

## Lab Report - 5

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Note:

The value of  $\alpha$  used in this document is given by:  $\alpha = 1 + \text{mod}(160; 3) = 2$

### Problem 1. (First-Order Model)

The propagation mechanism of an epidemic, such as the one caused by the SARS-CoV-2 virus, can be modelled, at least in its initial phase, as a process in which each infected individual will eventually transmit the disease to an average of  $R_0$  healthy people; these newly infected patients will, in turn, infect  $R_0$  healthy individuals each, and so on, creating a pernicious positive feedback in the system. The constant  $R_0$  is called the basic reproduction number for a virus. In signal processing terms, the infection mechanism is equivalent to a first-order recursive filter. Assume that each infected person spreads the virus over a single day and then recovers and assume that an initial patient zero appears at day  $n = 0$ . The number of newly infected people per day is described by the difference equation

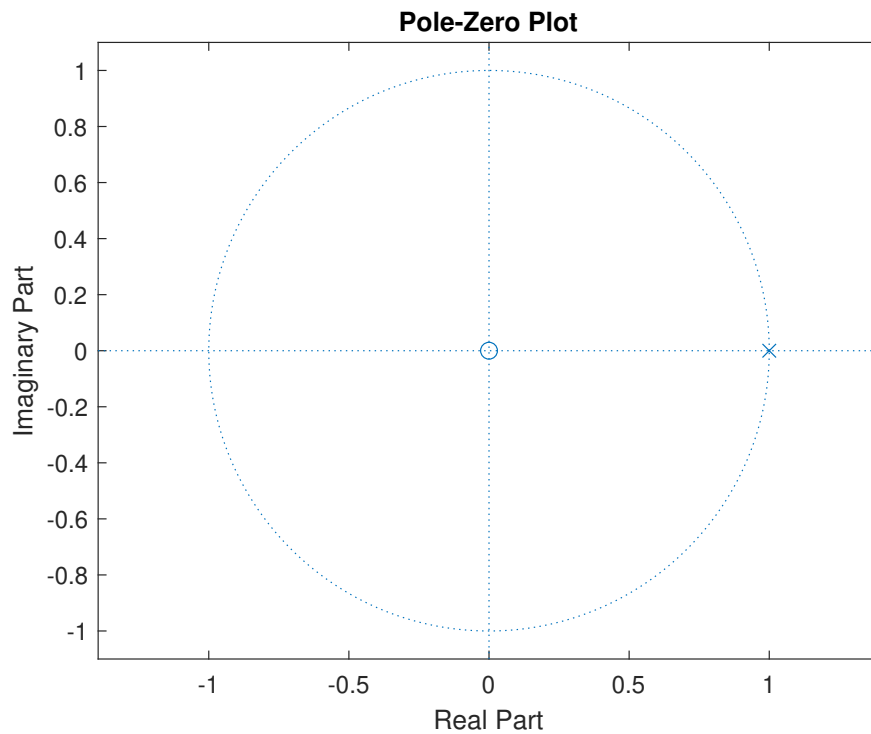
$$y[n] = \delta[n] + R_0 y[n - 1]$$

**(1. What is the transfer function  $H_1(z)$  of the above system? Plot the pole-zero plot of the system.)**

**(Solution)** The transfer function of the model is as follows,

$$H(z) = \frac{z}{z - R_0} \quad (0.1)$$

The pole-zero plot for  $R_0 = 1$  is as follows,



(2. Solve the difference equation and give the time-domain equation for the number of newly infected people. Note that this depends on the parameter  $R_0$ . Comment on the effect of the parameter  $R_0$ . Can this be inferred from the pole-zero plot?.)  
**(Solution)** After solving the difference equation we get the solution as -

$$y(n) = \delta(n) \quad (0.2)$$

When  $R_0 = 0$

$$y(n) = R_0^n \quad (0.3)$$

When  $R_0 \neq 0$

The stability of the system is influenced by  $R_0$ , which can be readily deduced from the pole-zero diagram. Since  $R_0$  functions as a pole, it must reside within the unit circle in the Pole-Zero plot for the system to maintain stability.

