The Zika Virus Transmission Model

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Abstract: A mathematical model of Zika virus is studied in this paper. Zika is caused by Zika virus, a flavivirus related to yellow fever, dengue. In 1952, first outbreak occurred in Uganda. In 1962, an epidemic was recognized as the first time in Thailand. In this study, we consider the transmission cycle between two population groups: human and mosquito. Using standard dynamical modeling method, the stability conditions of our model is considered by Routh-Hurwitz criteria. The numerical simulations are shown to support the analytical results.

Key words: Basic reproductive number, disease free steady state, endemic steady state, stability, Zika.

1. Introduction

On 6 July 2016, World Health Organization (since 2007) reported evidence of zika virus from 65 countries [1]. There are 61 countries reported mosquito-borne zika virus transmission. Guinea-Bissau is the last country where there is the report of mosquito-borne zika virus [1], [2]. Zika virus is caused by a virus transmitted by *Aedes* mosquitoes, Family *flavivirus*, related to yellow fever, dengue [3]. In 1947, Zika virus is initially found in monkey of Uganda. In 1952, Zika virus was found in human [4]-[7].

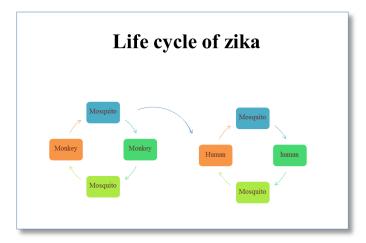


Fig. 1. Life cycle of zika [7]-[9].

In 1952, an epidemic was recognized as the first time in Thailand. The first case is the traveler, a female from Canadian. Between 21 January to 4 February, Thailand confirmed zika virus disease in a traveler.

Human can be infected by mosquito-borne from the Aedes genus. This is the same mosquito that transmits dengue, chikungunya and yellow fever [10]. In 2016, Gao et al. [4], studied mathematical model of zika virus as mosquito-borne disease. They considered sexual transmission, biting rate and mortality rate of mosquito. They used sensitivity analysis to analyze the basic reproductive number (R_0). Kucharski *et al.* [11] use a mathematical model of vector-borne infections, the transmission dynamical of zika virus in island population and examined the transmission dynamical of zika virus on six archipelagos in Frence in 2013-14. Bichare et al. [12] identified the impact of short term mobility between two idealized interconnected communities, using a Lagrangina mode land show simulations of short term mobility. In 2011, Naowawat, et al. [13], studied dynamical model for determining human susceptible to dengue fever. The basic reproductive number is explained as follows: $R_0 < 1$, the disease-free state is local stable. The endemic state is local stable for $R_0 > 1$. In this paper, we consider the transmission of zika virus. We divided the human population into susceptible, exposed, infectious and recovered classes (SEIR models). Mosquito population is separated into susceptible, exposed and infectious classes (SEI model). Using standard dynamical modeling method, the stability of model is determined by using Routh-Hurwitzcriteria. The Threshold parameter is found to separate the different behaviors of two steady states. The numerical simulations are shown to support the analytical results.

2. Mathematical Model

In our model, we consider the transmission of zika virus between two population, human population and mosquito population. The human population is divided into four sub-groups: susceptible, exposed, infected and recovered. The mosquito is divided into three sub-groups: susceptible, exposed and infected. The variables and parameters in our equations are described as follows:

- S_h is the size of susceptible human populations,
- E_h is the size of exposed human population,
- I_h is the size of infected human population,
- R_h is the size of recovered human population,
- S_v is the size of susceptible mosquito population,
- $E_{\rm v}~$ is the size of exposed mosquito population,
- $I_{\scriptscriptstyle V}~$ is the size of infected mosquito population,
- *B* is the birth rate of human population,
- d_v is the natural death rate of human populations,
- ε is the biting rate of mosquito population,
- β_h is the rate at which mosquito to human Zika is contracted,
- F_h is the incubation period of Zika virus in human,
- γ is the recovery rate of human populations,
- $N_{\scriptscriptstyle h}$ is the number of human populations,
- m is the number of other animals that the mosquito can feed on
- A is the recruitment rate of mosquitoes
- $d_{\scriptscriptstyle \rm V}~$ is the natural death rate of mosquito
- β_{v} is the rate at which human to mosquito Zika is contracted,
- $F_{\nu}~$ is the incubation period of Zika virus in mosquito,

The diagram of our dynamical systems can be described by following Fig. 2:

The dynamical rate of each population group is described by following equations.

