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RESEARCH ARTICLE



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On the dynamics of dengue virus type 2 with residence times and vertical transmission

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ABSTRACT

A two-patch mathematical model of Dengue virus type 2 (DENV-2) that accounts for vectors' vertical transmission and between patches human dispersal is introduced. Dispersal is modelled via a Lagrangian approach. A host-patch residence-times basic reproduction number is derived and conditions under which the disease dies out or persists are established. Analytical and numerical results highlight the role of hosts' dispersal in mitigating or exacerbating disease dynamics. The framework is used to explore dengue dynamics using, as a starting point, the 2002 outbreak in the state of Colima, Mexico.

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Vector-borne diseases; DENV-2 Asian genotype; dengue; residence times; multi-patch model; global stability

1. Introduction

Dengue, a re-emerging vector-borne disease, is caused by members of the genus *Flavivirus* in the family *Flaviviridae* with four active antigenically distinct serotypes: DENV-1, DENV-2, DENV-3 and DENV-4 (Deubel, Kinney, & Trent, 1988). The pathogenicity of dengue can range from asymptomatic, mild dengue fever (DF) to dengue hemorrhagic fever (DHF) and dengue shock syndrome (DSS) (Deubel et al., 1988; Halstead et al., 2002). Although infection with a dengue serotype does not usually protect against other serotypes, it is believed that secondary infections with a heterologous serotype increase the probability of DHF and DSS (Burke, Nisalak, Johnson, & Scott, 1988; Halstead, Nimmannitya, & Cohen, 1970). According to the World Health Organization, 40% of the global population is at risk for dengue infection with an estimate of 50–100 million infections yearly including 500,000 cases of DHF. It has been estimated that about 22,000 deaths, mostly children under 15 years of age, can be attributed to DHF (WHO, 2009). In the United States, approximately 5% or more of the Key West population in Florida was exposed to dengue during the 2009–2010 outbreak (CDC, 2010) while the Hawaii Department of Health reported 190 cases during the 2015 outbreak on Oahu, the first outbreak since 2011. Since dengue is

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not endemic in Hawaii, health authorities have suggested that the recent outbreak may have been started by infected visitors (State of Hawaii, 2015). Dengue is highly prevalent and endemic in South-East Asia, which has experienced a 70% increase in cases since 2004 (Kwok, 2010); Mexico, also an endemic country, reported during the 2002 outbreak over a million cases of DF and more than 17,000 cases of DHF (Guzman & Kouri, 2003; Morens & Fauci, 2008). Dengue is transmitted primarily by the vector Ae. aegypti, which is now found in most countries in the tropics and sub-tropics (Harris et al., 2000; Reiter & Gubler, 1997). The secondary vector, Ae. albopictus, has a range reaching farther north than Ae. aegypti with eggs better adapted to subfreezing temperatures (Hawley, Reiter, Copeland, Pumpuni, & Craig, Jr, 1987; Morens & Fauci, 2008). Differences in susceptibility and transmission of dengue infection (Arunachalam et al., 2008; Knox, Kay, Hall, & Ryan, 2003; Tewari et al., 2004) raise the possibility that some serotypes are either more successful at invading a host population, or more pathogenic, or both (Kyle & Harris, 2008). DENV-2 is the most associated with dengue outbreaks involving DHF and DSS cases (Montoya et al., 2003; Rico-Hesse et al., 1997; Sittisombut et al., 1997; Zhang et al., 2006), followed by DENV-1 and DENV-3 viruses (Balmaseda et al., 2006; Harris et al., 2000; Montoya et al., 2003). While infection with any of the four dengue serotypes could lead to DHF, the rapid displacement of DENV-2 American by DENV-2 Asian genotype has been linked to major outbreaks with DHF cases in Cuba, Jamaica, Venezuela, Colombia, Brazil, Peru and Mexico (Lewis et al., 1993; Montoya et al., 2003; Rico-Hesse et al., 1997; Rico-Hesse et al., 1998, 1997; Zhang et al., 2006). A possible mechanism involved in the dispersal and persistence of DENV-2 in nature is vertical transmission (transovarial transmission) via Ae. aegypti. Prior studies were unsuccessful in demonstrating vertical transmission via Ae. aegypti (Rodhain & Rosen, 1997). However, the use of advances in molecular biology has shown that vertical transmission involving Ae. aegypti and Ae. albopictus is possible in captivity and in the wild (Arunachalam et al., 2008; Bosio, Thomas, Grimstad, & Rai, 1992; Cecílio, Campanelli, Souza, Figueiredo, & Resende, 2009; Gunther, Martínez-Muñoz, Pérez-Ishiwara, & Salas-Benito, 2007; Rosen, Shroyer, Tesh, Freier, & Lien, 1983). Thus, assessing transmission dynamics and pathogenicity between the DENV-2 American and Asian genotypes' differences is one of the priorities associated with the study of the epidemiology of dengue. In short, dengue has an increasing recurrent presence putting a larger percentage of the global population at risk of dengue infection, a situation that has become the norm due to the growth of travel and tourism between endemic and non-endemic regions. The aim of this work is to better understand the impact of human mobility on dengue disease transmission, its impact on dengue dynamics, and the use of mobility-based strategies, standard control measures, in reducing the prevalence of dengue infections.

Mathematical models describing the dynamics of interaction between host and vector go back to Lotka (1923), Macdonald (1952) and Ross (1911); first used to study vector-host dynamics in the context of Malaria (Brauer & Castillo-Chávez, 2012; Gumel, Castillo-Chavez, Mickens, & Clemence, 2006; Shim, Feng, & Castillo-Chavez, 2012). Variations of such framework have been applied to dengue (for a review see Smith et al., 2012). Further applications of modelling variations in the context of Malaria include, (Forouzannia & Gumel, 2014; McKenzie & Samba, 2004; Ngwa, Niger, & Gumel, 2010; Niger & Gumel, 2008) and in the context of dengue (Castillo-Chávez, Sanchez, & Murillo, 2011; Chowell & Sanchez, 2006; Gumel et al., 2006; Murillo, Holechek, Murillo, Sanchez, & Castillo-Chavez, 2014; Nishiura, 2006).

The potential role of vertical transmission in dengue endemic regions or in fluctuating environments has been explored in Adams and Boots (2010), Esteva and Vargas (2000), Nishiura (2006). The role in the displacement of DENV-2 American via DENV-2 Asian vertical transmission has also been addressed (Murillo et al., 2014). The role of host movement has also been explored in the context of dengue Adams and Kapan (2009) in a formulation that does not account for the *effective population size*. In this paper, the role of vertical transmission and movement via residence times are explored via a two-patch model involving non-mobile vectors and mobile hosts. This paper is organized as follows: The derivation of the model is presented in Section 2; Analytical results are collected in Section 3; The results of numerical simulations are collected in Section 4; Section 5 explores the possible role of movement on joint dynamics of dengue in Colima and Manzanillo in the presence of host mobility; Concluding remarks are collected in Section 6.

2. Derivation of the model

A single patch model is derived and embedded into a two-patch model via a residencetimes matrix in order to study the impact of host mobility on dengue disease dynamics. Conditions for dengue eradication and persistence in the population are computed.

