

# Compartmental models on networks

Jayanth Kumar N

*Indian Institute of Science Education and Research, Pune*

June 29, 2017

## Abstract

In this article we try to figure out the relation between the total  $R_0$  of a country and the  $R_0$  of the individual cities how it varies with respect to migration between the cities. We model this using SIR and SEIR not graphs with vertices as cities and edges representing the migration. We try to figure out this relationship using numerically and analytically. To proceed numerically we use simulations to create data and calculate  $R_0$  using the relation derived here. Analytically we try to use next generation operator method to figure out the combined  $R_0$

## 1 Motivation

$H_1N_1$  commonly called Swine Flu is a human respiratory infection caused by an influenza strain, which nearly costs 284000 lives in 2009 alone globally. In *epidemiology* each disease can be modelled using compartmental models, one such commonly used models are SIR and SEIR models. Each disease can be given a characteristic value  $R_0$  known as basic reproductive number which represents the average number of infections an infected individual can cause over its infectious period in an uninfected population. A paper by Jesan, Menon and Sinha use the incidence data to calculate the  $R_0$  of swine flu in India assuming the SIR model. We see that

Table 1. Regional variation of basic reproduction number for 2009 influenza A(H1N1)v in India

Region/city	Period <sup>a</sup>	$\langle \lambda \rangle$	$R_0$
Pune	30/07–14/08	$0.25 \pm 0.04$	$1.74 \pm 0.14$
Mumbai	05/08–20/08	$0.22 \pm 0.06$	$1.65 \pm 0.18$
Delhi	13/08–28/08	$0.12 \pm 0.02$	$1.36 \pm 0.06$
Southern region <sup>b</sup>	15/08–30/08	$0.11 \pm 0.02$	$1.34 \pm 0.05$

$R_0$  is estimated by the method of exponential curve fitting from 5-day moving averages of incidence data for different regions/cities. In each case, bootstrap estimates yielded similar values.

<sup>a</sup>For each region/city, the time interval over which  $R_0$  is determined is chosen on the basis of exhibiting the highest rise in disease incidence.

<sup>b</sup>The Southern region comprises the cities of Bangalore, Chennai and Hyderabad.

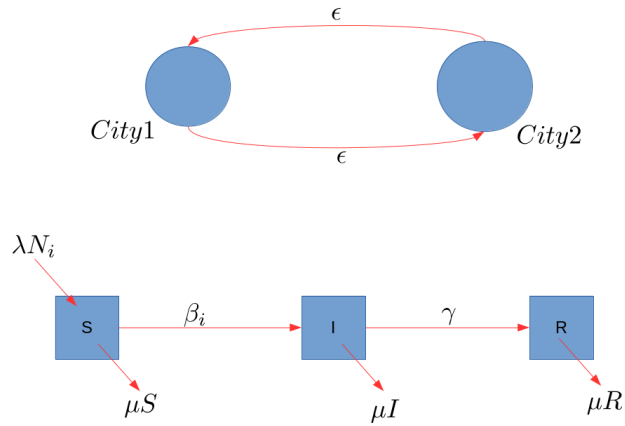
in the table from their paper that the  $R_0$  of India which was calculated to be 1.45 is not the average of the  $R_0$  of the individual cities. So here in this article we try to find out this relation between the  $R_0$  of the cities and the total  $R_0$ .

## 2 Methodology followed

In order to numerically figure out the relationship between total and individual city  $R_0$ , We try to model this scenario into simple case with just two vertices having 3 or 4 compartments each and 2 edges representing the

migration between the cities. We write down the differential equations for these systems, which are elaborately mentioned in the next section. We then apply Gillespie Algorithm on these equations to stochastically simulate the models. Using this data we try to figure out  $R_0$  via the relation between intrinsic/exponential growth rate of the infected individuals and  $R_0$ . This relation between intrinsic growth rate  $r$  and  $R_0$  depends upon the model used, we will derive the relation between them in each case in the following sections. We will then try to see how the total  $R_0$  of the system change from individual  $R_0$  of the cities as the transfer rates between the cities are increased. We will try to apply the above method on SIR and SEIR (with the restriction that the infected individuals are not allowed to migrate). Analytically we will try to use next generation operator method on the system of equation to derive the total  $R_0$  of the system in the last part of this article.

