

Mathematical model of Zika virus with vertical transmission



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ABSTRACT

Zika is a flavivirus transmitted to humans through either the bites of infected *Aedes* mosquitoes or sexual transmission. Zika has been linked to congenital anomalies such as microcephaly. In this paper, we analyze a new system of ordinary differential equations which incorporates human vertical transmission of Zika virus, the birth of babies with microcephaly and asymptotically infected individuals. The Zika model is locally and globally asymptotically stable when the *basic reproduction number* is less than unity. Our model shows that asymptomatic individuals amplify the disease burden in the community, and the most important parameters for ZIKV spread are the death rate of mosquitoes, the mosquito biting rate, the mosquito recruitment rate, and the transmission *per* contact to mosquitoes and to adult humans. Scenario exploration indicates that personal-protection is a more effective control strategy than mosquito-reduction strategy. It also shows that delaying conception reduces the number of microcephaly cases, although this does little to prevent Zika transmission in the broader community. However, by coupling aggressive vector control and personal protection use, it is possible to reduce both microcephaly and Zika transmission. 2000 Mathematics Subject Classifications: 92B05, 93A30, 93C15.

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1. Introduction

Zika virus (ZIKV) is a mosquito-borne disease transmitted to humans through the bites of infected *Aedes* mosquitoes, including *Aedes aegypti*, *Aedes africanus*, *Aedes apicoargenteus*, *Aedes furcifer*, *Aedes hensilli*, *Aedes luteocephalus* and *Aedes vitattus*. First identified in a rhesus macaque population in 1947 in the Zika forest of Uganda, ZIKV is from the *Spondweni* serocomplex of the *Flaviviridae* family of viruses. Historically, ZIKV was thought to cause mild symptoms in humans, including headaches, maculopapular rash, fever, malaise, conjunctivitis, and arthralgia, occurring three to twelve days after the bite from an infected mosquito. Recently, however, there have been reported increases in congenital anomalies (such as microcephaly), Guillain-Barre syndrome, and other neurological and autoimmune disorders in regions where ZIKV has been newly introduced (Cao-Lormeau et al., 2016; World Health Organization, 2015). Many researchers believe that ZIKV is responsible for these increases, suggesting that ZIKV is a more serious disease than initially realized.

In December 2015, the European Centre for Disease Prevention and Control issued a comprehensive update on the possible association between ZIKV, congenital microcephaly and Guillain-Barre (European Centre for Disease Prevention and Control,

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2015). Most evidence, however, is correlative. In Brazil, for example, 2782 cases of microcephaly were reported in the year following ZIKV introduction, as compared with 147 cases and 167 cases in the two years prior to ZIKV arrival (Romero, 2015). Retrospective analysis of data from French Polynesia similarly uncovered an unusual number of babies born with neural defects during the height of the ZIKV outbreak (Vogel, 2016). Over this same period, French Polynesia also saw a spike in Guillain-Barre syndrome (FauciMorens, 2016; Oehler et al., 2014), as well as increases in a range of other neurologic conditions including meningitis, meningoencephalitis, and myelitis (Talan, 2016). More recently, a series of Latin American countries, including Brazil, Colombia, and Venezuela have observed similar upticks in the incidence of Guillain-Barre (World Health Organization, 2016a, World Health Organization, 2016b), consistent with the proposed relationship between this disorder and ZIKV infection.

In addition to correlative support, several clinical and lab-based findings hint at potential mechanisms to explain the link between ZIKV and neural complications (Mlakar et al., 2016). In 1952, for example, Dick et al (Dick, Kitchen, & Haddow, 1952), demonstrated ZIKV tropism to the brain in intraperitoneally infected mice. Expanding on this finding, Bell, and colleagues (Bell, Field, & Narang, 1971) later showed that both neurons and glia could be infected by ZIKV. More recently, a number of studies, have demonstrated evidence of intrauterine infection with ZIKV (Oliveira Melo et al., 2016), including infection of the fetal brain (Martines, 2015; Rubin, Greene, & Baden, 2016). This latter finding, in particular, provides a direct path from maternal ZIKV infection to microcephaly – a rare neurological condition in which an infant’s brain develops abnormally in the womb or does not grow as it should after birth (Mayo Foundation for Medical Education and Research, 2016). Ultimately, microcephaly results in an infant’s head size being significantly smaller than the heads of other children of the same age and sex (Mayo Foundation for Medical Education and Research, 2016). Although microcephaly can range from mild to severe, cases currently associated with the ZIKV outbreak in Brazil are notable for the level of damage observed in the brains of affected infants (da Silva et al., 1953; Talan, 2016). Furthermore, congenital Zika usually come with a wide spectrum of clinical features (da Silva et al., 1953).

In this paper, we develop and analyze a mathematical model for ZIKV. Our focus is multi-fold. First, we consider overall ZIKV transmission in the adult population. Second, we consider ZIKV transmission to infants, either directly by mosquitoes or else prior to birth through vertical transmission from the mother. Infant ZIKV cases may be particularly severe because central nervous system (CNS) infections in young children can cause long-term damage to the developing brain (Bundy, 2014, p. 221). Finally, we consider microcephaly rates, which we assume occur as a result of vertical transmission of ZIKV to the fetus during the early stages of pregnancy. The paper is organized as follows. The model is formulated in Section 2 and we investigate the theoretical properties of the Zika model with mother-to-child vertical transmission in Section 3. In Section 4, we assess the impact of the asymptomatic classes and identify key parameters with the most impact on disease burden in Section 5. We conduct numerical exploration of three control strategies in Section 6. The study results are discussed in Section 7.

2. Model formulation

We model the transmission dynamics of ZIKV using a compartmental framework. We consider two human populations consisting of adults and newly born babies as well as the vector population. The population of newly born babies consists of

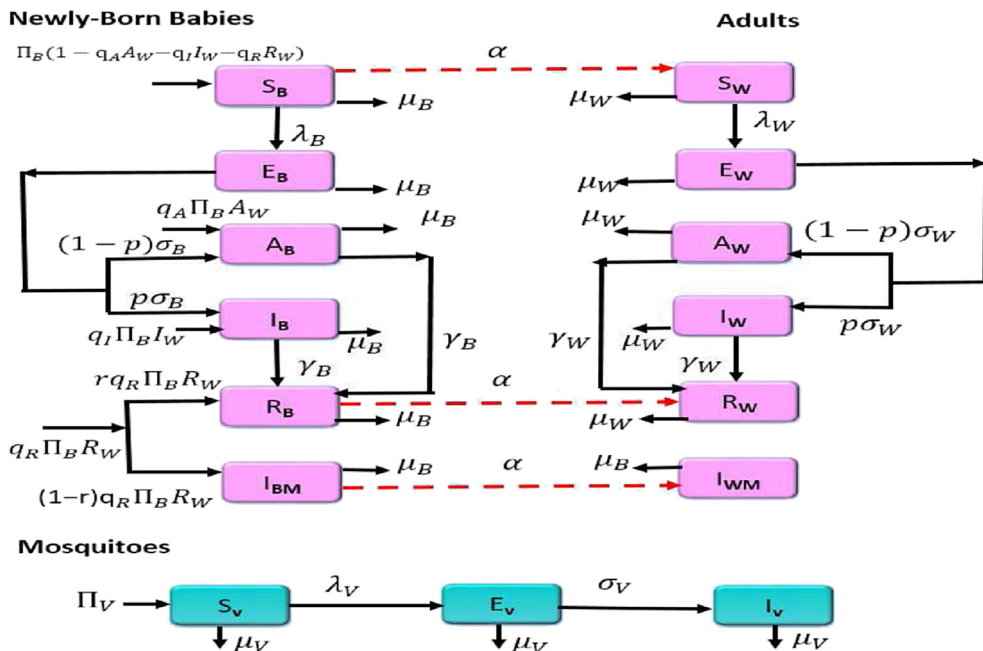


Fig. 1. Flow diagram of the Zika transmission model.

susceptible ($S_B(t)$), exposed ($E_B(t)$), asymptomatic ($A_B(t)$), symptomatic newly born without microcephaly, newly born with microcephaly ($I_{BM}(t)$) and recovered newly born babies ($R_B(t)$). The total population of adults, $N_W(t)$, at time t is split into mutually exclusive sub-populations of individuals who are susceptible ($S_W(t)$), exposed ($E_W(t)$), asymptomatic ($A_W(t)$), symptomatic ($I_W(t)$), adult with microcephaly ($I_{WM}(t)$) and recovered adults ($R_W(t)$). The population of the mosquitoes include the susceptible ($S_V(t)$), exposed ($E_V(t)$) and infected mosquitoes ($I_V(t)$). The total population for each group is given as:

$$\begin{aligned} N_B(t) &= S_B(t) + E_B(t) + A_B(t) + I_B(t) + I_{BM}(t) + R_B(t), \\ N_W(t) &= S_W(t) + E_W(t) + A_W(t) + I_W(t) + I_{WM}(t) + R_W(t), \\ N_V(t) &= S_V(t) + E_V(t) + I_V(t). \end{aligned}$$

and total human population is $N_H(t) = N_B + N_W$. Equations representing the mathematical model are given below. The flow diagram of the model is depicted in Fig. 1, and the associated state variables and parameters are described in Table 1.

$$\begin{aligned} S'_B(t) &= \pi_B - q_A \pi_B A_W(t) - q_I \pi_B I_W(t) - q_R \pi_B R_W(t) - \lambda_B(I_V, N_B) S_B(t) - (\alpha + \mu_B) S_B(t) \\ E'_B(t) &= \lambda_B(I_V, N_B) S_B(t) - (\alpha + \sigma_B + \mu_B) E_B(t) \\ A'_B(t) &= q_A \pi_B A_W(t) + (1 - p) \sigma_B E_B(t) - (\alpha + \gamma_B + \mu_B) A_B(t) \\ I'_B(t) &= q_I \pi_B I_W(t) + p \sigma_B E_B(t) - (\alpha + \gamma_B + \mu_B) I_B(t) \\ I'_{BM}(t) &= r q_R \pi_B R_W(t) - (\alpha + \mu_B) I_{BM}(t) \\ R'_B(t) &= (1 - r) q_R \pi_B R_W(t) + \gamma_B A_B(t) + \gamma_B I_B(t) - (\alpha + \mu_B) R_B(t) \\ S'_W(t) &= \alpha S_B(t) - \lambda_W(I_V, N_W) S_W(t) - \mu_W S_W(t) \\ E'_W(t) &= \lambda_W(I_V, N_W) S_W(t) - (\sigma_W + \mu_W) E_W(t) \\ A'_W(t) &= (1 - p) \sigma_W E_W(t) - (\gamma_W + \mu_W) A_W(t) \\ I'_W(t) &= p \sigma_W E_W(t) - (\gamma_W + \mu_W) I_W(t) \\ I'_{WM}(t) &= \alpha I_{BM}(t) - \mu_W I_{WM}(t) \\ R'_W(t) &= \alpha R_B(t) + \gamma_W A_W(t) + \gamma_W I_W(t) - \mu_W R_W(t) \\ S'_V(t) &= \pi_V - \lambda_V(A_B, I_B, A_W, I_W, N_B, N_W) S_V(t) - \mu_V S_V(t) \\ E'_V(t) &= \lambda_V(A_B, I_B, A_W, I_W, N_B, N_W) S_V(t) - (\mu_V + \sigma_V) E_V(t) \\ I'_V(t) &= \sigma_V E_V(t) - \mu_V I_V(t). \end{aligned} \tag{2.1}$$

where, (') represent derivative with respect to t , and

$$\begin{aligned} \lambda_W(I_V, N_W) &= \frac{\beta_W b_V I_V}{N_W}, \quad \lambda_B(I_V, N_B) = \frac{\eta \beta_B b_V I_V}{N_B}, \\ \lambda_V(A_B, I_B, A_W, I_W, N_B, N_W) &= \beta_V b_V \left[\frac{I_W + \rho_W A_W + \eta(I_B + \rho_B A_B)}{N_W + \eta N_B} \right], \end{aligned}$$

are the disease forces of infection rates, and all other parameters are as defined in Table 1. In particular, β_W, β_B and β_V are the transmission probability *per contact* in adults, newly born babies and mosquitoes, b_V is the mosquito biting rate, ρ_W and ρ_B are modification parameters modeling the infectivity of the asymptomatic babies and adults. The parameter η is a modification parameter that indicates that babies' exposure rate is different from that of adults. For instance, they may be protected from mosquito bites, making them less likely to get the infections, on the other hand, they may receive more mosquito bites if left unprotected; we assume that $\eta > 0$. We assume that the infection in the asymptomatic individuals might not be high enough to infect the susceptible mosquitoes or is the same level as for the infectious individuals, in which case the modification parameters are taken as $0 \leq \rho_B, \rho_W \leq 1$.

Zika virus is passed prenatally from a pregnant woman to her unborn fetus (Moore et al., 2017). For example, during the 2015 Zika outbreak in Brazil, Zika virus RNA was found in the amniotic fluid of two women whose fetuses were determined *via* prenatal ultrasound to have microcephaly (Schuler-Faccini et al., 2016). Depending on timing of infection in the womb, newborn babies can also be infected from birth (Besnard, Lastere, Teissier, Cao-Lormeau, & Musso, 2014). Thus, we assume that some babies are born with infected with the virus. The parameters q_A, q_I, q_R represent fractions of newly born babies who are infected due to vertical transmission. So that the fraction $\pi_B(1 - q_A A_W - q_I I_W - q_R R_W)$ are babies born healthy by infected and recovered mothers and the remaining fraction $q_A \pi_B A_W, q_I \pi_B I_W, q_R \pi_B R_W$ are born infected.