2.1. Single patch model

We consider a host population composed of susceptible (S_h) , exposed (E_h) , infectious (I_h) , and recovered (R_h) individuals interacting with a vector population composed of susceptible (S_{ν}) , exposed (E_{ν}) , and infected (I_{ν}) vectors. The dynamics of dengue follows an SEIR structure for the host population and an SEI type for the vector population. The birth rate for the host population is μ_h , assumed to be equal to the death rate, that is, hosts' demographic differentials are conveniently ignored, that is, the host population is assumed to be constant. Susceptible hosts are infected, by infectious mosquitoes, at the rate $a\beta_{vh}\frac{I_v}{N_h}$ where a is the biting rate and β_{vh} is the infectiousness of human to mosquitoes. The exposed population develops symptoms becoming infectious at the rate v_h . Infectious individuals recover at the per-capita rate γ . Susceptible mosquitoes become infected, via interactions with infectious hosts, at the rate $a\beta_{hv}\frac{l_h}{N_v}$. Recent studies place significant importance to the connection between DENV-2 and DHF cases (Chowell, Diaz-Dueñas, Chowell, et al., 2007; Espinoza-Gómez, Díaz-Dueñas, Torres-Lepe, Cedillo-Nakay, & Newton-Sánchez, 2005; Montoya et al., 2003; Rico-Hesse et al., 1997; Sittisombut et al., 1997; Zhang et al., 2006) and on DENV-2 vertical transmission (Martins et al., 2012). Hence, it is assumed that a fraction of the mosquitoes q are 'born' infected entering directly into the infectious class. The natural per-capita vector mortality is μ_v .

The model describing the dynamics of DENV-2 is given by the following system of differential equations:

$$\begin{split} \dot{S}_{h} &= \mu_{h}N_{h} - a\beta_{\nu h}S_{h}\frac{I_{\nu}}{N_{h}} - \mu_{h}S_{h} \\ \dot{E}_{h} &= \beta_{\nu h}S_{h}\frac{I_{\nu}}{N_{h}} - (\mu_{h} + \nu_{h})E_{h} \\ \dot{I}_{h} &= \nu_{h}E_{h} - (\mu_{h} + \gamma_{h})I_{h} \\ \dot{R}_{h} &= \gamma_{h}I_{h} - \mu_{h}R_{h} \\ \dot{S}_{\nu} &= \mu_{\nu}(N_{\nu} - qI_{\nu}) - a\beta_{h\nu}S_{\nu}\frac{I_{h}}{N_{h}} - \mu_{\nu}S_{\nu} \\ \dot{E}_{\nu} &= a\beta_{h\nu}S_{\nu}\frac{I_{h}}{N_{h}} - (\nu_{\nu} + \mu_{\nu})E_{\nu} \\ \dot{I}_{\nu} &= \nu_{\nu}E_{\nu} + q\mu_{\nu}I_{\nu} - \mu_{\nu}I_{\nu} \end{split}$$
(1)

In the absence of selection, that is, differences in birth and death rate and in the absence of vertical transmission, Model (1) turns out to be isomorphic to model considered by in Chowell, Diaz-Dueñas, Miller, et al. (2007). Model (1) is well defined supporting a sharp threshold property, namely, the disease dies out if the basic reproduction number \mathcal{R}_0 is less than unity, persisting whenever $\mathcal{R}_0 > 1$ where

$$\mathcal{R}_{0}^{2} = \frac{a^{2}\beta_{h\nu}\beta_{\nu h}N_{\nu}\nu_{h}\nu_{\nu}}{(1-q)N_{h}(\mu_{h}+\nu_{h})(\mu_{h}+\gamma_{h})(\mu_{\nu}+\nu_{\nu})\mu_{\nu}}$$

2.2. Heterogeneity through virtual dispersal

The single patch model is the building block for the two-patch model used in this study. Within each patch, in the absence of host mobility, dengue dynamics are modelled via System 1. A metapopulation approach, an Eulerian perspective, is most often applied to the study of vector-borne diseases involving host mobility (Adams & Kapan, 2009; Auger, Kouokam, Sallet, Tchuente, & Tsanou, 2008; Gao & Ruan, 2012). Here, a Lagrangian approach is used instead to model the movement of individuals between patches (see Bichara & Castillo-Chavez, 2015; Bichara, Kang, Castillo-Chavez, Horan, & Perrings, 2015). It is assumed that vectors do not move between patches since vectors *Ae. aegypti* and *Ae. albopictus* do not travel more than few tens of meters over their lifetime (Adams & Kapan, 2009; WHO, 2015); moving 400–600 m at most (Bonnet & Worcester, 1946; Niebylski & Craig, Jr. 1994), respectively. In short, we neglect vector's dispersal, which fits well with the simulations involving two cities in the state of Colima, Mexico.

The host resident of Patch 1, population size $N_{h,1}$, spends, on average, p_{11} proportion of its time in their own Patch 1 and p_{12} proportion of its time visiting Patch 2. Residents of Patch 2, population of size $N_{h,2}$, spend p_{22} proportion of their time in Patch 2 while spending $p_{21} = 1 - p_{22}$ visiting Patch 1. Thus, at time *t*, the *effective population* in Patch 1 is $p_{11}N_{h,1} + p_{21}N_{h,2}$ and the *effective population* in Patch 2 is $p_{12}N_{h,1} + p_{22}N_{h,2}$. The susceptible population of Patch 1 (S_1) could be infected by a vector in either Patch 1 ($I_{v,1}$) or Patch 2 by ($I_{v,2}$). Thus, the dynamics of the susceptible population in Patch 1 are given by

$$\dot{S}_{h,1} = \mu_h N_{h,1} - a_1 \beta_{\nu h} p_{11} S_{h,1} \frac{I_{\nu,1}}{p_{11} N_{h,1} + p_{21} N_{h,2}} - a_2 \beta_{\nu h} p_{12} S_{h,1} \frac{I_{\nu,2}}{p_{12} N_{h,1} + p_{22} N_{h,2}} - \mu_h S_{h,1}.$$
(2)

144 🕒 D. BICHARA ET AL.

And so, the *effective infectious* population in Patch 1 is $p_{11}I_{h,1} + p_{21}I_{h,2}$, and consequently the proportion of infectious individuals in Patch 1 is

$$\frac{p_{11}I_{h,1} + p_{21}I_{h,2}}{p_{11}N_{h,1} + p_{21}N_{h,2}}$$

The dynamics of susceptible mosquitoes in Patch 1 are modelled as follows:

$$\dot{S}_{\nu,1} = \mu_{\nu}(N_{\nu,i} - qI_{\nu,i}) - a_1\beta_{h\nu}S_{\nu,1}\frac{p_{11}I_{h,1} + p_{21}I_{h,2}}{p_{11}N_{h,1} + p_{21}N_{h,2}} - \mu_{\nu}S_{\nu,1}.$$
(3)

The complete dynamics of DENV-2, with the host moving between patches, is given by the following system:

$$\begin{cases} \dot{S}_{h,i} = \mu_h N_{h,i} - \beta_{\nu h} S_{h,i} \sum_{j=1}^2 a_j p_{ij} \frac{I_{\nu,j}}{p_{1j} N_{h,1} + p_{2j} N_{h,2}} - \mu_h S_{h,i}, \\ \dot{E}_{h,i} = \beta_{\nu h} S_{h,i} \sum_{j=1}^2 a_j p_{ij} \frac{I_{\nu,j}}{p_{1j} N_{h,1} + p_{2j} N_{h,2}} - (\mu_h + \nu_h) E_{h,i}, \\ \dot{I}_{h,i} = \nu_h E_{h,i} - (\mu_h + \gamma_i) I_{h,i}, \\ \dot{R}_{h,i} = \gamma_i I_{h,i} - \mu_h R_{h,i}, \\ \dot{S}_{\nu,i} = \mu_\nu (N_{\nu,i} - q I_{\nu,i}) - a_i \beta_{h\nu} S_{\nu,i} \frac{\sum_{j=1}^2 p_{ji} I_{h,j}}{\sum_{k=1}^2 p_{ki} N_{h,k}} - \mu_\nu S_{\nu,i}, \\ \dot{E}_{\nu,i} = a_i \beta_{h\nu} S_{\nu,i} \frac{\sum_{j=1}^2 p_{ji} I_{h,j}}{\sum_{k=1}^2 p_{ki} N_{h,k}} - (\mu_\nu + \nu_\nu) E_{\nu,i}, \\ \dot{I}_{\nu,i} = \nu_\nu E_{h,i} + q \mu_\nu I_{\nu,i} - \mu_\nu I_{\nu,i}, i = 1, 2. \end{cases}$$

