

Stochastic Multistrain Dengue model

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Multi-strain dynamics

In this paper we propose a stochastic 4-serotype SIR model with cross-immunity to describe multi-strain Dengue dynamics. The model permits up to 4 dengue infections with reduced susceptibility after the first Dengue episode due to cross immunity. Immunity to each serotype is considered complete and permanent. Figure 1 depicts all 48 possible states and 64 state-transitions included in the model.

Let S be individuals susceptible to all 4 types of dengue, I_i infectious with Dengue type i and R_i individuals recovered from Dengue type i . Infectious individuals already on their secondary and later Dengue infections are represented by multiple indices. For example, $I_{[23]1}$ is an individual which has had Dengues type 2 and 3 in the past – and therefore is immune to them – and is currently transmitting Dengue 1. The index outside the bracket denotes current infection. Recovered individuals indices denote their immunity, so for instance R_{123} is an individual which is immune to Dengue types 1, 2 and 3, but not to 4. Let $I_{*i} = \sum I_{[...]}i$ with $[...]$ representing exposure history of the infected individual which can vary from 0 to 3 in length. All individuals are born to the S state and birth and death rates are equal.

The possible state-transitions and their propensities are listed in table 1.

The model is implemented as a continuous time Markov jump process. Let

$$\vec{X}(t) = [S(t), I_1(t), I_2(t), \dots, R_{1234}(t)]$$

be the state of the system at the time t . $\sum X(t) = N, \forall t$ with N being the population size. The system is written as a forward Kolmogorov differential equation, which in matrix form looks like

$$\frac{dP_X(t)}{dt} = QP_X(t) \quad (1)$$

Where $P_X(t)$ is the matrix of transition probabilities (given in table 1) and Q is the generator matrix, whose non-zero values are also given in table 1 (in the state change column). The full formula and matrices are omitted due to their large sizes.

Expected Change and Covariance Matrix

It is useful to calculate the expected change and the covariance matrix for the changes $\Delta X = [\Delta S, \Delta I_1, \Delta I_2, \Delta I_3, \Delta I_4, \Delta R_1, \Delta R_2, \Delta R_3, \Delta R_4, \Delta I_{12}, \Delta I_{13}, \Delta I_{14}, \Delta I_{21}, \Delta I_{23}, \Delta I_{24}, \Delta I_{31},$

$\Delta I_{32}, \Delta I_{34}, \Delta I_{41}, \Delta I_{42}, \Delta I_{43}, \Delta R_{12}, \Delta R_{13}, \Delta R_{14}, \Delta R_{23}, \Delta R_{24}, \Delta R_{34}, \Delta I_{231}, \Delta I_{241}, \Delta I_{341}, \Delta I_{132}, \Delta I_{142}, \Delta I_{342}, \Delta I_{123}, \Delta I_{143}, \Delta I_{243}, \Delta I_{124}, \Delta I_{134}, \Delta I_{234}, \Delta R_{123}, \Delta R_{124}, \Delta R_{134}, \Delta R_{234}, \Delta I_{1234}, \Delta I_{1243}, \Delta I_{1342}, \Delta I_{2341}, \Delta R_{1234}]^T$.

Thus from the probabilities of table 1,

$$E(\Delta X) = \sum_{i=1}^{114} p_i(\Delta X)_i = \begin{pmatrix} -\beta S I_{*1} + \mu N - \mu S \\ \vdots \\ \beta S I_{*4} + \mu N - \mu S \\ \beta S I_{*1} - \sigma I_1 - \mu I_1 \\ \vdots \\ \beta S I_{*4} - \sigma I_4 - \mu I_4 \\ \sigma I_1 - \beta \delta R_1 I_{*j} - \mu R_1 \\ \vdots \\ \sigma I_4 - \beta \delta R_4 I_{*j} - \mu R_4 \\ \beta \delta R_i I_{[i]j} - \sigma I_{[i]j} - \mu I_{[i]j} \\ \sigma I_{[i]j} - \beta \delta R_{ij} I_{*k} - \mu R_{ij} \\ \beta \delta R_{ij} I_{*k} - \sigma I_{[ij]k} - \mu I_{[ij]k} \\ \sigma I_{[ij]k} - \beta \delta R_{ijk} I_{*l} - \mu R_{ijk} \\ \beta \delta R_{ijk} I_{*l} - \sigma I_{[ijk]l} - \mu I_{[ijk]l} \\ \sigma I_{[ijk]l} - \mu R_{1234} \end{pmatrix} \Delta t \quad (2)$$

The expected change, $E(\Delta X)$, is a 48×1 vector. The covariance matrix is a 48×48 matrix, $\Sigma(\Delta X) = \sum_{i=1}^{114} p_i(\Delta X)_i (\Delta X)_i^T$