Despite the fact that there is sufficient evidence to conclude that intrauterine Zika virus infection is a cause of microcephaly (Moore et al., 2017), not all newly born babies are born with microcephaly, although they may have other congenital abnormalities (da Silva et al., 1953). We assume that some babies are born recovered from the virus. Thus, the parameter r correspond to the fraction of the $q_R \pi_B R_W$ recovered babies born by recovered mothers, while the remaining portion $(1 - r) q_R \pi_B R_W$ are newly born babies who have microcephaly. The parameter p represent the fraction of adults and newly born babies who are asymptomatic and the remaining fraction $(1 - p)$ are adults and newly born babies who are infectious. The parameter α denotes the maturation rate. Microcephalic individuals experience profound developmental delay (Carter, Mirzaa, McDonnell, & Boycott, 2013); although their lifespan is not known, they live for a short period due to severe neurologic impairments (some have been known to live up to 9 years) (Carter et al., 2013). As a result, we assume that microcephalic adults do not reproduce.

Table 1
Description of the state variables and parameters of the Zika model (2.1).

| Variable | Description |
|------------------------|----------------------------------------------------------------------------------|
| $S_B(t), S_W(t)$ | Susceptible newly born babies and adults |
| $E_B(t), E_W(t)$ | Exposed newly born babies and adults |
| $A_B(t), A_W(t)$ | Asymptomatic newly born babies and adults |
| $I_B(t), I_W(t)$ | Symptomatic newly born without microcephaly and adults |
| $I_{BM}(t), I_{WM}(t)$ | Microcephalic newly born babies and adults |
| $R_B(t), R_W(t)$ | Recovered newly born babies and adults |
| $S_V(t)$ | Susceptible female mosquitoes |
| $E_V(t)$ | Exposed female mosquitoes |
| $I_V(t)$ | Infected female mosquitoes |
| Parameter | Description |
| π_B | Birth rate newly born babies |
| p | Fraction of adults and newly born babies who are asymptomatic |
| $1 - p$ | Remaining fraction of adults and newly born babies who are infectious |
| α | Maturation rate |
| r, q_A, q_I, q_R | Fractions of newly born babies who are infected and have microcephaly |
| $1 - r$ | Remaining fraction of newly born babies who have microcephaly |
| η | Modification parameter |
| β_W, β_B | Transmission probability <i>per</i> contact of adults and newly born babies |
| ρ_W, ρ_B | Infectivity modification parameters in asymptomatic adults and newly born babies |
| σ_W, σ_B | Progression rate of exposed adults and newly born babies |
| γ_W, γ_B | Recovery rate of asymptomatic and symptomatic adults and newly born babies |
| μ_W, μ_B | Natural death rate of adults and newly born babies |
| π_V | Recruitment rate of mosquitoes |
| β_V | Transmission probability <i>per</i> contact of susceptible mosquitoes |
| b_V | Mosquito biting rate |
| σ_V | Progression rate of exposed mosquitoes |
| μ_V | Natural death rate of mosquitoes |

2.1. Basic properties

We shall now explore the basic dynamical features of model (2.1). Since the model (2.1) describes both human and mosquito populations during a Zika epidemic, it will only be epidemiologically meaningful if all state variables are non-negative for $t \geq 0$. That is, its solution with positive initial data will remain positive for all time ($t > 0$).

Lemma 1. *Let the initial data $F(0) \geq 0$, where $F(t) = (S_B, E_B, A_B, I_B, I_{BM}, S_W, E_W, A_W, I_W, R_W, I_{WM}, R_B, S_V, E_V, I_V)$. Then the solutions $F(t)$ of model (2.1) are non-negative for all time $t > 0$. Furthermore*

$$\limsup_{t \rightarrow \infty} N_H(t) \leq \frac{\pi_B}{\mu_H}, \quad \limsup_{t \rightarrow \infty} N_V(t) = \frac{\pi_V}{\mu_V},$$

where $\mu_H = \min\{\mu_B, \mu_W\}$.
with,

$$\begin{aligned} N_B(t) &= S_B(t) + E_B(t) + A_B(t) + I_B(t) + I_{BM}(t) + R_B(t), \\ N_W(t) &= S_W(t) + E_W(t) + A_W(t) + I_W(t) + I_{WM}(t) + R_W(t), \\ N_V(t) &= S_V(t) + E_V(t) + I_V(t). \end{aligned}$$

The proof of Lemma 1 is given in [Appendix A](#).

Invariant regions

Model (2.1) will be analyzed in a biologically-feasible region as follows. Consider the feasible region

$$\Gamma = \Gamma_H \times \Gamma_V \subset \mathbb{R}_+^{12} \times \mathbb{R}_+^3$$

with,

$$\begin{aligned} \Gamma_H &= \left\{ S_B(t), E_B(t), A_B(t), I_B(t), I_{BM}(t), R_B(t), S_W(t), E_W(t), A_W(t), I_W(t), I_{WM}(t), R_W(t) \right. \\ &\quad \left. : N_H(t) = \frac{\pi_B}{\mu_H} \right\}, \end{aligned}$$

$$\Gamma_V = \left\{ S_V(t), E_V(t), I_V(t) : N_V(t) = \frac{\pi_V}{\mu_V} \right\}.$$

Lemma 2. The region $= \Gamma_H \times \Gamma_V \subset \mathbb{R}_+^{12} \times \mathbb{R}_+^3$ is positively invariant for the basic model (2.1) with non-negative initial conditions in \mathbb{R}_+^{15}

The proof of Lemma 2 is given in Appendix B.

In the next section the conditions for the stability of the disease-free equilibrium of model (2.1) are explored.

3. Analysis of the model

Model (2.1) with endogenous reactivation and exogenous reinfection is now analyzed to gain insight into its dynamical features.

3.1. Local stability of the disease-free equilibrium

The disease free equilibrium (DFE) of model (2.1), which is obtained by setting the right hand sides of the model equations to zero is given by:

$$\begin{aligned} \mathcal{E}_0 &= (S_B^*, E_B^*, A_B^*, I_B^*, I_{BM}^*, R_B^*, S_W^*, E_W^*, A_W^*, I_W^*, I_{WM}^*, R_W^*, S_V^*, E_V^*, I_V^*) \\ &= \left(\frac{\pi_B}{\alpha + \mu_B}, 0, 0, 0, 0, 0, \frac{\alpha \pi_B}{\mu_W(\alpha + \mu_B)}, 0, 0, 0, 0, 0, \frac{\pi_V}{\mu_V}, 0, 0 \right). \end{aligned} \tag{3.1}$$

The local asymptotic stability of \mathcal{E}_0 can be established using the next generation operator method on the system (2.1). Taking the infected compartments $(E_B^*, A_B^*, I_B^*, I_{BM}^*, E_W^*, A_W^*, I_W^*, I_{WM}^*, E_V^*, I_V^*)$ at the DFE and using the notation in (van den Driessche & Watmough, 2002), the Jacobian matrices F and V for the new infection terms and the remaining transfer terms are respectively given by,

$$F = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{\eta \beta_B b_V S_B^*}{N_B^*} \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{\beta_W b_V S_W^*}{N_W^*} \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & \frac{\eta \rho_B \beta_V b_V S_V^*}{S_W^* + \eta S_B^*} & \frac{\eta \beta_V b_V S_V^*}{S_W^* + \eta S_B^*} & 0 & 0 & \frac{\rho_W \beta_V b_V S_V^*}{S_W^* + \eta S_B^*} & \frac{\beta_V b_V S_V^*}{S_W^* + \eta S_B^*} & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

and

$$V = \begin{pmatrix} k_2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -(1-p)\sigma_B & k_3 & 0 & 0 & 0 & -q_A \pi_B & 0 & 0 & 0 & 0 & 0 \\ -p\sigma_B & 0 & k_4 & 0 & 0 & 0 & -q_I \pi_B & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & k_5 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & k_8 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & -(1-p)\sigma_W & k_9 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & -p\sigma_W & 0 & k_{10} & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & -\alpha & 0 & 0 & 0 & k_{12} & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & k_{14} & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & -\sigma_V & \mu_V \end{pmatrix},$$

where $k_1 = \alpha + \mu_B$, $k_2 = \alpha + \sigma_B + \mu_B$, $k_3 = \alpha + \gamma_B + \mu_B$, $k_4 = \alpha + \gamma_B + \mu_B$, $k_5 = \alpha + \mu_B$, $k_6 = \alpha + \mu_B$, $k_7 = \mu_W$, $k_8 = \sigma_W + \mu_W$, $k_9 = \gamma_W + \mu_W$, $k_{10} = \gamma_W + \mu_W$, $k_{11} = \mu_W$, $k_{12} = \mu_W$, $k_{13} = \mu_V$, $k_{14} = \mu_V + \sigma_V$.

It follows that the basic reproduction number of model (2.1), denoted by \mathcal{R}_0 , is given by;

$$\mathcal{R}_0 = \rho(FV^{-1}) = \sqrt{\mathcal{R}_V(\mathcal{R}_W + \mathcal{R}_B)} \tag{3.2}$$

where

$$\begin{aligned} \mathcal{R}_V &= \frac{S_V^* \beta_V b_V \sigma_V}{k_{14} \mu_V}, \\ \mathcal{R}_B &= \frac{\eta^2 \beta_B b_V \sigma_B [\rho_B (1-p) k_4 + p k_3]}{k_2 k_3 k_4 (S_W^* + \eta S_B^*)}, \\ \mathcal{R}_W &= \frac{\beta_W b_V \sigma_W}{k_8 (S_W^* + \eta S_B^*)} \left[\frac{(1-p)(k_3 \rho_W + \eta \rho_B q_A \pi_B)}{k_3 k_9} + \frac{p(k_4 + \eta q_I \pi_B)}{k_4 k_{10}} \right]. \end{aligned}$$

The following result is established using Theorem 2 in (van den Driessche & Watmough, 2002).

Lemma 3. *The DFE of model (2.1), given by \mathcal{E}_0 , is locally asymptotically stable (LAS) if $\mathcal{R}_0 < 1$, and unstable if $\mathcal{R}_0 > 1$.*

The epidemiological quantity, \mathcal{R}_0 gives the average number of ZIKV cases generated by a typical infected individual introduced into an entirely susceptible human population (Anderson and May, 1991; Diekmann, Heesterbeek, & Metz, 1990; Hethcote, 2000; van den Driessche & Watmough, 2002). Furthermore, the expression \mathcal{R}_B is the number of secondary infections in newly born babies by one introduced infectious mosquito, while the expression \mathcal{R}_W is the number of secondary infections in adults by one infectious mosquito. The expression \mathcal{R}_W consists of infections from newly born babies due to vertical transmission (mother-to-child infection) and infections due to horizontal transmissions from adults and infections from infants that have matured into adults. Lastly, the expression \mathcal{R}_V is the number of secondary infections in mosquitoes resulting from a newly introduced infectious adult woman and newly born baby. ZIKV can be adequately controlled in the community with adults and newly born babies if the threshold quantity (\mathcal{R}_0) can be reduced to (and maintained at) a value less than unity (i.e. $\mathcal{R}_0 < 1$).

3.2. Global asymptotic stability of the disease-free equilibrium

For Zika elimination to be independent of the initial sizes of the sub-populations of the model, the global asymptotic stability of the DFE must be established. This is what we consider next. Consider the feasible region

$$\Gamma_1 = \{X \in \Gamma : S_B \leq S_B^*, S_W \leq S_W^*, S_V \leq S_V^*\},$$

where, $X = R_W, S_B, E_B, A_B, I_B, I_{BM}, S_W, E_W, A_W, I_W, I_{WM}, R_B, S_V, E_V, I_V$.

Lemma 4. *The region Γ_1 is positively invariant for model (2.1)*

The proof of Lemma 4 is given in Appendix C.

Theorem 1. *The DFE, \mathcal{E}_0 , of model (2.1), is globally asymptotically stable (GAS) in Γ_1 whenever $\mathcal{R}_0 \leq 1$.*

The proof of Theorem 1 is given in Appendix D.

The above result shows that ZIKV will be eliminated from the community if the threshold quantity \mathcal{R}_0 can be brought to a value less than unity.

4. Assessing the impact of the asymptomatic classes

In this section we shall explore the impact of the asymptotically infected individuals since an estimated 80% of Zika infections do not show symptoms (Centers for Disease Control and Prevention, 2016; Duffy, Chen, Hancock, Powers, & Kool, 2009; Oster, 2016), and when infections lead to illness, the symptoms are usually mild.

Note, that we have elected to work with the square of the reproduction number, so our results remain tractable. The conclusion is not altered if the actual expression for the reproduction number is used.

Thus, differentiating the square of the basic reproduction number, \mathcal{R}_0^2 , given in (3.2), partially with respect to the asymptomatic modification parameters ρ_B and ρ_W , gives

$$\frac{\partial \mathcal{R}_0^2}{\partial \rho_B} = \frac{\sigma_V \beta_V b_V^2 S_V^* \eta (1-p) (\eta \sigma_B k_8 k_9 + q_A \pi_B \sigma_W \beta_W k_2)}{k_{14} \mu_V k_2 k_3 k_8 k_9 (S_W^* + \eta S_B^*)} > 0,$$

and

$$\frac{\partial \mathcal{R}_0^2}{\partial \rho_W} = \frac{\sigma_V \beta_V b_V^2 S_V^* (1-p) \beta_W \sigma_W}{k_{14} \mu_V k_8 k_9 (S_W + \eta S_B^*)} > 0.$$

This, implies, the square of *basic reproduction number*, \mathcal{R}_0^2 , is an increasing function of the parameters ρ_B and ρ_W . Thus, the disease burden in the community will increase as the infectivity of the asymptomatic individuals increases.

Furthermore, if we take the limit of \mathcal{R}_0^2 , as $\rho_B \rightarrow 1$ and $\rho_W \rightarrow 1$ (meaning, that infectivity of the asymptomatic individuals is the same as that of the infectious individuals), we have

$$\lim_{\rho_B \rightarrow 1, \rho_W \rightarrow 1} \mathcal{R}_0 = \sigma_V \beta_V b_V^2 S_V^* \left\{ \sigma_W \beta_W k_2 k_3 k_9 (k_4 + \eta q_1 \pi_B) p + \sigma_W \beta_W k_2 k_4 k_{10} (k_3 + \eta q_A \pi_B) (1-p) + \eta^2 \sigma_B \beta_B k_8 k_9 k_{10} [p k_3 + (1-p) k_4] \right\} / [k_{14} \mu_V k_2 k_3 k_4 k_8 k_9 k_{10} (S_W^* + \eta S_B^*)] > 0.$$

Thus, as the infectivity of the asymptomatic individuals increases, the disease burden increases, thereby increasing the number of Zika infected individuals in the community.

