

Notes for HRP 204

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1 Session 3 (April 14)

The idea of this session is to add demography data in a (relatively) simple way to the original SIR model.

1.1 Definitions

Survival curves. The survival curve is a function which maps the current age to its survival rate. (*i.e.*, it takes in an age and outputs the percentage of the population which survives to be greater than, or equal to this age. Another way of stating this is as the probability that the age of death X is greater than some given α , $P(X \geq \alpha)$, where α is the desired age.)

Mortality rate. In the case that this curve follows an exponential distribution, we will define the exponential parameter as the *mortality rate*, which we will write as $\mu \in \mathbf{R}_+$. This would then imply that the area under the graph is the life expectancy, and would be equal to $1/\mu$. This also gives us a simple way of estimating the average mortality rate for any given distribution by setting $\mu \approx 1/\mathbb{E}[X]$.

This comes from the fact that
 $\mathbb{E}[X] = \int P(X \geq x) dx$.

1.2 Simple models

The obvious ‘world’s simplest model’ is an easy one: let S be the number of susceptible individuals. In this model, $S(t)$ is always constant for all time (*i.e.*, by definition $\dot{S} = 0$), and of course has no interesting dynamics.

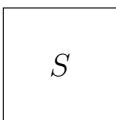


Figure 1: World's simplest model.

A slightly more complicated model (shown in figure 2) just includes the death rate as a ‘leaky state.’ In other words, the decay is given by $\dot{S}(t) = -\mu S(t)$, which has the obvious solution

$$S(t) = S_0 e^{-\mu t},$$

a decaying exponential.

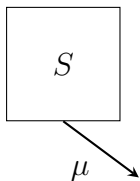


Figure 2: World's second simplest model.

Finally, an only slightly *more* complicated model comes from adding births (with rate $b \in \mathbf{R}_+$) to the previous one (shown in figure 3). The result is the differential equation $\dot{S}(t) = (b - \mu)S(t)$ with solution

$$S(t) = S_0 e^{(b-\mu)t}.$$

This is still an exponential (except when $b = \mu$, in which case the population is constant).

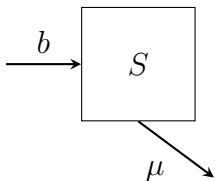


Figure 3: World's third simplest model.

1.2.1 Adding mortality in SIR

Using the previous simple idea, we can then easily add death rates to the SIR model.

Demography model. A simple, basic assumption when adding birth and death rates to the SIR model is to have the mortality rate be equal for any of the three states, and, to prevent issues with normalization we often assume that $\mu = b$. Note that this does not include any excess mortality rates from the disease itself (we will deal with this later).

This ensures that the population is always constant in our model, so it can be normalized.

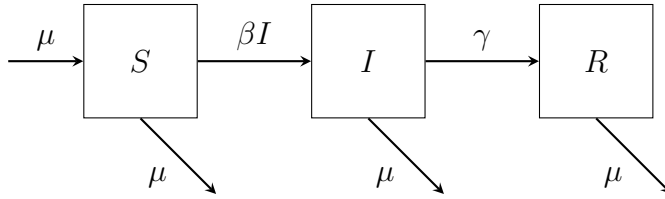


Figure 4: SIR incorporating deaths.

The new model equations (modeled in figure 4) then are

$$\begin{aligned}\dot{S} &= \mu - \beta IS - \mu S \\ \dot{I} &= \beta IS - \gamma I - \mu I \\ \dot{R} &= \gamma I - \mu R.\end{aligned}$$

Simple properties. We can give some simple thresholds and properties of the system by analyzing interesting cases. For example, if infections are ‘taking off,’ then we must have

$$\dot{I} \geq 0 \implies \beta IS - \gamma I - \mu I \geq 0.$$

Since $I \geq 0$ always, then we must have

$$\beta S - \gamma - \mu \geq 0 \implies S \geq \frac{\gamma + \mu}{\beta}, \quad (1)$$

i.e., that the susceptible proportion of the population must be greater than the rate of recovery *and* the rate of deaths and births divided by β . We should note that this is larger than the SIR model without demographics since $\mu > 0$.

Intuitively, this happens since we can view the total period of infectiousness as lasting roughly $1/(\gamma + \mu)$ rather than $1/\gamma$.

With demography, we can then derive R_0 as the inverse of the susceptible fraction

$$R_0 = \frac{\beta}{\gamma + \mu},$$

which is always smaller than that of the non-demographic model.

1.3 Equilibria

We say a condition is at an *equilibrium* if its relative proportions do not change over time, *i.e.*, if all time derivatives are zero. Additionally, we will say that a disease is *endemic* when it is at an equilibrium, but the infected proportion is strictly positive.

Equilibria of SIR with demography. To compute the equilibria, we simply find (S^*, I^*, R^*) such that

$$\dot{S}(t) = \dot{I}(t) = \dot{R}(t) = 0,$$

simultaneously at these points. A rather simple and not terribly exciting equilibrium is at $(S^*, I^*, R^*) = (1, 0, 0)$, *i.e.*, no infections have been introduced. We will call this the *disease-free equilibrium*.

A second important equilibrium is the *endemic equilibrium*, which happens when, in the same way as (1),

$$I^*(\beta S^* - \gamma - \mu) = 0,$$

but, since $I^* > 0$ by assumption, we must have

$$S^* = \frac{1}{R_0},$$

as defined above. This implies that the infected population must satisfy

$$0 = \mu - \beta I^* S^* - \mu S^* = \mu - \frac{\beta I^* - \mu}{R_0},$$

so

$$I^* = \frac{\mu(R_0 - 1)}{\beta}. \tag{2}$$

Stability of equilibria. While we will not prove this here, we can note that each of the two equilibria have certain ranges in which they are stable. If $R_0 < 1$, then the endemic equilibrium is infeasible (*i.e.*, there does not exist an equilibrium point which satisfies both $I^* > 0$ and $R_0 < 1$), which is easily seen from (2). The only remaining equilibrium is the disease-free equilibrium, in which the disease eradicates itself since its reproduction rate is smaller than one, and this equilibrium point is always stable.

