

## ENDEMIC EQUILIBRIUM POINT OF A MATHEMATICAL MODEL FOR DENGUE CHIKUNGUNYA VIRUS

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### Abstract

*The global stability analysis of endemic equilibria, for non-linear higher-dimensional problems are generally difficult. This paper attempts to determine the endemic equilibrium of the Mathematical model for Dengue virus and its co-existence with Chikungunya virus. The unique positive endemic equilibrium of the model is globally asymptotically stable when  $R_0 > 1$ .*

**Keywords:** Basic reproduction number, disease-free equilibrium, epidemic, dengue disease, infective, co-circulating, cross-immunity (CI), antibody-dependent enhancement (ADE)

#### 1.0 Introduction

Dengue virus is distributed by several species of mosquito of the Aedes Aegypti type which is the first. The second potential vector is Aedes Albopictus, third is Aedes Polynesiensis and the fourth is Aedes Scutellaris [1]. The urban-adapted, day-biting Aedes aegypti is the predominant mosquito vector of dengue although transmission may also occur through a secondary vector, Aedes albopictus [2]. Aedes aegypti is a cosmopolitan species distributed worldwide. Dengue fever is an international health problem in recent years [3], the number of cases of Dengue have increased sharply in the tropical and sub-tropical regions around the world [4, 3]. Chikungunya (CHIKV) is a genus of single-stranded virus that infect humans directly from bites of an Aedes spp. mosquitoes [5]. The manifestation of Chikungunya illness are fever without signs are discomfort and pain, skin rash, joint pain, headache, abdominal pain, nausea and vomiting, positive tourniquet test and leukopenia. The symptoms of Chikungunya are similar to those of Dengue diseases because they are spread by the same mosquitoes. Thus, there are no specific antiviral drugs to treat Chikungunya, so patient are advised to get enough rest, drink fluids and take drugs such as paracetamol to reduce fever and pain [6].

Mathematical model has been developed in [7], it was shown to exhibit the phenomenon of backward bifurcation at  $R_{DC} = 1$ , which was caused by the induced death for humans for both Dengue and Chikungunya [7]. The occurrence of backward bifurcation make effective disease control difficult, hence the need for the global asymptotic stability in the absence of induced death for humans for both Dengue and Chikungunya be shown in other to ascertain the successful eradication of the disease from the population.

#### 2.0 Model Formation

Consider the following model for the transmission dynamics of Dengue Chikungunya model [8] which consist of seventeen nonlinear ordinary differential equations given by,

$$\begin{aligned}
 \frac{dS_H}{dt} &= \Lambda_H - (\lambda_{DH} + \lambda_{CH} + \mu_H)S_H, \\
 \frac{dE_D}{dt} &= \lambda_{DH}S_H - (\gamma_D + \mu_H)E_D, \\
 \frac{dI_{D1}}{dt} &= \gamma_D E_D - (\delta_D + \mu_H + \theta_D)I_{D1}, \\
 \frac{dI_{D2}}{dt} &= \theta_D(1 - P_D)I_{D1} - (\alpha_D \delta_D + \tau_D + \mu_H)I_{D2} - \phi_D I_{WD}, \\
 \frac{dI_{WD}}{dt} &= \phi_D P_D I_{D1} - (\mu_H + \delta_D + \phi_D)I_{WD}, \\
 \frac{dR_D}{dt} &= \tau_D I_{D2} - \mu_H R_D, \\
 \frac{dE_C}{dt} &= \lambda_{CH}S_H - (\gamma_C + \mu_H)E_C, \\
 \frac{dI_{C1}}{dt} &= \gamma_C E_C - (\delta_C + \mu_H + \theta_C)I_{C1}, \\
 \frac{dI_{C2}}{dt} &= \theta_C(1 - P_C)I_{C1} - (\alpha_C \delta_C + \tau_C + \mu_H)I_{C2} - \phi_C I_{WC}, \\
 \frac{dI_{WC}}{dt} &= \phi_C P_C I_{C1} - (\mu_H + \delta_C + \phi_C)I_{WC}, \\
 \frac{dR_C}{dt} &= \tau_C I_{C2} - \mu_H R_C, \\
 \frac{dS_{MD}}{dt} &= \Lambda_{VD} - \lambda_{VD}S_{MD} - \mu_V S_{MD}, \\
 \frac{dE_{MD}}{dt} &= \lambda_{VD}S_{MD} - (\gamma_{MD} + \mu_V)E_{MD}, \\
 \frac{dI_{MD}}{dt} &= \gamma_{MD}E_{MD} - \mu_V I_{MD}, \\
 \frac{dS_{MC}}{dt} &= \Lambda_{VC} - \lambda_{VC}S_{MC} - \mu_V S_{MC}, \\
 \frac{dE_{MC}}{dt} &= \lambda_{VC}S_{MC} - (\gamma_{MC} + \mu_V)E_{MC}, \\
 \frac{dI_{MC}}{dt} &= \gamma_{MC}E_{MC} - \mu_V I_{MC},
 \end{aligned}$$

(1)

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Where the associated variable and parameters are described in Table 1.

