = \frac{\eta_1 s_1 p_1}{1 + \omega s_1} + \frac{\eta_2 s_1 y_1}{1 + \omega s_1}, \quad c p_1 = \pi y_1 - r x_1 p_1, \quad \lambda = m x_1 - \rho x_1 p_1.$ 

we get

$$\begin{aligned} \frac{\mathrm{d} \mathrm{U}_{1}}{\mathrm{d} \mathrm{t}} &= -\delta \frac{(\mathrm{s}-\mathrm{s}_{1})^{2}}{\mathrm{s}(1+\omega\mathrm{s}_{1})} + \left(1 - \frac{\mathrm{s}_{1}(1+\omega\mathrm{s}_{1})}{\mathrm{s}(1+\omega\mathrm{s}_{1})}\right) \left(\frac{\eta_{1}\mathrm{s}_{1}\mathrm{p}_{1}}{1+\omega\mathrm{s}_{1}} + \frac{\eta_{2}\mathrm{s}_{1}\mathrm{y}_{1}}{1+\omega\mathrm{s}_{1}}\right) \\ &- \frac{\eta_{1}\mathrm{s}_{1}\mathrm{p}_{1}}{1+\omega\mathrm{s}_{1}} \frac{\mathrm{s}\mathrm{p}\mathrm{y}_{1}(1+\omega\mathrm{s}_{1})}{\mathrm{s}_{1}\mathrm{p}\mathrm{p}\mathrm{y}(1+\omega\mathrm{s})} - \frac{\eta_{2}\mathrm{s}_{1}\mathrm{y}_{1}}{1+\omega\mathrm{s}_{1}} \frac{\mathrm{s}(1+\omega\mathrm{s}_{1})}{\mathrm{s}_{1}(1+\omega\mathrm{s})} + \frac{\eta_{1}\mathrm{s}_{1}\mathrm{p}_{1}}{1+\omega\mathrm{s}_{1}} + \frac{\eta_{2}\mathrm{s}_{1}\mathrm{y}_{1}}{1+\omega\mathrm{s}_{1}} \\ &- \frac{\eta_{1}\mathrm{s}_{1}\mathrm{p}_{1}}{1+\omega\mathrm{s}_{1}} \frac{\mathrm{p}\mathrm{s}_{1}\mathrm{y}}{\mathrm{p}\mathrm{y}_{1}} + \frac{\eta_{1}\mathrm{s}_{1}\mathrm{p}_{1}}{1+\omega\mathrm{s}_{1}} - 2\frac{\eta_{1}\mathrm{s}_{1}\mathrm{p}_{1}}{\pi\mathrm{y}_{1}(1+\omega\mathrm{s}_{1})}\mathrm{r}\mathrm{x}_{1}\mathrm{p}_{1} + \frac{\eta_{1}\mathrm{s}_{1}\mathrm{p}_{1}}{\pi\mathrm{y}_{1}(1+\omega\mathrm{s}_{1})}\mathrm{r}\mathrm{x}_{p}\mathrm{p}_{1} \\ &+ \frac{\eta_{1}\mathrm{s}_{1}\mathrm{p}_{1}}{\pi\mathrm{y}_{1}(1+\omega\mathrm{s}_{1})}\mathrm{r}\mathrm{x}_{1}\mathrm{p}_{1}\frac{\mathrm{x}_{1}}{\mathrm{p}} - \frac{\eta_{1}\mathrm{s}_{1}\mathrm{p}_{1}\mathrm{m}}{\mathrm{p}\mathrm{x}_{1}(1+\omega\mathrm{s}_{1})}\frac{(\mathrm{x}-\mathrm{x}_{1})^{2}}{\mathrm{x}}. \end{aligned} \tag{2.1}$$

Eq. (2.1) can be simplified as:

$$\begin{split} \frac{\mathrm{d} \mathrm{u}_{1}}{\mathrm{d} \mathrm{t}} &= -\delta \frac{(\mathrm{s}-\mathrm{s}_{1})^{2}}{\mathrm{s}(1+\omega\mathrm{s}_{1})} + \frac{\eta_{1}\mathrm{s}_{1}\mathrm{p}_{1}}{1+\omega\mathrm{s}_{1}} \bigg[ 3 - \frac{\mathrm{s}_{1}(1+\omega\mathrm{s})}{\mathrm{s}(1+\omega\mathrm{s}_{1})} - \frac{\mathrm{s}\mathrm{p}\mathrm{y}_{1}(1+\omega\mathrm{s}_{1})}{\mathrm{s}_{1}\mathrm{p}\mathrm{p}\mathrm{y}(1+\omega\mathrm{s})} - \frac{\mathrm{p}_{1}\mathrm{y}}{\mathrm{p}\mathrm{y}_{1}} \bigg] \\ &\quad + \frac{\eta_{2}\mathrm{s}\mathrm{n}\mathrm{y}_{1}}{1+\omega\mathrm{s}_{1}} \bigg[ 2 - \frac{\mathrm{s}_{1}(1+\omega\mathrm{s})}{\mathrm{s}(1+\omega\mathrm{s}_{1})} - \frac{\mathrm{s}(1+\omega\mathrm{s})}{\mathrm{s}_{1}(1+\omega\mathrm{s})} \bigg] - \frac{\eta_{1}\mathrm{s}\mathrm{n}\mathrm{p}\mathrm{p}}{\pi\mathrm{y}_{1}(1+\omega\mathrm{s}_{1})} \mathrm{r}\mathrm{x}_{1}\mathrm{p}_{1} \bigg[ 2 - \frac{\mathrm{x}}{\mathrm{x}_{1}} - \frac{\mathrm{x}_{1}}{\mathrm{x}} \bigg] \\ &\quad - \frac{\mathrm{m}_{1}\mathrm{s}\mathrm{n}\mathrm{p}\mathrm{n}\mathrm{m}}{\mathrm{p}\mathrm{m}\mathrm{y}_{1}(1+\omega\mathrm{s}_{1})} \frac{(\mathrm{x}-\mathrm{x}_{1})^{2}}{\mathrm{x}} \\ &= -\delta \frac{(\mathrm{s}-\mathrm{s}_{1})^{2}}{\mathrm{s}(1+\omega\mathrm{s}_{1})} - \frac{\eta_{1}\mathrm{s}\mathrm{n}\mathrm{p}\mathrm{n}}{\pi\mathrm{y}_{1}(1+\omega\mathrm{s}_{1})} \frac{\mathrm{r}\lambda}{\mathrm{p}\mathrm{x}_{1}} \frac{(\mathrm{x}-\mathrm{x}_{1})^{2}}{\mathrm{x}} + \frac{\eta_{1}\mathrm{s}\mathrm{n}\mathrm{p}\mathrm{n}}{1+\omega\mathrm{s}_{1}} \bigg[ 3 - \frac{\mathrm{s}_{1}(1+\omega\mathrm{s})}{\mathrm{s}(1+\omega\mathrm{s}_{1})} - \frac{\mathrm{s}\mathrm{p}\mathrm{y}_{1}(1+\omega\mathrm{s})}{\mathrm{s}_{1}\mathrm{p}\mathrm{n}\mathrm{y}(1+\omega\mathrm{s})} - \frac{\mathrm{p}\mathrm{n}\mathrm{y}}{\mathrm{p}\mathrm{y}\mathrm{n}} \bigg] \\ &\quad + \frac{\eta_{2}\mathrm{s}\mathrm{n}\mathrm{y}\mathrm{n}}{1+\omega\mathrm{s}\mathrm{n}} \bigg[ 2 - \frac{\mathrm{s}_{1}(1+\omega\mathrm{s})}{\mathrm{s}(1+\omega\mathrm{s}\mathrm{n})} - \frac{\mathrm{s}(1+\omega\mathrm{s}\mathrm{n})}{\mathrm{s}_{1}(1+\omega\mathrm{s})} \bigg]. \end{split}$$

Using the rule

$$\frac{1}{n}\sum_{i=1}^{n}a_{i}\geqslant\sqrt[n]{\prod_{i=1}^{n}a_{i}}, \quad \text{where,} \quad a_{i}\geqslant 0, i=1,2,\ldots,n,$$

we get

$$\frac{1}{3}\left(\frac{s_1(1+\omega s)}{s(1+\omega s_1)}+\frac{spy_1(1+\omega s_1)}{s_1p_1y(1+\omega s)}+\frac{p_1y}{py_1}\right) \geqslant 1, \quad \frac{1}{2}\left(\frac{s_1(1+\omega s)}{s(1+\omega s_1)}+\frac{s(1+\omega s_1)}{s_1(1+\omega s)}\right) \geqslant 1.$$

Therefore,  $\frac{dU_1}{dt} \leq 0$  for all s, y, p, x > 0 and  $\frac{dU_1}{dt} = 0$  if and only if  $s = s_1, y = y_1, p = p_1$  and  $x = x_1$ . It follows that the global stability of  $E_1$  is induced from LaSalle's invariance principle.