Fig. 2 shows a contour plot of the reproduction number \mathcal{R}_0 , as a function of the asymptomatic modification parameters ρ_B and ρ_W . This figure indicates that Zika burden in the community gets amplified as the level of infectivity of the asymptomatic individuals increases toward that of the infectious individuals.

The impact of the asymptomatic classes is further assessed by simulating the model (2.1) using various values for the modification parameters ρ_B and ρ_W and the same parameters in Table 1. The results obtained, depicted in Fig. 3, show, as expected, that the cumulative number of new cases generated by infectious mosquitoes to susceptible humans increases with increasing values of ρ_B and ρ_W (see Fig. 3(a)). Similarly, the cumulative number of new cases generated by infectious humans to susceptible mosquitoes increases with increasing values of the modification parameters (see Fig. 3(b)). Notice that 3(a) and 3(b) differ in the direction of spread, since one is infection from mosquito to host and the other is from host to mosquito.

Next, we evaluate the contributions of the asymptomatic and infectious individuals to the disease burden in the community. We observed from Table 2, that the percentage contributed by asymptomatic humans goes up as p_W and p_B go up.

5. Sensitivity analysis

The outputs of deterministic models are governed by the model input parameters, which may exhibit some uncertainty in their determination or selection. We employed a global sensitivity analysis to assess the impact of uncertainty and the sensitivity of the outcomes of the numerical simulations to variations in each parameter of the model (2.1) using Latin Hypercube Sampling (LHS) and partial rank correlation coefficients (PRCC). LHS is a stratified sampling without replacement technique which allows for an efficient analysis of parameter variations across simultaneous uncertainty ranges in each parameter (Blower & Dowlatabadi, 1994; Marino, Hogue, Ray, & Kirschner, 2008; McKay, Beckman, & Conover, 2000; Sanchez & Blower, 1997). PRCC measures the strength of the relationship between the model outcome and the parameters, stating the degree of the effect that each parameter has on the outcome (Blower & Dowlatabadi, 1994; Marino et al., 2008; McKay et al., 2000; Sanchez & Blower, 1997). Thus, sensitivity analysis determines the parameters with the most significant impact on the outcome of the numerical simulations of the model (Blower & Dowlatabadi, 1994; Marino et al., 2008; McLeod, Brewster, Gumel, & Slonowsky, 2006). To generate the LHS matrices, we assume that all the model parameters are uniformly

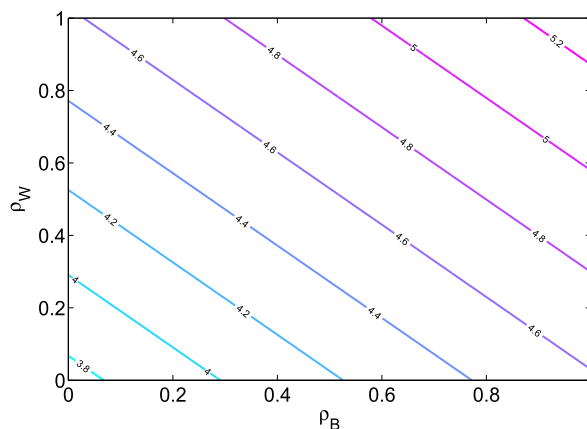


Fig. 2. Contour plot of the reproduction number (\mathcal{R}_0) of the Zika model (2.1) as a function of the asymptomatic modification parameters ρ_B and ρ_W . Parameter values used are as given in Table 3.

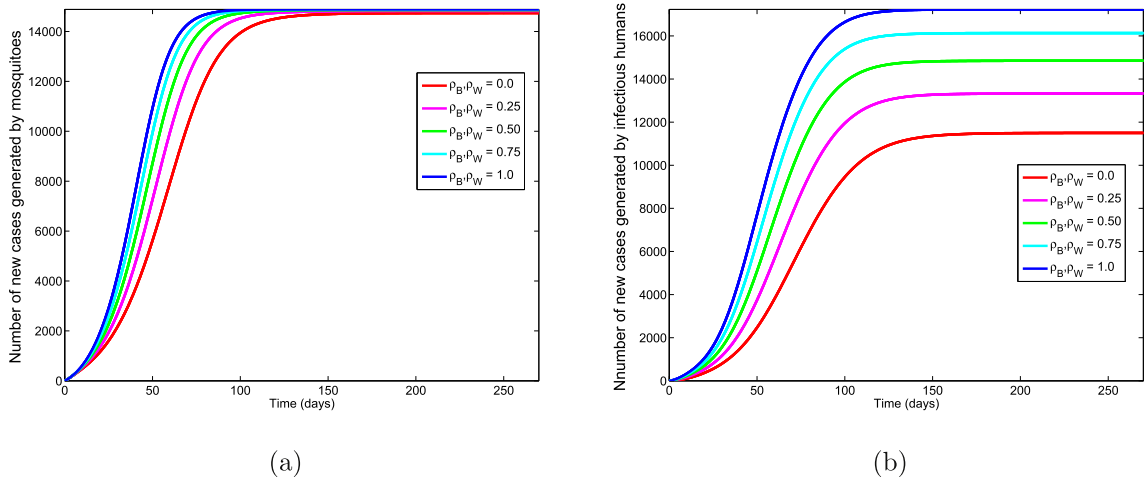


Fig. 3. Simulations of the Zika model (2.1) with different values of $\rho_B = \rho_W = 0, 0.25, 0.5, 0.75, 1.0$ (a). Cumulative number of new cases generated by infectious mosquitoes transmitting to susceptible humans. (b). Cumulative number of new cases generated by infectious humans transmitting to susceptible mosquitoes. Parameter values used are as given in Table 3.

Table 2

Contribution of the asymptomatic and infectious individuals to the mosquitoes' cumulative infections with various values of ρ_B and ρ_W .

| ρ_W, ρ_B | 0.25 | 0.50 | 0.75 | 1.0 |
|------------------|-------|-------|-------|-------|
| Asymptomatic | 20.5% | 34.1% | 43.7% | 50.9% |
| Infectious | 79.5% | 65.9% | 56.3% | 49.1% |

distributed. Then a total of 1000 simulations of the models *per* LHS run were carried out, using the ranges and baseline values tabulated in Table 3 (with the basic reproduction number, \mathcal{R}_0 , as the response function). From Fig. 4 it follows that the parameters that have the most influence on Zika transmission dynamics are the death rate of the mosquitoes (μ_V), the mosquito biting rate (b_V), mosquito recruitment rate (π_V), the transmission probability *per* contact to mosquitoes (β_V) and to adult humans (β_W), and the adult recovery rate (γ_W). Identification of these key parameters is important to the formulation of effective control strategies for combating the spread of disease. In particular, the results of this sensitivity analysis suggest that a strategy that reduces the transmission probability *per* contact to mosquitoes or to adult humans (i.e., reduces β_V or β_W respectively), will adequately reduce the spread of ZIKV in the community. Furthermore, a strategy that reduces the mosquito

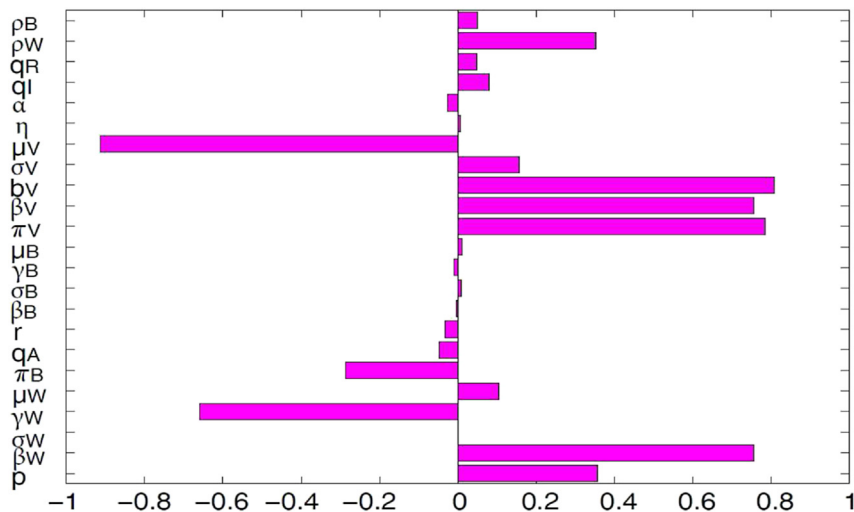


Fig. 4. PRCC values for the Zika model (2.1), using as response functions the reproduction number \mathcal{R}_0 . Parameter values (baseline) and ranges used are given in Table 3.

Table 3
Parameter values of model (2.1).

| Parameter | Values | Range | References |
|-----------------------|------------------------------|-------------------------------------------------------------|-------------------------------------------------------------------------------------------------------------------------------------------------------------|
| α | $\frac{1}{16 \times 365}$ | $\frac{1}{18 \times 365}$ – $\frac{1}{15 \times 365}$ | Assumed |
| ρ_B, ρ_W | 0.5 | 0.1–1 | Assumed |
| p, r, q_A, q_I, q_R | 0.5 | 0.1–1 | Assumed |
| β_W | 0.33 | 0.10–0.75 | (Manore, Hickmann, Xu, Wearing, & Hyman, 2014; Newton & Reiter, 1992; Paupy et al., 2010) |
| σ_W | $\frac{1}{7.5}$ | $\frac{1}{12}$ – $\frac{1}{3}$ | (Bewick, Fagan, Calabrese, & Augusto, 2016) |
| γ_W | 8.5 | $\frac{1}{14}$ – $\frac{1}{3}$ | (Bewick et al., 2016) |
| μ_W | $\frac{1}{70 \times 365}$ | $\frac{1}{76 \times 365}$ – $\frac{1}{68 \times 365}$ | (Manore et al., 2014) |
| π_B | $\frac{1}{15 \times 365}$ | $\frac{1}{18 \times 365}$ – $\frac{1}{12 \times 365}$ | (Wikipedia, 2015) |
| β_B | 0.33 | 0.001–0.54 | (Dumont & Chiroleu, 2010; Dumont, Chiroleu, & Domerg, 2008; Manore et al., 2014; Poletti et al., 2011; Turell, Beaman, & Tammariello, 1992) |
| σ_B | $\frac{1}{7.5}$ | $\frac{1}{12}$ – $\frac{1}{3}$ | (Bewick et al., 2016) |
| γ_B | 8.5 | $\frac{1}{14}$ – $\frac{1}{3}$ | (Bewick et al., 2016) |
| μ_B | $\frac{1}{18.60 \times 365}$ | $\frac{1}{14.88 \times 365}$ – $\frac{1}{22.32 \times 365}$ | (World Health Organization, 2016a, World Health Organization, 2016b) |
| π_V | 500 | 50–5000 | (Bewick et al., 2016) |
| β_V | 0.33 | 0.10–0.75 | (Manore et al., 2014; Newton & Reiter, 1992; Paupy et al., 2010) |
| b_V | 0.5 | 0.33–1.0 | (Manore et al., 2014; Putnam & Scott, 1995; Trpis & Haussermann, 1986) |
| σ_V | $\frac{1}{3.5}$ | $\frac{1}{6}$ – $\frac{1}{2}$ | (Dubrulle, Mousson, Moutailler, Vazeille, & Failloux, 2009; Dumont & Chiroleu, 2010; Moulay, Aziz-Alaoui, & Cadivel, 2011; Sebastian, Lodha, & Kabra, 2009) |
| μ_V | $\frac{1}{21}$ | $\frac{1}{42}$ – $\frac{1}{8}$ | (Sheppard, Macdonald, Tonn, & Grabs, 1969; Trpis & Haussermann, 1986; Trpis, Haussermann, & Craig, 1995) |

recruitment rate or the mosquito biting rate (i.e., reduces π_V or b_V , respectively) or else increases the death rate of the mosquitoes (i.e., increases μ_V) will be effective in curtailing the spread of ZIKV in the community.

6. Assessment of control strategies

Motivated by the sensitivity analysis in the previous section, we now explore some of the key model parameters to determine their effectiveness as targets for control strategies. In particular, we consider mosquito recruitment rate (π_V), which can be modified with larvicides or through effective management of mosquito breeding sites. We also consider the mosquito death rate, (μ_V), which can be modified with adulticides, and the mosquito biting rates, (b_V), which can be modified using repellants or behavioral avoidance (e.g. remaining in buildings with screened windows and air-conditioning). We then consider a combined strategy where all three management strategies are employed simultaneously. Finally, we consider delayed pregnancy, because this is a recommendation put forward by several Latin American and Caribbean governments as a means of reducing microcephaly.

For our analysis, moderate control levels correspond to the baseline parameter values used in our sensitivity analysis (Table 3, column 2) whereas low and high control levels correspond to the extreme parameter values (Table 3, column 3). The parameter values and initial conditions used in these simulations are theoretical in the sense that they are similar to comparable parameters for other mosquito-transmitted diseases, but are not specific to ZIKV, which is only poorly studied. The goal is thus to illustrate the control strategies proposed in this paper for relatively broad, generic parameter ranges. These predictions should then be retested as more information becomes available that is specific to ZIKV.

6.1. Epidemiological consequences of mosquito-reduction strategies

We consider the following control levels for this strategy:

1. Low mosquito-reduction strategy: $\pi_V = 500/day$, $\mu_V = \frac{1}{21}/day$;
2. Moderate mosquito-reduction strategy: $\pi_V = 250/day$, $\mu_V = \frac{1}{14}/day$;
3. High mosquito-reduction strategy: $\pi_V = 125/day$, $\mu_V = \frac{1}{8}/day$.