### 3 Modelling using SIR model



#### Equations

$$\frac{dS_1}{dt} = \lambda(S_1 + I_1 + R_1) - \mu S_1 - \beta_1 S_1 I_1 + \epsilon S_2 - \epsilon S_1$$

$$\frac{dI_1}{dt} = -\mu I_1 + \beta_1 S_1 I_1 + \epsilon I_2 - \epsilon I_1 - \gamma I_1$$

$$\frac{dR_1}{dt} = -\mu R_1 + \epsilon R_2 - \epsilon R_1 + \gamma I_1$$

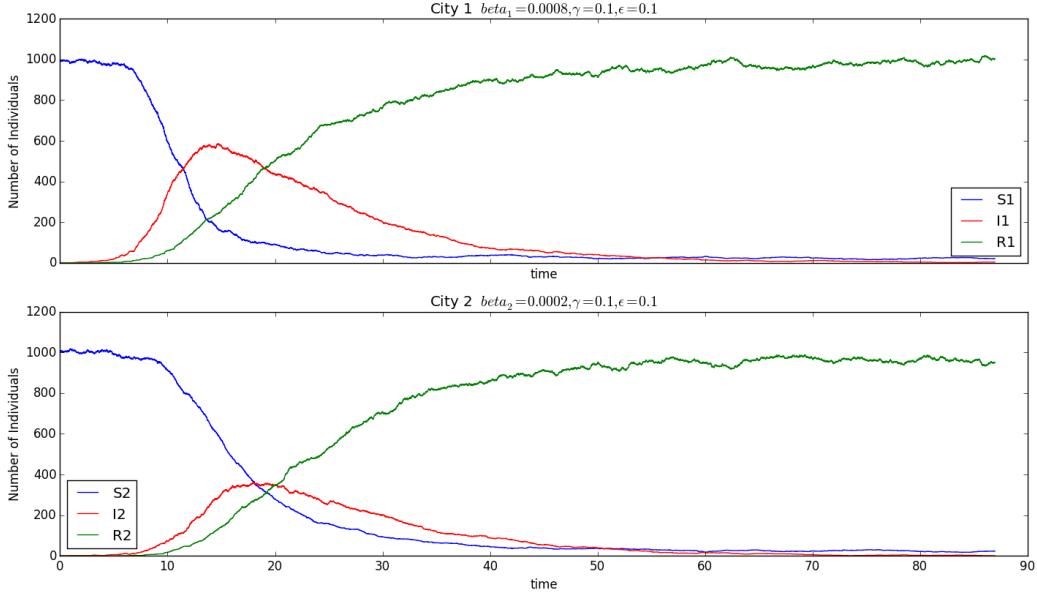
$$\frac{dS_2}{dt} = \lambda(S_2 + I_2 + R_2) - \mu S_2 - \beta_2 S_2 I_2 + \epsilon S_1 - \epsilon S_2$$

$$\frac{dI_2}{dt} = -\mu I_2 + \beta_2 S_2 I_2 + \epsilon I_1 - \epsilon I_2 - \gamma I_2$$

$$\frac{dR_2}{dt} = -\mu R_2 + \epsilon R_1 - \epsilon R_2 + \gamma I_2$$

where  $\lambda$  is the birth rate,  $\mu$  is the death rate,  $\gamma$  is the recovery rate and  $\beta_i$  is the contact rate in city  $i$ .

The above model was stochastically simulated using Gillespie algorithm. Below is the graph of one such simulation.