$$(4)$$

Since the total populations of hosts and vectors are constant in each patch, System (4) has the same qualitative dynamics as

$$\begin{aligned} \dot{S}_{h,i} &= \mu_h N_{h,i} - \beta_{\nu h} S_{h,i} \sum_{j=1}^2 a_j p_{ij} \frac{I_{\nu,j}}{p_{1j} N_{h,1} + p_{2j} N_{h,2}} - \mu_h S_{h,i}, \\ \dot{E}_{h,i} &= \beta_{\nu h} S_{h,i} \sum_{j=1}^2 a_j p_{ij} \frac{I_{\nu,j}}{p_{1j} N_{h,1} + p_{2j} N_{h,2}} - (\mu_h + \nu_h) E_{h,i}, \\ \dot{I}_{h,i} &= \nu_h E_{h,i} - (\mu_h + \gamma_i) I_{h,i}, \\ \dot{E}_{\nu,i} &= a_i \beta_{h\nu} (N_{\nu,i} - E_{\nu,i} - I_{\nu,i}) \frac{\sum_{j=1}^2 p_{ji} I_{h,j}}{\sum_{k=1}^2 p_{ki} N_{h,k}} - (\mu_\nu + \nu_\nu) E_{\nu,i}, \\ \dot{I}_{\nu,i} &= \nu_\nu E_{h,i} - (1 - q) \mu_\nu I_{\nu,i}. \end{aligned}$$
(5)

The parameters of Model 5 are described in Table 1.

We now show that the model is *biologically* well posed.

Lemma 2.1: The set

$$\Omega = \{ (S_{h,i}, E_{h,i}, I_{h,i}, E_{\nu,i}, I_{\nu,i}) \in \mathbb{R}^6_+ \mid S_{h,i} + E_{h,i} + I_{h,i} \le N_{h,i}, E_{\nu,i} + I_{\nu,i} \le N_{\nu,i} \}$$

is a compact positively invariant for the System (5).

Table 1. Description of the parameters used in System (5).

Parameters	Description
$\beta_{\rm vh}$	Infectiousness of human to mosquitoes
β_{hv}	Infectiousness of mosquitoes to humans
a _i	Biting rate in Patch i
$\dot{\mu}_{h}$	Humans' birth and death rate
vh	Humans' incubation rate
Υi	Recovery rate in Patch i
p _{ij}	Proportion of time residents of Patch <i>i</i> spend in Patch <i>i</i>
b _v	Vectors' natural birth rate
$\mu_{\rm V}$	Vectors' natural mortality rate
ν_V	Vectors' incubation rate

Proof: The positive orthant is clearly positively invariant. Since the host population is constant, then the inequality $S_{h,i} + E_{h,i} + I_{h,i} \le N_{h,i}$ is always satisfied. We have

$$\dot{E}_{\nu,i} + \dot{I}_{\nu,i} \mid_{E_{\nu,i}+I_{\nu,i}=N_{\nu,i}} = -\mu_{\nu}N_{\nu,i} + q\mu_{\nu}I_{\nu,i} \le -\mu_{\nu}(1-q)N_{\nu,i} \le 0$$

Hence, $E_{v,i} + I_{v,i} \leq N_{v,i}$ and the set Ω , an intersection of positively invariant sets (\mathbb{R}^7_+ , $\{S_{h,i}+E_{h,i}+I_{h,i} \leq N_{h,i}\}$, and $\{E_{v,i}+I_{v,i} \leq N_{v,i}\}$), is positively invariant; the set is a compact set.

3. Equilibria and stability analysis

This section characterizes the equilibrium dynamics of Model (5).

3.1. The disease-free equilibrium and the basic reproduction number

The disease-free equilibrium is

$$E_0 = (N_{h,1}, N_{h,2}, \mathbf{0}_{\mathbb{R}^8})$$

which is used to compute the basic reproduction number via the next generation method (Diekmann, Heesterbeek, & Metz, 1990; van den Driessche & Watmough, 2002). The basic reproduction number \mathcal{R}_0 is defined by the expression (See Appendix 1, for details), $\mathcal{R}_0^2 = \rho(M_{\nu h}M_{h\nu})$, that is, the spectral radius of the matrix of $M_{\nu h}M_{h\nu}$, where

$$M_{\nu h} = \begin{pmatrix} \frac{a_1 \beta_{\nu h} p_{11} N_{h,1} \nu_{\nu}}{(p_{11} N_{h,1} + p_{21} N_{h,2}) (\mu_{\nu} + \nu_{\nu}) (1-q) \mu_{\nu}} & \frac{a_2 \beta_{\nu h} p_{12} N_{h,1} \nu_{\nu}}{(p_{12} N_{h,1} + p_{22} N_{h,2}) (\mu_{\nu} + \nu_{\nu}) (1-q) \mu_{\nu}} \\ \frac{a_1 \beta_{\nu h} p_{21} N_{h,2} \nu_{\nu}}{(p_{11} N_{h,1} + p_{21} N_{h,2}) (\mu_{\nu} + \nu_{\nu}) (1-q) \mu_{\nu}} & \frac{a_2 \beta_{\nu h} p_{22} N_{h,2} \nu_{\nu}}{(p_{12} N_{h,1} + p_{22} N_{h,2}) (\mu_{\nu} + \nu_{\nu}) (1-q) \mu_{\nu}} \end{pmatrix}$$

and

$$M_{h\nu} = \begin{pmatrix} \frac{a_1\beta_{h\nu}p_{11}N_{\nu,1}\nu_h}{(p_{11}N_{h,1}+p_{21}N_{h,2})(\mu_h+\nu_h)(\mu_h+\gamma_1)} & \frac{a_1\beta_{h\nu}p_{21}N_{\nu,1}\nu_h}{(p_{11}N_{h,1}+p_{21}N_{h,2})(\mu_h+\nu_h)(\mu_h+\gamma_2)} \\ \frac{a_2\beta_{h\nu}p_{12}N_{\nu,2}\nu_h}{(p_{12}N_{h,1}+p_{22}N_{h,2})(\mu_h+\nu_h)(\mu_h+\gamma_1)} & \frac{a_2\beta_{h\nu}p_{21}N_{\nu,2}\nu_h}{(p_{12}N_{h,1}+p_{22}N_{h,2})(\mu_h+\nu_h)(\mu_h+\gamma_2)} \end{pmatrix}$$

The matrix $\begin{pmatrix} 0 & M_{\nu h} \\ M_{h\nu} & 0 \end{pmatrix}$ is called the host–vector network configuration (Iggidr, Sallet, & Souza, 2014). The result of local asymptotic stability if $\mathcal{R}_0^2 < 1$ and instability if $\mathcal{R}_0^2 > 1$ has

146 👄 D. BICHARA ET AL.

been established in van den Driessche & Watmough (2002). The following theorem gives the global result of the DFE.

Theorem 3.1: If $\mathcal{R}_0^2 \leq 1$, the DFE is globally asymptotically stable in the nonnegative orthant. If $\mathcal{R}_0^2 > 1$, the DFE is unstable.