This strategy combines the larviciding and adulticiding strategies discussed in Appendices E and F, respectively.

Fig. 5 shows for each of the three control levels, the cumulative number of new infections in adults, new infections in newly born babies and newly born babies with microcephaly. Comparing the three control levels in Table 4 shows a decrease in the cumulative number of new cases with increasing control.

6.2. Epidemiological consequences of the personal protection strategy

Personal protection reduces mosquito biting rates (b_V). Similar to mosquito control, we consider three different levels of protection:

1. Low-effectiveness personal-protection strategy: $b_V = 0.5/day$;

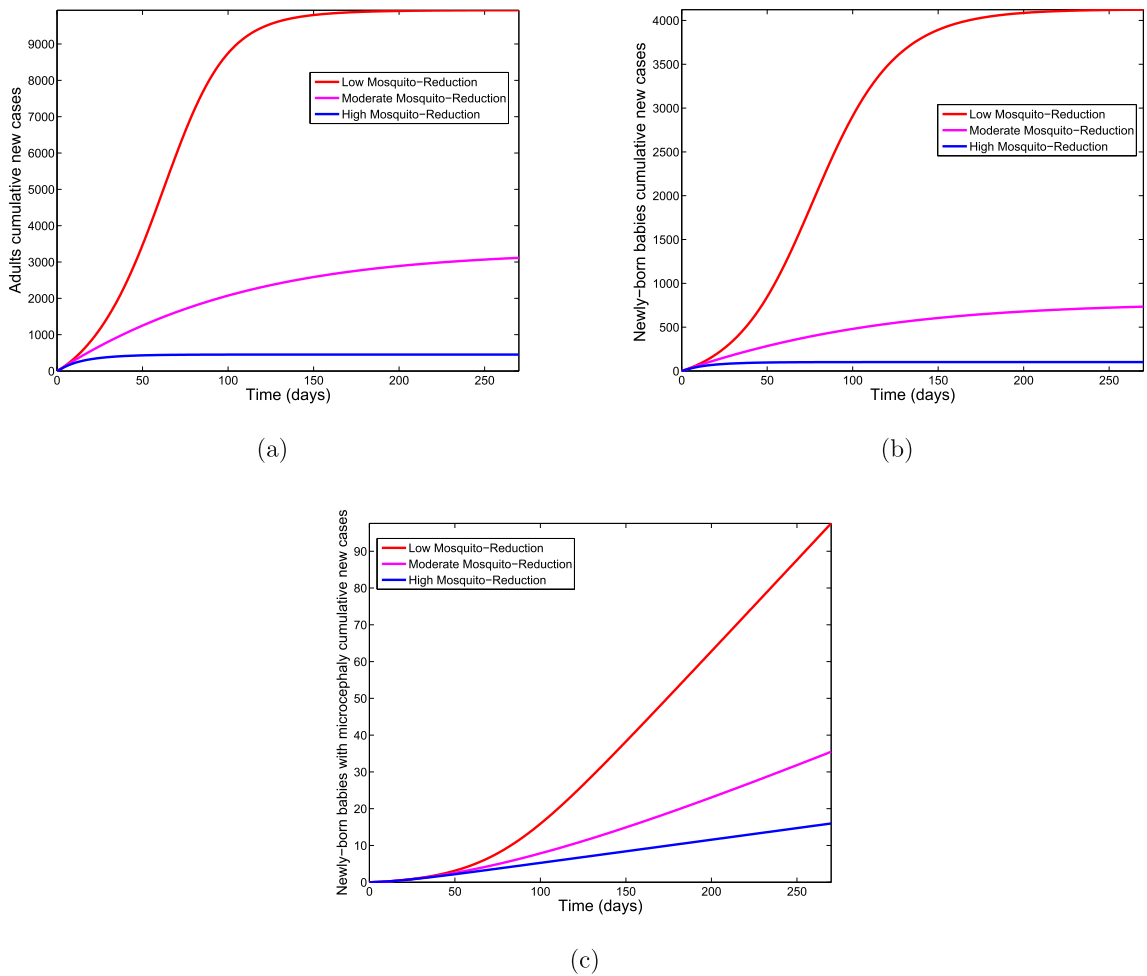


Fig. 5. Simulation of the Zika model (2.1) for various control levels of the mosquito reduction control strategy. (a). The cumulative number of new Zika cases in adults. (b). The cumulative number of new Zika cases in newly born babies. (c). The cumulative number of new cases of newly born babies with microcephaly. Parameter values used are as given in Table 3.

2. Moderate-effectiveness personal-protection strategy: $b_V = 0.250/day$;
3. High-effectiveness personal-protection strategy: $b_V = 0.125/day$.

Fig. 6 shows the cumulative number of new infections in adults, new infections in newly born babies and newly born babies with microcephaly for each of the three protection levels. The high effectiveness personal-protection strategy lead to a considerable reduction in the number of new cases compared to the moderate-effectiveness level (see Table 5) at the same time period. The low-effectiveness level performed the poorest producing the most number of new cases.

6.3. Combined mosquito control and personal protection strategy

The combined strategy (where both the mosquito reduction and personal protection strategies are implemented simultaneously) was assessed for the following three control levels:

1. Low-control strategy: $\pi_V = 500/day$, $\mu_V = \frac{1}{21}/day$, $b_V = 0.5/day$;
2. Moderate-control strategy: $\pi_V = 250/day$, $\mu_V = \frac{1}{14}/day$, $b_V = 0.25/day$;
3. High-control strategy: $\pi_V = 125/day$, $\mu_V = \frac{1}{8}/day$, $b_V = 0.125/day$.

Fig. 7 shows the cumulative number of new infections in adults, new infections in newly born babies, and babies with microcephaly for each control strategy. A comparison of the three control levels in Table 6 shows that higher levels of combined control are more effective for preventing new ZIKV cases.

Table 4
Simulation results of the Zika model (2.1) using the mosquito reduction control strategy.

| Humans | Low-Control | Moderate-Control | High-Control |
|------------------------------|--------------------|-------------------|-------------------|
| Adults | 2.1×10^6 | 5.8×10^5 | 1.2×10^5 |
| Newly born babies | 7.7×10^5 | 1.4×10^5 | 2.8×10^4 |
| Newly born with microcephaly | 10.0×10^3 | 3.9×10^3 | 2.0×10^3 |

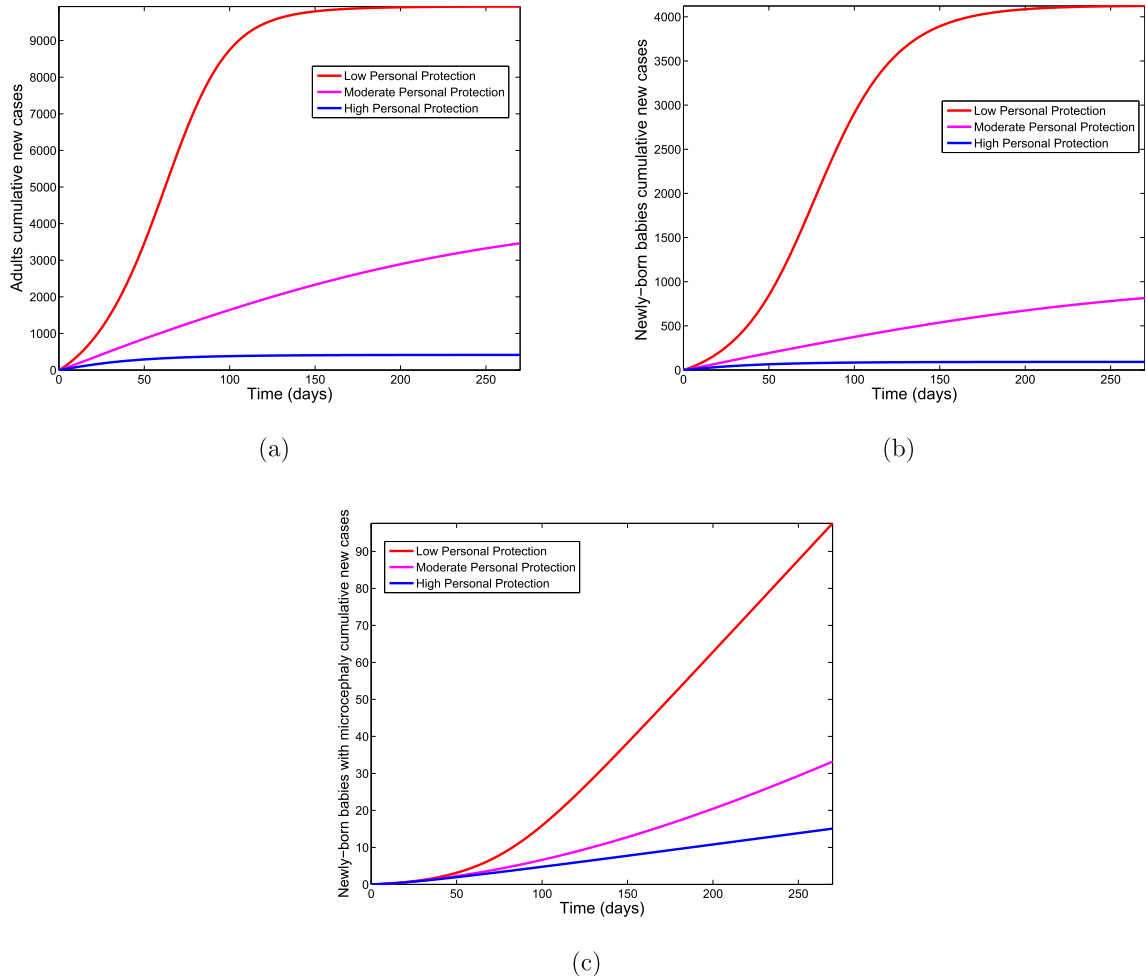


Fig. 6. Simulation of the Zika model (2.1) for various control levels of the personal-protection strategy. (a). The cumulative number of new Zika cases in adults. (b). The cumulative number of new Zika cases in newly born babies. (c). The cumulative number of new cases of newly born babies with microcephaly. Parameter values used are as given in Table 3.

A comparison across control strategies (larviciding, adulticiding, mosquito-reduction, personal-protection, and the combined strategy) in each group (see Table 7) shows, as expected, that the combined strategy is more effective than the other strategies implemented separately. Indeed, combining strategies results in anywhere from a 43% reduction to a 94% reduction as compared to single control strategies. With respect to single control strategies, personal protection is more effective than mosquito-reduction for reducing ZIKV and also for preventing microcephaly in newborns.

6.4. Delayed pregnancy

In light of the warnings issued by the Brazilian, Colombian, El Salvadorian, and Jamaican governments for reproductive women to delay conceiving (Ahmed, 2016; Darlington, 2016), we explore the impact that this will have on ZIKV transmission and the number of babies born with microcephaly. To consider delayed pregnancy, we adjusted the human birth rate π_B . As above, we consider three levels of delayed pregnancy:

Table 5

Simulation results of the cumulative number of new cases for the Zika model (2.1) using the personal-protection strategy.

| Humans | Low Control | Moderate Control | High Control |
|------------------------------|--------------------|-------------------|-------------------|
| Adults | 2.1×10^6 | 5.4×10^5 | 9.5×10^4 |
| Newly born babies | 7.7×10^5 | 1.2×10^5 | 2.1×10^4 |
| Newly born with microcephaly | 10.0×10^3 | 3.4×10^3 | 1.9×10^3 |

1. No women delay: $\pi_B = \frac{1}{15 \times 365} \text{ day}^{-1}$;
2. Some women delay: $\pi_B = \left(\frac{1}{15 \times 365}\right)/2 \text{ day}^{-1}$;
3. Many women delay: $\pi_B = \left(\frac{1}{15 \times 365}\right)/10 \text{ day}^{-1}$.

As expected, for all three scenarios, we observed a negligible difference in the cumulative number of new cases of ZIKV infections among adults and newly born babies. Indeed, there was even a small increase in the number of infected infants when pregnancy was delayed. However, there was a significant impact on the cumulative number of babies born with microcephaly, with the highly delayed rate producing the least number of babies with microcephaly (see Fig. 8). This is further demonstrated in Table 8, where we compared the three delayed conception rates levels.