Using this simulation we can obtain the time series of the infected individuals in each city and for the system (by adding the individual time series). We now try to figure out the relation between the intrinsic growth rate  $r$  and  $R_0$ . At initial conditions when  $S_i = N_i$  and  $I = 0$ ,

$$\frac{dI}{dt} = rI$$

integrating,

$$I(t) = ce^{rt}$$

where  $c$  is a constant. Hence to calculate  $r$  we can fit a straight line to  $\log(I)$  vs  $t$  at initial part and the slope of this line should give  $r$ .

### 3.1 Relation between $r$ and $R_0$

We start by deriving **Lotka-Euler** equation and then applying epidemiology to it.

#### Assumptions

1. There are only females in the population and can reproduce independent of males.
  2. time and age are measured in years.  $t = 0$  denotes the present time
  3. The population displays exponential growth rate at a fixed growth rate
  4. Age distribution of the population does not change with time.
- Total No. of births at time  $t = \sum$  all children born to **all ages of mothers** at a particular time  $t$ .
  - No. of births to mother of age  $a$  at time  $t = (\text{Total number of births at time } t - a) \times (\text{Expected No. of offsprings per year for mother of age } a)$

**Note:**(Total number of births at time  $t - a$ ) denotes the total number of mother of age  $a$  at time  $t$ , ie these many were born.

Combining above two equations we get,

$$b(t) = \int_{a=0}^{\infty} b(t-a)n(a)da$$

where  $b(t)$  denotes the number of birth at time  $t$  and  $n(a)$  denotes the expected number of offsprings a single mother of age  $a$  will produce.

Since population is growing exponentially with a constant exponential rate  $r$ ,

No. of birth at time  $t = (\text{No. of births at time } t - a) \times (\text{Exponential growth from time } a)$

$$b(t) = b(t-a)e^{ra}$$

From above equations we get

$$b(t) = \int_{a=0}^{\infty} b(t-a)e^{-ra}n(a)da \quad (1)$$

Define  $R$  as  $R = \int_{a=0}^{\infty} n(a)da$  total number of offsprings a single women will produce in her lifetime.

Define  $g(a)$  as  $\frac{n(a)}{R}$  this will denote the probability that a female will produce an offspring at age  $a$ , which is the PDF of the female producing the offspring over her life time. Now we relate this to epidemiology by considering birth of offspring=production of new infection and age  $a =$  time since infection.

From definition of  $R_0$ , Averde No. of secondary infections produced by a single infected individual over its infectious period.

$$R_0 = \int_0^{\infty} n(a)da \quad (2)$$

Multiplying eq(1) by  $\frac{1}{b(t)R_0}$  we, get

$$\frac{1}{R} = \int_{a=0}^{\infty} e^{-ra}g(a)da$$

We know that Moment generating function of a probabiltly distribution  $g(a)$  is  $M(z) = \int_0^{\infty} e^{(za)}g(a)da$  putting  $z = -r$  we get the relation