Proof: We use the comparison theorem (Smith & Waltman, 1995) to prove the GAS of the DFE. Since $S_{h,i} \leq N_{h,i}$ and $S_{v,i} \leq N_{v,i}$, we have that

$$\dot{E}_{h,i} \le \beta_{\nu h} N_{h,i} \sum_{j=1}^{2} a_{j} p_{ij} \frac{I_{\nu,j}}{p_{1j} N_{h,1} + p_{2j} N_{h,2}} - (\mu_{h} + \nu_{h}) E_{h,i}$$
(6)

and

$$\dot{E}_{\nu,i} \le a_i \beta_{h\nu} N_{\nu,i} \frac{\sum_{j=1}^2 p_{ji} I_{h,j}}{\sum_{k=1}^2 p_{ki} N_{h,k}} - (\mu_\nu + \nu_\nu) E_{\nu,i}.$$
(7)

We define an auxiliary system via the right-hand side of Equations (6) and (7) and the infected compartments of Equation (5) as follows:

$$\begin{pmatrix} \dot{E}_{h,i} \\ \dot{E}_{\nu,i} \\ \dot{I}_{h,i} \\ \dot{I}_{\nu,i} \end{pmatrix} = \begin{pmatrix} \beta_{\nu h} N_{h,i} \sum_{j=1}^{2} a_{j} p_{ij} \frac{I_{\nu,j}}{p_{1j} N_{h,1} + p_{2j} N_{h,2}} - (\mu_{h} + \nu_{h}) E_{h,i} \\ a_{i} \beta_{h\nu} N_{\nu,i} \frac{\sum_{j=1}^{2} p_{ji} I_{h,j}}{\sum_{k=1}^{2} p_{ki} N_{h,k}} - (\mu_{\nu} + \nu_{\nu}) E_{\nu,i} \\ \nu_{h} E_{h,i} - (\mu_{h} + \gamma_{i}) I_{h,i} \\ \nu_{\nu} E_{h,i} - (1 - q) \mu_{\nu} I_{\nu,i} \end{pmatrix} = (F + V) \begin{pmatrix} E_{h,i} \\ E_{\nu,i} \\ I_{h,i} \\ I_{\nu,i} \end{pmatrix};$$
(8)

where the matrices F and V in (8) were just generated using the next generation method. System (8) is linear and its dynamics are well known. Since V is a Metzler matrix and F a nonnegative matrix (Berman & Plemmons, 1979), then

$$\rho(-FV^{-1}) < 1 \iff \alpha(F+V) < 0$$

where $\alpha(F + V)$ is the stability modulus of F + V. Thus, if $\mathcal{R}_0 = \rho(-FV^{-1}) < 1$, all the eigenvalues of F + V are negative. Hence, the nonnegative solutions of (8) are such that

$$\lim_{t \to \infty} E_{h,i} = \lim_{t \to \infty} E_{\nu,i} = 0 \quad \text{and} \quad \lim_{t \to \infty} I_{h,i} = \lim_{t \to \infty} I_{\nu,i} = 0$$

Since, all the variables in System (5) are nonnegative, the use of a comparison theorem (Smith & Waltman, 1995) leads to

$$\lim_{t\to\infty} E_{h,i} = \lim_{t\to\infty} E_{\nu,i} = 0 \quad \text{and} \quad \lim_{t\to\infty} I_{h,i} = \lim_{t\to\infty} I_{\nu,i} = 0, \quad i = 1, 2.$$

Therefore, by using the asymptotic theory of autonomous systems (Castillo-Chávez & Thieme, 1995), System (5) has the qualitative dynamics of the following limit system:

$$\dot{S}_{h,i} = \mu_h N_{h,i} - \mu_h S_{h,i}$$

for which the equilibrium $(N_{h,1}, N_{h,2})$ is globally asymptotically stable. If $\mathcal{R}_0 > 1$, the instability of the DFE follows from Diekmann et al. (1990), van den Driessche & Watmough (2002).

Theorem 3.2: If $\mathcal{R}_0 > 1$, System (5) is uniformly persistent, that is, there exists $\eta > 0$ such that

$$\liminf_{t \to \infty} \{S_{h,i}, E_{h,i}, I_{h,i}, E_{\nu,i}, I_{\nu,i}\} > \eta$$

for any initial conditions satisfying $S_{h,i}(0) > 0$, $E_{h,i}(0) > 0$, $I_{h,i}(0) > 0$, $E_{v,i}(0) > 0$ and $I_{v,i}(0) > 0$ for i = 1, 2.

Proof: Let $X = \Omega$, $x = (S_{h,1}, S_{h,2}, E_{h,1}, E_{h,2}, E_{v,1}, E_{v,2}, I_{h,1}, I_{h,2}, I_{v,1}, I_{v,2})$ and $X_0 = \{x \in X \mid I_{v,1} + I_{v,2} > 0\}$. Hence, $\partial X_0 = X \setminus X_0 = \{x \in X \mid I_{v,1} = I_{v,2} = 0\}$. Let ϕ_t be semi-flow induced by the solutions of (5) and $M_{\partial} = \{x \in \partial X_0 \mid \phi_t x \in \partial X_0, t \ge 0\}$. By Lemma 2.1, we have $\phi_t X_0 \subset X_0$ and ϕ_t is bounded in X_0 . Therefore a global attractor for ϕ_t exists. The DFE is the unique equilibrium on the manifold ∂X_0 and is GAS on ∂X_0 . Moreover $\bigcup_{x \in M_{\partial}} \omega(x) = \{E_0\}$ and no subset of M forms a cycle in ∂X_0 . Finally since the DFE is unstable on X_0 if $\mathcal{R}_0 > 1$, we deduce that System (5) is uniformly persistent by using a result from Zhao (2013) (Theorem 1.3.1 and Remark 1.3.1).

Theorem 3.3: Whenever the host-vector configuration is irreducible and $\mathcal{R}_0^2 > 1$, System (5) has a unique endemic equilibrium.

Proof: We will use a result by Hethcote & Thieme (1985) to prove the uniqueness of the endemic equilibrium. An endemic equilibrium $(\bar{S}_{h,1}, \bar{S}_{h,2}, \bar{E}_{h,1}, \bar{E}_{h,2}, \bar{E}_{v,1}, \bar{E}_{v,2}, \bar{I}_{h,1}, \bar{I}_{h,2}, \bar{I}_{v,1}, \bar{I}_{v,2})$ satisfies:

$$\begin{cases} \mu_{h}N_{h,i} = \beta_{\nu h}\bar{S}_{h,i}\sum_{j=1}^{2}a_{j}p_{ij}\frac{\bar{I}_{\nu,j}}{p_{1j}N_{h,1}+p_{2j}N_{h,2}} + \mu_{h}\bar{S}_{h,i}, \\ (\mu_{h}+\nu_{h})\bar{E}_{h,i} = \beta_{\nu h}\bar{S}_{h,i}\sum_{j=1}^{2}a_{j}p_{ij}\frac{\bar{I}_{\nu,j}}{p_{1j}N_{h,1}+p_{2j}N_{h,2}}, \\ \nu_{h}\bar{E}_{h,i} = (\mu_{h}+\gamma_{i})\bar{I}_{h,i}, \\ (\mu_{\nu}+\nu_{\nu})\bar{E}_{\nu,i} = a_{i}\beta_{h\nu}(N_{\nu,i}-\bar{E}_{\nu,i}-\bar{I}_{\nu,i})\frac{\sum_{j=1}^{2}p_{ji}\bar{I}_{h,j}}{\sum_{k=1}^{2}p_{ki}N_{h,k}}, \\ (1-q)\mu_{\nu}I_{\nu,i} = \nu_{\nu}E_{\nu,i}. \end{cases}$$
(9)

The first equation of (9) implies that

$$\bar{S}_{h,i} = \frac{\mu_h N_{h,i}}{\beta_{\nu h} \sum_{j=1}^2 a_j p_{ij} \frac{\bar{I}_{\nu,j}}{p_{1j} N_{h,1} + p_{2j} N_{h,2}} + \mu_h}$$

