



SEIRD-LQ

Population Modeling of Multiple Contagious Viruses

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Abstract

We build a population model that quantitatively describes the expected trend in the spread of a contagious virus on a macroscopic level. The model calculates the rate of contagion with three central components that contribute to the spread, first of which is population category which can be classified into four types - susceptible, infectious, and removed (immune or deceased). These population types, combined with the possibility of catching the virus twice, results in models that are often abbreviated to be known as the SEIRS model. The second component is the intrinsic nature and properties of the virus such as its deadliness and symptom-free growth duration. The third central component is mesoscopic factors such as the percentage of people following social distancing and other precautions, which decides the rate of transmission or fluid contagiousness. We wanted to do further than that. So we decided to take this model further, forming a model we call the *SEIRD-LQ* model, which will be explained in further sections.

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Introduction

1.1 | Motivation

With the ongoing pandemic and an interest in mathematical modeling of systems, the SIR model is what first came to our minds. On further research done as a part of our literature review, we found that there models that predict not only more trends of contagious viruses but also does it more accurately. For this reason we moved on to the SEIR model. As a part of our literature review of contagiousness of viruses, we found out that there are many cases that are not included in the modeling of viruses. For this reason, we first went on to formulate a modified version of SEIR model.

1.2 | Aims and Objectives

We will have mainly two aims as a part of our project, the first being to develop the aforementioned modified version of SEIR model, namely the *SEIRD-LQ* model. The abbreviation stands for the following-

- S → Susceptible
- E → Exposed
- I → Infected
- R → Recovered
- D → Dead
- L → Lockdown
- Q → Quarantine

The second aim for the project is to form an interactive simulation for the model on MATLAB which gives a better visual trend display to the user under different circumstances.

Background & Literature Overview

2.1 | Compartmental Population Models in Epidemiology

A population model is a type of mathematical model that is applied to the study of population dynamics. Models allow a better understanding of how complex interactions and processes work. Modeling of dynamic interactions in nature can provide a manageable way of understanding how numbers change over time or in relation to each other. Many patterns can be noticed by using population modeling as a tool. [Worster \(1994\)](#)

Ecological population modeling is concerned with the changes in parameters such as population size and age distribution within a population. This might be due to interactions with the environment, individuals of their own species, or other species. [Singh and Uyenoyama \(2004\)](#)

Population models are used to determine maximum harvest for agriculturists, to understand the dynamics of biological invasions, and for environmental conservation. Population models are also used to understand the spread of parasites, viruses, and disease ([Singh and Uyenoyama \(2004\)](#)). Another way populations models are useful are when species become endangered. Population models can track the fragile species and work and curb the decline.

2.1.1 | SIR Model

The SIR model ([Harko et al. \(2014\)](#) and [Beckley et al. \(2013\)](#)) is one of the simplest compartmental models, and many models are derivatives of this basic form. The model consists of three compartments:

- **S** → The number of susceptible individuals. When a susceptible and an infectious individual come into "infectious contact", the susceptible individual contracts the disease and transitions to the infectious compartment.
- **I** → The number of infectious individuals. These are individuals who have been infected and are capable of infecting susceptible individuals.

- **R** → for the number of removed (and immune) or deceased individuals. These are individuals who have been infected and have either recovered from the disease and entered the removed compartment, or died. It is assumed that the number of deaths is negligible with respect to the total population. This compartment may also be called "recovered" or "resistant".

This model is reasonably predictive ([Yang et al. \(2020\)](#)) for infectious diseases that are transmitted from human to human, and where recovery confers lasting resistance, such as measles, mumps and rubella.

Spatial SIR model simulation. Each cell can infect its eight immediate neighbors. These variables (S, I, and R) represent the number of people in each compartment at a particular time. To represent that the number of susceptible, infectious and removed individuals may vary over time (even if the total population size remains constant), we make the precise numbers a function of t (time): $S(t)$, $I(t)$ and $R(t)$. For a specific disease in a specific population, these functions may be worked out in order to predict possible outbreaks and bring them under control. ([Yang et al. \(2020\)](#))

As implied by the variable function of t, the model is dynamic in that the numbers in each compartment may fluctuate over time. The importance of this dynamic aspect is most obvious in an endemic disease with a short infectious period, such as measles in the UK prior to the introduction of a vaccine in 1968. Such diseases tend to occur in cycles of outbreaks due to the variation in number of susceptibles ($S(t)$) over time. During an epidemic, the number of susceptible individuals falls rapidly as more of them are infected and thus enter the infectious and removed compartments. The disease cannot break out again until the number of susceptibles has built back up, e.g. as a result of offspring being born into the susceptible compartment.

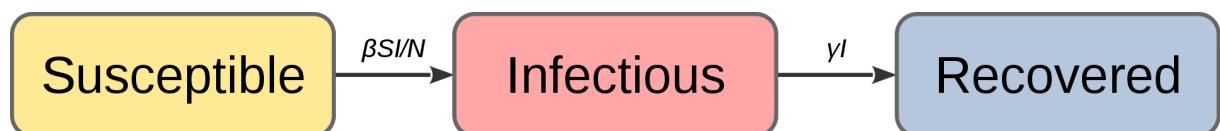


Figure 2.1: States in an SIR epidemic model and the rates at which individuals transition between them

2.1.1.1 | Transmission Rates

For the full specification of the model, the arrows should be labeled with the transition rates between compartments. Between S and I, the transition rate is assumed to be $\frac{d(S/N)}{dt} = -\beta \frac{SI}{N}$, where N is the total population, β is the average number of contacts per person per time, multiplied by the probability of disease transmission in a contact between a susceptible and an infec-

tious subject, and SI/N_2 is the fraction of those contacts between an infectious and susceptible individual which result in the susceptible person becoming infected. (This is mathematically similar to the law of mass action in chemistry in which random collisions between molecules result in a chemical reaction and the fractional rate is proportional to the concentration of the two reactants).

Between I and R, the transition rate is assumed to be proportional to the number of infectious individuals which is γI . This is equivalent to assuming that the probability of an infectious individual recovering in any time interval dt is simply γdt . If an individual is infectious for an average time period D , then $\gamma = \frac{1}{D}$. This is also equivalent to the assumption that the length of time spent by an individual in the infectious state is a random variable with an exponential distribution. The "classical" SIR model may be modified by using more complex and realistic distributions for the I-R transition rate (e.g. the Erlang distribution - [Krylova and Earn \(2013\)](#)).

For the special case in which there is no removal from the infectious compartment ($\gamma = 0$), the SIR model reduces to a very simple SI model, which has a logistic solution, in which every individual eventually becomes infected.

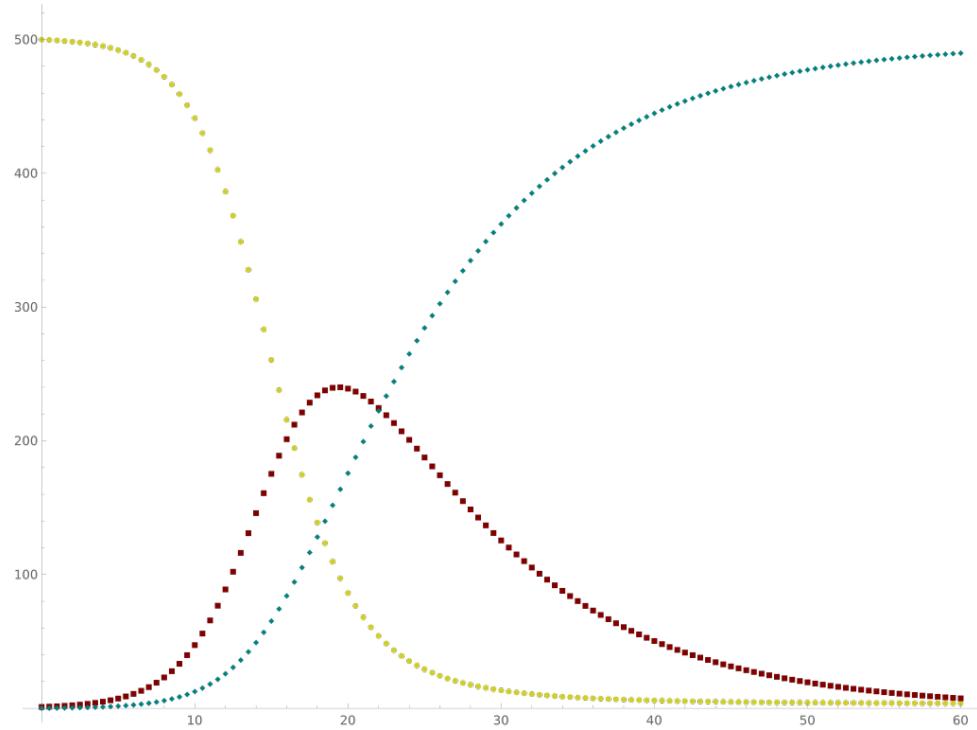


Figure 2.2: Yellow=Susceptible, Maroon=Infectious, Teal=Recovered

2.1.1.2 | Mathematical Modeling

The dynamics of an epidemic, for example, the flu, are often much faster than the dynamics of birth and death, therefore, birth and death are often omitted in simple compartmental models. The SIR system without so-called vital dynamics (birth and death, sometimes called demography) described above can be expressed by the following set of ordinary differential equations: ([Beckley et al. \(2013\)](#) and [Hethcote \(2000\)](#))

$$\begin{aligned}\frac{dS}{dt} &= -\frac{\beta IS}{N} \\ \frac{dI}{dt} &= \frac{\beta IS}{N} - \gamma I \\ \frac{dR}{dt} &= \gamma I\end{aligned}$$

where S is the stock of susceptible population, I is the stock of infected, R is the stock of removed population (either by death or recovery), and N is the sum of these three. This model was for the first time proposed by William Ogilvy Kermack and Anderson Gray McKendrick as a special case of what we now call Kermack-McKendrick theory, and followed work McKendrick had done with Ronald Ross. This system is non-linear, however it is possible to derive its analytic solution in implicit form. ([Harko et al. \(2014\)](#)) Firstly note that from:

$$\frac{dS}{dt} + \frac{dI}{dt} + \frac{dR}{dt} = 0$$

it follows that:

$$S(t) + I(t) + R(t) = \text{constant} = N$$

expressing in mathematical terms the constancy of population N . Note that the above relationship implies that one need only study the equation for two of the three variables. Secondly, we note that the dynamics of the infectious class depends on the following ratio:

$$R_0 = \frac{\beta}{\gamma}$$

the so-called basic reproduction number (also called basic reproduction ratio). This ratio is derived as the expected number of new infections (these new infections are sometimes called secondary infections) from a single infection in a population where all subjects are susceptible. ([Bailey et al. \(1975\)](#)) This idea can probably be more readily seen if we say that the typical time between contacts is $T_c = \beta^{-1}$, and the typical time until removal is $T_r = \gamma^{-1}$. From here it follows that, on average, the number of contacts by an infectious individual with others before the infectious has been removed is: T_r/T_c . By dividing the first differential equation by the third, separating the variables and integrating we get

$$S(t) = S(0)e^{-R_0(R(t)-R(0))/N}$$

where $S(0)$ and $R(0)$ are the initial numbers of, respectively, susceptible and removed subjects. Writing $s_0 = S(0)/N$ for the initial proportion of susceptible individuals, and $s_\infty = S(\infty)/N$ and $r_\infty = R(\infty)/N$ for the proportion of susceptible and removed individuals respectively in the limit $t \rightarrow \infty$, one has

$$s_\infty = 1 - r_\infty = s_0 e^{-R_0(r_\infty - r_0)}$$

(note that the infectious compartment empties in this limit). This transcendental equation has a solution in terms of the Lambert W function,^[9] namely

$$s_\infty = 1 - r_\infty = -R_0^{-1} W\left(-s_0 R_0 e^{-R_0(1-r_0)}\right)$$

This shows that at the end of an epidemic that conforms to the simple assumptions of the SIR model, unless $s_0 = 0$, not all individuals of the population have been removed, so some must remain susceptible. A driving force leading to the end of an epidemic is a decline in the number of infectious individuals. The epidemic does not typically end because of a complete lack of susceptible individuals. The role of both the basic reproduction number and the initial susceptibility are extremely important. In fact, upon rewriting the equation for infectious individuals as follows:

$$\frac{dI}{dt} = \left(R_0 \frac{S}{N} - 1\right) \gamma I$$

it yields that if:

$$R_0 \cdot S(0) > N$$

then:

$$\frac{dI}{dt}(0) > 0$$

i.e., there will be a proper epidemic outbreak with an increase of the number of the infectious (which can reach a considerable fraction of the population). On the contrary, if

$$R_0 \cdot S(0) < N$$

then

$$\frac{dI}{dt}(0) < 0$$

i.e., independently from the initial size of the susceptible population the disease can never cause a proper epidemic outbreak. As a consequence, it is clear that both the basic reproduction number and the initial susceptibility are extremely important.

2.1.1.3 | The force of infection

Note that in the above model the function:

$$F = \beta I$$

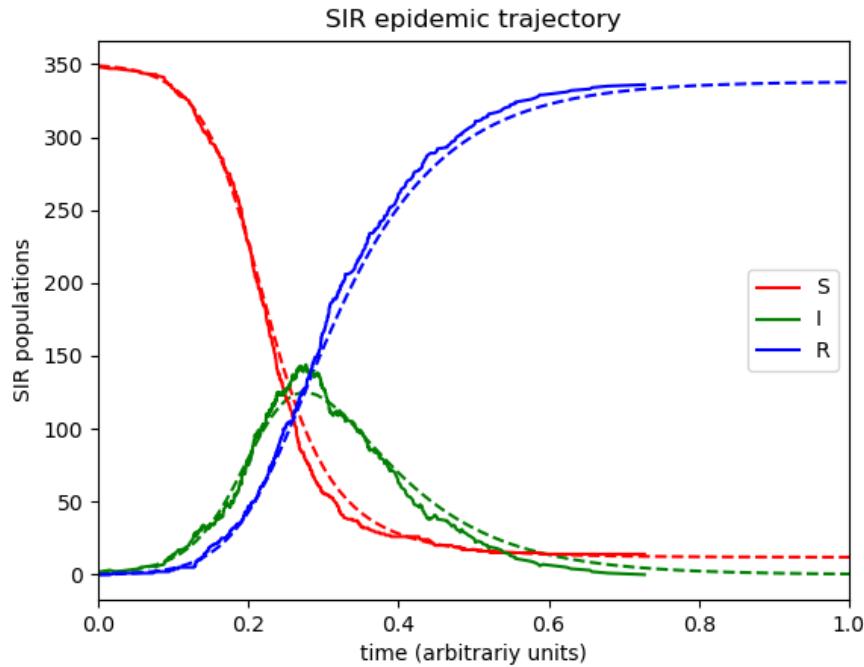


Figure 2.3: A single realization of the SIR epidemic as produced with an implementation of the Gillespie algorithm and the numerical solution of the ordinary differential equation system (dashed).

models the transition rate from the compartment of susceptible individuals to the compartment of infectious individuals, so that it is called the force of infection. However, for large classes of communicable diseases it is more realistic to consider a force of infection that does not depend on the absolute number of infectious subjects, but on their fraction (with respect to the total constant population N):

$$F = \beta \frac{I}{N}$$

Capasso [10] and, afterwards, other authors have proposed nonlinear forces of infection to model more realistically the contagion process.