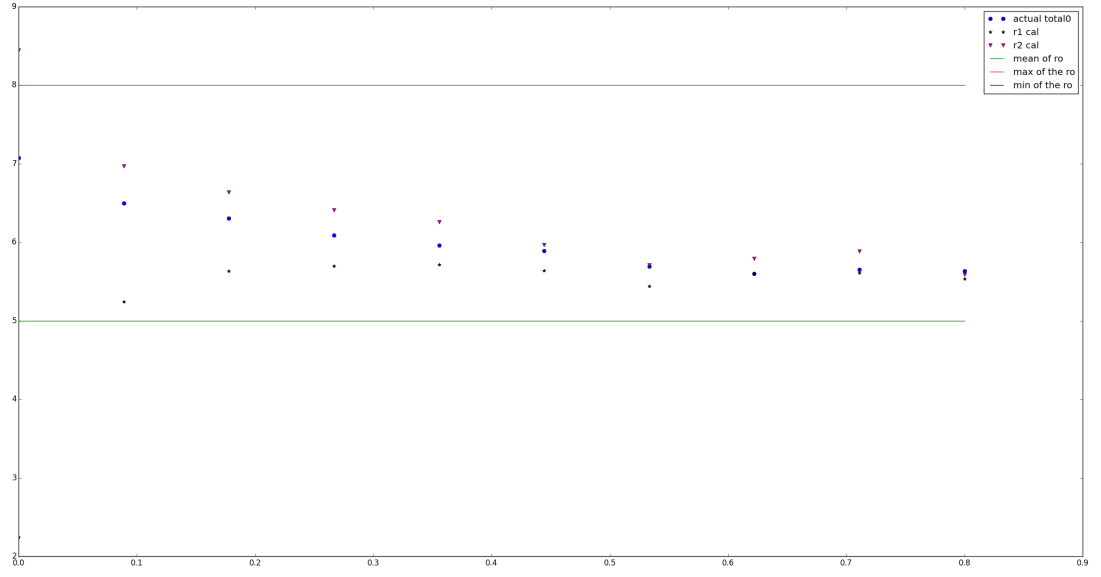
$$\frac{1}{R} = M(-r) \quad (3)$$

In our case  $g(a)$  is the generation interval distribution. Assuming it to be exponential with mean  $1/\text{gamma}$ . We find moment generating function of expoential distribution with mean  $\lambda$  to be  $M(t) = \frac{\lambda}{\lambda-t}$ . So

$$R_o = 1 + \frac{r}{\gamma} \quad (4)$$

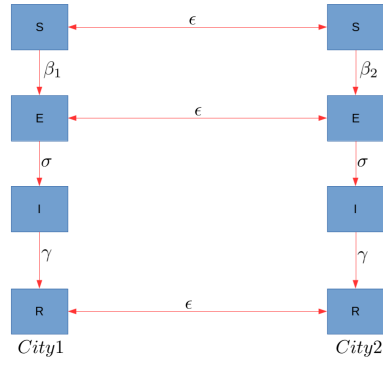
Hence using the above relation we can find the  $R_0$  of the system numerically.

## 3.2 Results



## 4 Modelling using SEIR model

### 4.1 Results



### Equations

$$\frac{dS_1}{dt} = \lambda(S_1 + E_1 + I_1 + R_1) - \mu S_1 - \beta_1 S_1 I_1 + \epsilon S_2 - \epsilon S_1$$

$$\frac{dE_1}{dt} = -\mu E_1 + \beta_1 S_1 I_1 + \epsilon E_2 - \epsilon E_1 - \sigma E_1$$

$$\frac{dI_1}{dt} = \sigma E_1 - \gamma I_1 - \mu I_1$$

$$\begin{aligned}
\frac{dR_1}{dt} &= -\mu R_1 + \epsilon R_2 - \epsilon R_1 + \gamma I_1 \\
\frac{dS_2}{dt} &= \lambda(S_2 + E_2 + I_2 + R_2) - \mu S_2 - \beta_2 S_2 I_2 + \epsilon S_1 - \epsilon S_2 \\
\frac{dE_2}{dt} &= -\mu E_2 + \beta_2 S_2 I_2 + \epsilon E_1 - \epsilon E_2 - \sigma E_2 \\
\frac{dI_2}{dt} &= \sigma E_2 - \gamma I_2 - \mu I_2 \\
\frac{dR_2}{dt} &= -\mu R_2 + \epsilon R_1 - \epsilon R_2 + \gamma I_2
\end{aligned}$$

Where  $\sigma$  denotes the rate of getting infectious from exposed, all other symbols denotes the same. As you can see we have restricted the migration of Infected individuals between the cities. So the infection can transfer between the cities only through migration of exposed individuals. Similar analysis as SIR model was used, but the relation between  $r$  and  $R_0$  has to be re-derived for this system.