**Table 1:** Description of state variables of the model (1)

State Variables	Description
$S_H(t)$	Population of susceptible individuals
$E_D(t)$	Population of humans exposed to dengue
$I_{D1}(t)$	Population of infectious humans with dengue
$I_{D2}(t)$	Population of infectious humans with dengue correctly diagnosed
$I_{DW}(t)$	Population of wrongly diagnosed dengue cases
$R_D(t)$	Population of humans who recovered from dengue
$E_C(t)$	Population of humans exposed to chikungunya
$I_{C1}(t)$	Population of infectious humans with chikungunya
$I_{C2}(t)$	Population of infectious humans with chikungunya correctly diagnosed
$I_{CW}(t)$	Population of wrongly diagnosed chikungunya cases
$R_C(t)$	Population of humans who recovered from chikungunya
$S_{MD}(t)$	Population of susceptible dengue vectors
$E_{MD}(t)$	Population of exposed dengue vectors
$I_{MD}(t)$	Population of infectious vectors with dengue
$S_{MC}(t)$	Population of susceptible chikungunya vectors
$E_{MC}(t)$	Population of exposed chikungunya vectors
$I_{MC}(t)$	Population of infectious vectors with chikungunya

**Table 2:** Description of parameters of model (1)

Parameter	Description
$\Lambda_H(t)$	Recruitment rate for humans.
$\mu_H(t)$	Natural mortality rate for humans.
$\beta_{DV}$	Probability of transmission of dengue from humans to vectors
$\beta_{DH}$	Probability of transmission of dengue from vectors to humans
$b_{DV}$	Biting rate of vectors that transmit dengue
$\eta_{D1}$	Modification parameter for reduced infectiousness of humans exposed to dengue
$\eta_{D2}$	Modification parameter for reduced infectiousness of humans rightly diagnosed for dengue
$\eta_{D3}$	Modification parameter for increased infectiousness of humans wrongly diagnosed for dengue
$\gamma_D$	Progression rate of humans exposed to dengue
$\delta_D$	Human disease induced death for dengue
$\theta_D$	Rate of diagnoses for dengue
$P_D$	Fraction of humans wrongly diagnosed for dengue
$\alpha_D$	Modification parameter for reduced mortality of humans rightly diagnosed for dengue
$\tau_D$	Recovery rate of humans from dengue
$\theta_D$	Rate of re-diagnoses for dengue
$\Lambda_V$	Recruitment rate of vectors that transmit dengue
$\mu_V$	Natural mortality rate for vector
$\gamma_{MD}$	Progression rate of vectors exposed to dengue
$\beta_{CV}$	Probability of transmission of chikungunya from humans to vectors
$\beta_{CH}$	Probability of transmission of chikungunya from vectors to humans

$b_{CV}$	Biting rate of vectors that transmit chikungunya
$\eta_{C1}$	Modification parameter for reduced infectiousness of humans exposed to chikungunya
$\eta_{C2}$	Modification parameter for reduced infectiousness of human rightly diagnosed for Chikungunya
$\eta_{C3}$	Modification parameter for increased infectiousness of humans wrongly diagnosed for chikungunya
$\gamma_C$	Progression rate of humans exposed to chikungunya
$\delta_C$	Human disease induced death for chikungunya
$\theta_C$	Rate of diagnoses for chikungunya
$P_C$	Fraction of humans wrongly diagnosed for chikungunya
$\alpha_c$	Modification parameter for reduced mortality of humans rightly diagnosed for chikungunya
$\tau_C$	Recovery rate of humans from chikungunya
$\phi_C$	Rate of re-diagnoses for chikungunya
$\Lambda_{VC}$	Recruitment rate of vectors that transmit chikungunya
$\gamma_{MC}$	Progression rate of vectors exposed to chikungunya

If in model (1)  $\delta_D = \delta_C = 0$ , special case is attained in which the disease induced death for humans for both dengue and chikungunya i.e., removing the parameters that cause the backward bifurcation. The resulting sub-model is given by