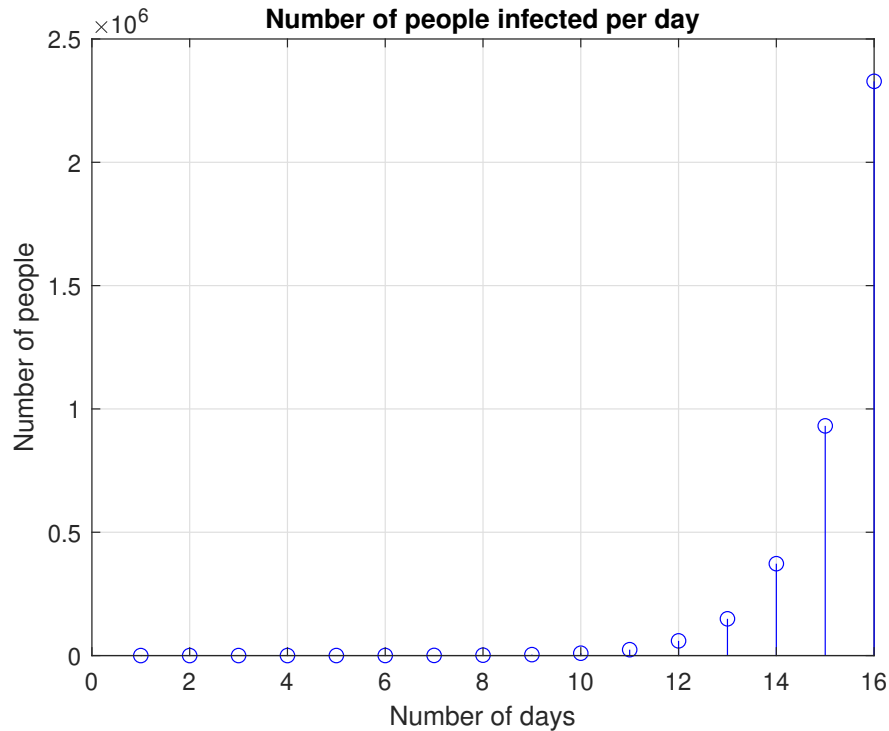
(3. Suppose  $R_0 = 2.5$ , how many days will it take to reach 1 million new daily

**infections?)** When  $R_0 = 2.5$ , the solution for this equation is,

$$y(n) = (5/2)^n \quad (0.4)$$

For the number of infections to reach 1 million,  $y(n) \geq 10^6$ .

Therefore,  $n = 16$  This is verified from the graph plotted for value  $R_0 = 2.5$ ,



**(4 :Using a similar one-point trick, and the data available from Covid-19 in India, estimate the value of  $R_0$  in the initial phase of the first wave of infections in India)**

This is the data on the newly infected individuals in India during the initial fifty days of the COVID-19 outbreak using the website <https://www.covid19india.org/>.

$Y = [2, 4, 5, 6, 10, 19, 65, 78, 114, 137, 140, 144, 173, 204, 225, 245, 254, 287, 307, 341, 368, 373, 387, 404, 423, 460, 506, 571, 596, 614, 658, 688, 755, 807, 835, 855, 1007, 1143, 1245, 1391, 1607, 1806, 1907, 2111, 2191, 2340, 2528, 2710, 2772, 2865]$