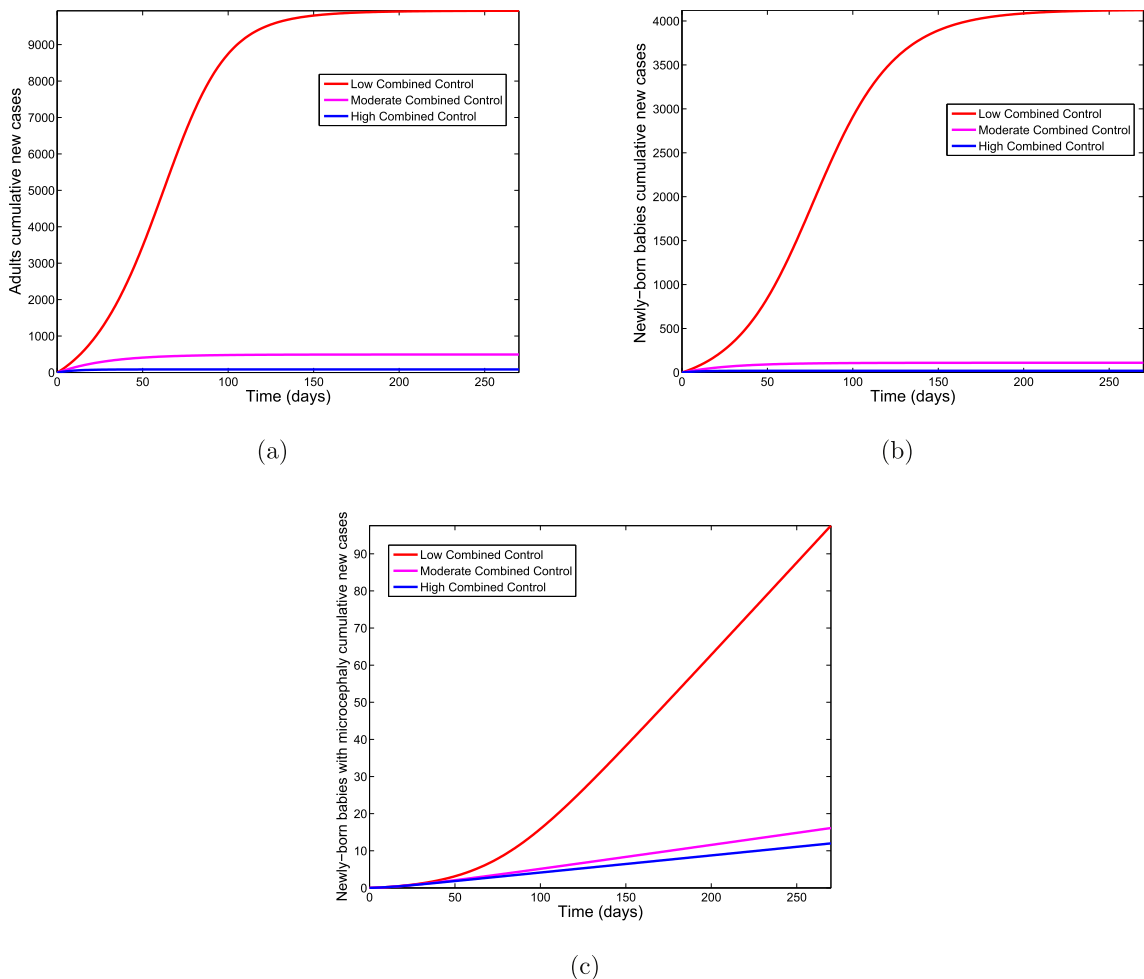


Fig. 7. Simulation of the Zika model (2.1) for various control levels of the combined control strategy. (a). The cumulative number of new Zika cases in adults. (b). The cumulative number of new Zika cases in newly born babies. (c). The cumulative number of new cases of newly born babies with microcephaly. Parameter values used are as given in Table 3.

Table 6

Simulation results of the cumulative number of new cases for the Zika model (2.1) using the combined control strategy.

| Humans | Low Control | Moderate Control | High Control |
|------------------------------|--------------------|-------------------|-------------------|
| Adults | 2.1×10^6 | 1.2×10^5 | 2.2×10^4 |
| Newly born babies | 7.7×10^5 | 2.6×10^4 | 4.9×10^3 |
| Newly born with microcephaly | 10.0×10^3 | 2.0×10^3 | 1.6×10^3 |

Table 7

Comparison of the cumulative number of new cases for the high-control levels of the various control strategies for the Zika model (2.1).

| Humans | Larviciding Control | Adulticiding Control | Mosquito Reduction | Personal Protection | Combined Control |
|------------------------------|---------------------|----------------------|--------------------|---------------------|-------------------|
| Adults | 7.5×10^5 | 2.6×10^5 | 1.2×10^5 | 9.5×10^4 | 2.2×10^4 |
| Newly born babies | 1.8×10^5 | 5.9×10^4 | 2.6×10^4 | 2.1×10^4 | 4.9×10^3 |
| Newly born with microcephaly | 4.4×10^3 | 2.7×10^3 | 2.0×10^3 | 1.9×10^3 | 1.6×10^3 |

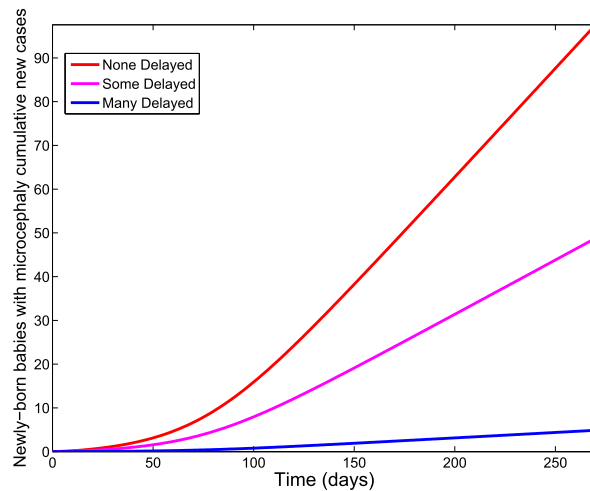


Fig. 8. Simulation of the Zika model (2.1) showing the cumulative number of new cases in newly born babies with microcephaly for various rates of delaying conception. Parameter values used are as given in Table 3.

As a final scenario, we implement delayed conception simultaneously with the combined mosquito control and personal protection strategies (see Section 6.3). Again, we consider low, moderate and high levels of mosquito control, personal-protection, and pregnancy delays. Unlike the findings with delayed pregnancy alone (see above), delayed pregnancy in combination with other control strategies result in an appreciable benefit on ZIKV transmission in adult and new-born populations (see Fig. 9 and Table 9). However, when large numbers of women delay conception, and this is combined with high levels of mosquito control and personal protection, there is a dramatic reduction in babies born with microcephaly (see Table 10). Thus, delayed pregnancy, particularly when combined with mosquito control and personal protection, appears to be beneficial for reducing microcephaly rates in regions with ongoing ZIKV outbreaks.

7. Discussion and conclusion

In this paper, we develop a new deterministic model to study the transmission dynamics of Zika virus. Our model incorporates mother-to-child transmission as well as the development of microcephaly in newly born babies. The analysis shows that the disease-free equilibrium of the model is locally and globally asymptotically stable whenever the associated reproduction number (\mathcal{R}_0), is less than unity and unstable otherwise. Sensitivity analysis further identifies parameters with

Table 8

Simulation results of the cumulative number of new cases for the Zika model (2.1) with various rates of delaying conception.

| Humans | None-delayed | Some-delayed | Many-delayed |
|------------------------------|--------------------|-------------------|-------------------|
| Adults | 2.1×10^6 | 2.1×10^6 | 2.1×10^6 |
| Newly born babies | 7.7×10^5 | 7.8×10^5 | 7.8×10^5 |
| Newly born with microcephaly | 10.0×10^3 | 5.0×10^3 | 499 |

Table 9

Simulation results of the cumulative number of new cases for the Zika model (2.1) using various rates of delaying conception, combined mosquito control and personal protection strategies.

| Humans | None-delayed & Low-Control | Some-delayed & Moderate-Control | Many-delayed & High-Control |
|------------------------------|----------------------------|---------------------------------|-----------------------------|
| Adults | 2.1×10^6 | 1.2×10^5 | 2.2×10^4 |
| Newly born babies | 7.7×10^5 | 2.6×10^4 | 4.8×10^3 |
| Newly born with microcephaly | 10.0×10^3 | 1.0×10^3 | 78.2 |

Table 10

Comparison of the combined control strategies (involving mosquito reduction and personal protection), and delayed pregnancy for the Zika model (2.1).

| Humans | Combined Control | Delayed Pregnancy | Delayed Pregnancy + Combined Control |
|------------------------------|-------------------|-------------------|--------------------------------------|
| Adults | 2.2×10^4 | 2.1×10^6 | 2.2×10^4 |
| Newly born babies | 4.9×10^3 | 7.8×10^5 | 4.8×10^3 |
| Newly born with microcephaly | 1.6×10^3 | 500 | 78.2 |

the strongest impact on model outcome (i.e., the basic reproduction number). These are the mosquito biting rate, the transmission probability *per* contact to mosquitoes and to human adults, the mosquito recruitment rate, and the mosquito death rate. Identification of these key parameters is vital to the formulation of effective ZIKV control strategies.

Based on our analysis, we consider various control strategies aimed at reducing mosquito biting rates, mosquito recruitment rates, and mosquito death rates to examine if these strategies will be effective in curtailing ZIKV spread in the community. We lack methods for reducing viral transmission probabilities; however, these would also be effective targets for control if

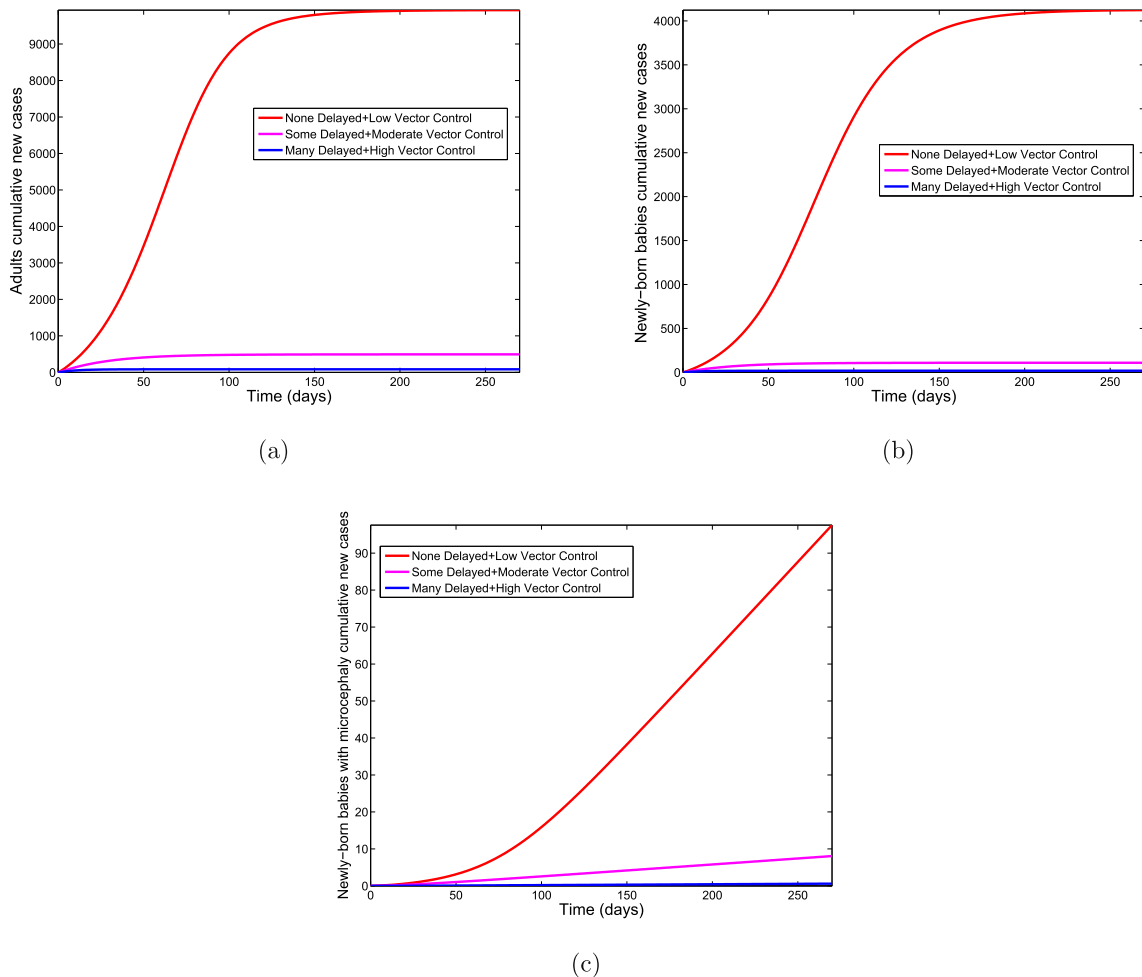


Fig. 9. Simulation of the Zika model (2.1) showing the cumulative number of new cases in newly born babies with microcephaly for various levels of delayed pregnancy. Parameter values used are as given in Table 3.

management strategies existed. Specifically, we implement a mosquito-reduction strategy, a personal-protection strategy, and a combined strategy each with three different control levels (low, moderate, high). Our results show that the cumulative number of new ZIKV cases generally decreases with increasing control as does the number of cases of microcephaly. As expected, the combined strategy is most effective across the board; this is followed by the personal protection strategy.

Some countries in the Western Hemisphere currently affected by ZIKV (including Brazil, El Salvador, Colombia and Jamaica (Ahmed, 2016; Darlington, 2016)) have issued warnings against women of reproductive age becoming pregnant. Using our model to numerically explore the effect of delayed pregnancy on disease transmission and microcephaly, we find that this strategy is highly effective for reducing the number of microcephaly cases, but does not impact levels of ZIKV transmission among either infants or adults. Coupling delayed pregnancy with mosquito control and personal protection, however, results in a considerable reduction in both ZIKV transmission and microcephaly. Thus, it appears that attacking ZIKV from all fronts, including aggressive mosquito control, strong adherence to repellent use, and delayed pregnancy, provides the best solution for ZIKV management, at least over the near term.

The issue of near-term versus long-term ZIKV management strategies is an important point that warrants further discussion. The model presented in the current paper is a highly detailed study of short-term ZIKV transmission. In particular, we restrict our analysis to a 9-month window beginning at the start of a ZIKV outbreak. Consequently, any babies born from women who are or who have ever been infected by ZIKV are at high-risk for microcephaly during this window. Over longer periods, however, ZIKV infection could occur prior to pregnancy. Consequently, a history of infection (i.e., recovery adults) in mothers would not necessarily indicate high-risk pregnancy. Quite the opposite - women who acquire infection before becoming pregnant may actually be protected during pregnancy. Thus, long-term predictions may differ substantially from predictions of the current model, and this is something that we explore in a separate paper (Bewick et al., 2016).