$$\begin{aligned}
\frac{dS_H}{dt} &= \Lambda_H - (\lambda_{DH} + \lambda_{CH} + \mu_H)S_H, \\
\frac{dE_D}{dt} &= \lambda_{DH}S_H - (\gamma_D + \mu_H)E_D, \\
\frac{dI_{D1}}{dt} &= \gamma_D E_D - (\mu_H + \theta_D)I_{D1}, \\
\frac{dI_{D2}}{dt} &= \theta_D(1 - P_D)I_{D1} - (\tau_D + \mu_H)I_{D2} - \phi_D I_{WD}, \\
\frac{dI_{WD}}{dt} &= \phi_D P_D I_{D1} - (\mu_H + \phi_D)I_{WD}, \\
\frac{dR_D}{dt} &= \tau_D I_{D2} - \mu_H R_D, \\
\frac{dE_C}{dt} &= \lambda_{CH}S_H - (\gamma_C + \mu_H)E_C, \\
\frac{dI_{C1}}{dt} &= \gamma_C E_C - (\mu_H + \theta_C)I_{C1}, \\
\frac{dI_{C2}}{dt} &= \theta_C(1 - P_C)I_{C1} - (\tau_C + \mu_H)I_{C2} - \phi_C I_{WC}, \\
\frac{dI_{WC}}{dt} &= \phi_C P_C I_{C1} - (\mu_H + \phi_C)I_{WC}, \\
\frac{dR_C}{dt} &= \tau_C I_{C2} - \mu_H R_C, \\
\frac{dS_{MD}}{dt} &= \Lambda_{VD} - \lambda_{VD}S_{MD} - \mu_V S_{MD}, \\
\frac{dE_{MD}}{dt} &= \lambda_{VD}S_{MD} - (\gamma_{MD} + \mu_V)E_{MD}, \\
\frac{dI_{MD}}{dt} &= \gamma_{MD}E_{MD} - \mu_V I_{MD}, \\
\frac{dS_{MC}}{dt} &= \Lambda_{VC} - \lambda_{VC}S_{MC} - \mu_V S_{MC}, \\
\frac{dE_{MC}}{dt} &= \lambda_{VC}S_{MC} - (\gamma_{MC} + \mu_V)E_{MC}, \\
\frac{dI_{MC}}{dt} &= \gamma_{MC}E_{MC} - \mu_V I_{MC},
\end{aligned}$$

(2)

The effective reproduction number corresponding to the mass action model (2) is obtain using the method of next generation matrix operator in [17] this is given by

$$\mathfrak{R}_T^m = \max[ \mathfrak{R}_{TD}^m, \mathfrak{R}_{TC}^m ], \tag{3}$$

where

$$\mathfrak{R}_{TD}^m = \sqrt{\frac{\beta_{DH}\beta_{DV}b_{DV}^2S_{MD}^*S_H^*\gamma_{MD}(g_2g_3g_4\eta_{D1} + g_3g_4\gamma_D + g_4\gamma_D\eta_{D2}\theta_D(1-P_D) + P_D\gamma_D\theta_Dg_3\eta_{D3} + P_D\gamma_D\theta_D\phi_D\eta_{D3})}{g_1g_2g_3g_4g_9\mu_V}},$$

$$\mathfrak{R}_{TC}^m = \sqrt{\frac{\beta_{CH}\beta_{CV}b_{CV}^2S_{MC}^*S_H^*\gamma_{MC}(g_5g_6g_8\eta_{C1} + g_7g_8\gamma_C + g_8\gamma_C\eta_{C2}\theta_C(1-P_C) + P_C\gamma_C\theta_Cg_3\eta_{C3} + P_C\gamma_C\theta_C\phi_C\eta_{C3})}{g_5g_6g_7g_8g_{10}\mu_V}}, \tag{4}$$

and  $g_1 = \gamma_D + \mu_H, g_2 = \mu_H + \theta_D, g_3 = \tau_D + \mu_H, g_4 = \phi_C + \mu_H, g_5 = \gamma_{MC} + \mu_H, g_6 = \mu_H + \theta_C, g_7 = \tau_C + \mu_H, g_8 = \phi_C + \mu_H, g_9 = \gamma_{MD} + \mu_V, g_{10} = \gamma_{MC} + \mu_V$

The global stability analysis of endemic equilibria, for non-linear higher-dimensional problems, such as the one in (2), is generally difficult to obtain [9]. For some special differential equation systems, one can be able to find a suitable Lyapunov function [10, 11], to prove the global stability. Unfortunately, there is no systemic way to construct or find Lyapunov functions, which hinders the application of this approach to more general model systems [9].

Theorem

The disease free equilibrium (DFE) of the mass action of the model (2), in the absence of disease-induced death for human ( $\delta_D$  and  $\delta_C$ ) is globally asymptotically stable (GAS) in D if  $R_T^m \leq 1$  and unstable if  $R_T^m \geq 1$ .

Proof:

Consider the following Lyapunov function

$$K = (K_1E_D + K_2I_{D1} + K_3I_{D2} + K_4I_{WD} + k_5E_C + K_6I_{C1} + K_7I_{C2} + K_8I_{WC} + K_9E_{MD} + K_{10}I_{MD} + K_{11}E_{MC} + K_{12}I_{MC}) \tag{5}$$

where

$$K_1 = \frac{\beta_{DV}b_{DV}S_{MD}^*\gamma_{MD}}{g_1g_2g_3g_4g_9\mu_V}(g_2g_3g_4\eta_{D1} + g_3g_4\gamma_D + \gamma_D\theta_D\eta_{D2}(g_4(1-P_D)P_D\phi_D) + P_D\gamma_D\theta_D\eta_{D3}g_3)$$

$$K_2 = \frac{\beta_{DV}b_{DV}S_{MD}^*\gamma_{MD}}{g_2g_3g_4g_9\mu_V}(g_3g_4 + \theta_D\eta_{D2}(g_4(1-P_D) + P_D\phi_D) + P_D\theta_D\eta_{D3}g_3)$$