Hence, we deduce that, from System (9), that

$$\begin{cases} \bar{E}_{h,i} = \frac{\beta_{\nu h}}{\mu_{h} + \nu_{h}} \frac{\mu_{h} N_{h,i}}{\beta_{\nu h} \sum_{j=1}^{2} a_{j} p_{ij} \frac{\bar{I}_{\nu,j}}{p_{1j} N_{h,1} + p_{2j} N_{h,2}} + \mu_{h}} \sum_{j=1}^{2} a_{j} p_{ij} \frac{\bar{I}_{\nu,j}}{p_{1j} N_{h,1} + p_{2j} N_{h,2}}, \\ \bar{I}_{h,i} = \frac{\nu_{h}}{\mu_{h} + \gamma_{i}} \bar{E}_{h,i}, \\ \bar{E}_{\nu,i} = \frac{a_{i} \beta_{h\nu}}{\mu_{\nu} + \nu_{\nu}} (N_{\nu,i} - \bar{E}_{\nu,i} - \bar{I}_{\nu,i}) \frac{\sum_{j=1}^{2} p_{ji} \bar{I}_{h,j}}{\sum_{k=1}^{2} p_{ki} N_{h,k}}, \\ \bar{I}_{\nu,i} = \frac{\nu_{\nu}}{(1-q)\mu_{\nu}} \bar{E}_{\nu,i}. \end{cases}$$
(10)

Let

$$F(x) = \begin{pmatrix} \frac{\beta_{vh}v_{v}}{(1-q)(\mu_{h}+v_{h})\mu_{v}} \frac{\mu_{h}N_{h,1}}{\beta_{vh}\sum_{j=1}^{2}\frac{a_{j}p_{1j}v_{v}}{(1-q)\mu_{v}} \frac{\bar{E}_{v,j}}{p_{1j}N_{h,1}+p_{2j}N_{h,2}} + \mu_{h}} \sum_{j=1}^{2} a_{j}p_{1j} \frac{\bar{E}_{v,j}}{p_{1j}N_{h,1}+p_{2j}N_{h,2}} \\ \frac{\beta_{vh}v_{v}}{(1-q)(\mu_{h}+v_{h})\mu_{v}} \frac{\mu_{h}N_{h,2}}{\beta_{vh}\sum_{j=1}^{2}\frac{a_{j}p_{2j}v_{v}}{(1-q)\mu_{v}} \frac{\bar{E}_{v,j}}{p_{1j}N_{h,1}+p_{2j}N_{h,2}} + \mu_{h}} \sum_{j=1}^{2} a_{j}p_{2j} \frac{\bar{E}_{v,j}}{p_{1j}N_{h,1}+p_{2j}N_{h,2}} \\ \frac{a_{1}\beta_{hv}v_{h}}{(\mu_{v}+v_{v})(\mu_{h}+\gamma_{1})} (N_{v,1}-\bar{E}_{v,1}-\frac{v_{v}}{(1-q)\mu_{v}}\bar{E}_{v,1}) \frac{\sum_{j=1}^{2}p_{j1}\bar{E}_{h,j}}{\sum_{k=1}^{2}p_{k1}N_{h,k}} \\ \frac{a_{2}\beta_{hv}v_{h}}{(\mu_{v}+v_{v})(\mu_{h}+\gamma_{2})} (N_{v,2}-\bar{E}_{v,2}-\frac{v_{v}}{(1-q)\mu_{v}}\bar{E}_{v,2}) \frac{\sum_{j=1}^{2}p_{j2}\bar{E}_{h,j}}{\sum_{k=1}^{2}p_{k2}N_{h,k}} \end{pmatrix}$$

where $x = (\bar{E}_{h,1}, \bar{E}_{h,2}, \bar{E}_{v,1}, \bar{E}_{v,2}, \bar{I}_{h,1}, \bar{I}_{h,2})$. The function F(x) is continuous, bounded, differentiable and $F(0_{\mathbb{R}^6}) = 0_{\mathbb{R}^6}$. The function F is monotone if the corresponding Jacobian matrix is Metzler, i.e. all off-diagonal entries are nonnegative. We have

$$DF(x) = \begin{pmatrix} 0 & 0 & \tilde{M}_{\nu h}(x) \\ 0 & 0 & & \\ \hline & \\ \tilde{M}_{h\nu}(x) & -a_1 \beta_{h\nu} \left(1 + \frac{\nu_{\nu}}{(1-q)\mu_{\nu}} \right) \frac{\sum_{k=1}^2 \frac{p_{k1}\nu_h E_{h,k}}{\mu_h + \gamma_k}}{p_{11}N_{h,1} + p_{21}N_{h,2}} & 0 \\ & \\ 0 & -a_2 \beta_{h\nu} \left(1 + \frac{\nu_{\nu}}{(1-q)\mu_{\nu}} \right) \frac{\sum_{k=1}^2 \frac{p_{k2}\nu_h E_{h,k}}{\mu_h + \gamma_k}}{p_{12}N_{h,1} + p_{22}N_{h,2}} \end{pmatrix}$$

where

$$\tilde{m}_{\nu h}^{ij}(x) = \frac{\beta_{\nu h}^{2} \mu_{h} \nu_{\nu} a_{j} p_{ij} N_{h,j}}{(1-q) \mu_{\nu} (\mu_{h} + \nu_{h}) \sum_{k=1}^{2} p_{kj} N_{h,k}} \cdot \frac{1}{\frac{\beta_{\nu h}}{(1-q) \mu_{\nu}} \sum_{k=1}^{2} \frac{a_{k} p_{jk} E_{\nu,k}}{p_{1k} N_{h,1} + p_{2k} N_{h,2}} + \mu_{h}} \\ \cdot \left[1 - \frac{\frac{\beta_{\nu h}}{(1-q) \mu_{\nu}} \sum_{k=1}^{2} \frac{a_{k} p_{jk} E_{\nu,k}}{p_{1k} N_{h,1} + p_{2k} N_{h,2}}}{\frac{\beta_{\nu h}}{(1-q) \mu_{\nu}} \sum_{k=1}^{2} \frac{a_{k} p_{jk} E_{\nu,k}}{p_{1k} N_{h,1} + p_{2k} N_{h,2}} + \mu_{h}} \right]$$
(11)

and

$$\tilde{m}_{h\nu}^{ij}(x) = a_i \beta_{h\nu} \left(N_{\nu,i} - E_{\nu,i} - \frac{\nu_{\nu} E_{\nu,i}}{(1-q)\mu_{\nu}} \right) \frac{\nu_h p_{ji}}{(\mu_h + \gamma_i) \sum_{k=1}^2 p_{ki} N_{h,k}}$$

Since, $\tilde{m}_{vh}^{ij} \ge 0$ and $\tilde{m}_{hv}^{ij} \ge 0$ for all i, j = 1, 2, hence all off diagonal entries of the Jacobian matrix are nonnegative and so, the function F(x) is monotone; moreover,

$$\mathrm{DF}(0_{\mathbb{R}^4}) = \begin{pmatrix} 0 & 0 & \tilde{M}_{\nu h}(0) \\ 0 & 0 & \tilde{M}_{h\nu}(0) & 0 & 0 \\ \tilde{M}_{h\nu}(0) & 0 & 0 & 0 \end{pmatrix}.$$

This matrix is irreducible whenever $\tilde{M}_{vh}(0)\tilde{M}_{hv}(0)$ and $\tilde{M}_{hv}(0)\tilde{M}_{vh}(0)$ are irreducible. The latter is guaranteed since $M_{vh}M_{hv}$ and $M_{hv}M_{vh}$ (from the next generation matrix) are both irreducible. Hence, an application of Theorem 2.1 in Hethcote & Thieme (1985) implies that Model (10) has a unique positive fixed point *if and only if* $\rho(DF(0_{\mathbb{R}^4})) = \mathcal{R}_0 > 1$, or equivalently $\mathcal{R}_0^2 > 1$.