2.1.1.4 | Exact analytical solutions to the SIR model

In 2014, Harko and coauthors derived an exact analytical solution to the SIR model. ([Harko et al. \(2014\)](#)) In the case without vital dynamics setup, for $\mathcal{S}(u) = S(t)$, etc, it corresponds to the following time parametrization

$$\begin{aligned}\mathcal{S}(u) &= S(0)u \\ \mathcal{I}(u) &= N - \mathcal{R}(u) - \mathcal{S}(u) \\ \mathcal{R}(u) &= R(0) - \rho \ln(u)\end{aligned}$$

for

$$t = \frac{N}{\beta} \int_u^1 \frac{du^*}{u^* \mathcal{I}(u^*)}, \quad \rho = \frac{\gamma N}{\beta}$$

with initial conditions $(S(1), \mathcal{I}(1), \mathcal{R}(1)) = (S(0), N - R(0) - S(0), R(0))$, $u_T < u < 1$ where u_T satisfies $\mathcal{I}(u_T) = 0$. By the transcendental equation for R_∞ above, it follows that $u_T = e^{-(R_\infty - R(0))/\rho}$ ($= S_\infty/S(0)$, if $S(0) \neq 0$) and $I_\infty = 0$. An equivalent analytical solution found by Miller [111112] yields

$$\begin{aligned} S(t) &= S(0)e^{-\xi(t)} \\ I(t) &= N - S(t) - R(t) \\ R(t) &= R(0) + \rho\xi(t) \\ \xi(t) &= \frac{\beta}{N} \int_0^t I(t^*) dt^* \end{aligned}$$

Here $\xi(t)$ can be interpreted as the expected number of transmissions an individual has received by time t . The two solutions are related by $e^{-\xi(t)} = u$. Effectively the same result can be found in the original work by Kermack and McKendrick. ([Kermack and McKendrick \(1927\)](#)) These solutions may be easily understood by noting that all of the terms on the right-hand sides of the original differential equations are proportional to I . The equations may thus be divided through by I , and the time rescaled so that the differential operator on the left-hand side becomes simply $d/d\tau$, where $d\tau = Idt$, i.e. $\tau = \int Idt$. The differential equations are now all linear, and the third equation, of the form $dR/d\tau = \text{const}$, shows that τ and R (and ξ above) are simply linearly related.

2.1.2 | Dynamic SIR Model

Consider a population characterized by a death rate μ and birth rate Λ , and where a communicable disease is spreading. ^[3] The model with mass-action transmission is: $\frac{dS}{dt} = \Lambda - \mu S - \frac{\beta IS}{N}$ $\frac{dI}{dt} = \frac{\beta IS}{N} - \gamma I - \mu I$ $\frac{dR}{dt} = \gamma I - \mu R$ for which the disease-free equilibrium (DFE) is:

$$(S(t), I(t), R(t)) = \left(\frac{\Lambda}{\mu}, 0, 0 \right)$$

In this case, we can derive a basic reproduction number:

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

which has threshold properties. In fact, independently from biologically meaningful initial values, one can show that.

$$\begin{aligned} R_0 \leq 1 \Rightarrow \lim_{t \rightarrow \infty} (S(t), I(t), R(t)) &= \text{DFE} = \left(\frac{\Lambda}{\mu}, 0, 0 \right) \\ R_0 > 1, I(0) > 0 \Rightarrow \lim_{t \rightarrow \infty} (S(t), I(t), R(t)) &= \text{EE} = \left(\frac{\gamma + \mu}{\beta}, \frac{\mu}{\beta} (R_0 - 1), \frac{\gamma}{\beta} (R_0 - 1) \right) \end{aligned}$$

The point EE is called the Endemic Equilibrium (the disease is not totally eradicated and remains in the population). With heuristic arguments, one may show that R_0 may be read as the

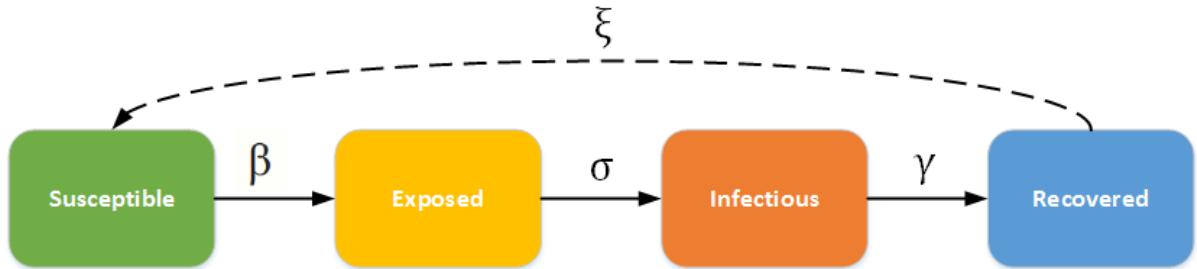


Figure 2.4: Compartmental view of SEIR model

average number of infections caused by a single infectious subject in a wholly susceptible population, the above relationship biologically means that if this number is less than or equal to one the disease goes extinct, whereas if this number is greater than one the disease will remain permanently endemic in the population.

2.1.2.1 | Steady State Solution

The expected duration of susceptibility will be $E[\min(T_L \mid T_S)]$ where T_L reflects the time alive (life expectancy) and T_S reflects the time in the susceptible state before becoming infected, which can be simplified ([Anderson et al. \(1992\)](#)) to:

$$E[\min(T_L \mid T_S)] = \int_0^\infty e^{-(\mu+\delta)x} dx = \frac{1}{\mu + \delta}$$

such that the number of susceptible persons is the number entering the susceptible compartment μN times the duration of susceptibility:

$$S = \frac{\mu N}{\mu + \lambda}$$

Analogously, the steady-state number of infected persons is the number entering the infected state from the susceptible state (number susceptible, times rate of infection $\lambda = \frac{\beta I}{N}$, times the duration of infectiousness $\frac{1}{\mu+v}$)

$$I = \frac{\mu N}{\mu + \lambda} \lambda \frac{1}{\mu + v}$$

2.1.3 | SEIR Model

Assuming that the incubation period is a random variable with exponential distribution with parameter a (i.e. the average incubation period is a^{-1}), and also assuming the presence of vital dynamics with birth rate Λ equal to death rate μ , we have the model: $\frac{dS}{dt} = \Lambda N - \mu S - \frac{\beta I S}{N}$ $\frac{dE}{dt} = \frac{\beta I S}{N} - (\mu + a)E$ $\frac{dI}{dt} = aE - (\gamma + \mu)I$ $\frac{dR}{dt} = \gamma I - \mu R$ We have $S + E + I + R = N$, but this

is only constant because of the simplifying assumption that birth and death rates are equal; in general N is a variable. For this model, the basic reproduction number is:

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

Similarly to the SIR model, also, in this case, we have a Disease-Free-Equilibrium ($(N, 0, 0, 0)$) and an Endemic Equilibrium EE, and one can show that, independently from biologically meaningful initial conditions:

$$(S(0), E(0), I(0), R(0)) \{ (S, E, I, R) \in [0, N]^4 : S \geq 0, E \geq 0, I \geq 0, R \geq 0, S + E + I + R = N \}$$

it holds that: $R_0 \leq 1 \Rightarrow \lim_{t \rightarrow +\infty} (S(t), E(t), I(t), R(t)) = DFE = (N, 0, 0, 0)$ $R_0 > 1, I(0) > 0 \Rightarrow \lim_{t \rightarrow +\infty} (S(t), E(t), I(t), R(t)) = EE$ In case of periodically varying contact rate $\beta(t)$ the condition for the global attractiveness of DFE is that the following linear system with periodic coefficients: $\frac{dE_1}{dt} = \beta(t)I_1 - (\gamma + a)E_1$

$$\frac{dI_1}{dt} = aE_1 - (\gamma + \mu)I_1$$

is stable.

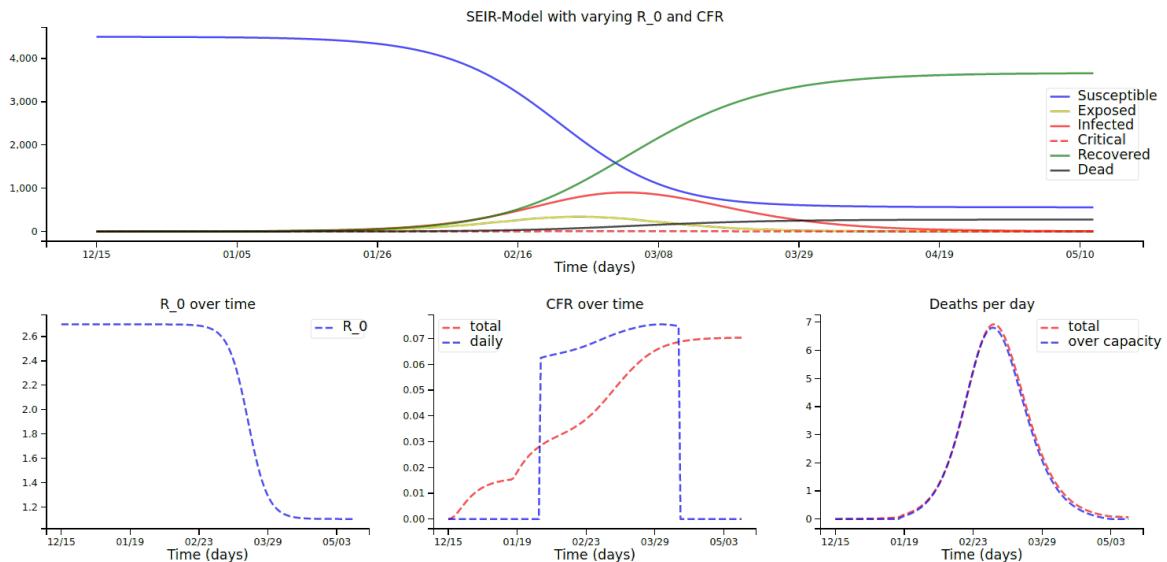


Figure 2.5: SEIR Graph

2.2 | MATLAB Overview

MATLAB is a proprietary multi-paradigm programming language and numerical computing environment developed by MathWorks. MATLAB allows matrix manipulations, plotting of

functions and data, implementation of algorithms, creation of user interfaces, and interfacing with programs written in other languages.

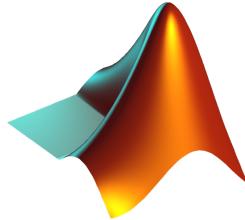


Figure 2.6: MATLAB Logo

2.2.1 | Functions

Both scripts and functions allow you to reuse sequences of commands by storing them in program files. Scripts are the simplest type of program, since they store commands exactly as you would type them at the command line. Functions provide more flexibility, primarily because you can pass input values and return output values. For example, this function named fact computes the factorial of a number (n) and returns the result (f).

```
1 function f = fact(n)
2     f = prod(1:n);
3 end
```

This type of function must be defined within a file, not at the command line. Often, you store a function in its own file. In that case, the best practice is to use the same name for the function and the file (in this example, fact.m), since MATLAB® associates the program with the file name. Save the file either in the current folder or in a folder on the MATLAB search path.

You can call the function from the command line, using the same syntax rules that apply to functions installed with MATLAB. For instance, calculate the factorial of 5.

```
1 x = 5;
2 y = fact(5)
```

```
y =
120
```

Starting in R2016b, another option for storing functions is to include them at the end of a script file. For instance, create a file named mystats.m with a few commands and two functions, fact

and perm. The script calculates the permutation of (3,2).

```

1 x = 3;
2 y = 2;
3 z = perm(x,y)
4
5 function p = perm(n,r)
6     p = fact(n)*fact(n-r);
7 end
8
9 function f = fact(n)
10    f = prod(1:n);
11 end

```

Call the script from the command line.

```
1 mystats
```

```
z =
6
```

2.2.1.1 | Function Definition Syntax

The first line of every function is the definition statement, which includes the following elements.

- **function keyword** (required)

Use lowercase characters for the keyword.

- **Function name** (required)

Valid function names follow the same rules as variable names. They must start with a letter, and can contain letters, digits, or underscores.

- **Output arguments** (optional)

If your function returns one output, you can specify the output name after the function keyword.

```
1 function myOutput = myFunction(x)
```

If function returns more than one output, enclose the output names in square brackets.

```
1 function [one,two,three] = myFunction(x)
```

If there is no output, you can omit it.

```
1 function myFunction(x)
```

Or you can use empty square brackets.

```
1 function [] = myFunction(x)
```

■ Input arguments (optional)

If your function accepts any inputs, enclose their names in parentheses after the function name. Separate inputs with commas.

```
1 function y = myFunction(one,two,three)
```

If there are no inputs, you can omit the parentheses.

2.2.2 | Plotting

There are many types of plots in MATLAB, as shown in figure 2.7

2.2.2.1 | 2D Plotting

`plot(X,Y)` creates a 2-D line plot of the data in Y versus the corresponding values in X.