$$K_3 = \frac{\beta_{DV}b_{DV}S_{MD}^*\gamma_{MD}\eta_{D2}}{g_3g_4\mu_V}, \quad K_4 = \frac{\beta_{DV}b_{DV}S_{MD}^*\gamma_{MD}}{g_3g_4g_9\mu_V}(\phi_D\eta_{D2} + g_3\eta_{D3})$$

$$K_5 = \frac{\beta_{CV}b_{CV}S_{MC}^*\gamma_{MC}}{g_1g_2g_3g_4g_9\mu_V}(g_6g_7g_8\eta_{C1} + g_7g_8\gamma_C + \gamma_C\theta_C\eta_{C2}(g_8(1-P_C)P_C\phi_C) + P_D\gamma_D\theta_D\eta_{D3}g_3),$$

$$K_6 = \frac{\beta_{CV}b_{CV}S_{MC}^*\gamma_{MC}}{g_6g_7g_8g_{10}\mu_V}(g_7g_8 + \theta_C\eta_{C2}(g_8(1-P_C) + P_C\phi_C) + P_C\theta_C\eta_{C3}g_7),$$

$$K_7 = \frac{\beta_{CV}b_{CV}S_{MC}^*\gamma_{MC}\eta_{D2}}{g_7g_8\mu_V}, \quad K_8 = \frac{\beta_{CV}b_{CV}S_{MC}^*\gamma_{MC}}{g_7g_8g_{10}\mu_V}(\phi_C\eta_{C2} + g_7\eta_{C3}),$$

$$K_9 = \frac{R_{DT}^m\gamma_{MD}}{g_9\mu_V}, \quad K_{10} = \frac{R_{DT}^m}{\mu_V}, \quad K_{11} = \frac{R_{CT}^m\gamma_{MC}}{g_{10}\mu_V}, \quad K_{12} = \frac{R_{CT}^m}{\mu_V}$$

With Lyapunov derivative (where a dot represents a time derivative)

$$\dot{K} = (K_1\dot{E}_D + K_2\dot{I}_{D1} + K_3\dot{I}_{D2} + K_4\dot{I}_{WD} + k_5\dot{E}_C + K_6\dot{I}_{C1} + K_7\dot{I}_{C2} + K_8\dot{I}_{WC} + K_9\dot{E}_{MD} + K_{10}\dot{I}_{MD} + K_{11}\dot{E}_{MC} + K_{12}\dot{I}_{MC}) \tag{6}$$

Substituting the right hand side of mass action version of the model (2) into (6) yields the following.

$$\dot{K} = K_1(\lambda_{DH}S_H - g_1E_D) + K_2(\gamma_DE_D - g_2I_{D1}) + K_3(\theta_D(1-P_D)I_{D1} - \phi_D I_{WD} - g_3I_{D2}) + K_4(\phi_DP_D I_{D1} - g_4I_{WD}) + k_5(\lambda_{CH}S_H - g_5E_C) + K_6(\gamma_CE_C - g_6I_{C1}) + K_7(\theta_C(1-P_C)I_{C1} - \phi_C I_{WC} - g_7I_{C2}) + K_8(\theta_CP_C I_{C1} + g_8I_{WC}) + K_9(\lambda_{VD}S_{MD} - g_9E_{MD}) + K_{10}(\gamma_{DM}E_{MD} - \mu_V I_{MD}) + K_{11}(\lambda_{VC}S_{MC} - g_{10}E_{MC}) + K_{12}(\gamma_{DM}E_{MD} - \mu_V I_{MC}) \tag{7}$$

Substituting values of  $K_1-K_{12}$  into (7) and after some algebraic simplifications, we have the following:

$$\begin{aligned} \dot{K} = & \mathfrak{R}_{DT}^m I_{MD} \left[ \frac{S_H}{S_H^*} \mathfrak{R}_{DT}^m - 1 \right] + \mathfrak{R}_{CT}^m I_{MC} \left[ \frac{S_H}{S_H^*} \mathfrak{R}_{CT}^m - 1 \right] + \left[ \frac{\beta_{DV} b_{DV} \gamma_{MD} S_{MD}^*}{g_9 \mu_V} \right] (\eta_{D1} E_D + I_{D1} + \eta_{D2} I_{D2} + \eta_{D3} I_{WD}) \left[ \frac{S_H}{S_H^*} \mathfrak{R}_{DT}^m - 1 \right] \\ & + \left[ \frac{\beta_{CV} b_{CV} \gamma_{MC} S_{MC}^*}{g_{10} \mu_V} \right] (\eta_{C1} E_C + I_{C1} + \eta_{C2} I_{C2} + \eta_{C3} I_{WC}) \left[ \frac{S_H}{S_H^*} \mathfrak{R}_{CT}^m - 1 \right] \end{aligned} \tag{8}$$