For human:

$$\frac{dS_h}{dt} = BN_h - \frac{\varepsilon\beta_h S_h I_v}{N_h + m} - d_h S_h$$
⁽¹⁾

$$\frac{dE_h}{dt} = \frac{\varepsilon \beta_h S_h I_v}{N_h + m} - (F_h + d_h) E_h$$
⁽²⁾

$$\frac{dI_h}{dt} = F_h E_h - (\gamma + d_h) I_h \tag{3}$$

$$\frac{dR_h}{dt} = \gamma I_h - d_h R_h \tag{4}$$

For mosquitoes:

$$\frac{dS_{\nu}}{dt} = A - \frac{\varepsilon \beta_{\nu} S_{\nu} I_{h}}{N_{h} + m} - d_{\nu} S_{\nu}$$
⁽⁵⁾

$$\frac{dE_{\nu}}{dt} = \frac{\varepsilon \beta_{\nu} S_{\nu} I_{h}}{N_{h} + m} - (F_{\nu} + d_{\nu}) E_{\nu}$$
(6)

$$\frac{dI_{\nu}}{dt} = F_{\nu}E_{\nu} - d_{\nu}I_{\nu}$$
⁽⁷⁾

where $N_h = S_h + E_h + I_h + R_h$ and $N_v = S_v + E_v + I_v$

We can be normalize our equations (1)-(7) by introducing the new variables:

$$s_{h} = \frac{S_{h}}{N_{h}}, e_{h} = \frac{E_{h}}{N_{h}}, i_{h} = \frac{R_{h}}{N_{h}}, r_{h} = \frac{R_{h}}{N_{h}}, s_{v} = \frac{S_{v}}{N_{v}} = \frac{S_{v}}{(A/d_{v})}, e_{v} = \frac{E_{v}}{N_{v}} = \frac{E_{v}}{(A/d_{v})} \text{ and } i_{v} = \frac{I_{v}}{N_{v}} = \frac{I_{v}}{(A/d_{v})}$$

Then we have the dynamical equations of the reduced equations as follows:

$$\frac{ds_{h}}{dt} = B - \frac{\varepsilon \beta_{h} A s_{h} i_{v}}{d_{v} (N_{h} + m)} - d_{h} s_{h}$$

$$\frac{de_{h}}{dt} = \frac{\varepsilon \beta_{h} A s_{h} i_{v}}{d_{v} (N_{h} + m)} - (F_{h} + d_{h}) e_{h}$$

$$\frac{di_{h}}{dt} = F_{h} e_{h} - (\gamma + d_{h}) i_{h}$$

$$\frac{ds_{v}}{dt} = d_{v} - \frac{\varepsilon \beta_{v} N_{h} s_{v} i_{h}}{N_{h} + m} - d_{v} s_{v}$$

$$\frac{di_{v}}{dt} = F_{v} (1 - i_{v} - s_{v}) - d_{v} i_{v}$$
(8)

with the condition $s_h + e_h + i_h + r_h = 1$ and $s_v + e_v + i_v = 1$

3. Analysis of Mathematical Model

3.1 Analytical Solutions

After we formulate the dynamical equations, the next step is to find the steady states of our equations. The standard dynamical modeling method is used for analysis of our model, a disease free steady state and

an endemic steady state are obtained. We set the right hand side of equations (8) to zero, then we obtain the steady states:

- i) Disease free steady state: $\overline{E}_0 = (s_h^*, e_h^*, i_h^*, s_v^*, i_v^*) = (1, 0, 0, 1, 0)$
- ii) Endemic steady state: $\overline{E}_1 = (s_h^*, e_h^*, i_h^*, s_v^*, i_v^*)$ where

$$s_{h}^{*} = \frac{Bd_{v}(N_{h} + m)}{d_{h}d_{v}(N_{h} + m) + \frac{AF_{v}N_{h}\beta_{h}\beta_{v}\varepsilon^{2}i_{h}^{*}}{(d_{v} + F_{v})(d_{v}(N_{h} + m) + \varepsilon N_{h}\beta_{v}i_{h}^{*})}}$$

$$e_{h}^{*} = \frac{ABF_{v}N_{h}\beta_{h}\beta_{v}\varepsilon^{2}i_{h}^{*}}{(d_{h} + F_{h})(AF_{v}N_{h}\beta_{h}\beta_{v}\varepsilon^{2}i_{h}^{*} + d_{h}d_{v}(d_{v} + F_{v})(N_{h} + m)(d_{v}(N_{h} + m) + N_{h}\beta_{v}\varepsilon i_{h}^{*}))}$$

$$i_{h}^{*} = -\frac{d_{h}d_{v}^{2}(d_{h} + F_{h})(d_{v} + F_{v})(N_{h} + m)^{2}(d_{h} + \gamma) - ABF_{h}F_{v}N_{h}\beta_{h}\beta_{v}\varepsilon^{2}}{(d_{h} + F_{h})N_{h}\beta_{v}(d_{h} + \gamma)\epsilon(d_{h}d_{v}(d_{v} + F_{v})(N_{h} + m) + AF_{v}\beta_{h}\varepsilon)}$$

$$s_{v}^{*} = \frac{d_{v}(N_{h} + m)}{d_{v}(N_{h} + m) + N_{h}\beta_{v}\varepsilon i_{h}^{*}}$$

$$i_{v}^{*} = \frac{F_{v}N_{h}\beta_{v}\varepsilon i_{h}^{*}}{(d_{v} + F_{v})(d_{v}(N_{h} + m) + N_{h}\beta_{v}\varepsilon i_{h}^{*})}$$

By using standard dynamic modeling method, the local stability of each steady state is defined of all eigenvalues. If all eigenvalues for each steady state produce the negative real part, we can conclude that steady state is local stability [14]. The eigenvalues are the solutions of the characteristic equation

$$\left|J_{\overline{F}_{i}}-\lambda I_{5}\right|=0, i=0,1$$

where $J_{\overline{E}_i}$ is the Jacobian matrix; i = 0,1 and I is the identity matrix.

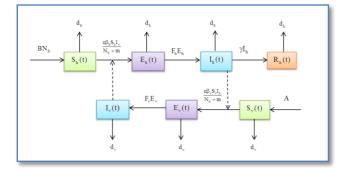


Fig.2 flowchart of the model.

Disease free steady state, the characteristic equation is $\overline{E}_0 = (s_h^*, e_h^*, i_h^*, s_v^*, i_v^*) = (1, 0, 0, 1, 0)$

$$\begin{vmatrix} -d_h - \lambda & 0 & 0 & 0 & -\frac{\varepsilon A \beta_h}{d_v (N_h + m)} \\ 0 & -(d_h + F_h) - \lambda & 0 & 0 & \frac{\varepsilon A \beta_h}{d_v (N_h + m)} \\ 0 & F_h & -(d_h + \gamma) - \lambda & 0 & 0 \\ 0 & 0 & 0 & -\frac{\varepsilon \beta_v N_h}{N_h + m} & -d_v - \lambda & 0 \\ 0 & 0 & 0 & 0 & -F_v & (-d_v - F_v) - \lambda \end{vmatrix} = 0$$

The characteristic equation of the above Jacobian matrix is

$$(-d_h - \lambda)(\lambda^4 + A_1\lambda^3 + A_2\lambda^2 + A_3\lambda + A_4) = 0$$

We can see all eigenvalues have negative real parts. We use Routh-Hurwitz criteria [15]. Where R₀<1

$$R_{0} = \frac{F_{h}F_{v}\beta_{h}\beta_{v}N_{h}\varepsilon^{2}}{d_{v}^{2}(d_{h}+F_{h})(d_{v}+F_{v})(N_{h}+m)^{2}(d_{h}+\gamma)}$$

Therefore, we can conclude that the endemic steady state is local stability for $R_0>1$ and the disease free steady state is local stability $R_0<1$.

