

Mathematical Modelling of a Classroom Disease Epidemic Simulation

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Introduction:

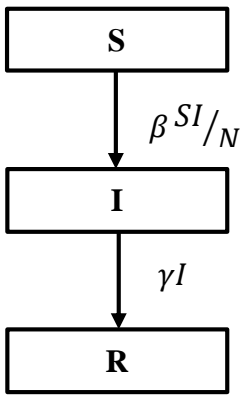
SIR Models (S = Susceptible, I = Infected, R = Recovered) can be utilized to describe the scenario where an acute infectious disease takes place. This project fits an SIR model (and its variations) to a classroom simulation of an infectious disease outbreak (The 'Handshake' Disease), which can be used in teaching infectious disease modelling to high-school and university level students.

SIR Model: Theoretical Background

Define: At time t:

- S: the number of susceptible people
- I: the number of infected people
- R: the number of recovered people

Schematic Diagram:



Where

- β = transmission rate
 - γ = recovery rate = $\frac{1}{\text{ave. infectious period}}$
 - N = total population
- are assumed to be constants.

SIR Equations:

Continuous Model:

$$\begin{cases} \frac{dS}{dt} = -\beta \frac{SI}{N} \\ \frac{dI}{dt} = \beta \frac{SI}{N} - \gamma I \\ \frac{dR}{dt} = \gamma I \end{cases}$$

Discrete Model:

$$\begin{cases} S_{n+1} = S_n - \frac{\beta S_n I_n}{N} \\ I_{n+1} = I_n + \frac{\beta S_n I_n}{N} - \gamma I_n \\ R_{n+1} = R_n + \gamma I_n \end{cases}$$

Acknowledgements:

The Vacation Scholarship Program offered me valuable experience and rewarding insights into what it is like when conducting mathematical research.

I would like to thank Dr. Anthony Morphett for his patient guidance, expert advice, and inspiring encouragement throughout this project.

REFERENCES:

- [1]. Keeling, M. J., & Rohani, P. (2011). *Modelling infectious diseases in humans and animals*. Princeton University Press.
- [2]. <http://go.unimelb.edu.au/267r>

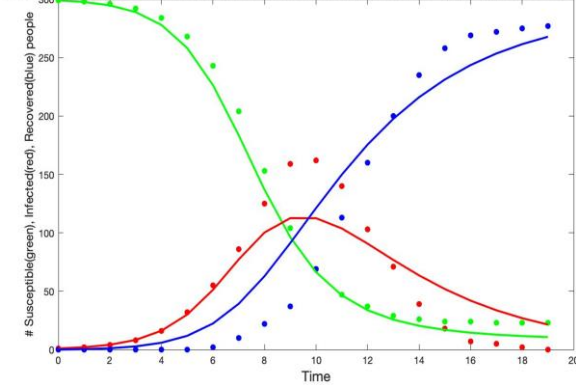
Classroom Disease Simulation: The 'Handshake' Disease

The 'handshake' disease is a simulated disease where the disease is transmitted by handshakes. For detailed information about the classroom activity, see reference [2].

Analysis: (Software Used: MATLAB_2019a)

PART 1: Fit the Standard SIR Model to One Typical Result from the 'Handshake' Game

SIR Model Stage1: Set Beta and Gamma as unknowns and solve for them to minimise the residual



Parameter Used:

- # handshakes = 10;
- Infectious period = 5;
- Total number of people = 300;
- Initial # infected people = 1.

