

# A model for in-host viral infection dynamics

Satvik Saha

## 1 Model descriptions

### 1.1 Agent based model

Let  $n, m \in \mathbb{N}$ . For each  $i \in \{1, \dots, n\}$ ,  $j \in \{1, \dots, m\}$ . Consider the following system.

$$\frac{dT_i}{dt} = b - \delta T_i - \frac{\kappa}{1 + \alpha A_i} T_i V_i, \quad (1)$$

$$\frac{dT_i^*}{dt} = \frac{\kappa}{1 + \alpha A_i} T_i V_i - q T_i^*, \quad (2)$$

$$\frac{dV_i}{dt} = p T_i^* - c V_i - c_A A_i V_i - X_i(t) + g \left( W_i + \sum_{j=1}^m \eta_{ij} Z_j \right), \quad (3)$$

$$\frac{dA_i}{dt} = b_A - \delta_A A_i + \kappa_A A_i (t - \tau) V_i(t - \tau), \quad (4)$$

$$\frac{dZ_j}{dt} = \sum_{i=1}^n \xi_{ij} V_i - \delta_Z Z_j, \quad (5)$$

$$W_i(t) = \zeta \sum_{k=1}^n Y_{ik}(\lfloor t \rfloor) V_k(\lfloor t \rfloor), \quad (6)$$

$$g(x) = x \mathbf{1}_{(v, \infty)}(x). \quad (7)$$

Here, we define random variables  $X_i(t) \sim \text{Exp}(\lambda)$ ,  $Y_{ik}(\lfloor t \rfloor) \sim \text{Bernoulli}(s_{ik} p_{\text{inf}})$ .

The parameters  $\xi_{ij}, \eta_{ij}$  are to be thought of as weights linking agents with their environments; the parameters  $s_{ik}$  are to be thought of as strengths of connections between agents forming a network.

The model state is described by  $(T, T^*, V, A, W, Z) \in \mathcal{S} \equiv \mathbb{R}_{\geq 0}^{3n} \times \mathcal{C}_\tau \times \mathcal{C}_\tau \times \mathbb{R}_{\geq 0}^m$ , with  $\mathcal{C}_\tau \equiv \mathcal{C}(0, \tau)$ ,  $T \equiv (T_1, \dots, T_n)$ , and so on.

The model parameters are  $(b, \delta, \kappa, q, p, c, b_A, \delta_A, \kappa_A, c_A, \alpha, \tau, \delta_Z, v, \lambda, p_{\text{inf}}, \zeta, \eta, \xi, s) \in \mathcal{P} \equiv \mathbb{R}_{\geq 0}^{17} \times \mathbb{R}_{\geq 0}^{mn} \times \mathbb{R}_{\geq 0}^{mn} \times [0, 1]^{n \times n}$ , with  $\eta \equiv [\eta_{ij}]_{ij}$ ,  $\xi \equiv [\xi_{ij}]_{ij}$ , and  $s \equiv [s_{ik}]_{ik}$ .

Table 1: Model state variables

Variable	Units	Interpretation
$T$	cells/ml	Concentration of target cells
$T^*$	cells/ml	Concentration of infected cells
$V$	copies/ml	Concentration of viral copies
$A$	imm/ml	Antibody/immunity level
$W$	copies/ml	Contact pressure of viral copies
$Z$	copies/m <sup>2</sup>	Environmental viral copies