Since  $S_H < S_H^*$ ,  $S_{MD} < S_{MD}^*$ ,  $S_{MC} < S_{MC}^*$ , we have that

$$\begin{aligned} \dot{K} \leq & \mathfrak{R}_{DT}^m I_{MD} [\mathfrak{R}_{DT}^m - 1] + \mathfrak{R}_{CT}^m I_{MC} [\mathfrak{R}_{CT}^m - 1] + \left[ \frac{\beta_{DV} b_{DV} \gamma_{MD} S_{MD}^*}{g_9 \mu_V} \right] (\eta_{D1} E_D + I_{D1} + \eta_{D2} I_{D2} + \eta_{D3} I_{WD}) [\mathfrak{R}_{DT}^m - 1] \\ & + \left[ \frac{\beta_{CV} b_{CV} \gamma_{MC} S_{MC}^*}{g_{10} \mu_V} \right] (\eta_{C1} E_C + I_{C1} + \eta_{C2} I_{C2} + \eta_{C3} I_{WC}) [\mathfrak{R}_{CT}^m - 1] \end{aligned} \tag{9}$$

Hence,  $\dot{K} \leq 0$ , whenever  $\mathfrak{R}_i^m \leq 1$ , with  $\dot{K} = 0$  if and only if

$$(E_D = I_{D1} = I_{D2} = I_{WD} = E_C = I_{C1} = I_{C2} = I_{WC} = E_{MD} = I_{MD} = E_{MC} = I_{MC} = 0) \tag{10}$$

Hence,  $\dot{K} \leq 0$ , is a Lyapunov functional defined in the domain  $D$ . Thus, it follows from *Lasalle's Invariance principle* [12] that:

$$\left( \begin{matrix} E_D(t), I_{D1}(t), I_{D2}(t), I_{WD}(t), E_C(t), I_{C1}(t), I_{C2}(t), I_{WC}(t), E_{MD}(t), \\ I_{MD}(t), E_{MC}(t), I_{MC}(t) \end{matrix} \right) \rightarrow (0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0) \tag{11}$$

as  $t \rightarrow \infty$ .

Hence, every orbit of the equations of the mass action version of model (2) with  $\delta_D = \delta_C = 0$ , approaches the DFE, as  $t \rightarrow \infty$  for  $R_i^m \leq 1$ .

This result shows that in a population where there are incidences of misdiagnosis of either disease with negligible disease induced mortality for humans, i.e.,  $\delta_D = \delta_C = 0$ , the DFE will be GAS whenever  $R_i^m \leq 1$ .

### 3.0 Numerical Simulations

In this section, we investigate the sensitivity of the parameters of the system (1) with respect to the reproduction number, the numerical simulation results is also shown to illustrates the impact of varying certain key parameter on the total disease incidence in a population. The work done in this section is based on parameter values in the table below.

**Table 3:** Baseline values and ranges of the parameters of our model (1)

Parameters	Baseline values	Ranges	References
$\Lambda VC$	5000 day <sup>-1</sup>	(3000, 50000)	[13]
$\beta CH$	0.60 day <sup>-1</sup>	(0.43, 0.79)	[14]
$\beta CV$	0.53 day <sup>-1</sup>	(0.60, 0.75)	[13]
$b CV$	0.53 day <sup>-1</sup>	(0.002, 0.54)	Assumed
$\eta C1$	0.5 day <sup>-1</sup>	(0, 1)	Assumed
$\eta C2$	0.8 day <sup>-1</sup>	(0, 1)	Assumed
$\eta C3$	1.2 day <sup>-1</sup>	(0, 1)	Assumed
$\gamma C$	0.071 day <sup>-1</sup>	(0, 1)	Assumed
$\delta C$	0.001 day <sup>-1</sup>	(0, 1)	[13]
$\theta C$	0.099 day <sup>-1</sup>	(0.3, 1.0)	Assumed
$\alpha C$	0.4 day <sup>-1</sup>	(0.005, 0.14284)	Assumed
$PC$	0.5 day <sup>-1</sup>	(0.002, 0.051)	Assumed
$\tau C$	0.15 day <sup>-1</sup>	(0.11, 0.17)	[15]
$\varphi C$	0.5 day <sup>-1</sup>	(0.002, 0.051)	Assumed
$\gamma MC$	0.3333 day <sup>-1</sup>	(0.02, 0.533)	Assumed

**Table 4:** Baseline values and ranges of the parameters of our model (1)

Parameters	Baseline values	Ranges	References
$\Lambda_H$	18250 day <sup>-1</sup>	$(\frac{1}{18300}, \frac{1}{18200})$	[16]
$\mu_H$	20075 day <sup>-1</sup>	$(\frac{1}{26,280}, \frac{1}{12775})$	[14]
$\Lambda V$	5000 day <sup>-1</sup>	(3000, 50000)	