- If X and Y are both vectors, then they must have equal length. The `plot` function plots Y versus X.
- If X and Y are both matrices, then they must have equal size. The `plot` function plots columns of Y versus columns of X.
- If one of X or Y is a vector and the other is a matrix, then the matrix must have dimensions such that one of its dimensions equals the vector length. If the number of matrix rows equals the vector length, then the `plot` function plots each matrix column versus the vector. If the number of matrix columns equals the vector length, then the function plots each matrix row versus the vector. If the matrix is square, then the function plots each column versus the vector.
- If one of X or Y is a scalar and the other is either a scalar or a vector, then the `plot` function plots discrete points. However, to see the points you must specify a marker symbol, for example, `plot(X,Y,'o')`.

```
1 % EXAMPLE 1: BASIC 2D PLOTTING
2
3 x = 0:pi/100:2*pi;
4 y1 = sin(x);
5 y2 = sin(x-0.25);
6 y3 = sin(x-0.5);
7
8 figure
9 plot(x,y1,x,y2,'--',x,y3,':')
```

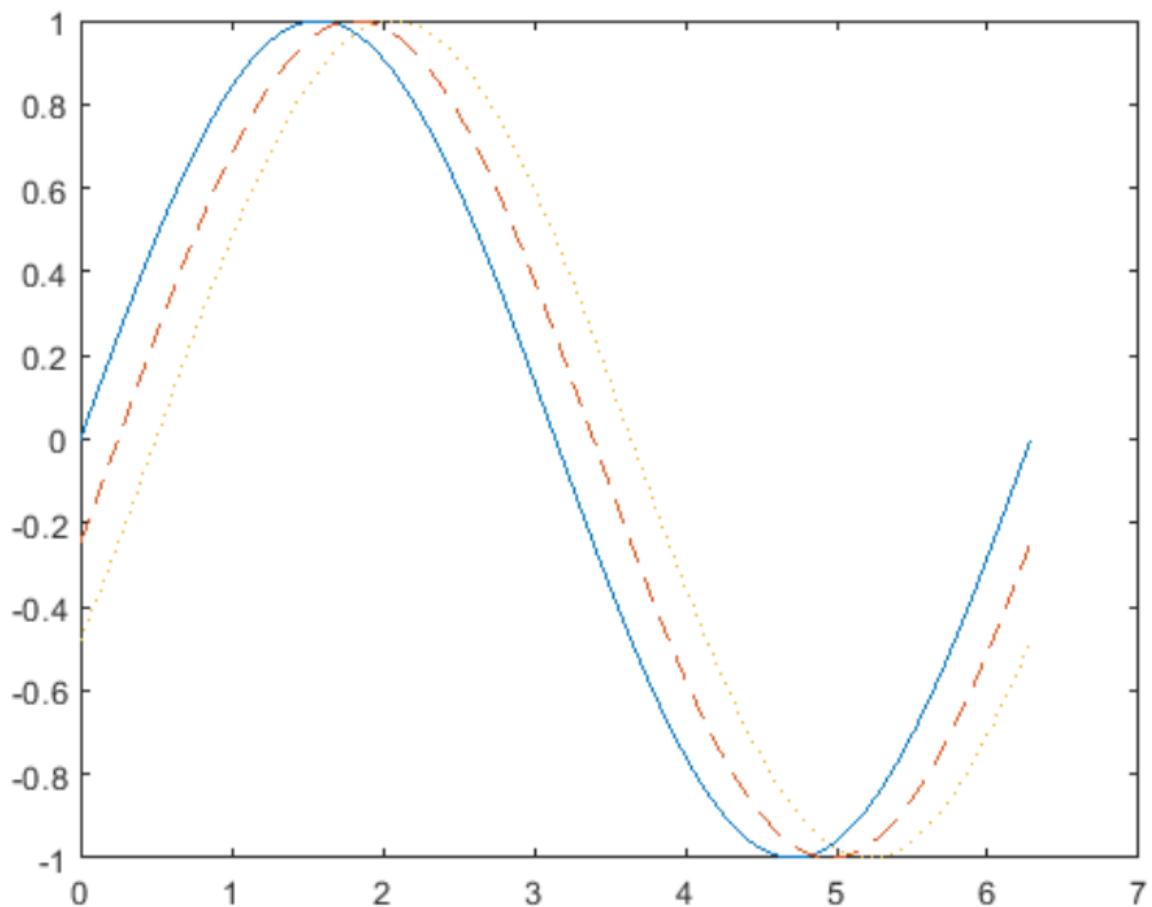


Figure 2.8: Example 1: Basic 2D Plotting

```
1 % EXAMPLE 2: PLOT FITTING
2 n=1:8;
3 x1(n)=[1 2 3 4 5 6 7 8];
```

```

4 y1(n)=[1 2 3.5 4.5 5 5.7 6.8 8];
5 plot(n,x1,'o',n,y1,'o','linewidth',2)
6 title('data points')
7 xlabel('n')
8 ylabel('x1 and y1')
9 grid on
10
11 A1=polyfit(x1,y1,1)
12 xi=linspace(1,8,50);
13 yi=polyval(A1,xi);
14 plot(x1,y1,'o',xi,yi)
15 title('polyfit')
16 xlabel('x1 and xi')
17 ylabel('y1 and yi')
18 grid on
19 A7=polyfit(x1,y1,7)
20 xi=linspace(1,8,50);
21 yi=polyval(A7,xi);
22 plot(x1,y1,'o',xi,yi)
23 title('polyfit')
24 xlabel('x1 and xi')
25 ylabel('y1 and yi')
26 grid on

```

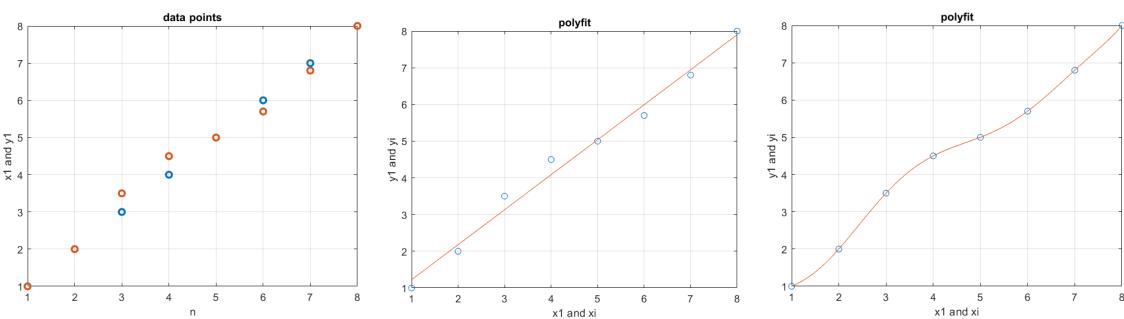


Figure 2.9: Example 2: 2D Plot Fitting

2.2.2.2 | 3D Plotting

■ Mesh Plot

The *mesh* function creates a wireframe mesh. By default, the color of the mesh is propor-

tional to the surface height.

■ Surface Plot

The *surf* function is used to create a 3-D surface plot.

Listing 2.1: Example: 3D Gaussian Plot

```
1 % EXAMPLE: 3D GAUSSIAN FUNCTION FLOT
2
3 a=1; % height of peak
4 b=1; % position of peak center
5 c=1; % standard deviation
6 x=-10:0.1:10;
7 y=x;
8
9 [X,Y]=meshgrid(x,y);
10 g3d=exp(-(X.^2+Y.^2));
11 surf(x,y,g3d)
12 xlim([-5 5])
13 ylim([-5 5])
14 zlim([0 1])
```

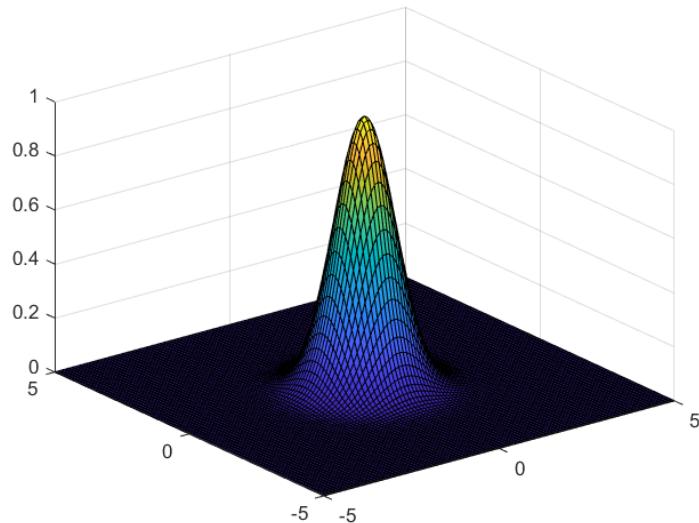


Figure 2.10: 3D Gaussian Function

2.2.3 | App Designer

App Designer lets you create professional apps without having to be a professional software developer. Drag and drop visual components to lay out the design of your graphical user interface (GUI) and use the integrated editor to quickly program its behavior. The corresponding files are saved in the `.mlapp` format.

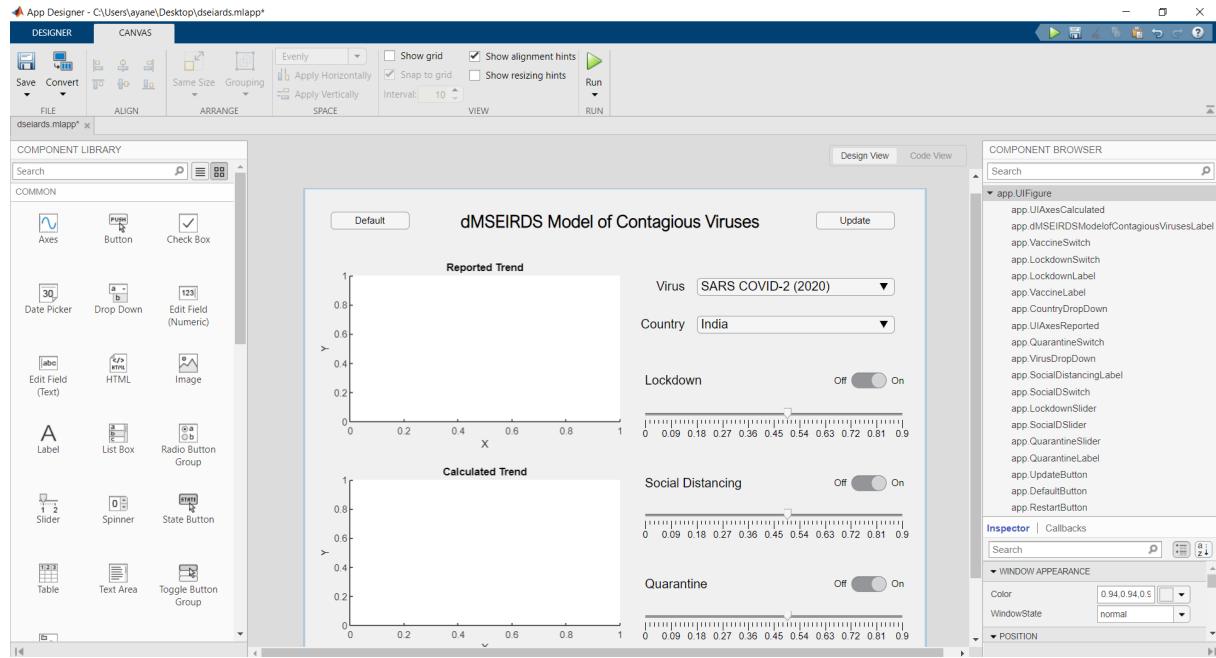


Figure 2.11: Screenshot of our first model plan, changed later

2.2.3.1 | Design a User Interface

Drag and drop visual components to the design canvas and use alignment hints to get a precise layout. App Designer automatically generates the object-oriented code that specifies the app's layout and design.

2.2.3.2 | Define App Behavior

Use the integrated version of the MATLAB Editor to define your app's behavior. App Designer can automatically check for coding problems using the Code Analyzer. You can view warning and error messages about your code as you're writing it, and modify your app based on the messages. You can also model the app behavior using a Stateflow® chart.

Line Plots	Data Distribution Plots	Discrete Data Plots	Geographic Plots	Polar Plots	Contour Plots	Vector Fields	Surface and Mesh Plots	Volume Visualization	Animation	Images
<code>plot</code>	<code>histogram</code>	<code>bar</code>	<code>geobubble</code>	<code>polarplot</code>	<code>contour</code>	<code>quiver</code>	<code>surf</code>	<code>streamline</code>	<code>animateline</code>	<code>image</code>
<code>plot3</code>	<code>histogram2</code>	<code>barh</code>	<code>geoplot</code>	<code>polarmhistogram</code>	<code>contourf</code>	<code>quiver3</code>	<code>surf3</code>	<code>streamslice</code>	<code>comet</code>	<code>imagesc</code>
<code>stairs</code>		<code>bar3</code>	<code>geocatter</code>	<code>polarscatter</code>	<code>contours</code>	<code>feather</code>	<code>surf1</code>	<code>streamparticles</code>	<code>comet3</code>	
<code>pie</code>		<code>bar3h</code>	<code>compass</code>		<code>contour3</code>		<code>ribbon</code>	<code>streamribbon</code>		
<code>errorbar</code>		<code>bar3n</code>	<code>contourslice</code>		<code>contourc</code>		<code>ribbon</code>	<code>streamtube</code>		
<code>area</code>		<code>pareto</code>	<code>epolar</code>		<code>contourf</code>		<code>pcolor</code>	<code>streamtube</code>		
<code>stackedplot</code>		<code>stem</code>	<code>ezpolar</code>		<code>fcontour</code>		<code>fsurf</code>	<code>coneplot</code>		
<code>loglog</code>	<code>scatteredhistogram</code>	<code>stem3</code>	<code>ezsurf</code>		<code>fimplicit</code>		<code>fsuif</code>	<code>coneplot</code>		
<code>semilogx</code>	<code>spy</code>	<code>scatter</code>	<code>ezsurf</code>		<code>implicit3</code>		<code>fsuif</code>	<code>coneplot</code>		
<code>semilogy</code>	<code>plotmatrix</code>	<code>scatter3</code>	<code>ezsurf3</code>		<code>isurface</code>		<code>fsuif</code>	<code>coneplot</code>		
<code>fplot</code>	<code>heatmap</code>	<code>stairs</code>	<code>ezsurf</code>		<code>mesh</code>		<code>fsuif</code>	<code>coneplot</code>		
<code>fplot3</code>	<code>worldcloud</code>		<code>ezsurf3</code>		<code>meshc</code>		<code>fsuif</code>	<code>coneplot</code>		
<code>fimplicit</code>	<code>parallelplot</code>		<code>ezsurf</code>		<code>meshz</code>		<code>fsuif</code>	<code>coneplot</code>		
			<code>ezsurf</code>		<code>waterfall</code>		<code>fsuif</code>	<code>coneplot</code>		
					<code>mesh</code>		<code>fsuif</code>	<code>coneplot</code>		

Figure 2.7: Types of Plots in MATLAB

Methodology

3.1 | Compartment Modification

The model we use as the base line for our model is the *SEIRD* model, a step ahead from *SEIR* which describes the epidemic trend fairly accurately. We first came across this model in [Fonseca i Casas et al. \(2020\)](#) which describes the *SEIRD* using multiple diagrams but the main jist of the model is given by [3.1](#).