In this paper, we formulated and analyzed a system of ordinary differential equations for transmission dynamics of ZIKV. Some of our theoretical and epidemiological findings are summarized below:

- (i). Model (2.1) is locally and globally asymptotically stable when $\mathcal{R}_0 < 1$ and unstable when $\mathcal{R}_0 > 1$;
- (ii). Disease burden in the community increases infectivity of the asymptomatic individuals. Similarly, the cumulative infections to mosquitoes from the asymptomatic increases as their infectivity level increases.
- (iii). Sensitivity analysis shows that the most important parameters for ZIKV spread are the death rate of the mosquitoes (μ_V), the mosquito biting rate (b_V), mosquito recruitment rate (π_V), the transmission probability per contact to mosquitoes and to adult humans (β_V and β_W respectively) and the adult recovery rate (γ_W);
- (iv). Numerical simulations using mosquito control indicate that personal protection is a better and more effective strategy than the mosquito-reduction in reducing the disease burden in the population. As expected, a combined strategy is the most effective for reducing the ZIKV disease burden in the community.
- (v). Additional numerical simulations suggest that delaying conception reduces the number of cases of microcephaly, although it does little to prevent ZIKV transmission in the broader community. Coupled with aggressive mosquito control level and personal-protection; however, it is possible to both reduce microcephaly and prevent ZIKV transmission.

Acknowledgments

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Appendix A. Proof of Lemma 1

Lemma 1. Let the initial data $F(0) \geq 0$, where $F(t) = (S_B, E_B, A_B, I_B, I_{BM}, S_W, E_W, A_W, I_W, R_W, I_{WM}, R_B, S_V, E_V, I_V)$. Then the solutions $F(t)$ of model (2.1) are non-negative for all time $t > 0$. Furthermore

$$\limsup_{t \rightarrow \infty} N_H(t) = \frac{\pi_B}{\mu_H}, \limsup_{t \rightarrow \infty} N_W(t) = \frac{\pi_V}{\mu_V},$$

where, $\mu_H = \min\{\mu_B, \mu_W\}$.
with,

$$\begin{aligned} N_B(t) &= S_B(t) + E_B(t) + A_B(t) + I_B(t) + I_{BM}(t) + R_B(t), \\ N_W(t) &= S_W(t) + E_W(t) + A_W(t) + I_W(t) + I_{WM}(t) + R_W(t), \\ N_V(t) &= S_V(t) + E_V(t) + I_V(t). \end{aligned}$$

Proof. Let $t_1 = \sup\{t > 0 : F(t) > 0\}$. Thus, $t_1 > 0$. Then it follows from the first equation of the Zika model (2.1) that

$$\frac{dS_B}{dt} = \pi_B[1 - q_A A_W(t) - q_I I_W(t) - q_R R_W(t)] - \lambda_B(I_V, N_B)S_B(t) - (\alpha + \mu_B)S_B(t).$$

which can be re-written as

$$\frac{d}{dt} \left\{ S_B(t) \exp \left[\int_0^{t_f} \lambda_B(I_V(u), N_H(u)) du + (\alpha + \mu_B)t \right] \right\} = \int_0^{t_f} [\pi_B - q_A \pi_B A_W(u) - q_I \pi_B I_W(u) - q_R \pi_B R_W(u)] \times \exp \left[\int_0^{t_f} \lambda_B(I_V(u), N_H(u)) du + (\alpha + \mu_B)t \right].$$

So that,

$$S_B(t_1) \exp \left[\int_0^{t_1} \lambda_B(I_V(u), N_H(u)) du + \mu_W t \right] - S_B(0) = \int_0^{t_f} [\pi_B - q_A \pi_B A_W(t) - q_I \pi_B I_W(t) - q_R \pi_B R_W(t)] \times \exp \left[\int_0^{t_f} \lambda_B(I_V(u), N_H(u)) du + (\alpha + \mu_B)t \right].$$

Hence,

$$S_B(t_1) = S_B(0) \exp \left[- \int_0^{t_1} \lambda_B(I_V(u), N_H(u)) du - (\alpha + \mu_B)t_1 \right] = \int_0^{t_f} [\pi_B - q_A \pi_B A_W(u) - q_I \pi_B I_W(u) - q_R \pi_B R_W(u)] \times \exp \left[- \int_0^{t_1} \lambda_B(I_V(u), N_H(u)) du - (\alpha + \mu_B)t_1 \right]. > 0.$$

It can similarly be shown that $F > 0$ for all $t > 0$. For the second part of the proof, it should be noted that $0 < S_B(t) \leq N_H(t), 0 < E_B(t) \leq N_H(t), 0 < A_B(t) \leq N_H(t), 0 < I_B(t) \leq N_H(t), 0 < I_{BM}(t) \leq N_H(t), 0 < R_B(t) \leq N_H(t), 0 < S_W(t) \leq N_H(t), 0 < E_W(t) \leq N_H(t), 0 < A_W(t) \leq N_H(t), 0 < I_W(t) \leq N_H(t), 0 < I_{WM}(t) \leq N_H(t), 0 < R_W(t) \leq N_H(t), 0 < S_V(t) \leq N_V(t), 0 < E_V(t) \leq N_V(t), 0 < I_V(t) \leq N_V(t)$

Adding the components of the model (2.1) gives

$$\frac{dN_H(t)}{dt} = \pi_B - \mu_B N_B(t) - \mu_W N_W(t)$$

$$\frac{dN_V(t)}{dt} = \pi_V - \mu_V N_V(t).$$

Thus,

$$\frac{dN_H(t)}{dt} = \pi_B - \mu_H N_H(t),$$

$$\frac{dN_V(t)}{dt} = \pi_V - \mu_V N_V(t).$$

where $\mu_H = \min\{\mu_B, \mu_W\}$.

Hence,

$$\frac{\pi_B}{\mu_H} = \liminf_{t \rightarrow \infty} N_H(t) = \limsup_{t \rightarrow \infty} N_H(t) \leq \frac{\pi_B}{\mu_H},$$

$$\frac{\pi_V}{\mu_V} = \liminf_{t \rightarrow \infty} N_V(t) \leq \limsup_{t \rightarrow \infty} N_V(t) = \frac{\pi_V}{\mu_V},$$

as required.

Appendix B. Proof of Lemma 2

Lemma 2. *The region $\Gamma = \Gamma_H \times \Gamma_V \subset \mathbb{R}_+^{12} \times \mathbb{R}_+^3$ is positively invariant for the basic model (2.1) with non-negative initial conditions in \mathbb{R}_+^{15}*

Proof. The following steps are followed to establish the positive invariance of Γ (i.e., solutions in Γ remain in Γ for all $t > 0$). The rate of change of the total population is obtained by adding the components of the model (2.1) to give

$$\frac{dN_H(t)}{dt} = \pi_B - \mu_B N_B(t) - \mu_W N_W(t)$$

$$\frac{dN_V(t)}{dt} = \pi_V - \mu_V N_V(t).$$

Thus,

$$\frac{dN_H(t)}{dt} = \pi_B - \mu_H N_H(t),$$

$$\frac{dN_V(t)}{dt} = \pi_V - \mu_V N_V(t).$$

where $\mu_H = \min\{\mu_B, \mu_W\}$. A standard comparison theorem (Lakshmikantham, Leela, & Martynyuk, 1989) can then be used to show that $N_H(t) \leq N_H(0)e^{-\mu_H t} + \frac{\pi_B}{\mu_H}(1 - e^{-\mu_H t})$, in particular, $N_H(t) \leq \frac{\pi_B}{\mu_H}$, if $N_H(0) \leq \frac{\pi_B}{\mu_B}$. And lastly, $N_V(t) \leq N_V(0)e^{-\mu_V t} + \frac{\pi_V}{\mu_V}(1 - e^{-\mu_V t})$, in particular, $N_V(t) \leq \frac{\pi_V}{\mu_V}$, if $N_V(0) \leq \frac{\pi_V}{\mu_V}$.

Thus, the region Γ is positively invariant. Hence, it is sufficient to consider the dynamics of the flow generated by (2.1) in Γ . In this region, the model is epidemiologically and mathematically well-posed (Hethcote, 2000). Thus, every solution of the model (2.1) with initial conditions in Γ remains in Γ for all $t > 0$. Therefore, the Γ -limit sets of the system (2.1) are contained in Γ . This result is summarized below.

Appendix C. Proof of Lemma 4

Lemma 4. *The region Γ_1 is positively invariant for model (2.1)*

Proof. It follows from the first equation of Zika model (2.1), (where $S_B^* = \frac{\pi_B}{\alpha + \mu_B}$), so that

$$\begin{aligned} \frac{dS_B(t)}{dt} &= \pi_B - q_A \pi_B A_W(t) - q_I \pi_B I_W(t) - q_R \pi_B R_W(t) - \frac{\eta \beta_B b_V S_B I_V}{N_H} - (\alpha + \mu_B) S_B(t) \\ &\leq \pi_B - (\alpha + \mu_B) S_B(t) \\ &\leq (\alpha + \mu_B) \left[\frac{\pi_B}{(\alpha + \mu_B)} - S_B(t) \right] \\ &= (\alpha + \mu_B) [S_B^* - S_B(t)]. \end{aligned} \tag{C-1}$$

Thus,

$$S_B(t) \leq S_B^* - [S_B^* - S_B(0)] e^{-(\alpha + \mu_B)t}.$$

Thus, if $S_B(0) \leq S_B^*$ for all $t \geq 0$, then $S_B(t) \leq S_B^*$ for all $t \geq 0$.

Similarly, it follows from the sixth equation of the Zika model (2.1) (where $S_W^* = \frac{\alpha \pi_B}{(\alpha + \mu_B) \mu_W}$), that

$$\begin{aligned}
 \frac{dS_W(t)}{dt} &= \alpha S_B - \frac{\beta_W b_V S_W I_V}{N_H} - \mu_W S_W(t), \\
 &\leq \alpha S_B - \mu_W S_W(t), \\
 &\leq \mu_W \left[\frac{\alpha S_B^*}{\mu_W} - S_W(t) \right], \\
 &= \mu_W \left[\frac{\alpha \pi_B}{\mu_W (\alpha + \mu_B)} - S_W(t) \right] \\
 &= \mu_W [S_W^* - S_W(t)].
 \end{aligned}
 \tag{C-2}$$

Hence,

$$S_W(t) \leq S_W^* - [S_W^* - S_W(0)]e^{-\mu_W t}.$$

Thus, if $S_W(0) \leq S_W^*$ for all $t \geq 0$, then $S_W(t) \leq S_W^*$ for all $t \geq 0$.

Finally, it follows from the twelfth equation of the Zika model (2.1), that

$$\begin{aligned}
 \frac{dS_V(t)}{dt} &= \pi_V - \beta_V b_V \left[\frac{A_W + I_W + \eta(A_B + I_B)}{N_H} \right] S_M - \mu_M S_M(t), \\
 &\leq \pi_V - \mu_V S_V(t), \\
 &\leq \mu_V \left[\frac{\pi_V}{\mu_V} - S_V(t) \right], \\
 &= \mu_V [S_V^* - S_V(t)].
 \end{aligned}
 \tag{C-3}$$

Thus,

$$S_V(t) \leq S_V^* - [S_V^* - S_V(0)]e^{-\mu_V t}.$$

Hence, if $N_M^* = \frac{\pi_V}{\mu_V}$ and $S_V(0) \leq S_V^*$ for all $t \geq 0$, then $S_V(t) \leq S_V^*$ for all $t \geq 0$.

Thus, in summary, it has been shown that the region Γ_1 is positively invariant and attracts all solutions in \mathbb{R}_+^{14} for the Zika model (2.1).

Appendix D. Proof of Theorem 1

Theorem 1. *The DFE, \mathcal{E}_0 , of model (2.1), is globally asymptotically stable (GAS) in Γ_1 whenever $\mathcal{R}_0 \leq 1$.*

Proof. To prove the global stability of the disease-free equilibrium, we will following the approach in (Augusto et al., 2013).

Let $X = (S_W, R_W, S_B, R_B, S_V)$ and $Z = (E_B, A_B, I_B, I_{BM}, E_W, A_W, I_W, I_{WM}, E_V, I_V)$ and group the system (2.1) into

$$\begin{aligned}
 \frac{dX}{dt} &= F(X, 0) \\
 \frac{dZ}{dt} &= G(X, Z),
 \end{aligned}
 \tag{D-1}$$

where $F(X, 0)$ is the right hand side of S_W, R_W, S_B, R_B and S_V with $E_B = A_B = I_B = I_{BM} = E_W = A_W = I_W = I_{WM} = E_V = I_V = 0$ and $G(X, Z)$ is the right hand side of $E_W, A_W, I_W, E_B, A_B, I_B, I_{BM}, E_V$ and I_V .

Next, consider the reduced system: $\frac{dX}{dt} = F(X, 0)$ given as:

$$\begin{aligned}
 \frac{dS_B}{dt} &= \pi_B - q_R \pi_B R_W(t) - (\alpha + \mu_B) S_B(t) \\
 \frac{dR_B}{dt} &= (1 - r) q_R \pi_B R_W(t) - (\alpha + \mu_B) R_B(t) \\
 \frac{dS_W}{dt} &= \alpha S_B(t) - \mu_W S_W(t) \\
 \frac{dR_W}{dt} &= \alpha R_B(t) - \mu_W R_W(t) \\
 \frac{dS_V}{dt} &= \pi_V - \mu_V S_V(t).
 \end{aligned}
 \tag{D-2}$$

Let

$$X^* = (S_B^*, R_B^*, S_W^*, R_W^*, S_V^*) = \left(\frac{\pi_B}{\alpha + \mu_B}, 0, \frac{\alpha\pi_B}{\mu_W(\alpha + \mu_B)}, 0, \frac{\pi_V}{\mu_V} \right).$$

be an equilibrium of the reduced system (D-2), we show that X^* is a globally stable equilibrium in Γ_1 .