#### 3. Numerical simulations

Using the values in Table 1, we consider two cases as follows: **Case 1**: We simulate system (1.1)-(1.4) with the following initial conditions:

IC1: s(0) = 14.0, y(0) = 1.0, p(0) = 1.5, and x(0) = 1.5; IC2: s(0) = 8.0, y(0) = 2.0, p(0) = 3.0, and x(0) = 4.0; IC3: s(0) = 4.0, y(0) = 3.5, p(0) = 6.0, and x(0) = 7.0.

We fix  $\omega = 0.09$  and consider the values of  $\eta_1$  and  $\eta_2$  as following sets:

Set (I): We let  $\eta_1 = \eta_2 = 0.001$ . Computing  $\Re_0 = 0.0857 < 1$ , Figures (1)-(4) show that,  $E_0 = (s_0, 0, 0, x_0)$  is globally asymptotically stable, where  $s_0 = \frac{\beta}{\delta} = 20$  and  $x_0 = \frac{\lambda}{m} = 1.4$ , which agrees with the result of (2.1). Set (II): We choose  $\eta_1 = \eta_2 = 0.05$ . Calculating  $\Re_0 = 4.2857 > 1$ , we compute the equilibria as  $E_0(20.0, 0, 0, 1.4)$  and  $E_1 = (6.66, 2.66, 3.73, 5.51)$ . We have observed that in Figures (1)-(4), when  $\Re_0 > 1$ , the solution of the system tend to  $E_1$  for IC1-IC3 and (2.2) is confirmed.

**Case 2.** We fixed the value  $\eta_1 = \eta_2 = 0.06$ , by using the following initial conditions s(0) = 7, y(0) = 2.0, p(0) = 3.0, and x(0) = 4.0, we can see from Figures (5)-(8) that the evolution of the system's states with different values of  $\omega$ . We have observed that  $\mathcal{R}_0 > 1$ , and the trajectory of the system converges to the equilibrium  $E_1$  for smaller values of  $\omega$  e.g.  $\omega = 0.0, 0.2, 0.4$ . Whereas,  $\mathcal{R}_0 \leq 1$ , and the system has one equilibrium  $E_0$  when  $\omega$  become larger e.g.  $\omega = 2, 5$ . Let  $\omega^{ct}$  be the critical value of the parameter  $\omega$ , such that

$$\Re_0 = \frac{(\eta_1 \pi \mathfrak{m} + \eta_2 \mathfrak{c} \mathfrak{m} + \eta_2 \mathfrak{r} \lambda)\beta}{\varepsilon(\mathfrak{c} \mathfrak{m} + \mathfrak{r} \lambda)(\delta + \beta \omega^{\mathfrak{ct}})} = 1.$$

Using the data given in Table 1, we obtain  $\omega^{ct} = 0.67$ . The variation of  $\mathcal{R}_0$  w.r.t.  $\omega$  are listed in Table 2. We can observed that as  $\omega$  is increased then  $\mathcal{R}_0$  is decreased. Moreover, we have the following cases:

- (i) if  $0 \le \omega < 0.67$ , then E<sub>1</sub> exists and it is globally asymptotically stable,
- (ii) if  $\omega \ge 0.67$ , then E<sub>0</sub> is globally asymptotically stable.

Parameter	Value	Parameter	Value
β	2	δ	0.1
$\eta_1$	varied	$\eta_2$	varied
π	4	с	0.1
r	0.5	λ	1.4
m	1	ρ	0.2
ω	varied	e	0.5

 Table 1: The value of the parameters of model (1.1)-(1.4).

Table 2: The value of  $\mathcal{R}_0$  for different values of  $\omega$ .

ω	Equilibria	$\mathcal{R}_0$
0.0	(3.77, 3.24, 3.91, 6.43)	14.3997
0.2	(8.93, 2.21, 3.54, 4.80)	2.8800
0.4	(15.93, 0.08, 2.25, 2.60)	1.6000
0.67	(20.00, 0.00, 0.00, 1.40)	1.00
1	(20.00, 0.00, 0.00, 1.40)	0.6857
5	(20.00, 0.00, 0.00, 1.40)	0.1426



Figure 1: Uninfected monocytes.



Figure 3: Free CHIKV particles.



Figure 5: Uninfected monocytes.



Figure 2: Infected monocytes.



Figure 4: Antibodies.



Figure 6: Infected monocytes.



Figure 7: Free CHIKV particles.

Figure 8: Antibodies.

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