If the host-vector configuration is not irreducible, that is, the graphs associated with the matrices $M_{vh}M_{hv}$ and $M_{hv}M_{vh}$ are not strongly connected, the dynamics of the disease

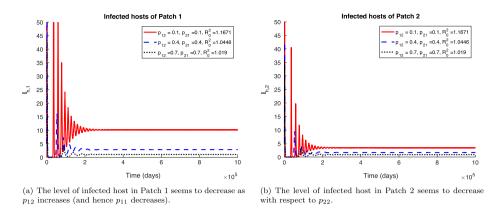


Figure 1. Dynamics of $I_{h,1}$ and $I_{h,2}$ for different values of p_{ij} .

within patches are either somehow independent or System (5) exhibits boundary equilibria. It is worthwhile noting that the irreducibility of residence times matrix \mathbb{P} does not imply the irreducibility of $M_{vh}M_{hv}$ and $M_{hv}M_{vh}$. Since the epidemiological and entomological parameters are all positive, the reducibility of the host–vector configuration happens only on the three following cases: (i) If the two patches are isolated, i.e. $p_{12} = p_{21} = 0$; (ii) residents of Patch 1 spend all their time in Patch 2 and residents of Patch 2 spend all their time in their own patch, i.e. $p_{12} = 1$ and $p_{21} = 0$; and (iii) the opposite scenario of (ii).

4. Simulations

Simulations are carried out in order to highlight the effects of residence times on disease dynamics. The simulations have a dual goal, first, to illustrate the theoretical results of this manuscript and secondly to illustrate the impact of host mobility across high- and low-risk dengue areas.

The basic reproduction number $\mathcal{R}_0(\mathbb{P})$ is a function of the residence times matrix \mathbb{P} . Simulation baseline values, except for those involving the entries of \mathbb{P} , are as follows:

$$\beta_{h\nu} = 0.5(0.001 - 0.54), \beta_{\nu h} = 0.41(0.3 - 0.9), \frac{1}{\mu_{\nu}} = 20(10 - 30) \text{ days},$$

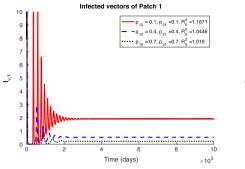
$$a_1 = 0.95 \text{ day}^{-1}, a_2 = 0.8 \text{ day}^{-1}, \frac{1}{\mu_h} = 60 \times 365 \text{ days}, \frac{1}{\gamma_1} = 7 \text{ days}, \frac{1}{\gamma_2} = 6 \text{ days}$$

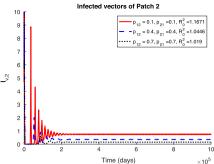
$$\frac{1}{\nu_h} = 5 \text{ days}, \frac{1}{\nu_{\nu}} = 7 \text{ days}.$$

The values of the parameters ν_h and ν_v are taken from Adams & Boots (2010), 2009 (2009). The infectiousness parameters ($\beta_{h\nu}$ and $\beta_{\nu h}$) and vector's natural mortality rate are taken from Chitnis, Hyman, and Manore (2013). Host and vector population are

$$N_{h,1} = 400,000, N_{h,2} = 300,000, N_{\nu,1} = 35,000, \text{ and } N_{\nu,2} = 30,000.$$

Patch 1 is the *high-risk* and Patch 2 is the *low-risk* and so, it is assumed that $a_1 > a_2$. Figure 1 represents the dynamics of Patch 1 (Figure 1(a)) and Patch 2 (Figure 1(b)) infected hosts while Figure 2 collects the vector dynamics in both patches. Since Patch 1 is high-risk, the number of infected host should decrease as p_{12} increases; see Figure 1(a). Figure 1(b)





(a) Asymptotically, the level of infected vectors in Patch 1 seems to decrease as p_{12} increases (and hence p_{11} decreases).

(b) Asymptotically, the level of infected vectors in Patch 2 seems to decrease with respect to p_{22} .

Figure 2. Dynamics of $I_{v,1}$ and $I_{v,2}$ for different values of p_{ij} .

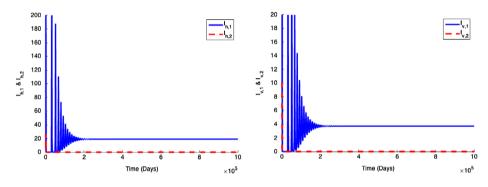


Figure 3. Dynamics of host and vectors if the host–vector configuration matrix is reducible.

shows the Patch 2 infected host population, which it is decreasing, as p_{21} and p_{12} increase. Disease prevalence among Patch 2 residents remains very small when compared to that in Patch 1. In Figure 2, Patch 1 (Figure 2(a)) and Patch 2 (Figure 2(b)) vector dynamics are seen to follow the hosts' endemicity pattern.

For all the different values of p_{ij} chosen in Figures 1 and 2, the host–vector configuration matrix

$$M = \begin{pmatrix} 0 & M_{\nu h} \\ M_{\nu h} & 0 \end{pmatrix}$$

or equivalently, the products $M_{h\nu}M_{\nu h}$ and $M_{\nu h}M_{h\nu}$, are irreducible. Moreover the basic reproduction number \mathcal{R}_0 is greater than one, hence the disease is, in both patches, at an endemic level.

Figure 3 displays the dynamics of the disease if the host-vector configuration matrix M is not irreducible. The disease dies out in Patch 2 where the basic reproduction number is $\mathcal{R}_{2,0}^2 = .8161$ and persists in Patch 1 for which $\mathcal{R}_{1,0}^2 = 1.1747$.

Figures 4 and 5 highlight, respectively, the effects of the vertical transmission on the dynamics of infected hosts and vectors in both patches. These figures also provide how the basic reproduction number changes with respect to the vertical transmission fraction *q*. By considering the same epidemic parameters as in Figures 1 and 2 and with residence

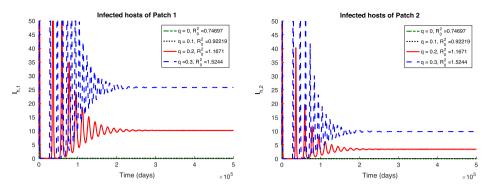


Figure 4. Dynamics of hosts in Patch 1 and Patch 2 with varying vertical transmission fraction q.

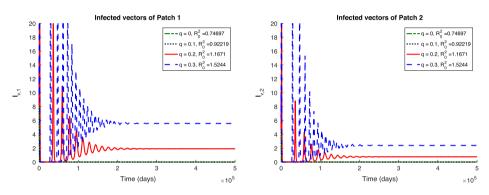


Figure 5. Dynamics of vectors in Patch 1 and Patch 2 with varying vertical transmission fraction q.

times fixed as $p_{11} = p_{22} = .9$ (or equivalently $p_{12} = p_{21} = .1$), the value q = .2 leads to an endemic steady state with $\mathcal{R}_0^2 = 1.1671$ as evidenced by Figures (1, 2) and Figures (4, 5), red solid lines. However, a 10% increase in the number of mosquitoes infected due to vertical transmission (q) causes a noticeable increase in the level of infected hosts and mosquitoes. And a 10% decrease in the value of q stirs the system from an endemic state to a disease-free state (see Figures (4) and (5), black dotted and dash-dot lines). These remarks showcase how reducing vertical transmission in the vector population, namely the use of larvicides could be an effective control strategy in mitigating or eliminating Dengue prevalence in endemic areas.