Table 2: Model parameters

Parameter	Units	Interpretation
$b$	cells ml <sup>-1</sup> day <sup>-1</sup>	Generation rate of target cells
$\delta$	day <sup>-1</sup>	Death rate of target cells
$\kappa$	cells <sup>-1</sup> ml day <sup>-1</sup>	Infection rate of target cells
$q$	day <sup>-1</sup>	Death rate of infected cells
$p$	copies cells <sup>-1</sup> day <sup>-1</sup>	Production rate of viral copies
$c$	day <sup>-1</sup>	Clearance rate of viral copies
$b_A$	imm ml <sup>-1</sup> day <sup>-1</sup>	Generation rate of antibodies
$\delta_A$	day <sup>-1</sup>	Clearance rate of antibodies
$\kappa_A$	copies <sup>-1</sup> ml day <sup>-1</sup>	Production rate of antibodies
$c_A$	imm <sup>-1</sup> ml day <sup>-1</sup>	Clearance rate of viral copies via antibodies
$\alpha$	imm <sup>-1</sup> ml	Inhibition of viral-target contact
$\tau$	day	Delay in antibody production
$\delta_Z$	day <sup>-1</sup>	Removal rate of viral copies
$v$	copies ml <sup>-1</sup>	Entry threshold of viral concentration
$\lambda$	copies <sup>-1</sup> ml day	Reciprocal of mean of stochastic viral removal
$p_{\text{inf}}$	–	Probability of viral load transfer
$\zeta$	–	Fraction of viral load transferred
$\eta_{ij}$	copies ml <sup>-1</sup> cells <sup>-1</sup> m <sup>2</sup> day <sup>-1</sup>	Environment-Agent transmission rate of virus
$\xi_{ij}$	copies <sup>-1</sup> ml cells m <sup>-2</sup> day <sup>-1</sup>	Viral shedding rate into environment
$s_{ik}$	–	Strength of contact between agents

After choosing thresholds  $V'$  and  $A'$ , we can count

$$S = \sum_{i=1}^n \mathbf{1}(A \leq A') \mathbf{1}(V \leq V') \quad (9)$$

$$I = \sum_{i=1}^n \mathbf{1}(V > V'), \quad (10)$$

$$R = n - S - I. \quad (11)$$

## 1.2 In-host submodel

Consider the in-host model described below.

$$\begin{aligned}
\frac{dT}{dt} &= b - \delta T - \frac{\kappa}{1 + \alpha A} TV, \\
\frac{dT^*}{dt} &= \frac{\kappa}{1 + \alpha A} TV - qT^*, \\
\frac{dV}{dt} &= pT^* - cV - c_A AV, \\
\frac{dA}{dt} &= b_A - \delta_A A + \kappa_A A(t - \tau)V(t - \tau).
\end{aligned}$$

Solving for an equilibrium, we demand

$$\frac{\kappa}{1 + \alpha A} TV = b - \delta T = qT^*, \quad pT^* = (c + c_A A)V, \quad b_A - \delta_A A = -\kappa_A AV.$$

Thus,

$$\frac{p}{q}(b - \delta T) = -\frac{(c + c_A A)(b_A - \delta_A A)}{\kappa_A A},$$

whence

$$T = \frac{b}{\delta} + \frac{q(c + c_A A)(b_A - \delta_A A)}{p\delta\kappa_A A}.$$

Furthermore,

$$1 + \alpha A = \frac{\kappa TV}{b - \delta T} = -T \frac{(b_A - \delta_A A)/\kappa_A}{(b - \delta T)/\kappa},$$

whence

$$\frac{b - \delta T}{\kappa T} = -\frac{b_A - \delta_A A}{\kappa_A(1 + \alpha A)}.$$

Thus,

$$T = \frac{q(c + c_A A)(1 + \alpha A)}{p\kappa A}.$$

This gives

$$\frac{b}{\delta} = \frac{q(c + c_A A)}{pA} \left[ \frac{1 + \alpha A}{\kappa} - \frac{b_A - \delta_A A}{\delta\kappa_A} \right].$$

Putting  $T_0 = b/\delta$ ,  $T = b_A/\delta_A$ , we have

$$pAT_0 = q(c + c_A A) \left[ \frac{1 + \alpha A}{\kappa} - \frac{\delta_A(A_0 - A)}{\delta\kappa_A} \right],$$

whence

$$p\kappa T_0 A = q(c + c_A A) \left[ 1 - \frac{\kappa/\delta}{\kappa_A/\delta_A} A_0 + \left( \alpha + \frac{\kappa/\delta}{\kappa_A/\delta_A} \right) A \right].$$