If $R_0 > 1$, then the disease-free equilibrium is always unstable. Introducing any nonzero proportion of infected causes the system to move towards the endemic equilibrium. In particular, the system will have oscillatory behavior, and tend to an equilibrium point is exactly the endemic equilibrium.

Oscillatory behavior. We will also not prove this here, but we can derive the approximate period of the oscillations of the system, when converging to the endemic equilibrium. In particular, we can show that the period, T , of each oscillation is approximately:

$$T \approx 2\pi\sqrt{AG},$$

where A is the average age of infection and G is the average duration of infectiousness. We can given an approximate value of A for an SIR + demography model, by using the fact that, in the endemic equilibrium, the average period spent as susceptible is around $\frac{1}{\beta I^*}$ (*i.e.*, this is the average age at which a person is infected). Using (2), we get the result.

Interesting results. If vaccination is implemented, yet the reduction is not enough to reach the critical threshold of $R_0 \ll 1$, we get the result that the *average age of infections increases*, since

$$A \approx \frac{1}{\mu(R_0 - 1)},$$

and reducing R_0 therefore increases A .

1.4 SIR with demography and excess mortality

There are several approaches to this. The simplest case is to add a term that includes the excess mortality (*i.e.*, mortality rate) at the infectious stage.

In other words, if the system is at some equilibrium point, ‘pushing’ the system by changing S , I , or R in some way will cause the system to return back to its original equilibrium state.

There should be a simple, Lyapunov-style argument for this, but I haven’t yet found it. Would be interesting to explore.

I’m curious about this. Do we assume an ansatz and derive the corresponding envelope?

A simpler (and the more common approach) is to include a term ρ which is approximately the probability of dying at the infectious stage before either recovering or dying of natural causes—in this case, ρ is called the *case fatality rate* or CFR.

Model and consequences. Adding mortality also has the side effect that the number of birth rates is no longer equal to the number of deaths. This implies that the model cannot simply be normalized by the population size. The resulting model, with the new addition of birth rates $\nu \in \mathbf{R}_+$ and case fatality rates $\rho \in \mathbf{R}_+$ is

$$\begin{aligned}\dot{S} &= \nu - \beta IS - \mu S \\ \dot{I} &= \beta IS - (\gamma + \mu)I - \frac{\rho}{1 - \rho}(\gamma + \mu)I \\ \dot{R} &= \gamma I - \mu R.\end{aligned}$$

Under the (rather common) assumption that mortality occurs late in the infectious period, we can simplify the above dynamics slightly to

$$\begin{aligned}\dot{S} &= \nu - \beta IS - \mu S \\ \dot{I} &= \beta IS - (\gamma + \mu)I \\ \dot{R} &= (1 - \rho)\gamma I - \mu R.\end{aligned}$$

Chronic infections. There are several simple special cases of this model. For example, taking $\rho = 1$, yields the case of infections which persist until death. This includes long-term infections (*e.g.*, HPV, HCV) or short-term infections with high fatality rates (*e.g.*, Ebola, Mad Cow Disease). This model has the endemic equilibrium given by

$$S^* = \frac{\nu}{\beta - \gamma}, \quad I^* = \frac{\nu(\beta - \gamma - \mu)}{(\beta - \gamma)(\gamma + \mu)},$$

when $R_0 > \beta/(\gamma + \mu)$.

1.5 Other models

Models without immunity. A simple model is one where infectious individuals return to the susceptible pool rather than a separate

It might be good to derive this at some point. Doesn't seem difficult, but may be a bit tedious.

‘recovered’ pool. This model is useful for infections which confer no immunity. The resulting equations are

$$\dot{S} = \gamma I - \beta SI, \quad \dot{I} = \beta SI - \gamma I,$$

with the endemic equilibrium given by

$$S^* = \frac{\gamma}{\beta}, \quad I^* = 1 - \frac{\gamma}{\beta}.$$

Model with rapidly waning immunity. There is a second model, which generalizes the model without immunity, that follows from adding a term to the SIR + demography model which sends those recovered back into the susceptible pool (see figure 5). Note that, by sending

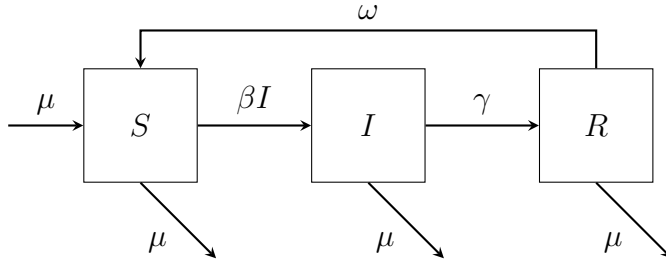


Figure 5: SIR incorporating deaths and waning immunity.

$\omega \rightarrow 0$ we get the SIR + demography model, while sending $\omega \rightarrow +\infty$, we recover the model without immunity. The differential equations for this model are

$$\begin{aligned} \dot{S} &= \mu - \beta IS - \mu S + \omega R \\ \dot{I} &= \beta IS - (\gamma + \mu)I \\ \dot{R} &= \gamma I - (\mu + \omega)R, \end{aligned}$$

as one might expect, the resulting oscillations are quite a bit more complex than those of the usual SIR + demography model.