## Study of Reproductive Number in SIR-SIS Model

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#### Abstract

The thresholds for mathematical epidemiology models specify the critical conditions for an epidemic to grow or die out. The reproductive number can provide significant insight into the transmission dynamics of a disease and can guide strategies to control its spread. We define the mean number of contacts, the mean duration of infection, and the mean transmission probability appropriately for certain epidemiological models, and construct a simplified formulation of the reproductive number as the product of these quantities. When the spread of the epidemic depends strongly upon the heterogeneity, and the expressions for the reproductive number become correspondingly more complex. In this article we formulate a model with different heterogeneous structures and demonstrate how to define the mean quantities for an explicit expression for the reproductive number. We derive an explicit formula for the reproductive number employing the spectral radius of the next generation operator.

**Keywords:** Reproductive number; Infection-free equilibrium; Spectral radius; M-matrix

## 1 Introduction

One of the fundamental questions of mathematical epidemiology is to find threshold conditions that determine whether an infectious disease will spread in a susceptible population when the disease is introduced into the population. The threshold conditions are characterized by the so called reproductive number, the reproduction number, the reproductive ratio, basic reproductive value, basic reproductive rate, or contact number, commonly denoted by  $R_0$  in mathematical epidemiology [1, 5, 14]. The concept of  $R_0$ , introduced by Ross in 1909 [20], is defined in epidemiological modeling such that if  $R_0 < 1$ , the modeled disease dies out, and if  $R_0 > 1$ , the disease spreads in the population. There have been intensive studies in the literature to calculate  $R_0$  for a wide class of epidemiological models of infectious disease [2, 3, 5, 6, 9, 10, 11, 12, 13]. In mathematical models, the reproductive number is determined by the spectral radius of the next generation operator in continuous models and, in particular, is determined by the dominant eigenvalue of the Jacobian matrix at the infection-free equilibrium for models in a finite-dimensional space [8]. It can also be obtained, in certain models, by suitable Lyapunov functions [11]. In this article, we use the SIS and SIR models as a basis [4, 7] and formulate a new heterogeneous models to demonstrate how different cases can be treated so that an appropriate reproductive number can be estimated.

## 2 Disease Transmission Model with *n* Patches

In Fig. (1) we consider an new model with different heterogeneous structures for HIV transmission in a population of individuals who are at high-risk for HIV.

The following system of ODEs describes this new model,

$$\frac{dS_{1}}{dt} = \mu S_{1}^{0} - \mu S_{1} - \delta_{1} S_{1} - \lambda_{1} S_{1} + \alpha_{1} I_{1}, 
\frac{dS_{i}}{dt} = \mu S_{i}^{0} - \mu S_{i} - \delta_{i} S_{i} - \lambda_{i} S_{i} + \delta_{i-1} S_{i-1} + \alpha_{i} I_{i} \quad ; \quad 2 \leq i \leq n-1, 
\frac{dS_{n}}{dt} = \mu S_{n}^{0} - \mu S_{n} - \lambda_{n} S_{n} + \delta_{n-1} S_{n-1} + \alpha_{n} I_{n}, 
\frac{dI_{1}}{dt} = \lambda_{1} S_{1} - (\mu + \alpha_{1} + \gamma_{1} + \epsilon_{1}) I_{1}, \qquad (1) 
\frac{dI_{i}}{dt} = \lambda_{i} S_{i} - (\mu + \alpha_{i} + \gamma_{i} + \epsilon_{i}) I_{i} + \epsilon_{i-1} I_{i-1} \quad ; \quad 2 \leq i \leq n-1, 
\frac{dI_{n}}{dt} = \lambda_{n} S_{n} - (\mu + \alpha_{n} + \gamma_{n}) I_{n} + \epsilon_{n-1} I_{n-1}, 
\frac{dR_{i}}{dt} = \gamma_{i} I_{i} - \mu R_{i} \quad ; \quad 1 \leq i \leq n,$$

where

$$\lambda_i = r\beta_i \frac{I_i}{I_i + S_i + R_i} \tag{2}$$

Figure (1) illustrates the system (1). This system is nonlinear due to the form of  $\lambda_i$ 's.