## 4.2 Relation between $r$ and $R_0$

In this model the generation interval distribution  $g(a) = \text{Exp}(\frac{1}{\gamma}) + \text{Exp}(\frac{1}{\sigma})$ , where  $\text{Exp}(\lambda)$  denotes exponential distribution with mean  $\lambda$ .

Convolution of two exponential distribution with mean  $\frac{1}{\lambda}$  and  $\frac{1}{\theta}$  has the following pdf

$$f(x) = \frac{\lambda\theta}{\lambda - \theta}(e^{-\lambda x} - e^{-\theta x})$$

with MGF as

$$M(t) = \frac{\theta\lambda}{(\lambda - t)(\theta - t)}$$

. So in our case  $R_0$  turns out to be

$$R_0 = (1 + \frac{r}{\gamma})(1 + \frac{r}{\sigma}) \quad (5)$$

Using Eq(5) instead of Eq(4) and similar analysis as SIR model gave us the following result.

## 5 Results

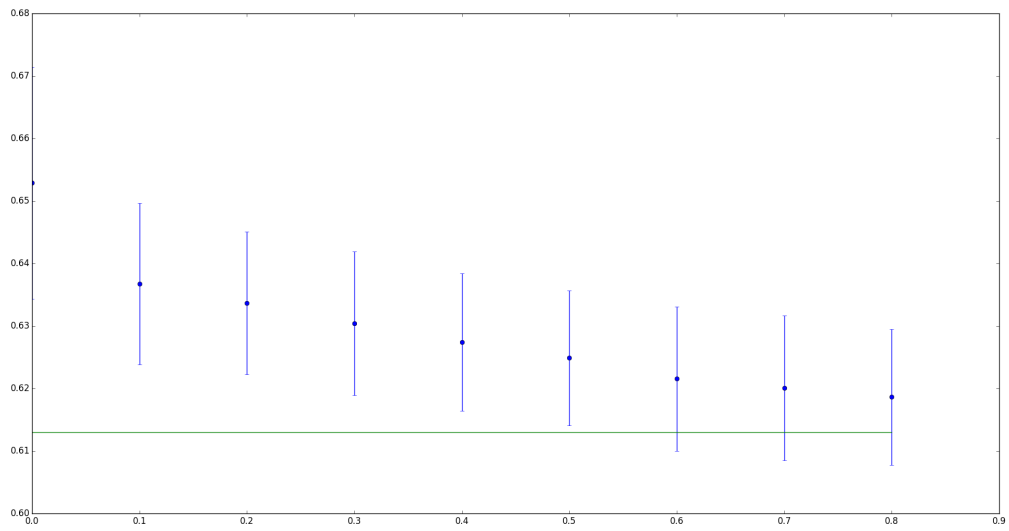


Figure 1: with infection starting in both the cities

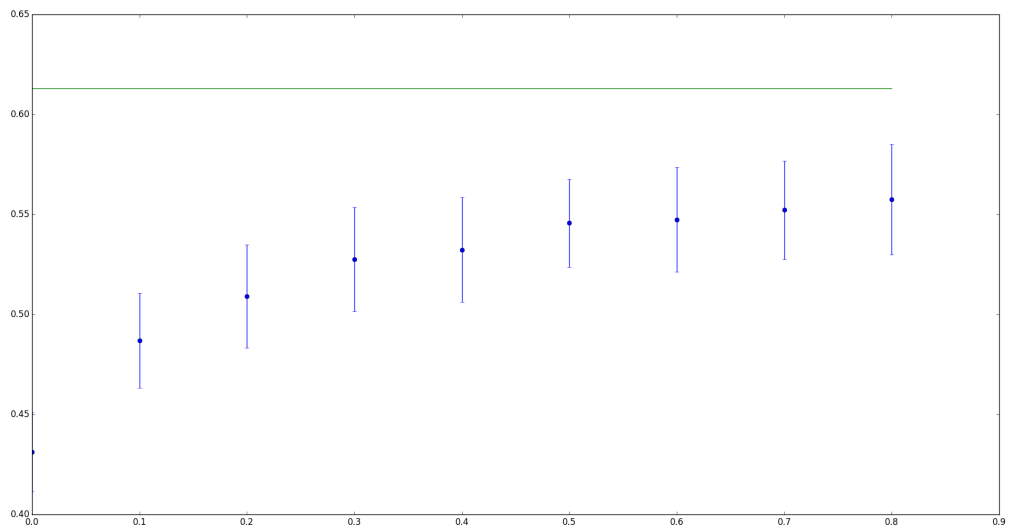


Figure 2: with infection starting at city1

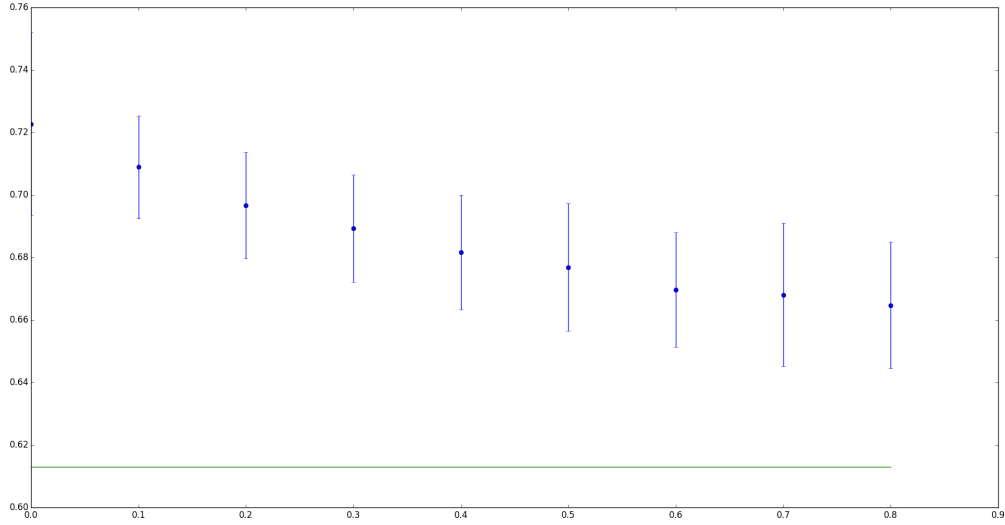


Figure 3: with infection staring at city2

## 6 Theoretical derivation

### 6.1 The Next generation operator method

$R_0$  is defined as the spectral radius of the 'next generation operator'. To find the next generation operator, we need to first identify the infected and non-infected compartments. Suppose there are  $n$  compartments of which  $m$  are infected. Let  $\bar{x} = (x_1, x_2, \dots, x_n)$  where each  $x_i$  denotes the number of individuals in  $i^{th}$  compartment. Let,  $F_i(x)$ , denote the rate of apperance of new infections in compartment  $i$ .  $V_i^-(x)$  be the rate of transfer of individuals into compartment  $i$  by all other means and  $V_i^+(x)$  be the rate of transfer of individuals out of compartment  $i$ .

$$\frac{dx_i}{dt} = F_i(x) - V_i(x) = f(x_i)$$

where  $V_i(x) = V_i^+(x) - V_i^-(x)$ .

We then construct  $\mathcal{F}$  and  $\mathcal{V}$  matrices by taking the partial derivatives of the  $F_i$  with respect to  $x_i$  and similarly for  $\mathcal{V}$  by taking partial derivatives of  $V_i$ . We define  $R_0$  to be the spectral radius of the  $\mathcal{F}\mathcal{V}^{-1}$ .