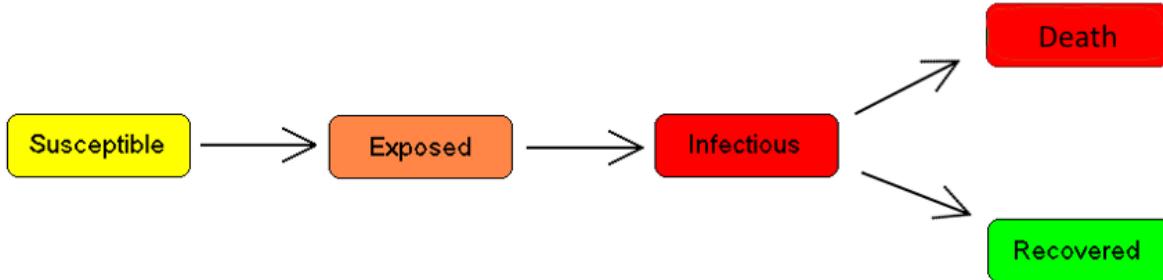


Figure 3.1: Diagrammatic view of the SEIRD model - from [Fonseca i Casas et al. \(2020\)](#)

[Figure 3.2](#) is the diagrammatic view of the compartmental view of our model, which includes an additional *L* and *Q* compartment which stand for

- $L \rightarrow$ Lockdown
- $Q \rightarrow$ Quarantine

as discussed in earlier sections. As visible in the figure, there are three sinks (*L*, *R* and *D*) in our *SEIRD* – *LQ* model compared to the two sinks in the *SEIRD* model (*R* and *D*). Sinks here mean that the population can be contained in those compartments without the requirement of them moving onto another compartment. One way to identify sinks is the one-way flow of population - they can come into the category/compartment, but not go out of it. The *death* compartment is a great example as it would be not just mathematically, but terrifyingly wrong

if someone were able to come back out from the *death* compartment. Hence D is a sink, in case of both *SEIRD* and *SEIRD-LQ*.

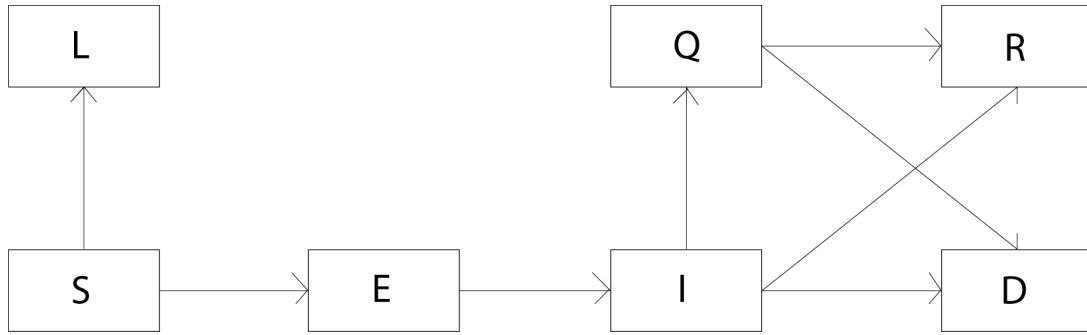


Figure 3.2: Diagrammatic view of our *SEIRD - LQ* model - made in *Adobe Illustrator*

3.2 | Mathematical Tweaking

3.2.1 | SEIRD Model

Let us take over each compartment of the *SEIRD* model one by one and represent its change mathematically.

$$S'(t) = -\frac{\beta S(t) I(t)}{N} \quad (3.1)$$

$$E'(t) = \frac{\beta S(t) I(t)}{N} - \alpha E \quad (3.2)$$

$$I'(t) = \alpha E(t) - \gamma I(t) \quad (3.3)$$

$$R'(t) = \gamma(1 - \mu)I(t) \quad (3.4)$$

$$D'(t) = \gamma\mu I(t) \quad (3.5)$$

$$N = S + E + I + R + D \quad (3.6)$$

3.2.2 | SEIRD-LQ Model

Let us take over each compartment of the *SEIRD-LQ* model one by one and represent its change mathematically.

$$\frac{dS}{dt} = -\beta (1 - C_Q) (1 - C_{SD}) IS - C_L S \quad (3.7)$$

$$\frac{dE}{dt} = \beta (1 - C_Q) (1 - C_{SD}) IS - \sigma E \quad (3.8)$$

$$\frac{dI}{dt} = \sigma E - (1 - C_Q) \gamma I - C_Q \gamma I \quad (3.9)$$

$$\frac{dR}{dt} = (1 - \Delta) \gamma I \quad (3.10)$$

$$\frac{dD}{dt} = \Delta \gamma I \quad (3.11)$$

$$S + E + I + R + D + L = N \quad (3.12)$$

3.3 | Data Collection

For this section, we looked not just into research papers but also the WHO website and other trusted health organization sites.

3.3.1 | Countries (C)

In our model, we are currently taking into account, four countries

Countries	Population (N)	Hospital Capacity (Hosp)
India	1.353e9	0.53
Russia	144.5e6	8.05
UAE	9.631e6	1.4
USA	328.2e6	2.77

3.3.2 | Viruses (V)

In our model, we are currently taking into account, two viruses

- SARS-COVID 2 (2020)
- Ebola (2014)

In case of Ebola, as the outbreak was mainly only bounded by Africa. We take global averages for datasets that do not have enough backup and/or data points.

Virus	Reproduction Rate	Death Rate	Incubation Period	Infectious Period
SARS COVID-2	2.2-2.9	1.5-2.5%	4.2	14
Ebola	1.8-2.1	30-90%	9	20

3.3.3 | Cross Product of C and V

India	N	Hosp	R_0	Death	T_Inc	T_Inf
SARS COVID-2	1.353e9	0.53	2.9	1.49%	4.2	14
Ebola	1.353e9	0.53	1.8	50%	9	20

Russia	N	Hosp	R_0	Death	T_Inc	T_Inf
SARS COVID-2	144.5e6	8.05	2.2	1.72%	4.2	14
Ebola	144.5e6	8.05	1.8	50%	9	20

UAE	N	Hosp	R_0	Death	T_Inc	T_Inf
SARS COVID-2	9.631e6	1.4	2.8	0.37%	4.2	14
Ebola	9.631e6	1.4	1.8	50%	9	20

USA	N	Hosp	R_0	Death	T_Inc	T_Inf
SARS COVID-2	328.2e6	2.7	2.2	4%	4.2	14
Ebola	328.2e6	2.7	1.8	50%	9	20

3.4 | Simulating the Model

The code written by us for our *SEIRD – LQ* model has been shown in [Appendix B](#) and is the second main work of our project. Please click on the hyperlink in previous sentence in blue, in order to view the code. Below is a screenshot [3.3](#) showing the outline of our application. There are more screenshots of our application at work, discussing the visual features and results derived from the application.

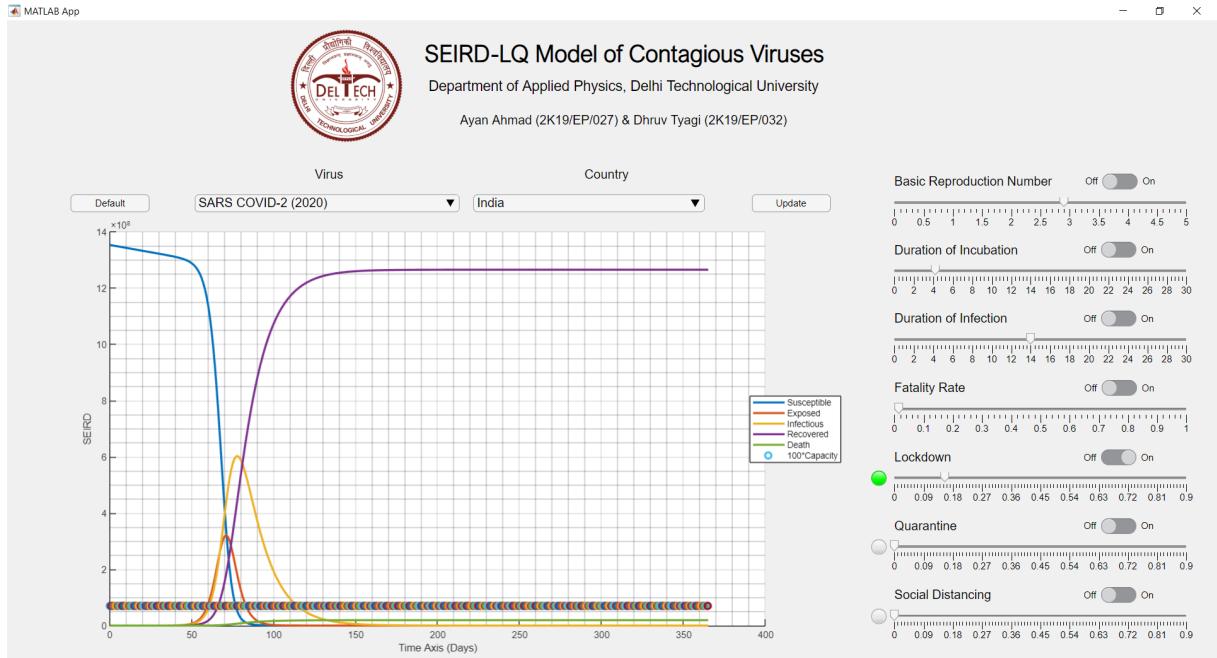


Figure 3.3: A look at the our SEIRD-LQ Application

4

Results

In this section we will be discussing the functionality of each component

- Buttons
- Switches
- Sliders
- Dropdown Menus

via the captions of the screenshots. 4.1 is the basic outline of our application. The screen the user will see on its launch.

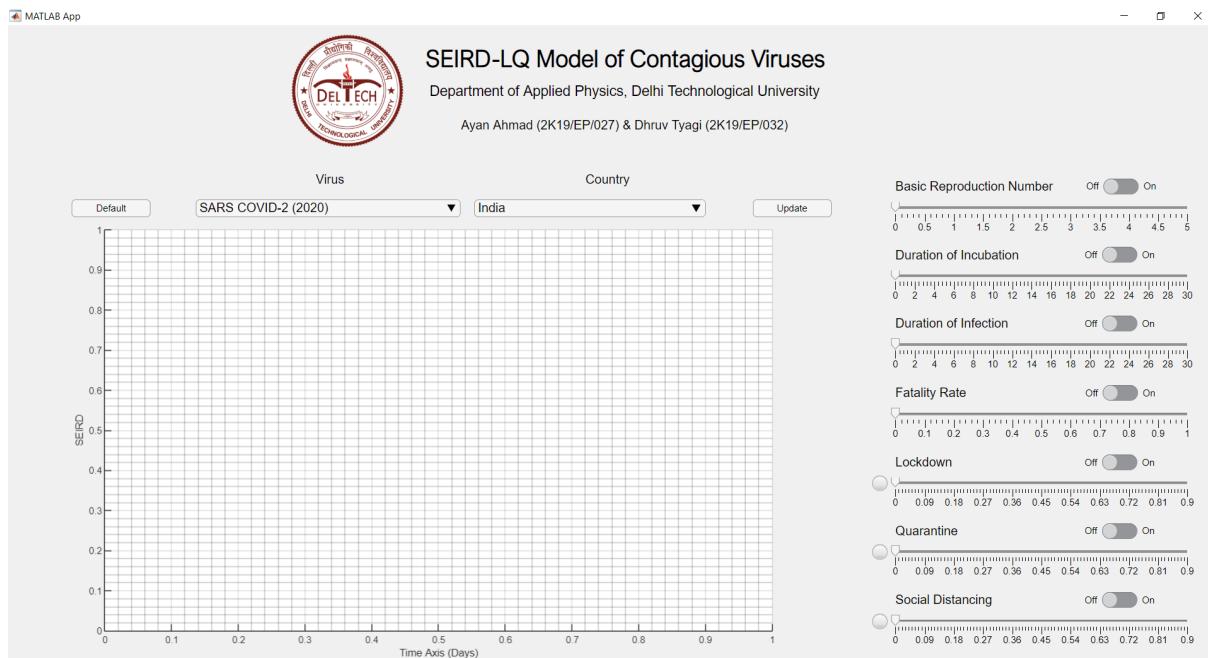


Figure 4.1: Basic outline of the application, startup screen

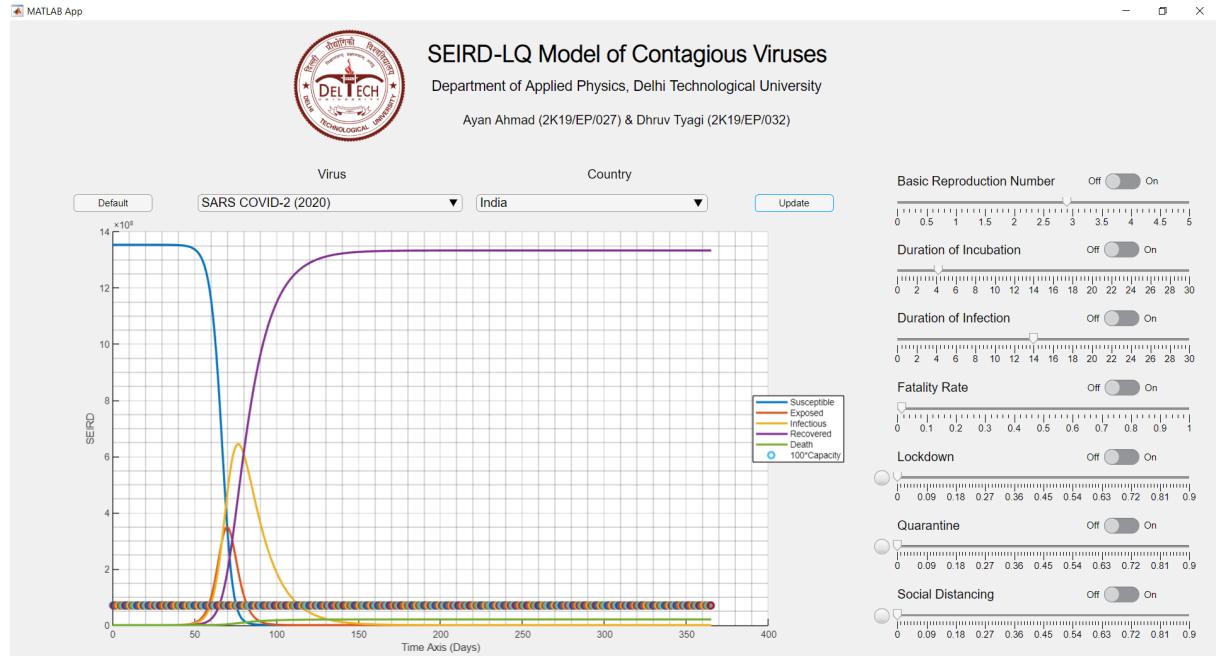


Figure 4.2: The first run requires the user to press the update button, the default values are assigned to each slider and the epidemic trend is plotted.

