Compartmental epidemic models with nonlinear recovery

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1 The model

Assume the recovery rate, $\mathcal{K}(I)$, is a smooth, non-negative function on [0,1].

$$\frac{\mathrm{d}S}{\mathrm{d}\tau} = \varepsilon (1 - S) - \rho SI \tag{1a}$$

$$\frac{\mathrm{d}I}{\mathrm{d}\tau} = \rho SI - \mathcal{K}(I)I \tag{1b}$$

$$\frac{\mathrm{d}R}{\mathrm{d}\tau} = I - \varepsilon(1 - S) \tag{1c}$$

where S + I + R = 1. In terms of the usual SIR parameters, the dimensionless parameters

16 above are

$$\varepsilon = \frac{\mu}{\gamma + \mu} \,, \tag{2a} \quad \{\mathtt{E:eps}\}$$

$$\rho = \frac{\beta N}{\gamma + \mu} \,. \tag{2b} \quad \{\texttt{E:rho}\}$$

If $\mathcal{K}(I) \equiv 1$ then ε is the mean time in the infectious class as a fraction of the of the mean lifetime and ρ is \mathcal{R}_0 . Note that $0 \leq \varepsilon < 1$ necessarily. I will write initial conditions as (S_i, I_i, R_i) .

23 **Equilibria**

There are typically two equilibria:

$_{\scriptscriptstyle 25}$ 2.1 Disease free equilibria (DFE)

26 As for the standard KM SIR model:

Unique if $\varepsilon > 0$: (S, I) = (1, 0).

Continuum of DFEs if $\varepsilon = 0$: $(S_0, 0)$ is a DFE for any $S_0 \in [0, 1]$.

29 2.2 Endemic equilibria (EE)

30 $(S, I) = (\hat{S}, \hat{I}).$

$$0 = \varepsilon (1 - \hat{S}) - \rho \hat{S} \hat{I} \tag{3} \quad \{E:\}$$

$$0 = \rho \hat{S} - \mathcal{K}(\hat{I}) \tag{4}$$

34 Hence

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$$\hat{S} = \frac{\mathcal{K}(\hat{I})}{\rho} \tag{5} \quad \{\texttt{E:Shat}\}$$

$$\mathcal{K}(\hat{I}) = \frac{\varepsilon}{\hat{I} + (\varepsilon/\rho)} \tag{6}$$

To solve for \hat{I} we need a specific form for $\mathcal{K}(I)$. In principle there can be multiple EEs.

For the standard KM SIR model,

$$\hat{I} = \varepsilon \left(1 - \frac{1}{\rho} \right), \qquad \mathcal{K}(I) \equiv 1. \tag{7} \quad \{\texttt{E:IhatKM}\}$$

If the recovery rate is density-dependent, it is most natural to assume $\mathcal{K}(I)$ is decreasing,

since this means that higher prevalence reduces recovery rates and hence lengthens infectious

periods. Here are some examples:

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$$\hat{I} = \begin{cases} \frac{\varepsilon}{\rho} \cdot \frac{a\rho - 1}{1 - b\varepsilon} & \mathcal{K}(I) = \frac{1}{a + bI}, & \rho > \frac{1}{a}, & \varepsilon < \frac{1}{b} \\ \frac{a}{2} - \frac{\varepsilon}{2\rho} - \sqrt{\left(\frac{a}{2} - \frac{\varepsilon}{2\rho}\right)^2 - \varepsilon\left(1 - \frac{a}{\rho}\right)}, & \mathcal{K}(I) = a - I, & a \ge 1 \\ -\frac{1}{b}W(-b\varepsilon e^{-b\varepsilon/\rho}) - \frac{\varepsilon}{\rho} & \mathcal{K}(I) = e^{-bI}, & b > 0 \end{cases}$$

$$(8) \quad \{\text{E:Ihat}\}$$

Here, W is Lambert's W function. For the above cases, **Figure 1** shows the dependence of the EE on ρ for $\varepsilon = 0.1$.

The (singular) recovery function $\mathcal{K}(I) = I^{-1/2}$ has been used a fair bit in the literature even though it blows up at I = 0, so I'm mentioning it. This has two EEs if $\rho > 4/\varepsilon$ and none if $\rho < 4/\varepsilon$:

$$\hat{I} = \frac{\varepsilon}{2\rho} \left(\rho \varepsilon - 2 \pm \sqrt{\rho \varepsilon (\rho \varepsilon - 4)} \right), \qquad \mathcal{K}(I) = I^{-1/2}, \qquad \rho > 4/\varepsilon \tag{9} \quad \{\text{E:Ihat-1/2}\}$$

This is quite peculiar and surely of mathematical interest only.

In case it is somehow helpful, here's an example with $\mathcal{K}(I)$ increasing:

$$\hat{I} = \varepsilon \left(\sqrt{\frac{1}{\varepsilon} + \frac{1}{4\rho^2}} - 1 \right), \qquad \mathcal{K}(I) = I$$
 (10) {E:Ihatinc}

$_{\scriptscriptstyle 4}$ 3 Infectious period distribution at EE

55 At the EE, prevalence is constant, so there is no change in the recovery rate over time.

6 Consequently, individuals recover according to

$$\frac{\mathrm{d}I}{\mathrm{d}\tau} = -\mathcal{K}(\hat{I})I, \qquad (11) \quad \{\mathbf{E}:\}$$

i.e., the infectious period distribution is exponential with mean $1/\mathcal{K}(\tilde{I})$ rather than mean 1.

4 Infectious period distribution near DFE (invasion)

In this limit, we can use the linearized system to obtain an approximation $I_{\rm lin}(au)$ initially.

Then, for as long as this approximation is valid, the recovery rate is $\mathcal{K}(I_{\text{lin}}(\tau))$, so we can

approximate the initial distribution of infectious by solving

$$\frac{\mathrm{d}I}{\mathrm{d}\tau} = -\mathcal{K}(I_{\mathrm{lin}}(\tau))I, \qquad (12) \quad \{\mathtt{E:invlim}\}$$

5 Infectious period distribution at any time

More generally, between the limits of invasion and endemic equilibrium, we could plug in the exact numerically computed $I(\tau)$ rather than $I_{\text{lin}}(\tau)$ into Equation (12) and compute the distribution of infectious periods at any time. However, this is a pain, not just because it is numerical but because we really need to integrate to $\tau = \infty$. Still, this is an alternative to agent-based simulations.

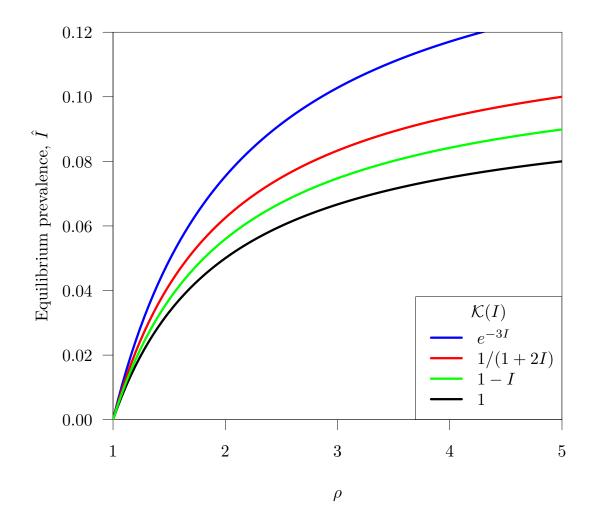


Figure 1: Equilibrium prevalence \hat{I} as a function of ρ for several recovery functions $\mathcal{K}(I)$, with $\varepsilon = 0.1$.

70 References