5. Colima City and Manzanillo dengue inspired simulation study

Ae. aegypti was declared eradicated in Mexico in 1963. Not surprisingly, all four dengue serotypes (DENV-1, DENV-2, DENV-3 and DENV-4) re-emerged two years after local the 1963 eradication (Díaz et al., 2006). Further, DHF cases have steadily increased since 1994 (Navarrete-Espinosa, Gómez-Dantés, Germán Celis-Quintal, & Vázquez-Martínez, 2005) Dengue is endemic in Mexico with approximately 60% of year-round cases reported in the southern part of the country; a region characterized by a warm and humid climate (Colón-González, Lake, & Bentham, 2011). Colima, located on the central Pacific Coast



Figure 6. The state of Colima is located on the central Pacific coast of Mexico.

Notes: It has a tropical climate, a surface of 5,455 km² and a population of approximately 488,028 inhabitants. The state of Colima is divided in 10 municipalities. Manzanillo, where the 2002 outbreak began, and Colima City are labelled in the map. $p_{11} = 0.99$, $p_{22} = 1.0$ with Manzanillo being represented with Patch 1 and Colima with Patch 2.

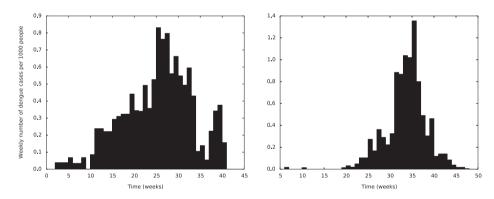


Figure 7. Incidence of dengue cases per weekly during the 2002 dengue epidemic diagnosed at the hospitals of the Mexican Institute of Public Health (IMSS) (Chowell, Diaz-Dueñas, Miller, et al., 2007) in Manzanillo (left) and Colima City (right), respectively.

(see Figure 6), is also a reservoir of Dengue. In 2002, the State of Colima reported 4,040 cases dengue in all of its 10 municipalities; 495 progressing to DHF (Chowell, Diaz-Dueñas, Chowell, et al., 2007; Espinoza-Gómez et al., 2005). DENV-2 was isolated from patients during this outbreak (Espinoza-Gómez et al., 2005). The increase in DHF cases in Mexico has been linked to the introduction of DENV-2 Asian, previously isolated in 2000 and again in 2002 (Lorono-Pino et al., 2004).

The dynamics of dengue are explored in the context of this 2002 State of Colima outbreak. The first reported (index) case was identified as that of a 10-year-old female

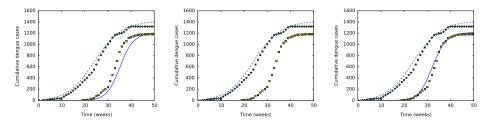


Figure 8. Circles: cumulative dengue cases reported on Manzanillo, dotted line: model prediction of Manzanillo cases, squares: cumulative dengue cases reported on Colima city, solid line: model prediction of cumulative cases in Colima city.

Notes: Left: $p_{11} = 1.0$, $p_{22} = 0.9996$, Centre: $p_{11} = 1.0$, $p_{22} = 1.0$, Right: $p_{11} = p_{22} = 0.9996$ Patch 1 represents Manzanillo and Patch 2 Colima.

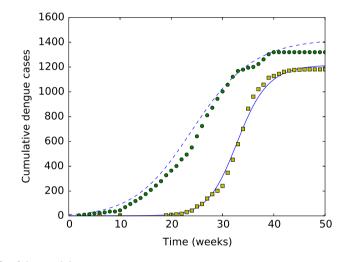


Figure 9. Best fit of the model $p_{11} = 0.9996$, $p_{22} = 1.0$. Notes: Circles: cumulative dengue cases reported on Manzanillo, dotted line: model prediction of Manzanillo cases, squares: cumulative dengue cases reported on Colima City, solid line: model prediction of cumulative cases in Colima City.

in the municipality of Manzanillo on 11 January 2002. Dengue infection spread throughout the whole state with the most affected municipalities being Colima city, the capital of the state, and Manzanillo, an important tourist destination in the coast (Espinoza-Gómez et al., 2005). The city of Colima reported approximately 1,167 dengue cases, with 169 cases progressing to DHF while Manzanillo, reported 1,334 dengue cases, with 123 progressing to DHF in 2002 (Chowell, Diaz-Dueñas, Miller, et al., 2007). The city of Colima and Manzanillo are linked via high levels of travel and tourism. Both cities account for approximately 47% of the state population. We apply a two-patch model to explore the role that movement, modelled via the matrix p_{ij} , may have had on dengue disease transmission during this 2002 outbreak. The estimated population of Manzanillo and Colima City was $N_{h,1} = 1,355$ and $N_{h,2} = 1,184$, respectively, and the initial mosquito populations were choosen to best fit the data. They were approximately 308 and 738 in Manzanillo and Colima City, respectively. Note that the host population is not the actual population of the cities but rather the population at risk in each of the corresponding cities. The population at risk is much smaller that the actual population because in the same city there are

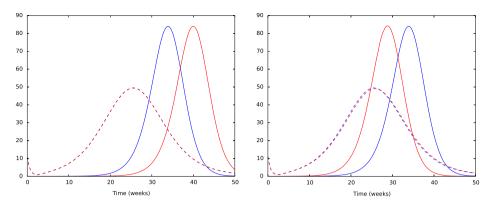


Figure 10. Dengue cases predicted by the model for Manzanillo (dotted line) and Colima City (solid line). Notes: In the first scenario (on the left), the blue lines represent no transit control and the red lines represent a reduction of 90% in movement from Manzanillo City to Colima. In the second scenario (on the right), the blue lines represent no movement control and the red lines represent an increment of movement from Colima to Manzanillo City of 1%.

social groups practically disconnected to others by geographic, cultural and social factors. Entomological parameters were estimated using Yang, Macoris, Galvani, and Andrighetti (2011) and taking into account the mean temperature in each region (Chowell, Diaz-Dueñas, Miller, et al., 2007). The remaining parameters used to study the outbreak in Colima, Mexico were obtained from the literature (Adams & Boots, 2010; Chitnis et al., 2013; García Rivera & Rigau-Pérez, 2006; Yang et al., 2011):

$$\beta_{h\nu}a_1 = 0.43 \text{ days}^{-1}, \beta_{h\nu}a_2 = 0.34 \text{ days}^{-1}, \mu_{\nu} = 0.036 \text{ (Colima)},$$

0.030 (Manzanillo) days⁻¹, $\frac{1}{\mu_h} = 60 \times 365 \text{ days}, \gamma_1 = 0.2 \text{ days}^{-1}, \gamma_2 = 0.2 \text{ days}^{-1},$
 $\nu_h = 0.18 \text{ days}^{-1}, \nu_{\nu} = 0.1 \text{ days}^{-1}.$

In order to assess, within our staged scenarios, the impact of migration during the 2002 dengue outbreak, we fit the two-patch model using the incidence data for Manzanillo and the city of Colima reported by the Mexican Social Security Institute (IMSS) during the outbreak (see Figure 7). The data fitting for cumulative dengue cases given by the model using 'scipy.optimize.curve_fit' library of python v2.7 programming language, is shown in Figure 9. Model results show that dengue spreads more quickly in the city of Colima when the proportion of visits from Manzanillo's infected residents is high, see the left panel of Figure 8 compared with Figure 9. Alternatively, susceptible Colima City residents would acquire dengue infections over a longer time frame in Manzanillo, introducing the disease over a slower time scale in their home residence, the city of Colima. Of course, the absence of movement leads to no dengue cases in Manzanillo; an outbreak occurring only in Colima $p_{11} = p_{22} = 1.0$, see centre panel of Figure 8; equal movement, $p_{11} = p_{22}$, would cause the outbreak in Colima to grow faster, as can be seen in the right panel of Figure 8. Hence, limiting the movement of the Manzanillo population seems like a good strategy while limiting the movement of the Colima population wouldn't be as effective. In the latest scenario, the economic cost would be high since Manzanillo is a tourist destination.