Setting  $\beta = (\kappa/\delta)/(\kappa_A/\delta_A)$ ,  $r = p/q$ ,  $\gamma = c_A/c$ , we have

$$\kappa r T_0 A = c(1 + \gamma A)[1 - \beta A_0 + (\alpha + \beta)A].$$

Thus,

$$\gamma(\alpha + \beta)A^2 + [\gamma(1 - \beta A_0) + (\alpha + \beta) - \kappa r T_0/c]A + (1 - \beta A_0) = 0,$$

or

$$A^2 + \left[ \frac{1 - \beta A_0}{\alpha + \beta} + \frac{1}{\gamma} - \frac{\kappa r T_0}{c\gamma(\alpha + \beta)} \right] A + \frac{1 - \beta A_0}{\gamma(\alpha + \beta)} = 0.$$

### 1.3 Multiscale model

$$\frac{dS}{dt} = -\beta_I(V, I)SI - \beta_Z(Z)SZ + \mu R, \quad (12)$$

$$\frac{dI}{dt} = \beta_I(V, I)SI + \beta_Z(Z)SZ - \gamma I, \quad (13)$$

$$\frac{dR}{dt} = \gamma I - \mu R, \quad (14)$$

$$\frac{dZ}{dt} = \xi I - \delta_Z Z, \quad (15)$$

$$\frac{dT}{dt} = b - \delta T - \frac{\kappa}{1 + \alpha A}TV, \quad (16)$$

$$\frac{dT^*}{dt} = \frac{\kappa}{1 + \alpha A}TV - qT^*, \quad (17)$$

$$\frac{dV}{dt} = \eta Z + pT^* - cV - c_A AV, \quad (18)$$

$$\frac{dA}{dt} = b_A - \delta_A A + \kappa_A A(t - \tau)V(t - \tau). \quad (19)$$

Here,

$$\beta_I(V, I) = \frac{\beta_{I0} + C_0 V}{1 + C_1 I}, \quad \beta_Z(Z) = \frac{\beta_{Z0}}{1 + C_2 Z}.$$

### 1.4 SIRS model

$$\frac{dS}{dt} = -\frac{\beta SI}{N} + \mu R, \quad (22)$$

$$\frac{dI}{dt} = \frac{\beta SI}{N} - \gamma I, \quad (23)$$

$$\frac{dR}{dt} = \gamma I - \mu R. \quad (24)$$

Here,  $N = S + I + R$ .

## 2 Objectives

1. Compare the  $S, I, R$  curves with those obtained from a simplified model with one agent and one environment.
2. Identify/interpret infection phases ( $S, I, R$ ) using the in-host variables ( $T, T^*, V, A$ ).
3. Investigate the effects of heterogeneity in the agents and their contact network. For instance,
  - (a) In-host parameters may be varied across agents, forming two or more groups.
  - (b) Groups of agents may be vaccinated.
4. Investigate the effect of the stochastic term  $X_i(t)$  in the in-host model.

### 3 Observations

1. The agent based model (1.1) is capable of producing infection curves with multiple waves/peaks.
2. Averaged infection curves from model (1.1) also show multiple peaks; the curve up to the first peak fits well against the SIRS model (1.4). Individual infection curves from model (1.1) up to the first peak also fit well against model (1.4).
3. There is a narrow range of  $p_{\text{inf}}$ , with all other parameters in model (1.1) fixed, in which a significant proportion of infection curves display multiple prominent peaks without damping. For lower  $p_{\text{inf}}$ , infection curves become more stochastic in nature.
4. Individuals in model (1.1) become ‘infected’ when a pulse is applied on  $W_i$ . The viral load  $V_i$  rapidly increases, which after a short delay leads to a rapid increase in the antibody/immunity  $A_i$ . This forces  $V_i$  to fall sharply to zero, after which  $A_i$  gradually drops back to its baseline level. A sufficiently elevated  $A_i$  confers ‘immunity’ to the individual, preventing reinfection. The probability of reinfection, as a function of time since infection, can be calculated.