To find  $R_0(n)$ ,

$$R_0(n) = \frac{y(n) - \delta(n)}{y(n-1)} \quad (0.5)$$

The average value of  $R_0$  calculated from the above data is,

$$R_0 = 1.17$$

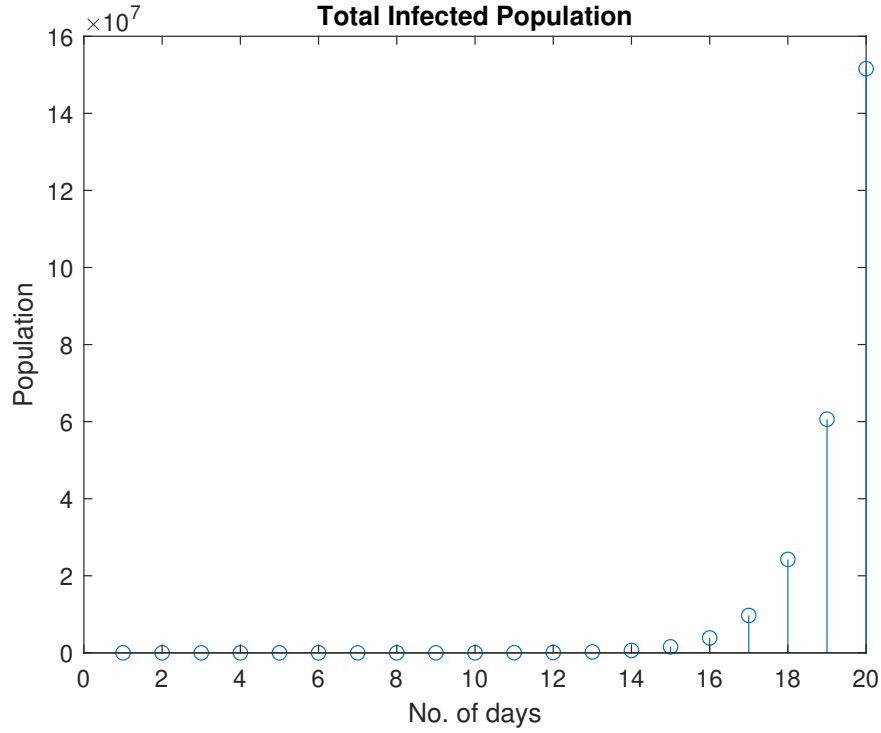
**(5. With  $R_0 = 2.5$ , plot the new daily infections for the first  $n = 20$  days. Design an integrator filter. Use this filter to obtain the total number of infections for the**

first  $n = 20$  days)

**(Solution)** Transfer function of an integrator filter is as follows,

$$H(z) = \frac{1}{1 - (1/z)} \quad (0.6)$$

By using this filter, we get the following plot for total number of people infected after  $n=20$  days,



## 2. Increasing the Complexity

The transfer function for the model is as follows,

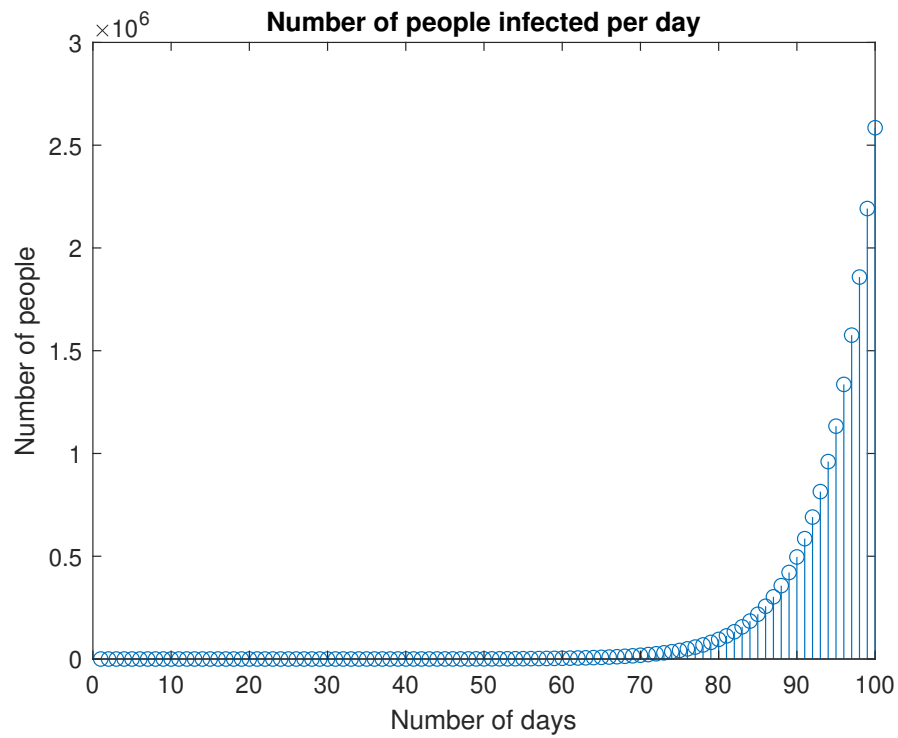
$$H(z) = \frac{1}{1 - \sum_{k=1}^M a_k z^{-k}}, M = 12 \quad (0.7)$$