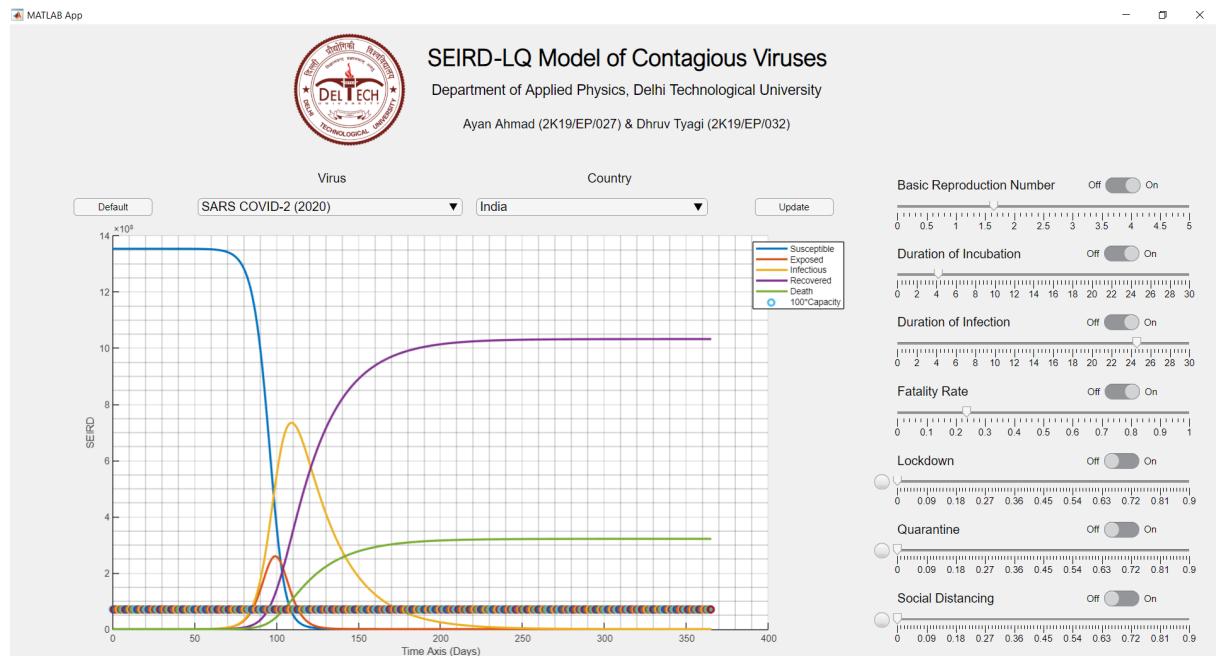


Figure 4.3: The user can switch on the parameter switches in order to be able to change them to their will and see how the curves depend on each parameter in our model.

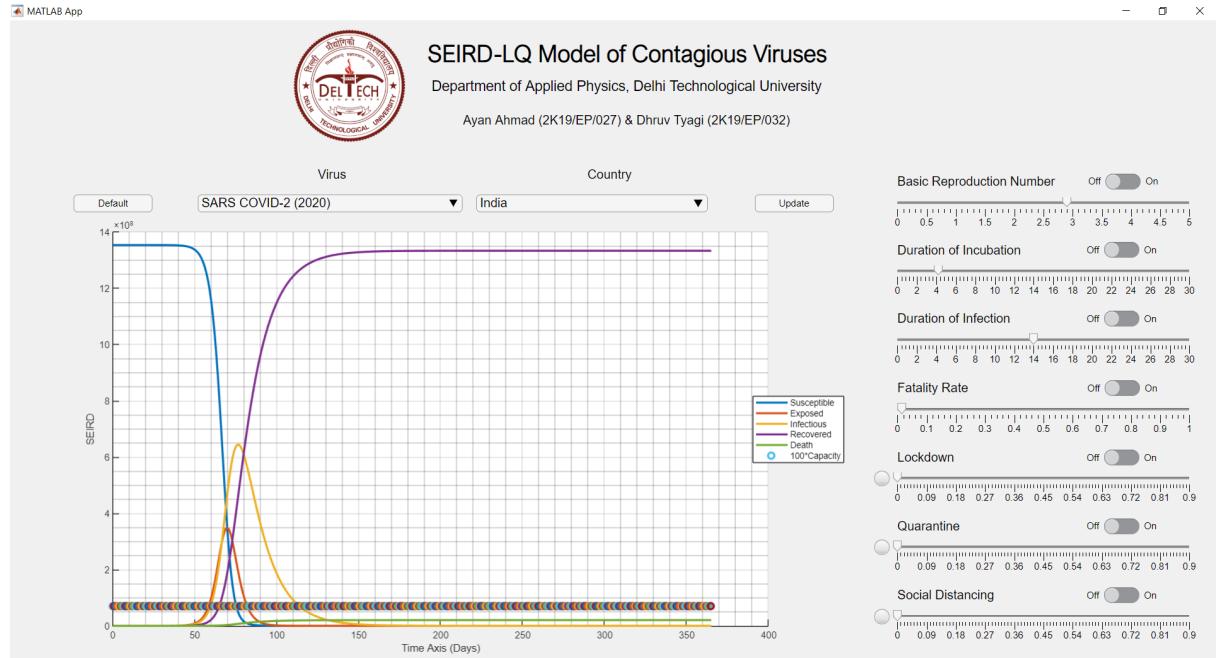


Figure 4.4: On pressing the default, and the update button, all values (sliders and switches) will go back to their original configuration for the particular dropdown combination.

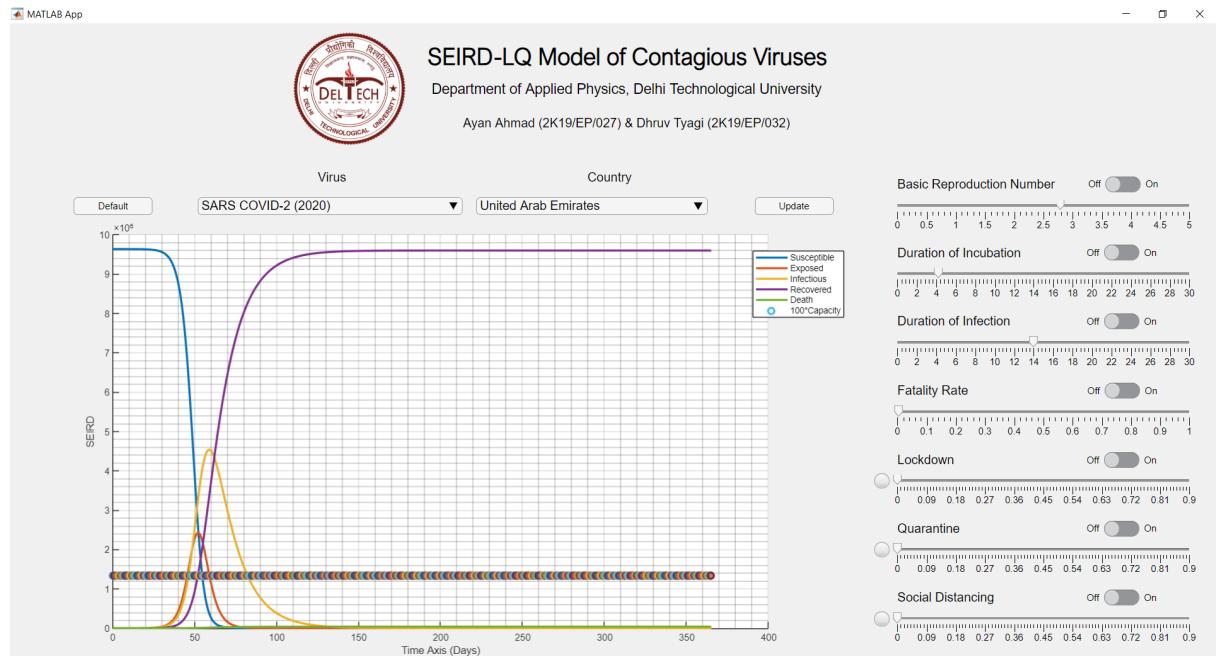


Figure 4.5: The user can select a different country from the country dropdown list. The values will be reset to default for the new combination (stored in code).

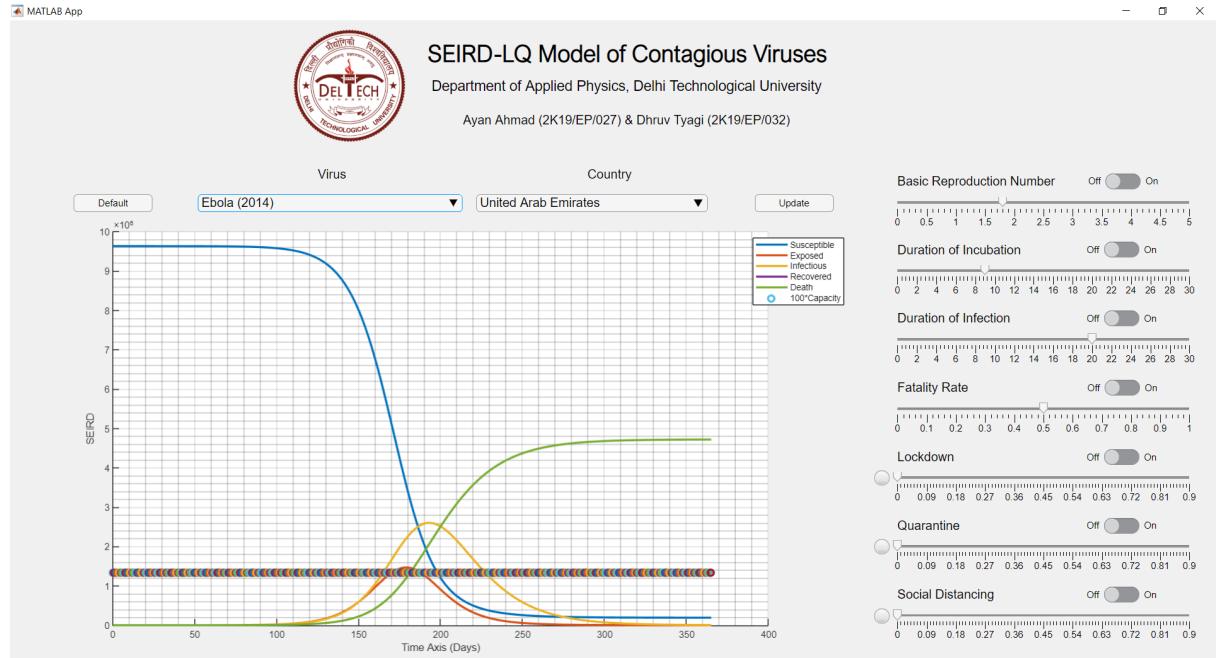


Figure 4.6: The user can select a different virus from the virus dropdown list. The values will be reset to default for the new combination (stored in code).

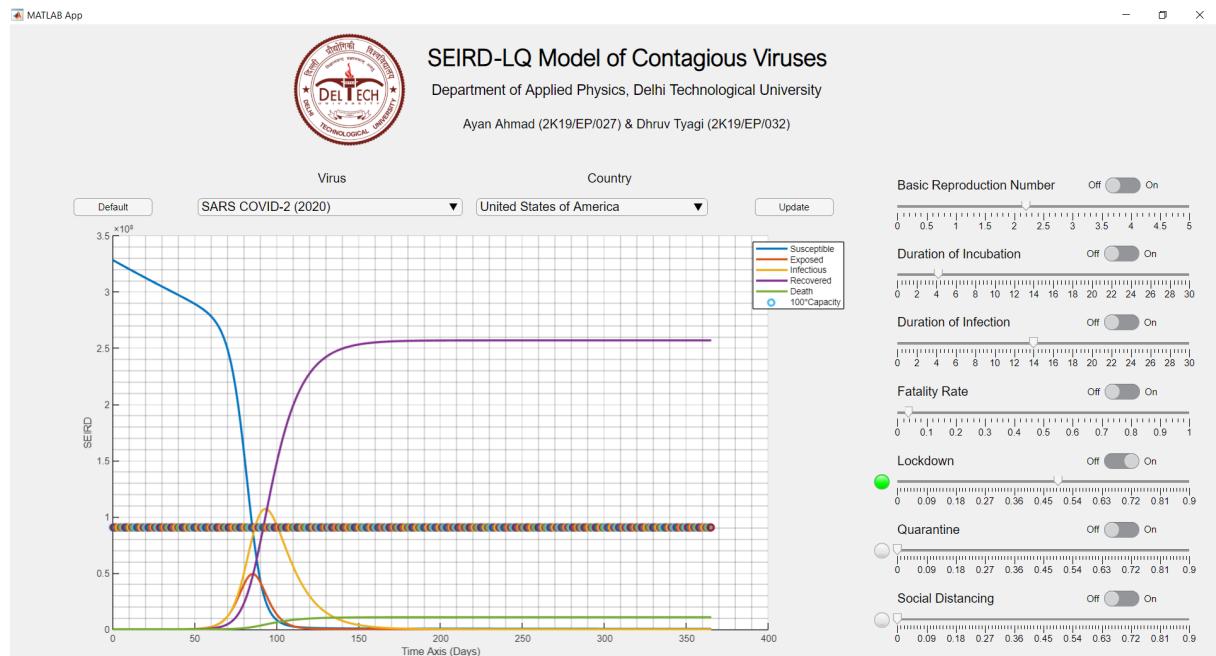


Figure 4.7: This is the change in curves when we set the lockdown quality to roughly 0.5 while keeping other values to their default. The lamp beside lockdown slider goes green.