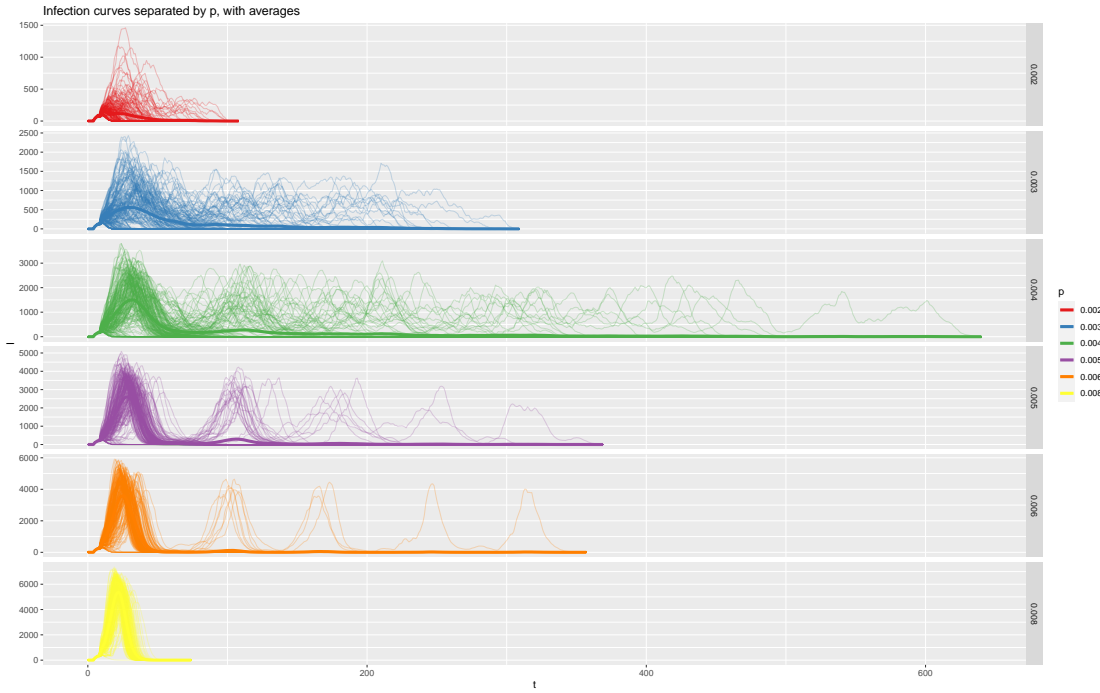


Figure 1: Infection curves, by varying infection probabilities  $p_{\text{inf}}$ .