$$a_k = [.1, .15, .25, .26, .34, .42, .25, .2, .15, .1, .1, .1] \quad (0.8)$$

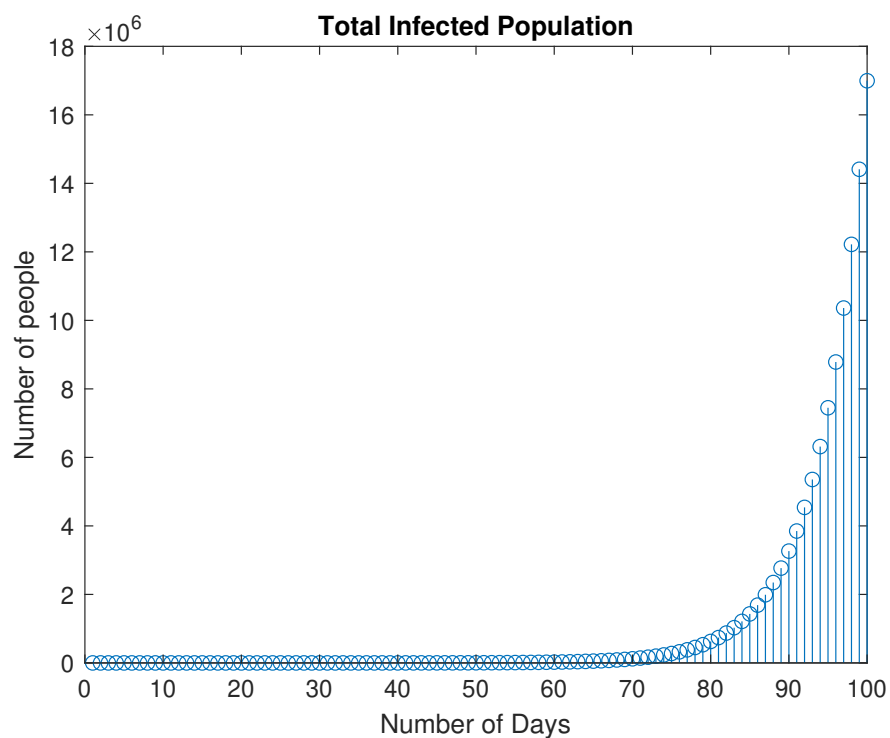
(1. Plot the new daily infections for the first  $n = 100$  days by implementing the filter given in Equation 1.8 with the Kronecker delta as the input. Use an integrator filter to obtain the total number of infections for  $n = 100$  days.)

**(Solution)**

After graphing the new daily infections for the initial  $n = 100$  days by applying the mentioned filter and using the Kronecker delta as input, the obtained results were -



After using an integrator filter, we get the total infection as the follows,



(2. Comment on the differences between the trends that are obtained with the first-order model. How many days will it take to reach 1 million new daily infections?)  
*(Solution)*

To figure out how long it would take to hit 1 million new daily infections, I looked at a graph and did some math. Both methods gave me the same answer: 95 days. On day 94, we had about 960,580 infections, and by day 95, it reached 1,132,900. So, I concluded it takes 95 days to get to 1 million new daily infections.

Comparing it to the First-Order method, this method's slope isn't as steep, showing a slower increase in infections. Also, it takes a bit longer to hit 1 million infections per day compared to the First-Order method.

**(3.Bonus: Comment on a reliable technique to estimate the coefficients  $(a_k)_{(k=1)}^M$ )**

**(Solution)** In the provided figure in the question, the values of  $(a_k)_{(k=1)}^M$  represent the number of contagions per day on a specific day, where  $k$  indicates the day number, and  $M$  equals 12.