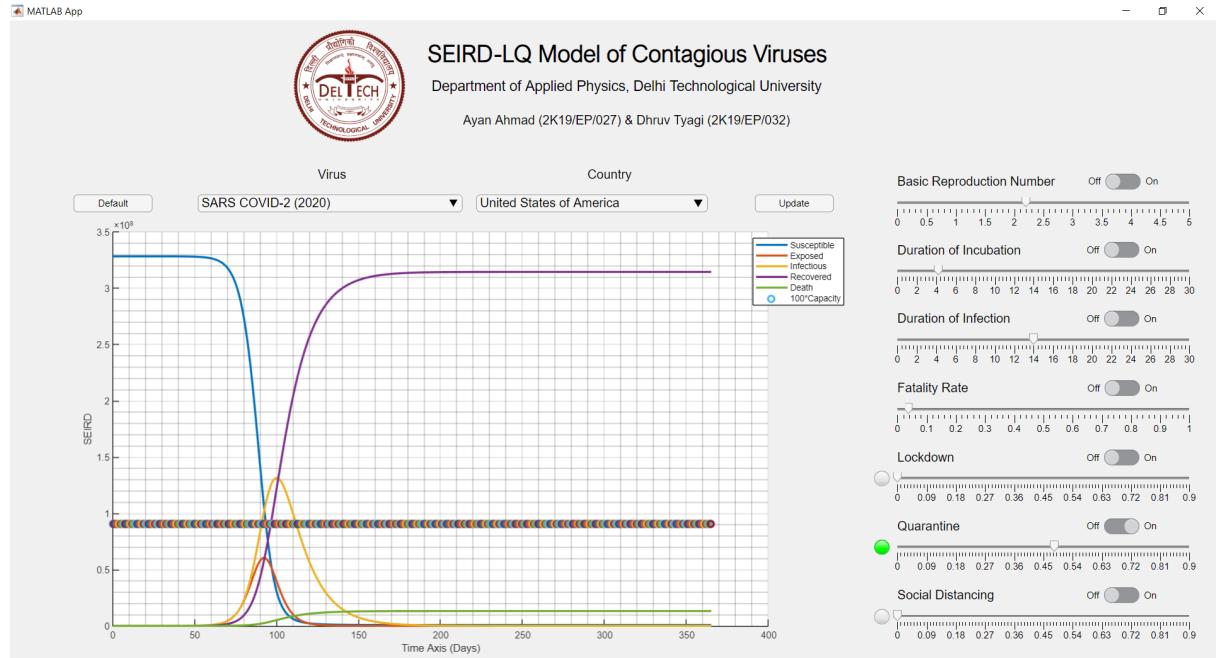


Figure 4.8: This is the change in curves when we set the quarantine quality to roughly 0.5 while keeping other values to their default. The lamp beside quarantine slider goes green.

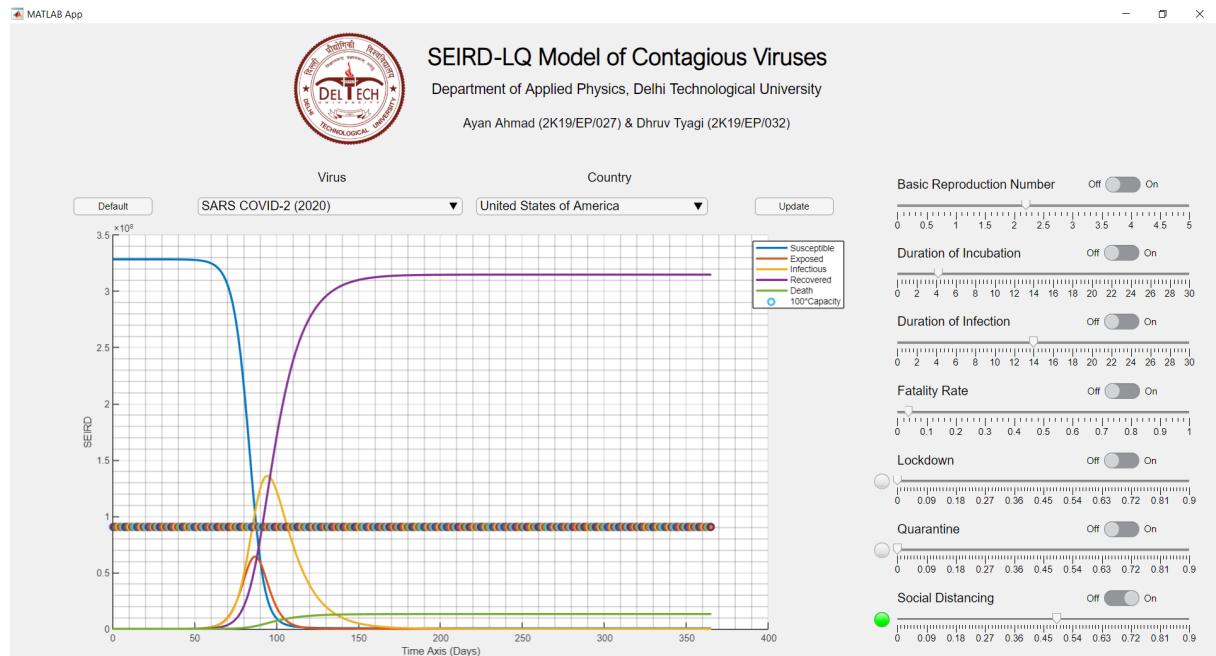


Figure 4.9: This is the change in curves when we set the social distancing quality to roughly 0.5 while keeping other values to their default. The lamp beside distancing slider goes green.

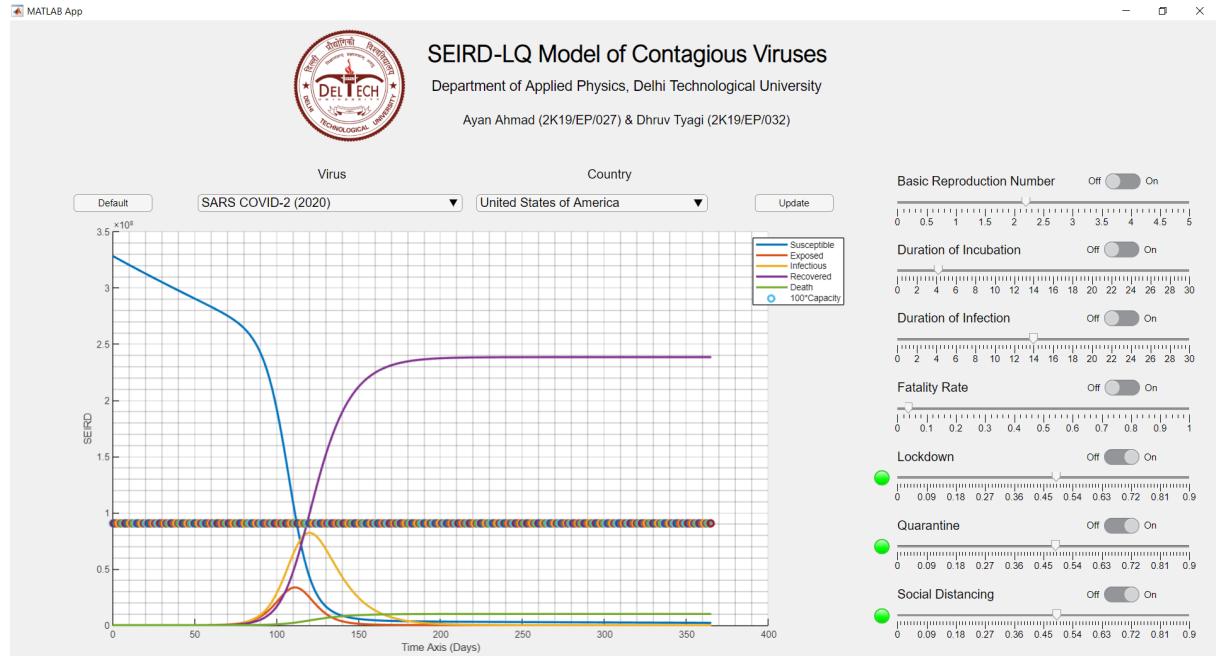


Figure 4.10: This is the change in curves when we set the all prevention quality to roughly 0.5 while keeping other values to their default. The three lamps all go green.

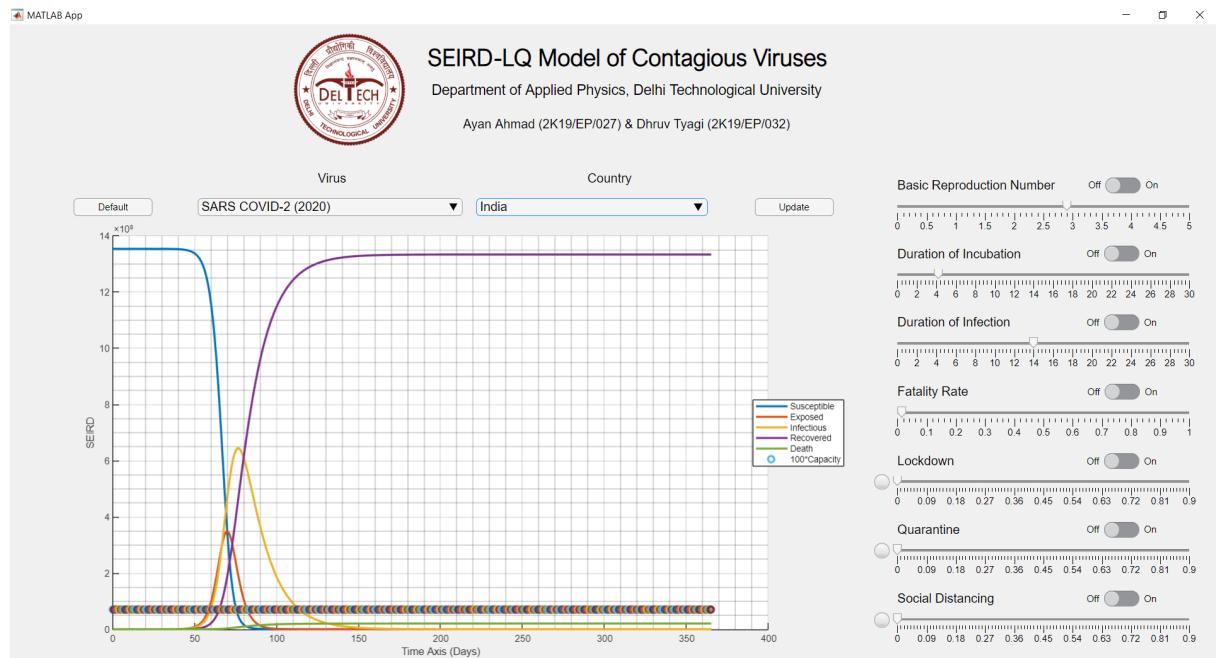


Figure 4.11: Choosing another combination from the dropdown menus automatically sets the values to new default rather than keeping it how it was.

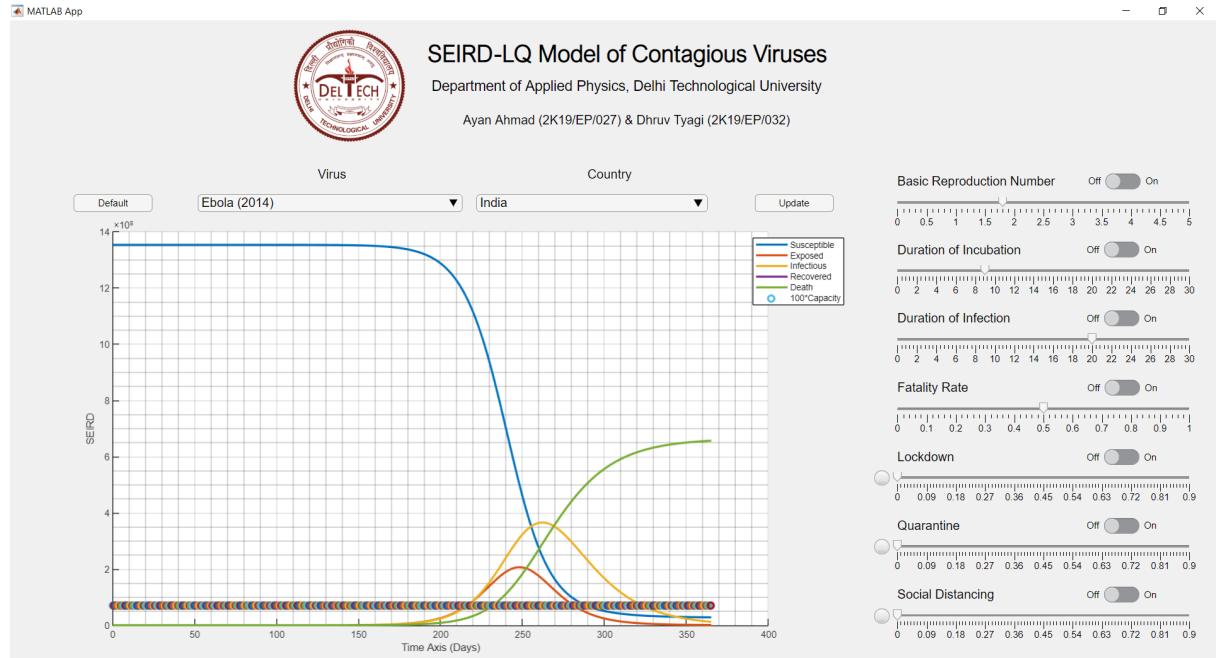


Figure 4.12: For the same combinations and prevention, Ebola causes less infections than Covid 2020.

Discussion & Conclusions

Besides the basic observations of other models, following are the conclusions of our application and modified model

5.0.1 | Effective methods for flattening the curve

Lockdown is the most effective method to flatten the curve. The order of effectiveness in terms of flattening the curve in order from greatest to lowest is as follows (for CoVid 2020)

- Lockdown
- Quarantine
- Social Distancing

5.0.2 | Position and arrival of peak

Smaller populations can result in an early peak as the graphs will have different proportion. The proportion can be described as the length of y axis in the plot and the peak value. For a country with lower population, the peak is expected to arise early assuming same contact density.

5.0.3 | Incubation and Infectious Period

For constant counterpart, T_{inc} is inversely proportional to the infectious peak. This makes sense as longer the incubation time per infection time, lesser the number of exposed people at every point in time. And longer the infectious duration, higher the chance of exposure and transmission as the infection lasts longer for every person contracting it.

5.0.4 | Fatality rate and death count

Higher fatality rate does not necessarily mean higher death count. Although the fatality rate for Ebola was high and it did cause higher deaths in our model. It is not impossible in our model to image a virus that can have that perfect match of contagion and fatality rate so as to decrease death toll by

- Being deadly but not very contagious
- Being highly deadly and medium-level contagious
- Not being deadly but being contagious of any level

Future Work

During our time working on the project, we have realised that we've just seen the tip of the iceberg and there's so much more to epidemics and MATLAB than our current knowledge holds. So as a part of our future work, we have two proposals to suggest and work on beyond the span of the project deadline.

6.1 | Library of Contagious Viruses

We have purposefully designed our code and interface with dropdown lists for country and virus so that we can collection more data as it is published and effortlessly be able to add the constants of model to the list of countries (which determine R_0 , N , $Death_{Rate}$) and viruses (which determine R_0 , $Death_{Rate}$, T_{inc} , T_{inf}) in order to output specific trends in the epidemic.

6.2 | Addition of Parameters and Compartments

It is no surprise that any model, be it *SEIRD* or our *SEIRD – LQ*, do not accurately describe the reported trend for a long enough time interval. This is because on a microscopic scale, there are many other factors that come into play such as asymptotic cases where the person is less likely to be quarantined and so on. If we were to take into account every small detail, we would have too many parameter constants in our equations which would need a lot of data, that too on a microscopic level, which can be difficult considering we must distance ourselves as well. So whenever there is enough data available on the virus in a particular microregion, a parameter can be added to the equation in order to achieve more accurate, non-ideal results.