$\beta_{DH}$	0.60 day <sup>-1</sup>	(0.43, 0.79)	[14]
$\beta_{DV}$	0.43 day <sup>-1</sup>	(0.60, 0.75)	[13]
$b_{DV}$	0.53 day <sup>-1</sup>	(0.002, 0.54)	Assumed
$\eta_{D1}$	0.5 day <sup>-1</sup>	(0, 1)	Assumed
$\eta_{D2}$	0.8 day <sup>-1</sup>	(0, 1)	Assumed
$\eta_{D3}$	1.2 day <sup>-1</sup>	(0, 1)	Assumed
$\gamma_D$	0.071 day <sup>-1</sup>	(0, 1)	Assumed
$\delta_D$	0.001day <sup>-1</sup>	(0, 1)	[13]
$\theta_D$	0.099 day <sup>-1</sup>	(0.3, 1.0)	Assumed
$\alpha_D$	0.4 day <sup>-1</sup>	(0.005, 0.14284)	Assumed
$P_D$	0.5 day <sup>-1</sup>	(0.002, 0.051)	Assumed
$\tau_D$	0.11 day <sup>-1</sup>	(0.11, 0.15)	[13]
$\varphi_D$	0.5 day <sup>-1</sup>	(0.002, 0.051)	Assumed
$\gamma_{MD}$	0.3333 day <sup>-1</sup>		Assumed

Table 5: Provides the elasticity indices of R0D and R0C to the 36 parameter values for the parameters of the model (1) using the reproduction number

Parameters	R0D	Parameters	R0D
$\beta_{DH}$	<b>0.5</b>	$\theta_D$	-0.238239
$\beta_{DV}$	<b>0.5</b>	$\eta_{D1}$	0.1065
$b_{DV}$	<b>1.00</b>	$\eta_{D2}$	0.0972003
$\Delta_H$	<b>-0.5</b>	$\eta_{D3}$	0.159365
$\Delta_{VD}$	<b>0.5</b>		
$\mu_H$	0.499364		
$\mu_V$	-1.21983		
$\gamma_{MD}$	-0.110927		
$\gamma_D$	-0.110927		
$\varphi_D$	-0.155139		
$\alpha_D$	-0.0000336831		
$\delta_D$	-0.00902911		
$\tau_D$	-0.00926286		
$P_D$	-0.238239		

Parameters	R0C	Parameters	R0C
$\beta_{CH}$	<b>0.5</b>	$\theta_C$	-0.238239
$\beta_{CV}$	<b>0.5</b>	$\eta_{C1}$	0.1065
$b_{CV}$	<b>1.00</b>	$\eta_{C2}$	0.0972003
$\Delta_H$	<b>-0.5</b>	$\eta_{C3}$	0.159365
$\Delta_{VC}$	<b>0.5</b>		
$\mu_H$	0.499364		
$\mu_V$	-1.21983		
$\gamma_{MC}$	-0.110927		
$\gamma_C$	-0.110927		
$\varphi_C$	-0.155139		
$\alpha_C$	-0.0000336831		
$\delta_C$	-0.00902911		
$\tau_C$	-0.00926286		
$P_C$	-0.238239		

The system (1) is numerically simulated so as to investigate the impact of certain important parameters which relates to the effective reproduction number. The value of the parameters in Table 3 and 4 are used for simulation. The contour plot of figure 1 shows that with increased infectiousness of humans wrongly diagnosed for dengue ( $\eta_{D3}$ ) is at 55 percent, the rate of diagnosed for Dengue ( $\theta_D$ ) must be 60 percent and above for the disease to be eradicated from the population. The contour plot of figure 2 shows that with increased infectiousness of humans wrongly diagnosed for dengue ( $\eta_{D3}$ ), it will require a very high rate of re-diagnosis for Dengue ( $\phi_D$ ) for proper eradication of the disease. This implies that high percentage of infectiousness require a higher percentage of proper re-diagnosis for eradication of Dengue disease from the population.

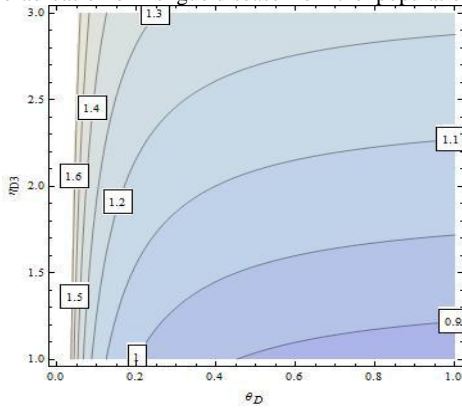


Figure 1: Contour plot of  $R_{0D}$  as a function of  $\eta_{D3}$  and  $\theta_D$ , other parameters are at baseline.

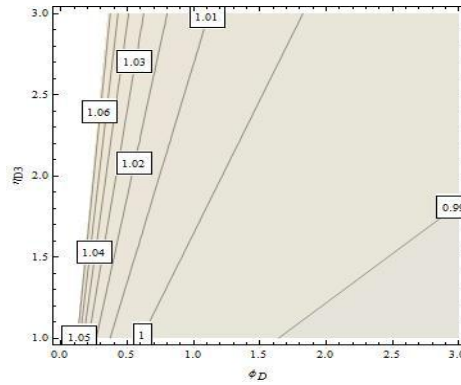


Figure 2: Contour plot of  $R_{0D}$  as a function of  $\eta_{D3}$  and  $\phi_D$ , other parameters are at baseline.

The contour plot of figure 3 shows that with increased infectiousness of humans wrongly diagnosed for chikungunya  $\eta_{C3}$  is at 63 percent, the rate of diagnosed for chikungunya  $\phi_C$  must be 80 percent and above for the disease to be eradicated from the population. The contour plot of figure 4 shows that with increased infectiousness of humans wrongly diagnosed for chikungunya ( $\eta_{C3}$ ), the disease can be eradicated from the population in the presence of re-diagnoses for chikungunya ( $\phi_C$ ).