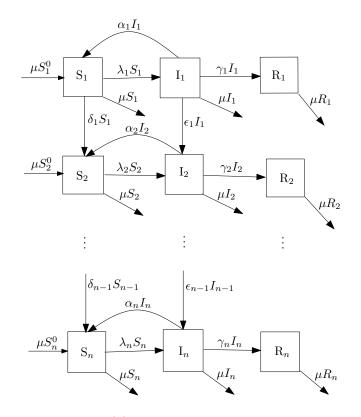


Figure 1: A schematic of system (1). Here  $S_i$ 's are the susceptibles,  $I_i$ 's are the infectives,  $R_i$ 's are the removeds,  $\mu > 0$ , a constant, is the death rate,  $\mu S_i^0 > 0$ , constants, are the migration terms,  $\gamma_i$ , constants, are the removal rate from  $I_i$  to  $R_i$ ,  $\alpha_i$ , constants, are the removal rate from  $I_i$  to  $S_i$ ,  $\delta_i$ , constants, are the removal rate from  $I_i$  to  $S_i$ ,  $\delta_i$ , constants, are the removal rate from  $I_i$  to  $I_{i+1}$ , and  $\lambda_i = r\beta_i \frac{I_i}{S_i + I_i + R_i}$  are the infection rate.

### 2.1 Reproductive Number

We derive an explicit formula for the reproductive number of infection by determining the spectral radius of the next generation operator of system (1) with (2) as follows.

System (1) has an infection-free equilibrium, given by,

$$E_0 = (S_1, S_2, S_3, ..., S_{n-1}, S_n, I_1 = 0, ..., I_n = 0).$$

Where  $S_1 = \frac{\mu S_1^0}{\mu + \delta_1}$ , and  $S_n = \frac{\mu S_n^0 + \delta_{n-1} S_{n-1}}{\mu}$ . Linearizing system (1) around the infection-free equilibrium, we have the Jacobian matrix,

$$J = \left[ \begin{array}{cc} A & B \\ 0 & C \end{array} \right]$$

where

$$A = \begin{bmatrix} -\mu - \delta_1 & 0 & 0 & \cdots & 0 & 0\\ \delta_1 & -\mu - \delta_2 & 0 & \cdots & 0 & 0\\ 0 & \delta_2 & -\mu - \delta_3 & \cdots & 0 & 0\\ \vdots & \vdots & \ddots & \ddots & \vdots & \vdots\\ 0 & 0 & 0 & \delta_{n-2} & -\mu - \delta_{n-1} & 0\\ 0 & 0 & 0 & \cdots & \delta_{n-1} & -\mu \end{bmatrix}$$

and

$$B = \begin{bmatrix} -r\beta_1 + \alpha_1 & 0 & 0 & \cdots & 0 \\ 0 & -r\beta_2 + \alpha_2 & 0 & \cdots & 0 \\ 0 & 0 & -r\beta_3 + \alpha_3 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & -r\beta_n + \alpha_n \end{bmatrix}$$

and

$$C = \begin{bmatrix} r\beta_1 - a_1 & 0 & 0 & \cdots & 0 & 0 \\ \epsilon_1 & r\beta_2 - a_2 & 0 & \cdots & 0 & 0 \\ 0 & \epsilon_2 & r\beta_3 - a_3 & \cdots & 0 & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \epsilon_{n-2} & r\beta_{n-1} - a_{n-1} & 0 \\ 0 & 0 & 0 & \cdots & \epsilon_{n-1} & r\beta_n - a_n \end{bmatrix}$$

with  $a_i = \mu + \alpha_i + \gamma_i + \epsilon_i$ ;  $1 \le i < n$  and  $a_n = \mu + \alpha_n + \gamma_n$ .

The stability of the Jacobian matrix at the infection-free equilibrium is completely determined by the stability of C. Define matrices F and V as,

$$F = [A_1]_{1 \times 1}$$
 and  $V = [A_2]_{1 \times 1}$ 

where,

$$A_{1} = \begin{bmatrix} r\beta_{1} & 0 & 0 & \cdots & 0 & 0 \\ 0 & r\beta_{2} & 0 & \cdots & 0 & 0 \\ 0 & 0 & r\beta_{3} & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & r\beta_{n-1} & 0 \\ 0 & 0 & 0 & \cdots & 0 & r\beta_{n} \end{bmatrix}$$

and

$$A_{2} = \begin{bmatrix} a_{1} & 0 & 0 & \cdots & 0 & 0 \\ -\epsilon_{1} & a_{2} & 0 & \cdots & 0 & 0 \\ 0 & -\epsilon_{2} & a_{3} & \cdots & 0 & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & -\epsilon_{n-2} & a_{n-1} & 0 \\ 0 & 0 & 0 & \cdots & -\epsilon_{n-1} & a_{n} \end{bmatrix}$$