**Spectral radius of square matrix** It is the largest absolute value of the matrix's eigen value.

### 6.2 Assumptions

1. If  $\bar{x} \geq 0$ , then  $F_i, V_i^+, V_i^- \geq 0 \forall i$
2. If  $\bar{x} = 0$ , then  $V_i^- = 0$
3.  $F_i = 0$  if  $i \geq m$
4. If  $\bar{x} \in X_s$ , where  $X_s$  is set of all Disease Free states. Then  $F_i, V_i^+ = 0$
5. It is assumed that a disease free equilibrium exists and it is a locally asymptotically stable solution of the disease free model. Thus if  $x_0$  denotes a disease free equilibrium of the system, then if  $\mathcal{F}(x)$  is set to zero, then all eigenvalues of  $Df(x_0)$  have negative real parts.



### 6.3 SIR model with birth and death

The following is the derivation of  $R_0$  for SIR model with birth and death.

**Equations**

$$\begin{aligned}\frac{dS_1}{dt} &= \lambda(S_1 + I_1 + R_1) - \mu S_1 - \beta_1 S_1 I_1 + \epsilon S_2 - \epsilon S_1 \\ \frac{dI_1}{dt} &= -\mu I_1 + \beta_1 S_1 I_1 + \epsilon I_2 - \epsilon I_1 - \gamma I_1 \\ \frac{dR_1}{dt} &= -\mu R_1 + \epsilon R_2 - \epsilon R_1 + \gamma I_1\end{aligned}$$

Similarly for City 2

- **For single city in isolation with no transfer rates**

$\epsilon = 0$  ,  $n = 3$  and  $m = 1$

$F_i$  is the rate of apperance of infection at compartment  $i$ , It is enough to consider the  $I$  compartment, Since only  $I$  contributes for the infection.

$$\begin{aligned}\frac{dI}{dt} &= \beta SI - \gamma I - \mu I \\ F &= \beta SI, V = -\gamma I - \mu I\end{aligned}$$

$$\frac{dF}{dI} = \beta S, \frac{dV}{dI} = -\gamma - \mu$$

at Disease Free equilibrium  $S_1 = N_0$  where  $N_0$  is the initial number of people

$$\left. \frac{dF}{dI} \right|_{DFE} = \beta N_0, \left. \frac{dV}{dI} \right|_{DFE} = -\gamma - \mu$$

$$\mathcal{F} = \left. \frac{dF}{dI} \right|_{DFE}, \mathcal{V} = \left. \frac{dV}{dI} \right|_{DFE}$$

Since  $R_0$  equals the spectral radius of  $\mathcal{F}\mathcal{V}^{-1}$ , We find that

$$R_0 = \frac{\beta N}{(\gamma + \mu)}$$

- **For two cities with transfer rates**

$n = 6$  and  $m = 2$  where  $I_1$  and  $I_2$  are the two infected compartments.

$$\begin{aligned}F_1 &= \beta_1 S_1 I_1, F_2 = \beta_2 S_2 I_2 \\ V_1 &= -\gamma I_1 - \mu I_1 - \epsilon I_1 + \epsilon I_2, V_2 = -\gamma I_2 - \mu I_2 - \epsilon I_2 + \epsilon I_1\end{aligned}$$

Now,

$$\mathcal{F} = \begin{bmatrix} \beta_1 N_1 & 0 \\ 0 & \beta_2 N_2 \end{bmatrix}$$

$$\mathcal{V} = \begin{bmatrix} -\gamma - \mu - \epsilon & \epsilon \\ \epsilon & -\gamma - \mu - \epsilon \end{bmatrix}$$

The Eigen values of  $\mathcal{F}\mathcal{V}^{-1}$  was calculated using sage,

The spectral radius defined as the absolute value of the largest eigen value turns out to be

$$R_0 = \frac{X + \sqrt{X^2 - 4\beta_1\beta_2N_1N_2Y}}{2Y}$$

where  $X = (\beta_1N_1 + \beta_2N_2)(\epsilon + \mu + \gamma)$  and  $Y = (2\epsilon\gamma + \gamma^2 + 2\epsilon\mu + 2\mu\gamma + \mu^2)$