6.3 | Mathematical Analysis

A more rigorous mathematical analysis and formulation of models for epidemiology.

6.4 | Localization

To dissect the countries further into smaller regions for more accurate data on microscopic changes and more accurate model.

MATLAB SIR Code with Gillespieal Algorithm (Literature Review)

Listing A.1: Gillespieal Algorithm for SIR Model - MATLAB Code

```

1 % SIRS EPIDEMIC MODEL - LITERATURE REVIEW
2
3 % This is not an original code, it had been analyzed by us to better
4 % understand epidemic simulating techniques in MATLAB. Our original
5 % (SEIRD-LQ) code is in the next appendix. The credit for this
6 % particular code goes to Stochastic simulation of epidemics.
7
8 clear all
9 for kk=1:2:3
10    gam=0.1; alpha=.05; delta=.01; beta=0.3; i0=2; % Parameters
11    if kk==1
12        N=100
13        time=120;
14    else
15        N=500
16        time=300;
17    end
18
19 % Eulers method to solve ODE
20
21 % Initial values
22 dt=.05; tim=time/dt;
23 i(1)=i0; s(1)=N-i0;r(1)=0;
24 ssum=0;
25 for tt=1:tim
26    nn=s(tt)+i(tt)+r(tt);
27    i(tt+1)=i(tt)+dt*((beta/nn)*i(tt)*s(tt)-gam*i(tt)-alpha*i(tt));
28    s(tt+1)=s(tt)+dt*(-(beta/nn)*i(tt)*s(tt)+delta*r(tt));
29    r(tt+1)=r(tt)+dt*(gam*i(tt)-delta*r(tt));

```

```

27     ssum=ssum+dt*(beta/nn)*i(tt)*s(tt);
28 end
29 TotalCases=round(ssum)+i0 % Cumm Cases up to time
30
31 subplot(2,2,kk)
32 plot([0:dt:time],i, k -- , linewidth ,2);
33 xlabel( Time );
34 ylabel( Infectives );
35 axis([0,time,0,.3*N])
36 hold on
37
38 subplot(2,2,kk+1)
39 plot([0:dt:time],i, k -- , linewidth ,2);
40 xlabel( Time );
41 ylabel( Infectives );
42 axis([0,20,0,20]);
43 hold on
44
45 for k=1:4 % Sample paths for MC, Gillespie algorithm
46     clear t s i r
47     t(1)=0; i(1)=i0; s(1)=N-i0;r(1)=0;
48     j=1;
49     while i(j)>0 & t(j)<time % Stop hits zero or at time
50         nn=s(j)+i(j)+r(j);
51         u1=rand;u2=rand;
52         den=((beta/nn)*i(j)*s(j)+gam*i(j)+alpha*i(j)
53             +delta*r(j));
54         t(j+1)=-log(u1)/den+t(j) % Time to next event
55         e1=(beta/nn)*s(j)*i(j)/den;
56         e2=e1+gam*i(j)/den;
57         e3=e2+alpha*i(j)/den;
58         e4=e3+delta*r(j)/den;
59         if (u2<=e1)
60             s(j+1)=s(j)-1;
61             i(j+1)=i(j)+1;
62             r(j+1)=r(j);
63         elseif (u2>e1 & u2<=e2)
64             s(j+1)=s(j);

```

```
65      i(j+1)=i(j)-1;
66      r(j+1)=r(j)+1;
67      40 A MatLaB Programs
68      elseif (u2>e2 & u2<=e3)
69      s(j+1)=s(j);
70      i(j+1)=i(j)-1;
71      r(j+1)=r(j);
72      else
73      s(j+1)=s(j)+1;
74      i(j+1)=i(j);
75      r(j+1)=r(j)-1;
76      end
77      j=j+1;
78  end
79
80 if k==1
81     subplot(2,2,kk)
82     stairs(t,i, y - , linewidth ,2);
83     hold on
84     subplot(2,2,kk+1)
85     stairs(t,i, y - , linewidth ,2);
86     hold on
87 end
88 if k==2
89     subplot(2,2,kk)
90     stairs(t,i, b - , linewidth ,2);
91     hold on
92     subplot(2,2,kk+1)
93     stairs(t,i, b - , linewidth ,2);
94     hold on
95 end
96 if k==3
97     subplot(2,2,kk)
98     stairs(t,i, g - , linewidth ,2);
99     hold on
100    subplot(2,2,kk+1)
101    stairs(t,i, g - , linewidth ,2);
102    hold on
```

```

103 end
104 if k==4
105 subplot(2,2,kk)
106 stairs(t,i, r - , linewidth ,2);
107 hold off
108 subplot(2,2,kk+1)
109 stairs(t,i, r - , linewidth ,2);
110 hold off
111 end
112 end
113
114 % Estimate Probability of Epidemic Extinction
115 count=0; tots=10000;
116 for k=1:tots % Number of sample paths
117 clear t s i r
118 i=i0; s=N-i0; r=0;
119 j=1;
120 outb=min(25,.25*N);
121 sumi=i0;
122 while i>0 & sumi<outb %Stop hits zero or cumm cases=outb
123 u1=rand; u2=rand;
124 nn=s+i+r;
125 den=((beta/nn)*i*s+gam*i+alpha*i+delta*r);
126 e1=(beta/nn)*s*i/den;
127 e2=e1+gam*i/den;
128 e3=e2+alpha*i/den;
129 e4=e3+delta*r/den;
130 if (u2<=e1)
131 i=i+1;
132 s=s-1;
133 r=r;
134 sumi=sumi+1;
135 elseif (u2>e1 & u2<=e2)
136 s=s;
137 i=i-1;
138 r=r+1;
139 elseif (u2>e2 & u2<=e3)
140 s=s;

```

```
141      i=i-1;
142      r=r;
143      else
144      s=s+1;
145      i=i;
146      r=r-1;
147      end
148      j=j+1;
149      end
150      if i==0
151      count=count+1;
152      end
153 end
154 probextSP=count/tots % Ext Approx from Sample Paths
155 probextBP=((gam+alpha)/beta)^(i0) % Ext Est from BP
156 end
```

MATLAB Code for SEIRD-LQ App (Methodology)

```

1 classdef SEIRD_LQ < matlab.apps.AppBase
2
3 % Properties that correspond to app components
4 properties (Access = public)
5     UIFigure                      matlab.ui.Figure
6     UIAxesCalculated              matlab.ui.control.UIAxes
7     SEIRDLQModelofContagiousVirusesLabel matlab.ui.control.Label
8     CountryDropDown                matlab.ui.control.DropDown
9     VirusDropDown                 matlab.ui.control.DropDown
10    UpdateButton                  matlab.ui.control.Button
11    DefaultButton                 matlab.ui.control.Button
12    AyanAhmad2K19EP027DhruvTyagi2K19EP032Label matlab.ui.control
13        .Label
14    Image                         matlab.ui.control.Image
15    DepartmentofAppliedPhysicsDelhiTechnologicalUniversityLabel
16        matlab.ui.control.Label
17    SocialDSwitch                 matlab.ui.control.Switch
18    SocialDistancingLabel         matlab.ui.control.Label
19    SocialDSlider                 matlab.ui.control.Slider
20    SocialDLamp                   matlab.ui.control.Lamp
21    QuarantineSwitch             matlab.ui.control.Switch
22    QuarantineLabel              matlab.ui.control.Label
23    QuarantineSlider             matlab.ui.control.Slider
24    QuarantineLamp               matlab.ui.control.Lamp
25    LockdownSwitch               matlab.ui.control.Switch
26    LockdownLabel_5              matlab.ui.control.Label
27    LockdownSlider               matlab.ui.control.Slider
28    LockdownLamp                 matlab.ui.control.Lamp
29    FatalityRateLabel            matlab.ui.control.Label
30    FatalitySlider               matlab.ui.control.Slider
31    DurationofInfectionLabel    matlab.ui.control.Label

```

```
30      DinfSlider           matlab.ui.control.Slider
31      DurationofIncubationLabel matlab.ui.control.Label
32      DincSlider            matlab.ui.control.Slider
33      VirusDropDownLabel_2   matlab.ui.control.Label
34      CountryDropDown_2Label_2 matlab.ui.control.Label
35      BasicReproductionNumberLabel matlab.ui.control.Label
36      BRNSlider             matlab.ui.control.Slider
37      DincSwitch            matlab.ui.control.Switch
38      DinfSwitch            matlab.ui.control.Switch
39      FatalitySwitch        matlab.ui.control.Switch
40      BRNSwitch             matlab.ui.control.Switch
41  end
42
43 % Callbacks that handle component events
44 methods (Access = private)
45
46 % Callback function: BRNSlider, BRNSwitch, CountryDropDown,
47 % DincSlider, DincSwitch, DinfSlider, DinfSwitch,
48 % FatalitySlider, FatalitySwitch, LockdownSlider,
49 % LockdownSwitch, QuarantineSlider, QuarantineSwitch,
50 % SocialDSlider, SocialDSwitch, UpdateButton, VirusDropDown
51 function UpdateButtonPushed(app, event)
52
53 %% GLOBAL VARIABLES
54
55 global N;
56 global R_0;
57 global death;
58 global T_inc;
59 global T_inf;
60
61 global hosp;
62
63 global beta;
64 global gamma;
65 global sigma;
```

```
68 %% PARAMETERS W.R.T DROP DOWN MENU SELECTIONS (LIBRARY  
69 % SECTION)  
70 % The parameters (or constants) of the equations to be used  
71 % will have different values for a given region and virus.  
72  
73 switch char(app.CountryDropDown.Value)  
74 case char('India')  
75 N=1.353e9;  
76 hosp=(N/1000)*0.53;  
77 switch char(app.VirusDropDown.Value)  
78 case char('SARS COVID-2 (2020)')  
79 R_0=2.9;  
80 death=1-0.9851;  
81 T_inc=4.2;  
82 T_inf=14;  
83 case char('Ebola (2014)')  
84 R_0=1.8;  
85 death=1-0.5;  
86 T_inc=9;  
87 T_inf=20;  
88 end  
89  
90 case char('Russia')  
91 N=144.5e6;  
92 hosp=(N/1000)*8.05;  
93 switch char(app.VirusDropDown.Value)  
94 case char('SARS COVID-2 (2020)')  
95 R_0=2.2;  
96 death=1-0.9828;  
97 T_inc=4.2;  
98 T_inf=14;  
99 case char('Ebola (2014)')  
100 R_0=1.8;  
101 death=1-0.5;  
102 T_inc=9;  
103 T_inf=20;  
end
```

```

104     case char('United Arab Emirates')
105         N=9.631e6;
106         hosp=(N/1000)*1.4;
107         switch char(app.VirusDropDown.Value)
108             case char('SARS COVID-2 (2020)')
109                 R_0=2.8;
110                 death=1-0.9963;
111                 T_inc=4.2;
112                 T_inf=14;
113             case char('Ebola (2014)')
114                 R_0=1.8;
115                 death=1-0.5;
116                 T_inc=9;
117                 T_inf=20;
118         end
119
120     case char('United States of America')
121         N=328.2e6;
122         hosp=(N/1000)*2.77;
123         switch char(app.VirusDropDown.Value)
124             case char('SARS COVID-2 (2020)')
125                 R_0=2.2;
126                 death=1-0.96;
127                 T_inc=4.2;
128                 T_inf=14;
129             case char('Ebola (2014)')
130                 R_0=1.8;
131                 death=1-0.5;
132                 T_inc=9;
133                 T_inf=20;
134         end
135
136     end
137
138
139 %% FUNCTIONALITY OF SWITCHES
140 % If the switch is off for a particular quantity, its value
141 % will be 0, else its value will take the default value

```

```
    assigned above.  
141  
142     if isequal(app.BRNSwitch.Value, 'On')  
143         R_0=app.BRNSlider.Value;  
144     else  
145         app.BRNSlider.Value=R_0;  
146     end  
147     if isequal(app.DincSwitch.Value, 'On')  
148         T_inc=app.DincSlider.Value;  
149     else  
150         app.DincSlider.Value=T_inc;  
151     end  
152     if isequal(app.DinfSwitch.Value, 'On')  
153         T_inf=app.DinfSlider.Value;  
154     else  
155         app.DinfSlider.Value=T_inf;  
156     end  
157     if isequal(app.FatalitySwitch.Value, 'On')  
158         death=app.FatalitySlider.Value;  
159     else  
160         app.FatalitySlider.Value=death;  
161     end  
162  
163  
164     if isequal(app.LockdownSwitch.Value, 'Off')  
165         app.LockdownSlider.Value=0;  
166         app.LockdownLamp.Color=[0.90,0.90,0.90];  
167     else  
168         app.LockdownLamp.Color=[0,1,0];  
169     end  
170     if isequal(app.SocialDSwitch.Value, 'Off')  
171         app.SocialDSlider.Value=0;  
172         app.SocialDLamp.Color=[0.90,0.90,0.90];  
173     else  
174         app.SocialDLamp.Color=[0,1,0];  
175     end  
176     if isequal(app.QuarantineSwitch.Value, 'Off')  
177         app.QuarantineSlider.Value=0;
```

```

178         app.QuarantineLamp.Color=[0.90,0.90,0.90];
179     else
180         app.QuarantineLamp.Color=[0,1,0];
181     end
182
183
184 %% FORMULAS
185
186 % BASIC
187 beta=R_0/(N*T_inc);
188 gamma = 1/T_inf;
189 sigma = 1/T_inc;
190
191 % ODE
192 t = 0:1:365;
193 y0 = [N-28, 0, 25, 3, 0];
194 [t,y]=ode45(@(t,y) ode_solve(t,y), t, y0);
195
196
197 %% PLOTTING
198
199 plot(app.UIAxesCalculated,t,y,t,100*hosp,'o','LineWidth', 2)
200 legend(app.UIAxesCalculated, 'Susceptible', 'Exposed',
201       'Infectious', 'Recovered', 'Death', '100*Capacity', 'Location',
202       'Best')
203
204 %% ODE SOLVER FUNCTION
205
206 function dydt = ode_solve(t,y)
207
208 S = y(1);
209 E= y(2);
210 I = y(3);
211
212 dS = -(beta).*(1-app.QuarantineSlider.Value/3).*(1-app.
213 SocialDSlider.Value/5).*I.*S - (app.LockdownSlider.Value
214 /200).*S;