Then F is a nonnegative matrix and V is a nonsingular M-matrix. Hence the reproductive number,  $R_0$ , is equal to the spectral radius of the next generation operator  $FV^{-1}$  [15],  $R_0 = \rho(FV^{-1})$ . To determine the spectral radius of  $FV^{-1}$ , we first represent the inverse of V by the following lower triangular matrix,  $V^{-1} = [A_3]_{1 \times 1}$ , where

$$A_{3} = \begin{bmatrix} \frac{1}{a_{1}} & 0 & 0 & \cdots & 0 & 0\\ m_{2,1} & \frac{1}{a_{2}} & 0 & \cdots & 0 & 0\\ m_{3,1} & m_{3,2} & \frac{1}{a_{3}} & \cdots & 0 & 0\\ \vdots & \vdots & \ddots & \ddots & \vdots & \vdots\\ m_{n-1,1} & m_{n-1,2} & m_{n-1,3} & \cdots & \frac{1}{a_{n-1}} & 0\\ m_{n,1} & m_{n,2} & m_{n,3} & \cdots & m_{n,n-1} & \frac{1}{a_{n}} \end{bmatrix}$$

with

$$m_{i,j} = -\frac{\prod_{k=j}^{i-1} \epsilon_k}{\prod_{k=j}^{i} a_k}.$$

Now we are ready to derive an explicit formula for the reproductive number  $R_0$ . Then we have,  $R_0 = \rho(A_4)$ , where,

$$A_4 = \begin{bmatrix} r\beta_1 \frac{1}{a_1} & 0 & 0 & \cdots & 0 & 0\\ 0 & r\beta_2 \frac{1}{a_2} & 0 & \cdots & 0 & 0\\ 0 & 0 & r\beta_3 \frac{1}{a_3} & \cdots & 0 & 0\\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots\\ 0 & 0 & 0 & 0 & r\beta_{n-1} \frac{1}{a_{n-1}} & 0\\ 0 & 0 & 0 & \cdots & 0 & r\beta_n \frac{1}{a_n} \end{bmatrix}$$

Therefore

$$R_0 = \max_{i=1,\dots,n} \frac{r\beta_i}{a_i} = \max_{i=1,\dots,n-1} \{ \frac{r\beta_i}{\mu + \alpha_i + \gamma_i + \epsilon_i}, \frac{r\beta_n}{\mu + \alpha_n + \gamma_n} \}.$$
 (3)

In summary, we have the following theorem:

#### 2.1.1 Theorem

Define the reproductive number  $R_0$  as

$$R_0 = \max_{i=1,\dots,n} \frac{r\beta_i}{a_i} = \max_{i=1,\dots,n-1} \{ \frac{r\beta_i}{\mu + \alpha_i + \gamma_i + \epsilon_i}, \frac{r\beta_n}{\mu + \alpha_n + \gamma_n} \}.$$
 (4)

If  $R_0 < 1$  the infection-free equilibrium is locally asymptotically stable, and if  $R_0 > 1$  the infection-free equilibrium is unstable.

For simple mathematical epidemiological models, the formula for  $R_0$  can be interpreted as the product of the number of contacts per unit of time, the infectivity of infection, and the duration of infection. For the more complex model (1), the explicit formula (4) for the reproductive number  $R_0$  can also be interpreted as the product of the mean number of contacts per unit of time, the total mean infectivity of infection, and total mean duration of infection. We define the mean duration of infection in each staged-progression-chain as,  $\bar{\tau}_i = \frac{1}{a_i}$ , i = 1, 2, ..., n. Then, the total mean duration of infection for the model (1) is,  $\bar{\tau} = \max_{i=1,...,n} \bar{\tau}_i$ . Define the mean infectivity for each staged-progression-chain as,  $\bar{\beta}_i = \frac{1}{\bar{\tau}_i} \frac{\beta_i}{a_i}$ . Then, the total mean infectivity for the model (2.1) is,  $\bar{\beta} = \frac{1}{\bar{\tau}_{i=1,...,n}} \bar{\beta}_i \bar{\tau}_i$ . Therefore, the reproductive number  $R_0$  can be written as,  $R_0 = r\bar{\beta}\bar{\tau}$ .