3.2 Numerical Solutions

The values of parameter in the model are shown in Table 1.

Table 1. Values of Parameter			
parameter	Values	Source	
ε	0.5	[4,10]	
eta_{v}	0.05	[4,10]	
F_{v}	1/5	[4,7,10]	
$d_{_V}$	1/14	[4,7,10]	
F_h	1/5	[6]	
γ	1/20	[11]	
В	1/(65×365)	[16]	
d_h	1/(75×365)	[16]	
eta_h	0.4	Assumed	
N_{h}	1	Assumed	
m	1	Assumed	
Α	10	Assumed	
Disease fre	ee R ₀ =0.0001	R ₀ =0.000114	
Endemic	Endemic R ₀ =2.8557		

0.2 Infected human Infected mosquite 0.2 Infected human and mosquito 0.15 0.1 0.05 0 20 40 60 100 140 160 180 200 80 120 Time(days)

Fig. 3. Time series solution of infected human and infected mosquito.

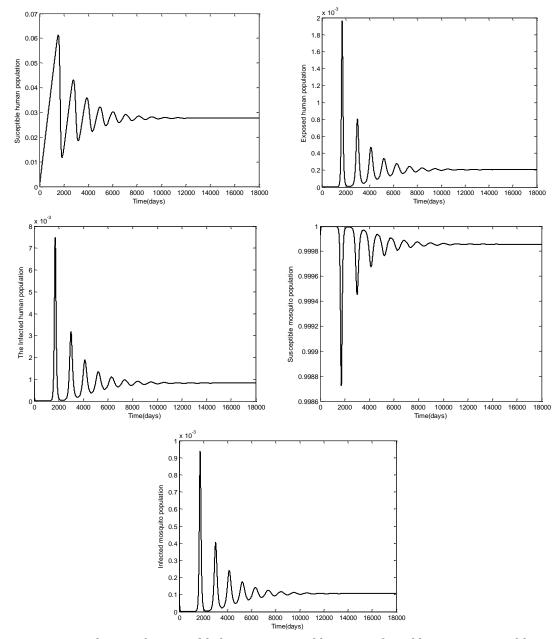


Fig. 4. Time series solution of susceptible human, exposed human, infected human, susceptible mosquito and infected mosquito on endemic steady state, the solutions converge to (0.0277, 0.0002, 0.0008, 0.9999, 0.0001) where R_0 =2.8557.

4. Conclusions

In this study, we constructed the dynamical transmission model of Zika virus. The model has two steady state, disease free steady and endemic steady state. The basic reproductive number [16], [17] is denoted by R_0 where

$$R_{0} = \frac{F_{h}F_{v}\beta_{h}\beta_{v}N_{h}\varepsilon^{2}}{d_{v}^{2}(d_{h}+F_{h})(d_{v}+F_{v})(N_{h}+m)^{2}(d_{h}+\gamma)}$$

Fig. 2 is time series solution of infected human and infected mosquito, where R_0 <1. And Fig. 3 is time series solution of susceptible human, exposed human, infected human, susceptible mosquito and infected

mosquito on endemic steady state, the solutions converge to (0.0277, 0.0002, 0.0008, 0.9999, 0.0001) where R_0 =2.8557.

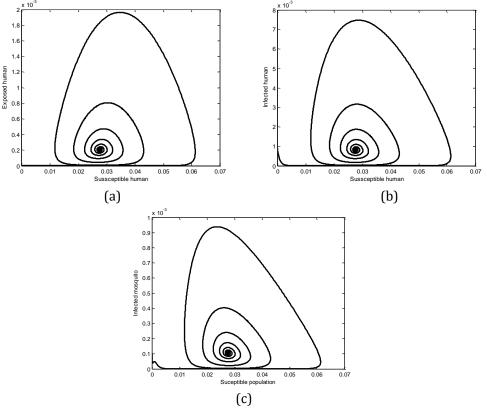


Fig. 5. Numerical solution of our dynamical equations. where $R_0 > 1$.

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