### 6.3.1 Plotting

## 6.4 SEIR Model

### Equations

$$\frac{dS_1}{dt} = \lambda(S_1 + E_1 + I_1 + R_1) - \mu S_1 - \beta_1 S_1 I_1 + \epsilon S_2 - \epsilon S_1$$

$$\frac{dE_1}{dt} = -\mu E_1 + \beta_1 S_1 I_1 + \epsilon E_2 - \epsilon E_1 - \sigma E_1$$

$$\frac{dI_1}{dt} = \sigma E_1 - \gamma I_1 - \mu I_1$$

$$\frac{dR_1}{dt} = -\mu R_1 + \epsilon R_2 - \epsilon R_1 + \gamma I_1$$

Similarly for City 2

- **For single city in isolation with no transfer rates**

$\epsilon = 0$  ,  $n = 4$  and  $m = 2$ ,  $E_1$  and  $I_1$  are considered as infective compartments

$$F_1 = \beta_1 S_1 I_1, F_2 = 0$$

$$V_1 = -\mu E_1 - \sigma E_1, V_2 = \sigma E_1 - \gamma I_1 - \mu I_1$$

Now, calculating  $\mathcal{F}$  and  $\mathcal{V}$  at DFE  $S_1 = N_1$

$$\mathcal{F} = \begin{bmatrix} 0 & \beta_1 N_1 \\ 0 & 0 \end{bmatrix}$$

$$\mathcal{V} = \begin{bmatrix} -\sigma - \mu & 0 \\ \sigma & -\gamma - \mu \end{bmatrix}$$

The Spectral radius of  $\mathcal{F}\mathcal{V}^{-1}$  was calculated using mathematica.

$$R_0 = \frac{\beta_1 N_1 \sigma}{(\gamma + \mu)(\mu + \sigma)}$$

- **For two cities with transfer rates**

$n = 8$  and  $m = 4$ ,  $E_1, I_1, E_2$  and  $I_2$  are considered as infective compartments.

$$F_1 = \beta_1 S_1 I_1, F_2 = 0$$

$$F_3 = \beta_2 S_2 I_2, F_4 = 0$$

$$V_1 = -\mu E_1 + \epsilon E_2 - \epsilon E_1 - \sigma E_1, V_2 = \sigma E_1 - \gamma I_1 - \mu I_1$$

$$V_3 = -\mu E_2 + \epsilon E_1 - \epsilon E_2 - \sigma E_2, V_4 = \sigma E_2 - \gamma I_2 - \mu I_2$$

Now, calculating  $\mathcal{F}$  and  $\mathcal{V}$  at DFE  $S_1 = N_1$  and  $S_2 = I_2$

$$\mathcal{F} = \begin{bmatrix} 0 & \beta_1 N_1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & \beta_2 N_2 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\mathcal{V} = \begin{bmatrix} -\sigma - \mu - \epsilon & 0 & \epsilon & 0 \\ \sigma & -\gamma - \mu & 0 & 0 \\ \epsilon & 0 & -\sigma - \epsilon - \mu & 0 \\ 0 & 0 & \sigma & -\mu - \gamma \end{bmatrix}$$

The Spectral radius of  $\mathcal{FV}^{-1}$  was calculated using sage.

$$R_0 = \frac{X + \sqrt{X^2 - 4\beta_1\beta_2N_1N_2\sigma^2Y}}{2(\gamma + \mu)Y}$$

where  $X = (\beta_1 N_1 \sigma + \beta_2 N_2 \sigma)(\epsilon + \mu + \sigma)$  and  $Y = (2\epsilon\mu + \mu^2 + 2\epsilon\sigma + 2\mu\sigma + \sigma^2)$

#### 6.4.1 Plotting