```

```

212     dL = (app.LockdownSlider.Value/200).*S;
213     dE = (beta).*(1-app.QuarantineSlider.Value/3).*(1-app.
214         SocialDSlider.Value/5).*I.*S - sigma.*E;
215     dQ = (app.QuarantineSlider.Value/3).*I - (1-app.
216         QuarantineSlider.Value/3).*gamma*(1-death)*I - (app.
217         QuarantineSlider.Value/3).*(death)*gamma*I;
218     dI = sigma.*E - (1-app.QuarantineSlider.Value/3).*gamma.*I - (
219         app.QuarantineSlider.Value/3).*gamma.*I;
220     dR = gamma*(1-death)*I;
221     dD = gamma*(death)*I;
222
223     dydt = [dS; dE; dI; dR; dD];
224 end
225 end
226
227 % Button pushed function: DefaultButton
228 function DefaultButtonPushed(app, event)
229
230     %% DEFAULT SETTING
231     % Puts all the switches in 'off' configuration and that
232     % triggers the update button which causes all the
233     % sliders to take default value as in library section.
234
235     app.BRNSwitch.Value='Off';
236     app.DincSwitch.Value='Off';
237     app.DinfSwitch.Value='Off';
238     app.FatalitySwitch.Value='Off';
239
240     end
241 end
242
243
244 % Component initialization
245 methods (Access = private)
246
247     % Create UIFigure and components
248     function createComponents(app)
249

```

```

243 % Create UIFigure and hide until all components are
244 % created
245 app.UIFigure = uifigure('Visible', 'off');
246 app.UIFigure.Position = [100 100 1363 869];
247 app.UIFigure.Name = 'MATLAB App';
248
249 % Create UIAxesCalculated
250 app.UIAxesCalculated = uiaxes(app.UIFigure);
251 title(app.UIAxesCalculated, '')
252 xlabel(app.UIAxesCalculated, 'Time Axis (Days)')
253 ylabel(app.UIAxesCalculated, 'SEIRD')
254 app.UIAxesCalculated.PlotBoxAspectRatio =
255 [1.66614420062696 1 1];
256 app.UIAxesCalculated.MinorGridLineStyle = '-';
257 app.UIAxesCalculated.XGrid = 'on';
258 app.UIAxesCalculated.XMinorGrid = 'on';
259 app.UIAxesCalculated.YGrid = 'on';
260 app.UIAxesCalculated.YMinorGrid = 'on';
261 app.UIAxesCalculated.Position = [13 9 941 577];
262
263 % Create SEIRDLQModelofContagiousVirusesLabel
264 app.SEIRDLQModelofContagiousVirusesLabel = uilabel(app.
265 UIFigure);
266 app.SEIRDLQModelofContagiousVirusesLabel.FontName =
267 'Arial Nova Light';
268 app.SEIRDLQModelofContagiousVirusesLabel.FontSize = 30;
269 app.SEIRDLQModelofContagiousVirusesLabel.FontWeight =
270 'bold';
271 app.SEIRDLQModelofContagiousVirusesLabel.Position = [531
272 776 511 59];
273 app.SEIRDLQModelofContagiousVirusesLabel.Text = 'SEIRD-LQ
274 Model of Contagious Viruses';
275
276 % Create CountryDropDown
277 app.CountryDropDown = uidropdown(app.UIFigure);
278 app.CountryDropDown.Items = {'India', 'Russia', 'United
279 Arab Emirates', 'United States of America'};

```

```

272 app.CountryDropDown.ValueChangedFcn = createCallbackFcn(
273     app, @UpdateButtonPushed, true);
274 app.CountryDropDown.FontSize = 16;
275 app.CountryDropDown.Position = [507 597 251 22];
276 app.CountryDropDown.Value = 'India';
277
278 % Create VirusDropDown
279 app.VirusDropDown = uidropdown(app.UIFigure);
280 app.VirusDropDown.Items = {'SARS COVID-2 (2020)', 'Ebola
281 (2014)'};
282 app.VirusDropDown.ValueChangedFcn = createCallbackFcn(app
283     , @UpdateButtonPushed, true);
284 app.VirusDropDown.FontSize = 16;
285 app.VirusDropDown.Position = [239 597 251 22];
286 app.VirusDropDown.Value = 'SARS COVID-2 (2020)';
287
288 % Create UpdateButton
289 app.UpdateButton = uibutton(app.UIFigure, 'push');
290 app.UpdateButton.ButtonPushedFcn = createCallbackFcn(app,
291     @UpdateButtonPushed, true);
292 app.UpdateButton.Position = [818 597 100 22];
293 app.UpdateButton.Text = 'Update';
294
295 % Create DefaultButton
296 app.DefaultButton = uibutton(app.UIFigure, 'push');
297 app.DefaultButton.ButtonPushedFcn = createCallbackFcn(app
298     , @DefaultButtonPushed, true);
299 app.DefaultButton.Position = [82 597 100 22];
300 app.DefaultButton.Text = 'Default';
301
302 % Create AyanAhmad2K19EP027DhruvTyagi2K19EP032Label
303 app.AyanAhmad2K19EP027DhruvTyagi2K19EP032Label = uilabel(
304     app.UIFigure);
305 app.AyanAhmad2K19EP027DhruvTyagi2K19EP032Label.FontSize =
306     16;
307 app.AyanAhmad2K19EP027DhruvTyagi2K19EP032Label.Position =
308     [576 707 422 24];

```

```

301 app.AyanAhmad2K19EP027DhruvTyagi2K19EP032Label.Text = 'Ayan Ahmad (2K19/EP/027) & Dhruv Tyagi (2K19/EP/032)';
302
303 % Create Image
304 app.Image = uiimage(app.UIFigure);
305 app.Image.Position = [360 692 144 143];
306 app.Image.ImageSource = 'DTU.png';
307
308 % Create
309 DepartmentofAppliedPhysicsDelhiTechnologicalUniversityLabel
310
311 app.
312     DepartmentofAppliedPhysicsDelhiTechnologicalUniversityLabel
313         = uilabel(app.UIFigure);
314
315 app.
316     DepartmentofAppliedPhysicsDelhiTechnologicalUniversityLabel
317         .FontSize = 18;
318
319 app.
320     DepartmentofAppliedPhysicsDelhiTechnologicalUniversityLabel
321         .Position = [536 747 503 30];
322
323 app.
324     DepartmentofAppliedPhysicsDelhiTechnologicalUniversityLabel
325         .Text = 'Department of Applied Physics, Delhi
Technology University';

% Create SocialDSwitch
app.SocialDSwitch = uiswitch(app.UIFigure, 'slider');
app.SocialDSwitch.ValueChangedFcn = createCallbackFcn(app
, @UpdateButtonPushed, true);
app.SocialDSwitch.Position = [1261 78 45 20];

% Create SocialDistancingLabel
app.SocialDistancingLabel = uilabel(app.UIFigure);
app.SocialDistancingLabel.FontSize = 16;
app.SocialDistancingLabel.Position = [999 77 129 22];
app.SocialDistancingLabel.Text = 'Social Distancing';

% Create SocialDSlider

```

```
326 app.SocialDSlider = uislider(app.UIFigure);
327 app.SocialDSlider.Limits = [0 0.9];
328 app.SocialDSlider.ValueChangedFcn = createCallbackFcn(app
329     , @UpdateButtonPushed, true);
330 app.SocialDSlider.Position = [999 59 326 3];
331
332 % Create SocialDLamp
333 app.SocialDLamp = uilamp(app.UIFigure);
334 app.SocialDLamp.Position = [969 50 20 20];
335 app.SocialDLamp.Color = [0.902 0.902 0.902];
336
337 % Create QuarantineSwitch
338 app.QuarantineSwitch = uiswitch(app.UIFigure, 'slider');
339 app.QuarantineSwitch.ValueChangedFcn = createCallbackFcn(
340     app, @UpdateButtonPushed, true);
341 app.QuarantineSwitch.Position = [1261 170 45 20];
342
343 % Create QuarantineLabel
344 app.QuarantineLabel = uilabel(app.UIFigure);
345 app.QuarantineLabel.FontSize = 16;
346 app.QuarantineLabel.Position = [999 169 84 22];
347 app.QuarantineLabel.Text = 'Quarantine';
348
349 % Create QuarantineSlider
350 app.QuarantineSlider = uislider(app.UIFigure);
351 app.QuarantineSlider.Limits = [0 0.9];
352 app.QuarantineSlider.ValueChangedFcn = createCallbackFcn(
353     app, @UpdateButtonPushed, true);
354 app.QuarantineSlider.Position = [999 151 326 3];
355
356 % Create QuarantineLamp
357 app.QuarantineLamp = uilamp(app.UIFigure);
358 app.QuarantineLamp.Position = [969 142 20 20];
359 app.QuarantineLamp.Color = [0.902 0.902 0.902];
360
361 % Create LockdownSwitch
362 app.LockdownSwitch = uiswitch(app.UIFigure, 'slider');
```

```
360 app.LockdownSwitch.ValueChangedFcn = createCallbackFcn(  
361     app, @UpdateButtonPushed, true);  
362 app.LockdownSwitch.Position = [1261 261 45 20];  
363  
364 % Create LockdownLabel_5  
365 app.LockdownLabel_5 = uilabel(app.UIFigure);  
366 app.LockdownLabel_5.FontSize = 16;  
367 app.LockdownLabel_5.Position = [999 260 78 22];  
368 app.LockdownLabel_5.Text = 'Lockdown';  
369  
370 % Create LockdownSlider  
371 app.LockdownSlider = uislider(app.UIFigure);  
372 app.LockdownSlider.Limits = [0 0.9];  
373 app.LockdownSlider.ValueChangedFcn = createCallbackFcn(  
374     app, @UpdateButtonPushed, true);  
375 app.LockdownSlider.Position = [999 242 326 3];  
376  
377 % Create LockdownLamp  
378 app.LockdownLamp = uilamp(app.UIFigure);  
379 app.LockdownLamp.Position = [969 233 20 20];  
380 app.LockdownLamp.Color = [0.902 0.902 0.902];  
381  
382 % Create FatalityRateLabel  
383 app.FatalityRateLabel = uilabel(app.UIFigure);  
384 app.FatalityRateLabel.FontSize = 16;  
385 app.FatalityRateLabel.Position = [999 352 95 22];  
386 app.FatalityRateLabel.Text = 'Fatality Rate';  
387  
388 % Create FatalitySlider  
389 app.FatalitySlider = uislider(app.UIFigure);  
390 app.FatalitySlider.Limits = [0 1];  
391 app.FatalitySlider.ValueChangedFcn = createCallbackFcn(  
392     app, @UpdateButtonPushed, true);  
393 app.FatalitySlider.Position = [999 334 326 3];  
394  
395 % Create DurationofInfectionLabel  
396 app.DurationofInfectionLabel = uilabel(app.UIFigure);  
397 app.DurationofInfectionLabel.FontSize = 16;
```

```
395 app.DurationofInfectionLabel.Position = [999 444 149 22];
396 app.DurationofInfectionLabel.Text = 'Duration of
397     Infection';
398
399 % Create DinfSlider
400 app.DinfSlider = uislider(app.UIFigure);
401 app.DinfSlider.Limits = [0 30];
402 app.DinfSlider.ValueChangedFcn = createCallbackFcn(app, @
403     UpdateButtonPushed, true);
404 app.DinfSlider.Position = [999 426 326 3];
405
406 % Create DurationofIncubationLabel
407 app.DurationofIncubationLabel = uilabel(app.UIFigure);
408 app.DurationofIncubationLabel.FontSize = 16;
409 app.DurationofIncubationLabel.Position = [999 535 162
410     22];
411 app.DurationofIncubationLabel.Text = 'Duration of
412     Incubation';
413
414 % Create DincSlider
415 app.DincSlider = uislider(app.UIFigure);
416 app.DincSlider.Limits = [0 30];
417 app.DincSlider.ValueChangedFcn = createCallbackFcn(app, @
418     UpdateButtonPushed, true);
419 app.DincSlider.Position = [999 517 326 3];
420
421 % Create VirusDropDownLabel_2
422 app.VirusDropDownLabel_2 = uilabel(app.UIFigure);
423 app.VirusDropDownLabel_2.HorizontalAlignment = 'right';
424 app.VirusDropDownLabel_2.FontSize = 16;
425 app.VirusDropDownLabel_2.Position = [345 635 40 22];
426 app.VirusDropDownLabel_2.Text = 'Virus';
```

```

427     app.CountryDropDown_2Label_2.Position = [604 635 58 22];
428     app.CountryDropDown_2Label_2.Text = 'Country';
429
430 % Create BasicReproductionNumberLabel
431 app.BasicReproductionNumberLabel = uilabel(app.UIFigure);
432 app.BasicReproductionNumberLabel.FontSize = 16;
433 app.BasicReproductionNumberLabel.Position = [999 625 207
434     22];
435 app.BasicReproductionNumberLabel.Text = 'Basic
436     Reproduction Number';
437
438 % Create BRNSlider
439 app.BRNSlider = uislider(app.UIFigure);
440 app.BRNSlider.Limits = [0 5];
441 app.BRNSlider.ValueChangedFcn = createCallbackFcn(app, @
442     UpdateButtonPushed, true);
443 app.BRNSlider.Position = [999 607 326 3];
444
445 % Create DincSwitch
446 app.DincSwitch = uiswitch(app.UIFigure, 'slider');
447 app.DincSwitch.ValueChangedFcn = createCallbackFcn(app, @
448     UpdateButtonPushed, true);
449 app.DincSwitch.Position = [1261 536 45 20];
450
451 % Create DinfSwitch
452 app.DinfSwitch = uiswitch(app.UIFigure, 'slider');
453 app.DinfSwitch.ValueChangedFcn = createCallbackFcn(app, @
454     UpdateButtonPushed, true);
455 app.DinfSwitch.Position = [1261 445 45 20];
456
457 % Create FatalitySwitch
458 app.FatalitySwitch = uiswitch(app.UIFigure, 'slider');
459 app.FatalitySwitch.ValueChangedFcn = createCallbackFcn(
460     app, @UpdateButtonPushed, true);
461 app.FatalitySwitch.Position = [1262 353 45 20];
462
463 % Create BRNSwitch
464 app.BRNSwitch = uiswitch(app.UIFigure, 'slider');

```

```
459     app.BRNSwitch.ValueChangedFcn = createCallbackFcn(app, @
        UpdateButtonPushed, true);
460     app.BRNSwitch.Position = [1263 626 45 20];
461
462     % Show the figure after all components are created
463     app.UIFigure.Visible = 'on';
464 end
465 end
466
467 % App creation and deletion
468 methods (Access = public)
469
470     % Construct app
471     function app = SEIRD_LQ
472
473         % Create UIFigure and components
474         createComponents(app)
475
476         % Register the app with App Designer
477         registerApp(app, app.UIFigure)
478
479         if nargout == 0
480             clear app
481         end
482     end
483
484     % Code that executes before app deletion
485     function delete(app)
486
487         % Delete UIFigure when app is deleted
488         delete(app.UIFigure)
489     end
490 end
491 end
```

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