To do this, solve the first and second equations of (D-2), this gives

$$S_B(t) = \left[\int_0^t e^{(\alpha+\mu_B)u} \pi_B [1 - q_R R_W(u)] du + S_B(0) \right] e^{-(\alpha+\mu_B)t} \tag{D-3}$$

$$R_B(t) = \left[\int_0^t (1-r) q_R \pi_W R_W(u) e^{(\alpha+\mu_B)u} du + R_B(0) \right] e^{-(\alpha+\mu_B)t}.$$

Integrating $S_B(t)$ in (D-3) gives:

$$S_B(t) = \frac{\pi_B [1 - e^{-(\alpha+\mu_B)t}]}{(\alpha + \mu_B)} - \left[\int_0^t e^{(\alpha+\mu_B)u} \pi_B q_R R_W(u) du + S_B(0) \right] e^{-(\alpha+\mu_B)t}. \tag{D-4}$$

Taking the limit of $S_B(t)$ in (D-4) and $R_B(t)$ in (D-3) as $t \rightarrow \infty$, we have

$$\lim_{t \rightarrow \infty} S_B(t) = \frac{\pi_B}{(\alpha + \mu_B)} \text{ and } \lim_{t \rightarrow \infty} R_B(t) = 0.$$

Next, solving the third and fourth equations of (D-2) $S_W(t)$ and $R_W(t)$ gives

$$S_W(t) = \left[\int_0^t \alpha S_B(z) e^{\mu_W z} dz + S_W(0) \right] e^{-\mu_W t} \text{ and } R_W(t) = \left[\int_0^t R_B(z) e^{\mu_B z} dz + R_W(0) \right] e^{-\mu_W t}.$$

Integrating $S_W(t)$ and $R_W(t)$ and substituting $S_B(t)$ in (D-4) and $R_B(t)$ in (D-3) gives

$$S_W(t) = \left\{ \int_0^t \alpha \left[\frac{\pi_B (1 - e^{k_1 u})}{k_1} - \left(\int_0^t e^{k_1 z} \pi_B q_R R_W(z) dz \right) + S_B(0) \right] e^{-k_1 u} e^{k_7 t} dt + S_W(0) \right\} e^{-k_7 t},$$

$$R_W(t) = \left\{ \int_0^t \left[\alpha \left(\int_0^s R q \pi_W R_W(u) e^{k_6 u} du + R_B(0) \right) e^{-k_6 u} e^{k_{12} s} \right] dt + R_W(0) \right\} e^{-k_{12} t},$$

where, $k_1 = \alpha + \mu_B, k_6 = \alpha + \mu_B, k_7 = \mu_W, k_{12} = \mu_W$.

Taking the limit as $t \rightarrow \infty$,

$$\lim_{t \rightarrow \infty} S_W(t) = \frac{\alpha\pi_B}{\mu_W(\alpha + \mu_B)} \text{ and } \lim_{t \rightarrow \infty} R_W(t) = 0.$$

Similarly, solving for $S_V(t)$ in (D-2) gives $S_V(t) = \frac{\pi_V}{\mu_V} + e^{-\mu_V t} \left[S_V(0) - \frac{\pi_V}{\mu_V} \right]$ which converges to $\frac{\pi_V}{\mu_V}$, as $t \rightarrow \infty$.

These asymptotic dynamics are independent of initial conditions in Γ . Hence, the convergence of solutions of (D-2) is global in Γ_1 . Next, following (Castillo-Chavez, Blower, van den Driessche, Kirschner, & Yakubu, 2002), we require $G(X, Z)$ to satisfy the two stated conditions:

- (i). $G(X, 0) = 0$ and
- (ii). $G(X, Z) = D_Z G(X^*, 0)Z - \widehat{G}(X, Z), \widehat{G}(X, Z) \geq 0,$

where $(X^*, 0) = \left(\frac{\pi_B}{\alpha + \mu_B}, 0, \frac{\alpha \pi_B}{\mu_W(\alpha + \mu_B)}, 0, \frac{\pi_V}{\mu_V}, 0, 0, 0, 0, 0, 0, 0, 0, 0 \right)$ and $D_Z G(X^*, 0)$ is the Jacobian of $G(X, Z)$ taken with respect to $(E_B, A_B, I_B, I_{BM}, E_W, A_W, I_W, I_{WM}, E_V, I_V)$ and evaluated at $(X^*, 0)$, which is an M-matrix (the diagonal elements are nonnegative). Thus,

$$D_Z G(X^*, 0) = \begin{pmatrix} -k_2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \eta \Psi_B \\ (1-p)\sigma_B & -k_3 & 0 & 0 & 0 & q\pi_W & 0 & 0 & 0 & 0 \\ p\sigma_B & 0 & -k_4 & 0 & 0 & 0 & q\pi_W & 0 & 0 & 0 \\ 0 & 0 & 0 & -k_5 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & -k_8 & 0 & 0 & 0 & 0 & \Psi_W \\ 0 & 0 & 0 & 0 & (1-p)\sigma_W & -k_9 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & p\sigma_W & 0 & -k_{10} & 0 & 0 & 0 \\ 0 & 0 & 0 & \alpha & 0 & 0 & 0 & -k_{12} & 0 & 0 \\ 0 & \eta\rho_B \Psi_V & \eta \Psi_V & 0 & 0 & \rho_W \Psi_V & \Psi_V & 0 & -k_{14} & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \sigma_V & -\mu_V \end{pmatrix},$$

where, $\Psi_B = \beta_B b_V \frac{S_B^*}{N_B^*}$, $\Psi_W = \frac{\beta_W b_V S_W^*}{N_W^*}$, $\Psi_V = \frac{\beta_V b_V S_V^*}{N_H^*}$ and

$$\widehat{G}(X, Z) = \begin{pmatrix} \Phi_B I_V \\ 0 \\ 0 \\ 0 \\ \Phi_W I_V \\ 0 \\ 0 \\ 0 \\ \eta\rho_B \Phi_V A_W + \eta \Phi_V I_W + \rho_W \Phi_V A_B + \Phi_V I_B \\ 0 \end{pmatrix}$$

where, $\Phi_B = \frac{\beta_B b_V S_B^*}{N_B^*} \left(1 - \frac{S_B}{N_B} \frac{N_B^*}{S_B^*} \right)$, $\Phi_W = \frac{\beta_W b_V S_W^*}{N_W^*} \left(1 - \frac{S_W}{N_W} \frac{N_W^*}{S_W^*} \right)$ and $\Phi_V = \frac{\beta_V b_V S_V^*}{N_H^*} \left(1 - \frac{S_V}{N_H} \frac{N_H^*}{S_V^*} \right)$.

Furthermore, $S_W^* = \frac{\alpha \pi_B}{\mu_W(\alpha + \mu_B)}$, $S_B^* = \frac{\pi_B}{\alpha + \mu_B}$, $N_H^* = S_W^* + \eta S_B^*$ and $S_V^* = \frac{\pi_V}{\mu_V}$. We have in Γ_1 that, $S_W \leq S_W^*$, $S_B \leq S_B^*$, and $S_V \leq S_V^*$. Therefore, it follows that $N_H \leq N_H^*$.

Hence, if the human population is at equilibrium level, we have that $\left(1 - \frac{N_W^* S_W}{S_W^* N_W} \right) > 0$, $\left(1 - \frac{N_B^* S_B}{S_B^* N_B} \right) > 0$ and $\left(1 - \frac{S_V N_H^*}{N_H S_V^*} \right) > 0$; thus, $\widehat{G}(X, Z) \geq 0$.

Thus,

$$G(X, Z) = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \Phi_B I_V \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \Phi_W I_V \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & \eta\rho_B \Phi_V A_W & \eta \Phi_V I_W & 0 & 0 & \rho_W \Phi_V A_B & \Phi_V I_B & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

Therefore, the disease-free equilibrium is globally asymptotically stable by the theorem in (Castillo-Chavez et al., 2002) (page 246).

E. Mosquito-Larviciding Strategy

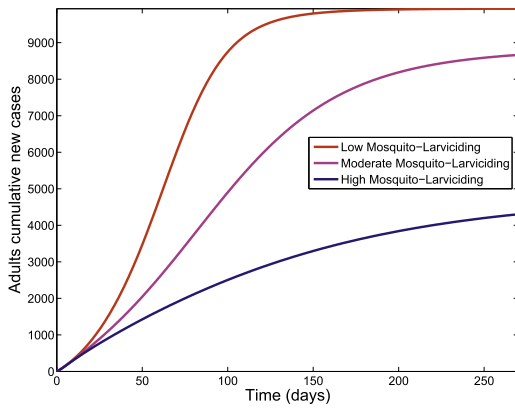
We consider the following three control levels:

1. Low mosquito-larviciding strategy: $\pi_V = 500/day$;
2. Moderate mosquito-larviciding strategy: $\pi_V = 250/day$;
3. High mosquito-larviciding strategy: $\pi_V = 125/day$.

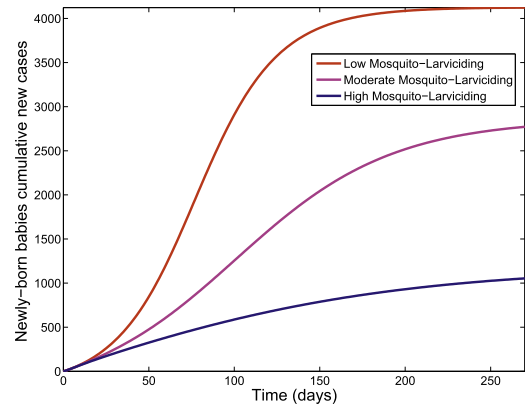
Fig. 10 shows the cumulative number of new infections in adults, new infections in newly born babies and newly born babies with microcephaly for each of the three control levels. Table 11 shows a clear decrease in the cumulative number of new cases with increasing control when comparing the three control levels.

Table 11
Simulation results of the cumulative number of new cases days for the Zika model (2.1) using mosquito-larviciding strategy.

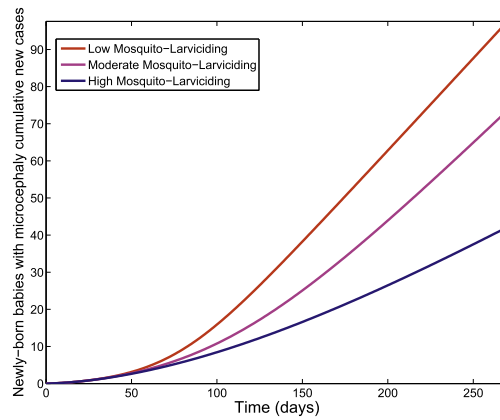
| Humans | Low-Control | Moderate-Control | High-Control |
|------------------------------|--------------------|-------------------|-------------------|
| Adults | 2.1×10^6 | 1.5×10^6 | 7.5×10^5 |
| Newly born babies | 7.7×10^5 | 4.4×10^5 | 1.8×10^5 |
| Newly born with microcephaly | 10.0×10^3 | 7.1×10^3 | 4.4×10^3 |



(a)



(b)



(c)

Fig. 10. Simulation of the Zika model (2.1) for various control levels of the larviciding strategy. (a). The cumulative number of new cases in adults (b). The cumulative number of new cases in newly born babies. (c). The cumulative number of new cases of newly born babies with microcephaly. Parameter values used are as given in Table 3.

F. Mosquito-Adulticiding Strategy

1. Low mosquito-adulticiding strategy: $\mu_V = \frac{1}{21}/day$;
2. Moderate mosquito-adulticiding strategy: $\mu_V = \frac{1}{14}/day$;
3. High mosquito-adulticiding strategy: $\mu_V = \frac{1}{8}/day$.

The cumulative number of new infections in adults, new infections in newly born babies and newly born babies with microcephaly for each of the three control levels is depicted in Fig. 11. A comparison of the three control levels in Table 12 shows a decrease in the cumulative number of new cases with increasing control.

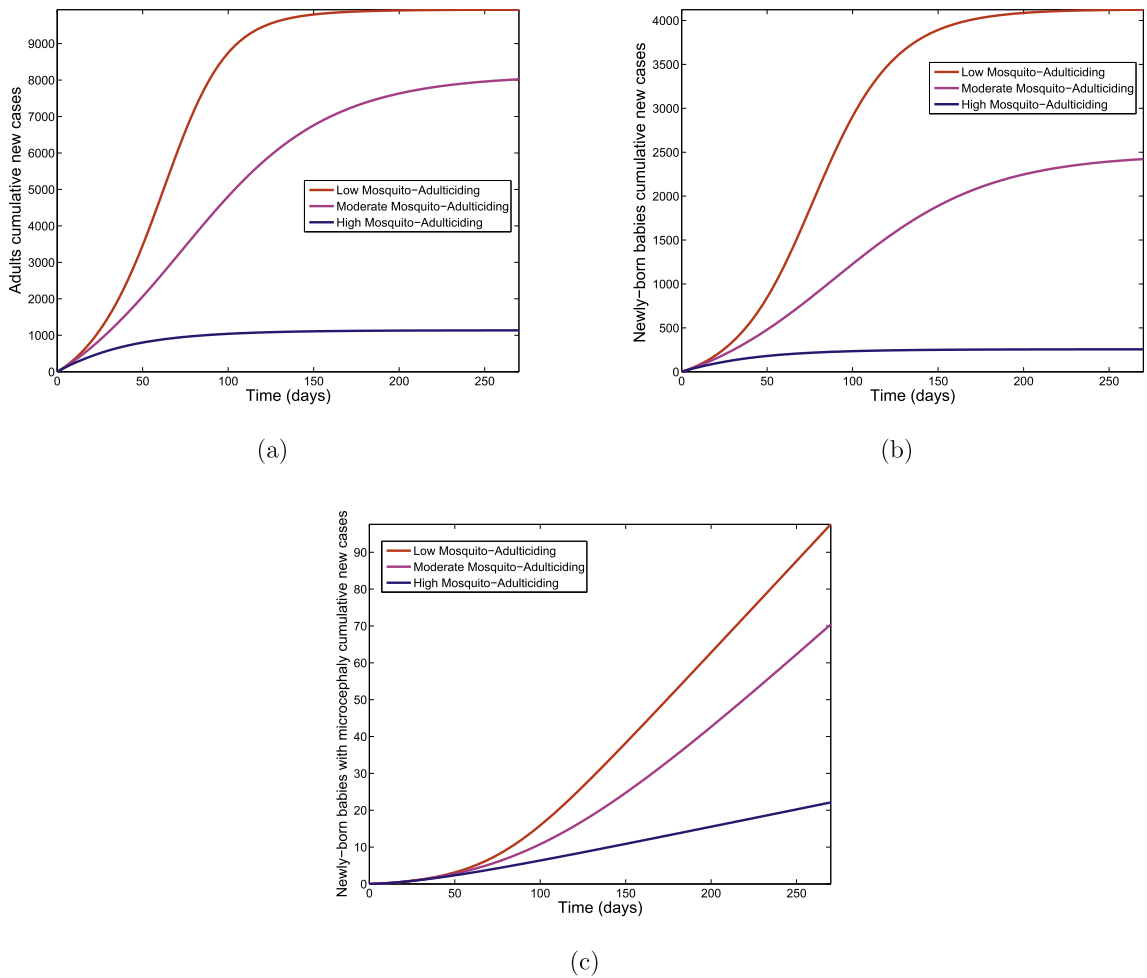


Fig. 11. Simulation of the Zika model (2.1) for various control levels of the adulticiding strategy. (a). The cumulative number of new cases in adults (b). The cumulative number of new cases in newly born babies. (c). The cumulative number of new cases of newly born babies with microcephaly. Parameter values used are as given in Table 3.