We can also observe in Figure 10 (on the left), that the effect of reducing the transit from Manzanillo to Colima city led only to a delay in the appearance of the outbreak in

Colima. This indicates that the outbreak in Colima followed its own local dynamics and that transit between these two cities only led to delays in the introduction of the dengue virus without affecting the local outbreak dynamics. When the average visiting time spent in a place where the disease prevalence is low (small value of p_{ij} , $i \neq j$) then the only way of reducing an outbreak would require strict migration control, that is, complete travel avoidance to the high risk zone. In Figure 10 (on the right), we see that with only a small fraction of visitors from Manzanillo to Colima, the outbreaks in both cities occur almost simultaneously. Model simulations re-affirm the views that the rate of host movement and time spent in endemic geographic regions are important for the spread of dengue between two patches. The question then becomes, why aren't then these residence times estimated?

6. Conclusion

The persistence of vector-borne diseases, such as dengue, is connected to factors that include the presence of ecological conditions that favour high vector densities, vector-host interactions, the spatial movement of humans, and of course, the effectiveness of control measures (Martens & Hall, 2000; Sutherst, 2004). In this paper, a two-patch host-vector model was used to study the role of movement on the transmission dynamics of dengue, especially DENV-2. We focus on the applications of our framework to scenarios where dengue is endemic and where vertical transmission has been documented. A residence times matrix \mathbb{P} is used to model host mobility. This modelling approach provides a framework for exploring spatial vector-borne disease dynamics and control within relatively 'close' environments. Analytical results were derived and the conditions for which the disease dies out or persists have been identified; conditions that depend on whether the basic reproduction number $\mathcal{R}_0(\mathbb{P})$ is less or greater than unity and the connectivity of patches.

Using data from the 2002 DENV-2 outbreak in Colima, Mexico, we compare the overall prevalence in the cities of Colima and Manzanillo as a function of pre-selected \mathbb{P} matrices. Our model shows that reducing travelling from to Colima ity, considered high-risk and the place of the 2002 outbreak onset, caused a slight delay in the spread of the disease. In order to completely prevent an outbreak in Colima City, migration between Colima city and Manzanillo must be stopped. Manzanillo a tourist destination implies that transit from Colima City to Manzanillo is expected to peak during certain seasons. The model suggests that dengue would become endemic in both patches almost simultaneously. The two-patch model highlights the role of human spatial movement on disease transmission and control. The strength of this effect depends on the proportion of time commuters to high or low risk spend in each patch.

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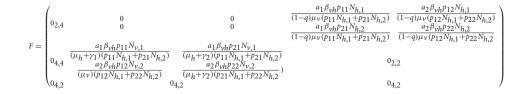
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Appendix 1. The basic reproduction number

Let $x = (E_{h,1}, E_{h,2}, E_{\nu,1}, E_{\nu,2}, I_{h,1}, I_{h,2}, I_{\nu,1}, I_{\nu,2})$ and so the relevant \mathcal{F} and \mathcal{V} are

$$\mathcal{F} = \begin{pmatrix} \frac{a_{1}\beta_{vh}p_{11}S_{h,1}I_{v,1}}{p_{11}N_{h,1}+p_{21}N_{h,2}} + \frac{a_{2}\beta_{vh}p_{12}S_{h,1}I_{v,2}}{p_{12}N_{h,1}+p_{22}N_{h,2}} \\ \frac{a_{1}\beta_{vh}p_{21}S_{h,2}I_{v,1}}{p_{11}N_{h,1}+p_{21}N_{h,2}} + \frac{a_{2}\beta_{vh}p_{22}S_{h,2}I_{v,2}}{p_{12}N_{h,1}+p_{22}N_{h,2}} \\ a_{1}\beta_{hv}S_{v,1}\frac{p_{11}I_{h,1}+p_{21}N_{h,2}}{p_{11}N_{h,1}+p_{21}N_{h,2}} \\ a_{2}\beta_{hv}S_{v,2}\frac{p_{12}I_{h,1}+p_{22}N_{h,2}}{p_{12}N_{h,1}+p_{22}N_{h,2}} \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} \text{ and } \mathcal{V} = \begin{pmatrix} -(\mu_{h}+\nu_{h})E_{h,1} \\ -(\mu_{h}+\nu_{h})E_{h,2} \\ -(\mu_{\nu}+\nu_{\nu})E_{\nu,2} \\ \nu_{h}E_{h,1} - (\mu_{\mu}+\gamma_{i})I_{h,1} \\ \nu_{h}E_{h,2} - (\mu_{\mu}+\gamma_{i})I_{h,2} \\ \nu_{\nu}E_{h,1} - (1-q)\mu_{\nu}I_{\nu,2} \end{pmatrix}.$$

Let $F \equiv D\mathcal{F}$ and $V \equiv D\mathcal{V}$ evaluated at the DFE. We obtain



and

$$V = \begin{pmatrix} -\mu_h - \nu_h & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & -\mu_h - \nu_h & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & -\mu_\nu - \nu_\nu & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & -\mu_\nu - \nu_\nu & 0 & 0 & 0 & 0 \\ \nu_\nu & 0 & 0 & 0 & -\mu_h - \gamma_1 & 0 & 0 & 0 \\ 0 & \nu_\nu & 0 & 0 & 0 & \mu_h - \gamma_h & 0 & 0 \\ 0 & 0 & \nu_\nu & 0 & 0 & 0 & -(1-p)\mu_\nu & 0 \\ 0 & 0 & 0 & \nu_\nu & 0 & 0 & 0 & -(1-p)\mu_\nu \end{pmatrix}$$

The basic reproduction number is the spectral radius of the matrix

where

and

$$M_{\nu h} = \begin{pmatrix} \frac{a_1 \beta_{\nu h} p_{11} N_{h,1} v_{\nu}}{(p_{11} N_{h,1} + p_{21} N_{h,2})(\mu_{\nu} + v_{\nu})(1-q)\mu_{\nu}} & \frac{a_2 \beta_{\nu h} p_{12} N_{h,1} v_{\nu}}{(p_{12} N_{h,1} + p_{22} N_{h,2})(\mu_{\nu} + v_{\nu})(1-q)\mu_{\nu}} \\ \frac{a_1 \beta_{\nu h} p_{21} N_{h,2} v_{\nu}}{(p_{11} N_{h,1} + p_{21} N_{h,2})(\mu_{\nu} + v_{\nu})(1-q)\mu_{\nu}} & \frac{a_2 \beta_{\nu h} p_{22} N_{h,2} v_{\nu}}{(p_{12} N_{h,1} + p_{22} N_{h,2})(\mu_{\nu} + v_{\nu})(1-q)\mu_{\nu}} \end{pmatrix}$$
$$M_{h\nu} = \begin{pmatrix} \frac{a_1 \beta_{h\nu} p_{11} N_{\nu,1} v_h}{(p_{11} N_{h,1} + p_{21} N_{h,2})(\mu_h + v_h)(\mu_h + \gamma_1)} & \frac{a_1 \beta_{h\nu} p_{21} N_{\nu,1} v_h}{(p_{11} N_{h,1} + p_{22} N_{h,2})(\mu_h + v_h)(\mu_h + \gamma_2)} \\ \frac{a_2 \beta_{\mu \nu} p_{22} N_{\nu,2} v_h}{(p_{12} N_{h,1} + p_{22} N_{h,2})(\mu_h + v_h)(\mu_h + \gamma_1)} & \frac{a_1 \beta_{\mu \nu} p_{21} N_{\nu,1} v_h}{(p_{12} N_{h,1} + p_{22} N_{\nu,2})(\mu_h + v_h)(\mu_h + \gamma_2)} \end{pmatrix}$$

The basic reproduction number \mathcal{R}^2_0 is defined by the expression,

$$\mathcal{R}_0^2 = \rho(M_{\nu h} M_{h\nu}).$$