### 2.1.2 Corollary

Suppose we do not have travel from *i* patch to i + 1 patch or we do not have travel from  $I_i$  to  $I_{i+1}$ , then we have,  $R_0 = \max_{i=1,\dots,n} \frac{r\beta_i}{\mu + \alpha_i + \gamma_i}$ .

### 2.1.3 Corollary

Suppose we do not have travel from *i* patch to i + 1 patch and from  $I_i$  to  $S_i$ , then we have,  $R_0 = \max_{i=1,\dots,n} \frac{r\beta_i}{\mu + \gamma_i}$ 

### 2.2 Example

In this Example for explain formula (4) we use the following model parameters,  $n = 4, S_1^0 = 4000, S_2^0 = 3000, S_3^0 = 2000, S_4^0 = 1000, S_1(0) = 3500, S_2(0) = 3000, S_3(0) = 2500, S_4(0) = 2000, I_1(0) = 40, I_2(0) = 30, I_3(0) = 20, I_4(0) = 10, R_1(0) = R_2(0) = R_3(0) = R_4(0) = 0, \mu = 0.001, \alpha_1 = 0.04, \alpha_2 = 0.03, \alpha_3 = 0.02, \alpha_4 = 0.01, \delta_1 = 0.006, \delta_2 = 0.005, \delta_3 = 0.004, \gamma_1 = 0.03, \gamma_2 = 0.02, \gamma_3 = 0.01, \gamma_4 = 0.005, \epsilon_1 = 0.05, \epsilon_2 = 0.04, \epsilon_3 = 0.03, \text{ and } \beta_1 = 0.0012, \beta_2 = 0.0009, \beta_3 = 0.0006, \beta_4 = 0.0006, \text{ for } R_0 < 1 \text{ and } \beta_1 = 0.014, \beta_2 = 0.0012, \beta_3 = 0.0008, \beta_4 = 0.0007, \text{ for } R_0 > 1.$ 

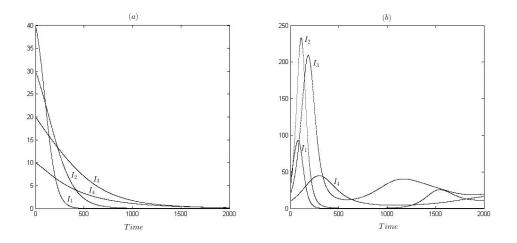


Figure 2: (a):  $R_0 = \max\{0.9917, 0.9890, 0.9836, 0.9836\} = 0.9890 < 1$  and (b):  $R_0 = \max\{1.1570, 1.3187, 1.3115, 1.1475\} = 1.5770 > 1$ .

### 2.3 Example

Suppose in example (2.2) we have,  $\epsilon_1 = \epsilon_2 = \epsilon_3 = 0$ . We take,  $\beta_1 = 0.0012, \beta_2 = 0.0009, \beta_3 = 0.0006, \beta_4 = 0.0006$ , for  $R_0 < 1$  and  $\beta_1 = 0.0014, \beta_2 = 0.0012, \beta_3 = 0.0008, \beta_4 = 0.0007$ , for  $R_0 > 1$ .

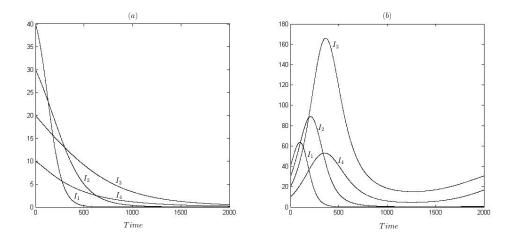


Figure 3: (a):  $R_0 = \max\{0.9859, 0.9804, 0.9677, 0.9836\} = 0.9859 < 1$  and (b):  $R_0 = \max\{1.1268, 1.1765, 1.2903, 1.1475\} = 1.2903 > 1$ .

# 3 Conclusions

One of the fundamental questions of mathematical epidemiology is to find threshold conditions that determine whether an infectious disease will speared in a susceptible population when the disease is introduced into this population, and the threshold conditions are usually characterized by the reproductive number  $R_0$ . The reproductive number plays an important role in understanding transmission dynamics of epidemics and predicting epidemics spread. In this article we used the spectral radius of the next generation operator of infection in finding a formula for the reproductive number.

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