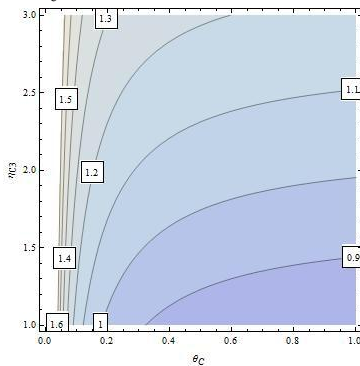


Figure 3: Contour plot of  $R_{0C}$  as a function of  $\eta_{C3}$  and  $\theta_C$ , other parameters are at baseline.

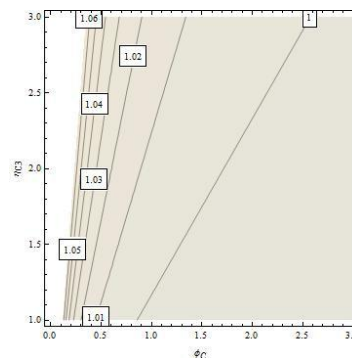
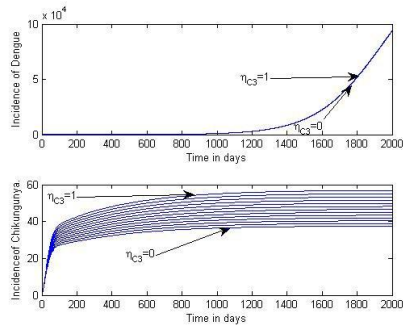
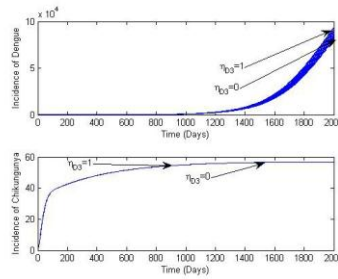


Figure 4: Contour plot of  $R_{0C}$  as a function of  $\eta_{C3}$  and  $\phi_C$ , other parameters are at baseline..

Figure 5 depicts the cumulative incidence of Dengue and Chikungunya as the rate of increased infectiousness of wrongly diagnosed for Chik  $\eta_{C3}$  was varied from 0 to 1. The result indicate that the burden of Chik disease will be higher with more than 20 new cases due to the presence of increased infectiousness of wrongly diagnosed for Chik. Figure 6 depicts the cumulative incidence of Dengue and Chikungunya as the rate of increased infectiousness of wrongly diagnosed for Dengue  $\eta_{D3}$  was varied from 0 to 1. The result shows that the burden of Dengue disease increased with 30,000 new cases in 2,000 days due to the presence of wrong diagnosis.

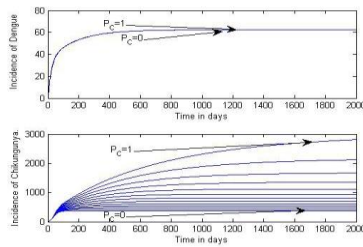


**Figure 5** (a) Incidence Dengue with variation in increased infectiousness of human wrongly diagnosed of Chikungunya (b) Incidence Chikungunya with variation in increased infectiousness of human wrongly diagnosed of Chikungunya.

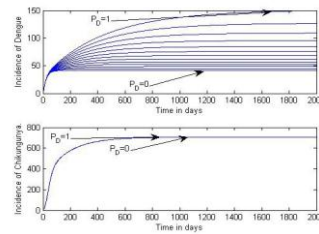


**Figure 6:** (a) Incidence Dengue with variation in increased infectiousness of human wrongly diagnosed of Dengue (b) Incidence Chikungunya with variation in increased infectiousness of human wrongly diagnosed of Dengue

Figure 7 depicts the cumulative incidence of Dengue and Chikungunya as the fraction of humans wrongly diagnosed for Chikungunya ( $p_C$ ) was varied from 0 to 1. The result shows that 1,750 new cases was detected which is great significance of the disease burden on the human population. Figure 8 depicts the cumulative incidence of Dengue and Chikungunya as the fraction of humans wrongly diagnosed for Dengue ( $P_D$ ) was varied from 0 to 1. The result shows that 110 new cases was detected which is also a great significance of the disease burden on the human population.