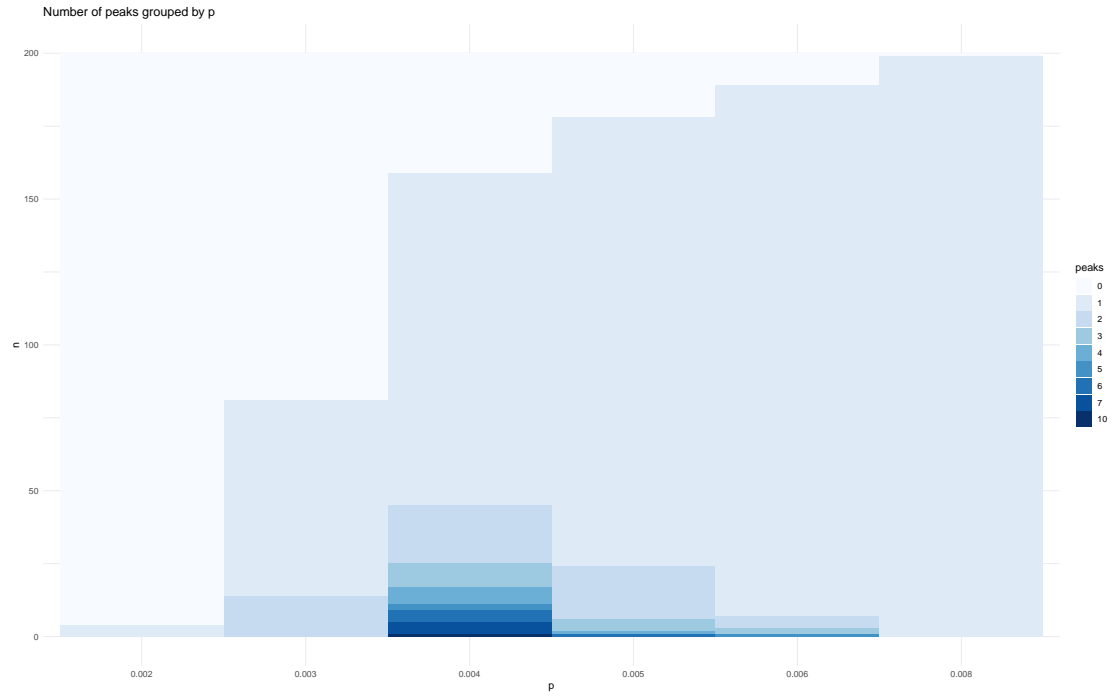


Figure 2: Number of runs (out of 200) with  $n$  peaks in the infection curve, by varying infection probabilities  $p_{\text{inf}}$ .

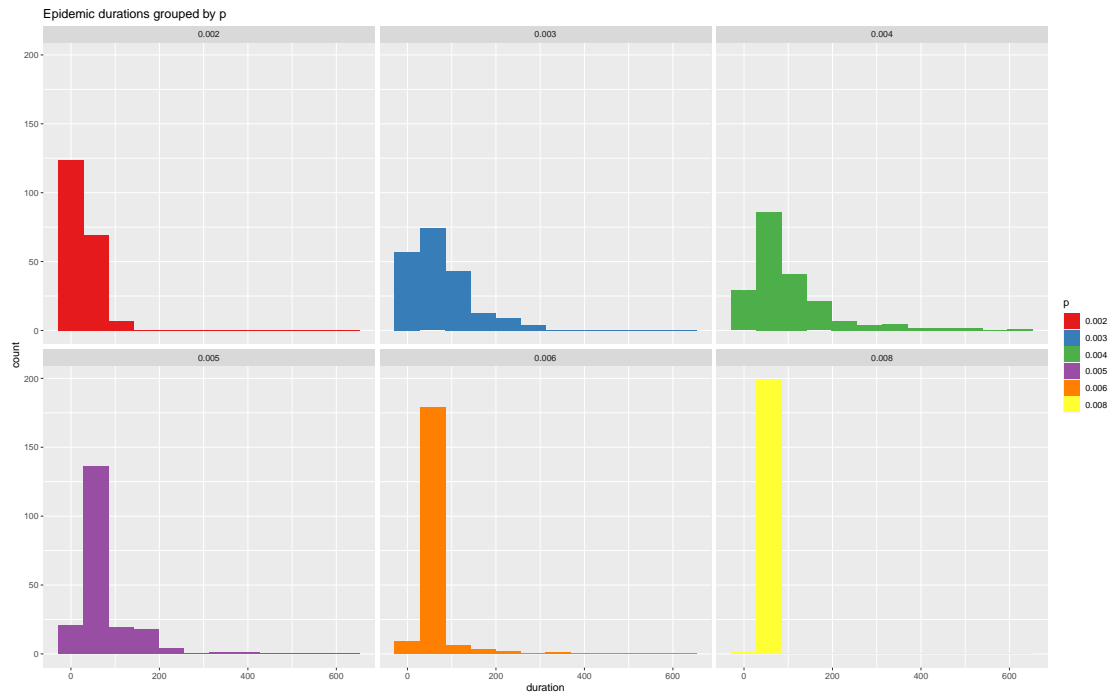


Figure 3: Distribution of epidemic durations, by varying infection probabilities  $p_{\text{inf}}$ .