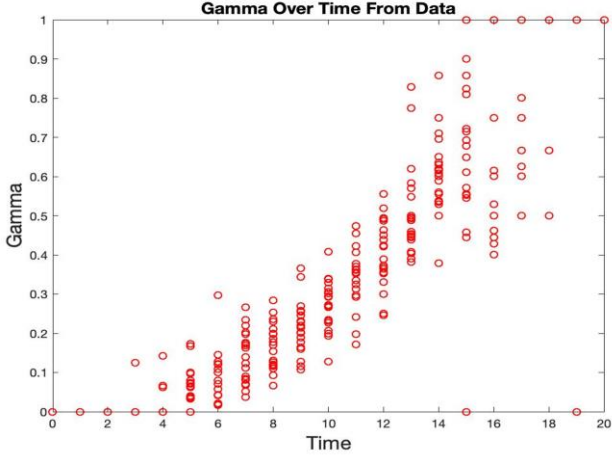
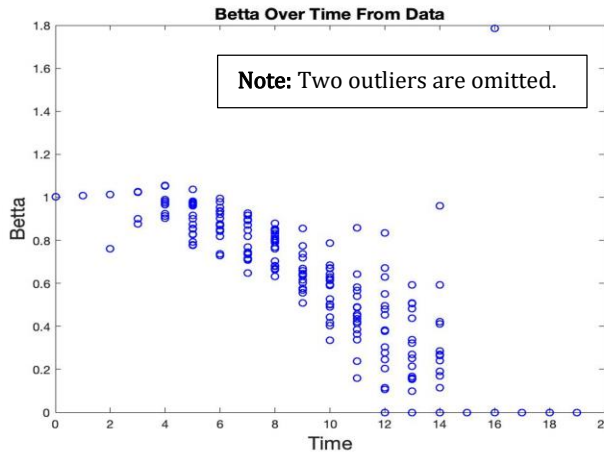
Result:

- $\beta = 0.9836$;
- $\gamma = 0.2639$;
- residual = 13432.

Conclusion: Setting β and γ as constants does not give a good fit quantitatively.

PART 2: The Behavior of β and γ Over Time Obtained from Discrete SIR Model

Based on the same parameter configuration as in Part 1 and 20 data sets, I got:

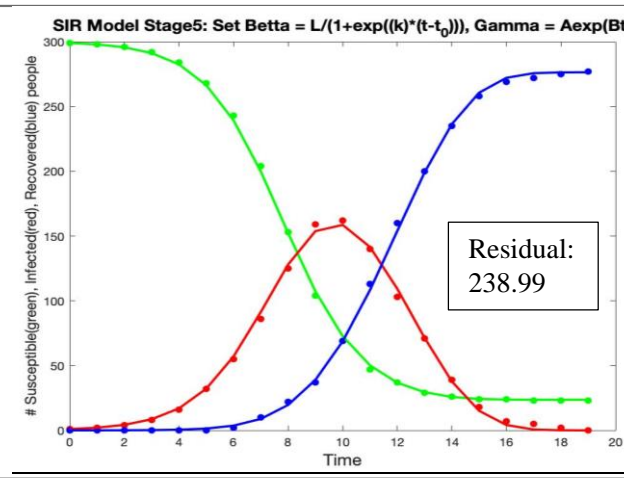


Conclusion: It can be seen that neither β nor γ keeps constant in the 'Handshake' disease. This is the reason why the standard SIR model could not give a good fit.

PART 3: Fit Some Variations of SIR Model to the 'Handshake' Disease Data

After trying a sequence of models of different forms with time-varying β and γ based on the information in Part 2, I figured out: $\beta(t) = \frac{L}{1 + e^{k(t-t_0)}}$ (logistic function) and $\gamma(t) = Ae^{Bt}$ provides the best fit. For example, for the same set of data in Part 1:

$$\begin{aligned} \beta(t) &= \frac{0.7375}{1 + e^{1.1816(t-21.1466)}} \\ \gamma(t) &= 0.0082e^{0.3294t} \end{aligned}$$



Discussion:

However, note that $\beta(t) = \frac{0.7375}{1 + e^{1.1816(t-21.1466)}}$ remains almost constant throughout the valid time span, $0 \leq t \leq 19$. Therefore, one can argue that setting $\beta = 0.7375$ also gives a relatively accurate result. In fact, I have verified this idea using MATLAB and a more general conclusion is: $\beta = \text{const.}$ and $\gamma(t) = Ae^{Bt}$ also gives a good fit. This model is simpler and much more efficient compared to the one setting β as a logistic function. Hence, for educational purposes, setting $\beta = \text{const.}$ and $\gamma(t) = Ae^{Bt}$ can be seen as an optimum choice.