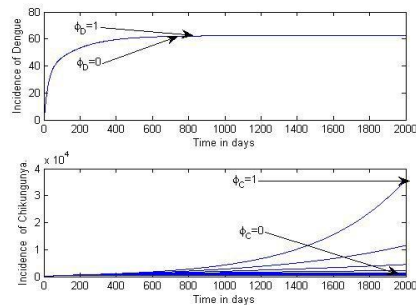


**Figure 7:** (a) Incidence Dengue with variation in Fraction of humans wrongly diagnosed for Chikungunya (b) incidence Chikungunya with variation in Fraction of humans wrongly diagnosed for Chikungunya

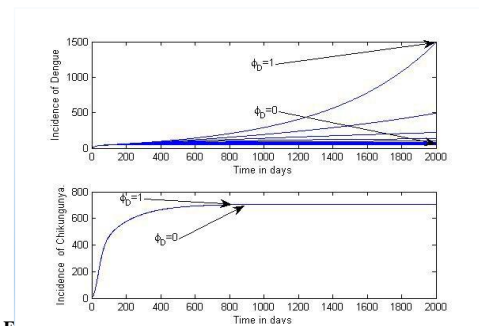


**Figure 8:** (a) Incidence Dengue with variation in Fraction of humans wrongly diagnosed for Dengue (b) Incidence Chikungunya with variation in Fraction of humans wrongly diagnosed for Dengue

Figure (9) depicts the cumulative incidence of Dengue with the rate of re-diagnosis for Chikungunya ( $\phi_C$ ) was varied from 0 to 1. The incidence increase as ( $\phi_C$ ) tend to 1, this implies that there is a great significance in the number of persons that will recover due to re-diagnosis of Chikungunya. Figure (10) depicts the cumulative incidence of Dengue and Chikungunya with the rate of re-diagnosis for Dengue ( $\phi_D$ ) was varied from 0 to 1. The incidence increase as ( $\phi_D$ ) tend to 1, this implies that there is a great significance in the number of persons that will recover due to re-diagnosis of Dengue.



**Figure 9:** (a) Incidence Dengue with variation in Rate of re-diagnoses for Chikungunya (b) Incidence chikungunya with variation in Rate of re-diagnoses for Chikungunya



**Figure 10:** (a) Incidence Dengue with variation in Rate of re-diagnoses for Dengue (b) Incidence Chikungunya with variation in Rate of re-diagnoses for Dengue

Figure (11) depicts the cumulative incidence of Dengue and Chikungunya with the rate of diagnosis for Dengue ( $\theta_D$ ) varied from 0 to 1. The incidence increase as ( $\theta_D$ ) tend to 1, this implies that there is a great significance in the number of persons that will



recover due to proper diagnosis for Dengue. Figure (12) depicts the cumulative incidence of Dengue and Chikungunya with the rate of diagnosis for Dengue ( $\theta_D$ ) varied from 0 to 1. The incidence increase as ( $\theta_C$ ) tend to 1, this implies that there is a great significance in the number of persons that will recover due to proper diagnosis for Chikungunya. This result means that both disease can be eradicated from the population if there is a proper diagnose.

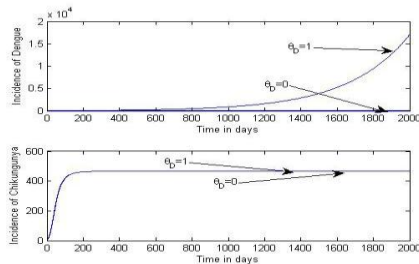


Figure 11: (a) Incidence Dengue with variation in Rate of diagnoses for dengue (b) Incidence Chikungunya with variation in Rate of diagnoses for dengue

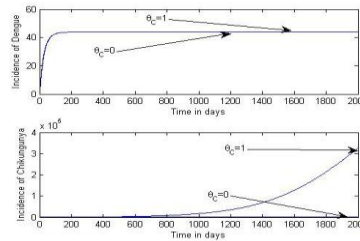


Figure 12: (a) Incidence of Dengue with variation in Rate of diagnoses for Chikungunya (b) Incidence of Chikungunya with variation in Rate of diagnoses for Chikungunya

#### 4.0 CONCLUSION

A special case of the model system (2) was shown to be globally asymptotically stable (GAS), when the associated effective reproduction number was below unity. This result shows that in a population where there are incidences of misdiagnosis of either disease with negligible disease induced mortality for humans ( $\delta_D = \delta_C = 0$ ), the DFE will be GAS whenever ( $R_T^m \leq 1$ ). Numerical study of the effective reproduction number showed that, the rate of diagnosed for Dengue must be 60 percent above and the rate of diagnosed for chikungunya ( $\theta_C$ ) must be 80 percent and above for the disease to be eradicated from the population. It also shows that with increased infectiousness of humans wrongly diagnosed for dengue and Chikungunya, it will require a very high rate of re-diagnosis for proper eradication of both disease. This implies that high percentage of infectiousness require a higher percentage of proper re-diagnosis for eradication of Dengue and Chikungunya disease from the population. Local sensitivity analysis was carried out on key parameter using the effective reproduction number as a response function. Ten parameters that had the highest index were: the transmission probability from humans to vectors with Dengue and Chikungunya ( $\beta_{DV}$  and  $\beta_{CV}$ ), the transmission probability from vectors to humans with Dengue and Chikungunya ( $\beta_{DH}$  and  $\beta_{CH}$ ), death rate for the vectors  $\mu_V$ , the biting rate of vector with Chikungunya and Dengue ( $b_{DV}$  and  $b_{CV}$ ), human recruitment rate and vectors recruitment rate ( $\Lambda_H, \Lambda_{VD}$  and  $\beta_{VC}$ ).

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