Table 12

Simulation results of the cumulative number of new cases for the Zika model (2.1) using mosquito-adulticiding strategy.

| Humans | Low-Control | Moderate-Control | High-Control |
|------------------------------|--------------------|-------------------|-------------------|
| Adults | 2.1×10^6 | 1.4×10^6 | 2.6×10^5 |
| Newly born babies | 7.7×10^5 | 4.0×10^5 | 5.8×10^4 |
| Newly born with microcephaly | 10.0×10^3 | 6.9×10^3 | 2.7×10^3 |

References

Agusto, F. B., Del Valle, S. Y., Blayneh, K. W., Ngonghala, C. N., Goncalves, M. J., Li, N., et al. (2013). The impact of bed-net use on malaria prevalence. *Journal of Theoretical Biology*, 320, 58–65.

Ahmed, A. (2016). *El salvadors advice on zika virus: Dont have babies*. New York Times. http://www.nytimes.com/2016/01/26/world/americas/el-salvadors-advice-on-zika-dont-have-babies.html?_r=0.

Anderson, R. M., & May, R. (1991). *Infectious diseases of humans*. New York: Oxford University Press.

Bell, T. M., Field, E. J., & Narang, H. K. (1971). Zika virus infection of the central nervous system of mice. *Archiv für die gesamte Virusforschung*, 35, 183–193.

Besnard, M., Lastere, S., Teissier, A., Cao-Lormeau, V. M., & Musso, D. (2014). Evidence of perinatal transmission of zika virus, French Polynesia, december 2013 and february 2014. *Euro Surveillance*, 19(13), 20751.

Bewick, S., Fagan, W., Calabrese, J., & Agosto, F. (2016). *Zika virus: Endemic versus epidemic dynamics and implications for disease spread in the Americas*.

Blower, S. M., & Dowlatabadi, H. I. (1994). Sensitivity and uncertainty analysis of complex models of disease transmission: An hiv model, as an example. *International Statistical Review/Revue Internationale de Statistique*, 229–243.

Bundy, D. A. P. (2014). *The impact of infectious disease on cognitive development. Environmental Effects on Cognitive Abilities*.

Cao-Lormeau, V. M., Blake, A., Mons, S., Lastère, S., Roche, C., Vanhomwegen, J., et al. (2016). Guillain-barre syndrome outbreak associated with zika virus infection in French Polynesia: A case-control study. *The Lancet*, 387(10027), 1531–1539.

Carter, M. T., Mirzaa, G., McDonnell, L. M., & Boycott, K. M. (2013). *Microcephaly-capillary malformation syndrome*.

- Castillo-Chavez, C., Blower, S., van den Driessche, P., Kirschner, D., & Yakubu, A.-A. (2002). *Mathematical approaches for emerging and reemerging infectious diseases*. New York: Springer-Verlag.
- Centers for Disease Control and Prevention. (2016). *Zika virus*.
- da Silva, A. A. M., Ganz, J. S. S., da Silva Sousa, P., Doriqui, M. J. R., Ribeiro, M. R. C., Branco, M. R. F. C., et al. (1953). Early growth and neurologic outcomes of infants with probable congenital zika virus syndrome. *Emerging Infectious Diseases*, 22(11), 2016.
- Darlington, S. (2016). *Cnn Brazil warns against pregnancy due to spreading virus shasta darlington-profile-image*. Cable News Network. Turner Broadcasting System, Inc. <http://www.cnn.com/2015/12/23/health/brazil-zika-pregnancy-warning/>
- Dick, G. W. A., Kitchen, S. F., & Haddow, A. J. (1952). Zika virus (i). isolations and serological specificity. *Transactions of the Royal Society of Tropical Medicine and Hygiene*, 46(5), 509–520.
- Diekmann, O., Heesterbeek, J. A. P., & Metz, J. A. P. (1990). On the definition and computation of the basic reproduction ratio r_0 in models for infectious diseases in heterogeneous populations. *Journal of Mathematical Biology*, 28, 503–522.
- Dubrulle, M., Mousson, L., Moutailler, S., Vazeille, M., & Failloux, A. B. (2009). Chikungunya virus and aedes mosquitoes: Saliva is infectious as soon as two days after oral infection. *PLoS One*, 4(6), e5895.
- Duffy, M. R., Chen, T. H., Hancock, W. T., Powers, A. M., & Kool, J. L. (2009). Zika virus outbreak on yap island, Federated States OF Micronesia. *New England Journal of Medicine*, 360(24), 2536–2543.
- Dumont, Y., & Chiroleu, F. (2010). Vector control for the chikungunya disease. *Mathematical Biosciences and Engineering*, 7(2), 315–348.
- Dumont, Y., Chiroleu, F., & Domerg, C. (2008). On a temporal model for the chikungunya disease: Modeling, theory and numerics. *Mathematical Biosciences*, 213(1), 80–91.
- European Centre for Disease Prevention and Control. (2015). *Zika virus epidemic in the americas: Potential association with microcephaly and guillain-barré syndrome*. Rapid risk assessment <http://ecdc.europa.eu/en/publications/Publications/zika-virus-americas-association-with-microcephaly-rapid-risk-assessment.pdf>, 10(14).
- Fauci, A. S., & Morens, D. M. (2016). Zika virus in the americas yet another arbovirus threat. *The New England Journal of Medicine*, 374(2).
- Hethcote, H. W. (2000). The mathematics of infectious diseases. *SIAM Review*, 42(4), 599–653.
- Lakshmikantham, V., Leela, S., & Martynuk, A. A. (1989). *Stability analysis of nonlinear systems*. New York and Basel: Marcel Dekker, Inc.
- Manore, C., Hickmann, J., Xu, S., Wearing, H., & Hyman, J. (2014). Comparing dengue and chikungunya emergence and endemic transmission in a. aegypti and a. albopictus. *Journal of Theoretical Biology*, 356(7), 174–191.
- Marino, S., Hogue, I. B., Ray, C. J., & Kirschner, D. E. (2008). A methodology for performing global uncertainty and sensitivity analysis in systems biology. *Journal of Theoretical Biology*, 254(1), 178–196.
- Martines, R. B. (2015). Notes from the field: Evidence of zika virus infection in brain and placental tissues from two congenitally infected newborns and two fetal losses brazil. *MMWR. Morbidity and Mortality Weekly Report*, 65, 2016.
- Mayo Foundation for Medical Education and Research. (2016). *Microcephaly*. <http://www.mayoclinic.org/diseases-conditions/microcephaly/basics/definition/con-20034823>.
- McKay, M. D., Beckman, R. J., & Conover, W. J. (2000). A comparison of three methods for selecting values of input variables in the analysis of output from a computer code. *Technometrics*, 42(1), 55–61.
- McLeod, R. G., Brewster, J. F., Gumel, A. B., & Slonowsky, A. (2006). Sensitivity and uncertainty analyses for a sars model with time-varying inputs and outputs. *Mathematical Biosciences and Engineering*, 3(3), 527.
- Mrakar, J., Korva, M., Tul, N., Popović, M., Poljsak-Prijatelj, M., Mraz, J., et al. (2016). Zika virus associated with microcephaly. *New England Journal of Medicine*, 374(10), 951–958.
- Moore, C. A., Staples, J. E., Dobyns, W. B., Pessoa, A., Ventura, C. V., Da Fonseca, E. B., et al. (2017). Characterizing the pattern of anomalies in congenital zika syndrome for pediatric clinicians. *JAMA Pediatrics*, 171(3), 288–295.
- Moulay, D., Aziz-Alaoui, M. A., & Cadivel, M. (2011). The chikungunya disease: Modeling, vector and transmission global dynamics. *Mathematical Biosciences*, 229(1), 50–63.
- Newton, E. A. C., & Reiter, P. (1992). A model of the transmission of dengue fever with an evaluation of the impact of ultra-low volume (ulv) insecticide applications on dengue epidemics. *American Journal of Tropical Medicine and Hygiene*, 47, 709–720.
- Oehler, E., Watrin, L., Larre, P., Leparc-Goffart, I., Lasterer, S., Valour, F., et al. (2014). Zika virus infection complicated by guillain-barre syndrome—case report, French Polynesia, December 2013. *Eurosurveillance*, 19(9), 20720.
- Oliveira Melo, A. S., Malinger, G., Ximenes, R., Szejnfeld, P. O., Alves Sampaio, S., & Bispo de Filippis, A. M. (2016). Zika virus intrauterine infection causes fetal brain abnormality and microcephaly: Tip of the iceberg? *Ultrasound in Obstetrics & Gynecology*, 47(1), 6–7.
- Oster, A. M. (2016). Interim guidelines for prevention of sexual transmission of zika virus United States, 2016. *MMWR. Morbidity and Mortality Weekly Report*, 65.
- Paupy, C., Ollomo, B., Kamgang, B., Moutailler, S., Rousset, D., Demanou, M., et al. (2010). Comparative role of aedes albopictus and aedes aegypti in the emergence of dengue and chikungunya in central africa. *Vector-borne and Zoonotic Diseases*, 10(3), 259–266.
- Poletti, P., Messeri, G., Ajelli, M., Vallorani, R., Rizzo, C., & Merler, S. (2011). Transmission potential of chikungunya virus and control measures: The case of Italy. *PLoS One*, 6(5), e18860.
- Putnam, J. L., & Scott, T. W. (1995). Blood feeding behavior of dengue-2 virus-infected aedes aegypti. *American Journal of Tropical Medicine and Hygiene*, 55, 225–227.
- Romero, S. (2015). *Alarm spreads in Brazil over a virus and a surge in malformed infants*. New York Times. http://www.nytimes.com/2015/12/31/world/americas/alarm-spreads-in-brazil-over-a-virus-and-a-surge-in-malformed-infants.html?smid=nytcore-ipad-share&smprod=nytcore-ipad&_r=1.
- Rubin, E. J., Greene, M. F., & Baden, L. R. (2016). Zika virus and microcephaly. *New England Journal of Medicine*, 374(10), 984–985.
- Sanchez, M. A., & Blower, S. M. (1997). Uncertainty and sensitivity analysis of the basic reproductive rate: Tuberculosis as an example. *American Journal of Epidemiology*, 145(12), 1127–1137.
- Schuler-Faccini, L., Ribeiro, E. M., Feitosa, I. M. L., Horovitz, D. D. G., Cavalcanti, D. P., Pessoa, A., et al. (2016). Possible association between zika virus infection and microcephaly—Brazil, 2015. *MMWR. Morbidity and Mortality Weekly Report*, 65.
- Sebastian, M. R., Lodha, R., & Kabra, S. K. (2009). Chikungunya infection in children. *Indian Journal of Pediatrics*, 76(2), 185–189.
- Sheppard, P. M., Macdonald, W. M., Tonn, R. J., & Grabs, B. (1969). The dynamics of an adult population of aedes aegypti in relation to dengue haemorrhagic fever in bangkok. *Journal of Animal Ecology*, 38, 661–701.
- Talan, J. (2016). Epidemiologists are tracking possible links between zika virus, microcephaly, and guillain–barré syndrome. *Neurology Today*, 16(4), 1–18.
- Trpis, M., & Haussermann, W. (1986). Dispersal and other population parameters of aedes aegypti in an african village and their possible significance in epidemiology of vector-borne diseases. *American Journal of Tropical Medicine and Hygiene*, 35, 1263–1279.
- Trpis, M., Haussermann, W., & Craig, G. B. (1995). Estimates of population size, dispersal, and longevity of domestic aedes aegypti by mark-release-recapture in the village of shauri moyo in eastern Kenya. *Journal of Medical Entomology*, 32, 27–33.
- Turell, M. J., Beaman, J. R., & Tammariello, R. F. (1992). Susceptibility of selected strains of aedes aegypti and aedes albopictus (diptera: Culicidae) to chikungunya virus. *Journal of Medical Entomology*, 29(1), 49–53.
- van den Driessche, P., & Watmough, J. (2002). Reproduction numbers and sub-threshold endemic equilibria for compartmental models of disease transmission. *Mathematical Biosciences*, 180, 29–48.
- Vogel, G. (2016). A race to explain Brazil's spike in birth defects. *Science*, 351(6269), 110–111.
- Wikipedia. (2015). *List of sovereign states and dependent territories by birth rate*.

- World Health Organization. (2015). *Neurological syndrome, congenital malformations, and zika virus infection. implications for public health in the Americas. Epidemiological Alert*. http://www.paho.org/hq/index.php?option=com_docman&task=doc_view&Itemid=270&gid=32405&lang=en.
- World Health Organization. (2016a). *Microcephaly - Brazil: Disease outbreak news*. <http://www.who.int/csr/don/archive/disease/microcephaly/en/>.
- World Health Organization. (2016b). *Zika situation report*. <http://www.who.int/emergencies/zika-virus/situation-report/who-zika-report-12-02-2016.pdf?ua=1>.