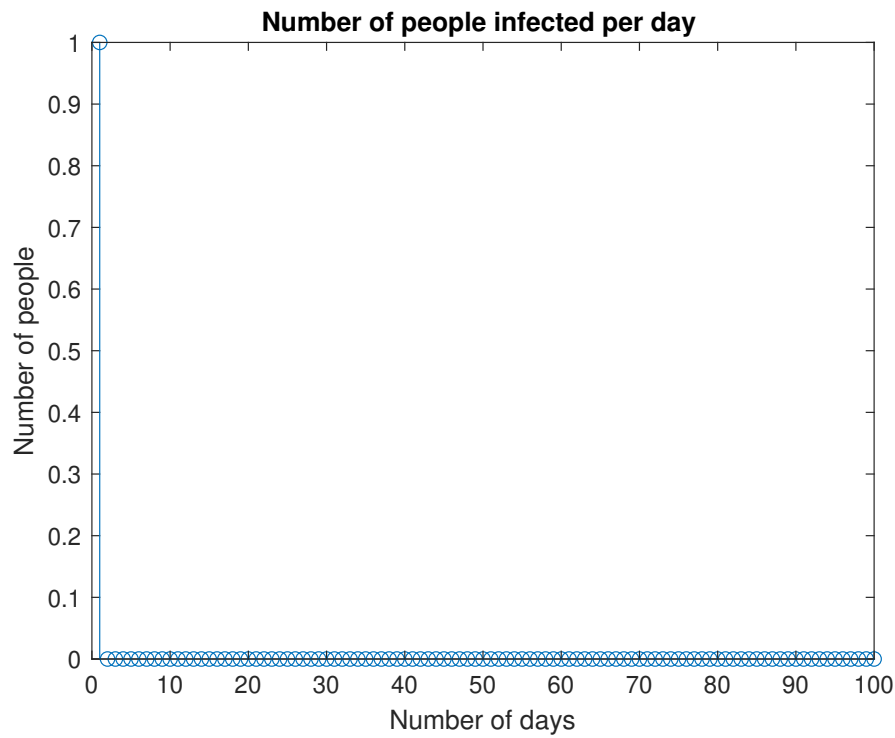
**Social Distancing** The transfer function for the model is as follows,

$$H(z) = \frac{1}{1 - (1 - \rho) \sum_{k=1}^M a_k z^{-k}}, M = 12 \quad (0.9)$$

**(1.Comment on the role of  $\rho$ . What does  $\rho = 1$  indicate?)**

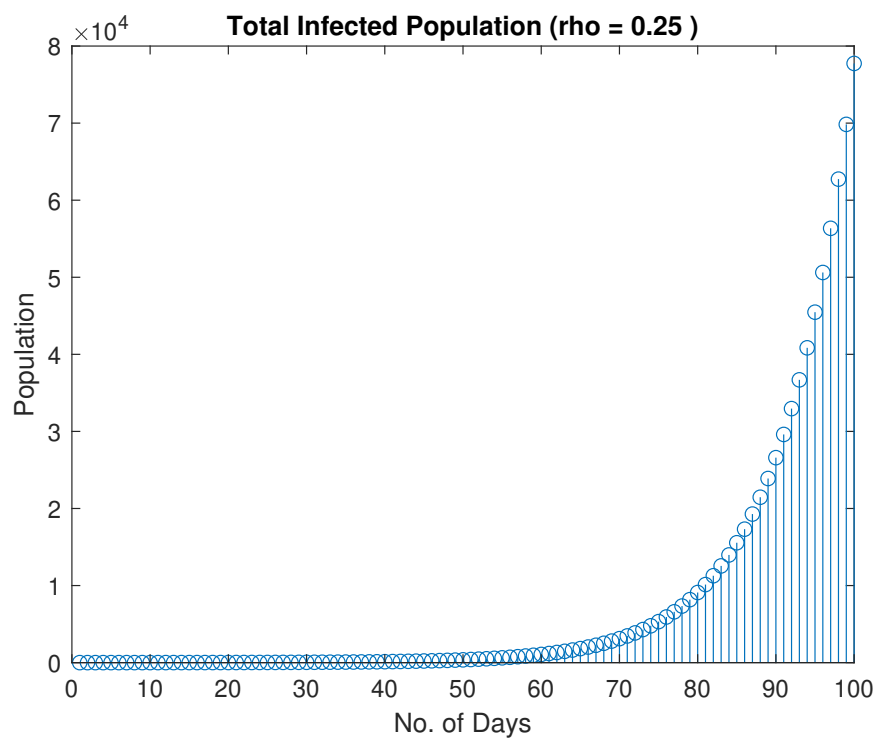
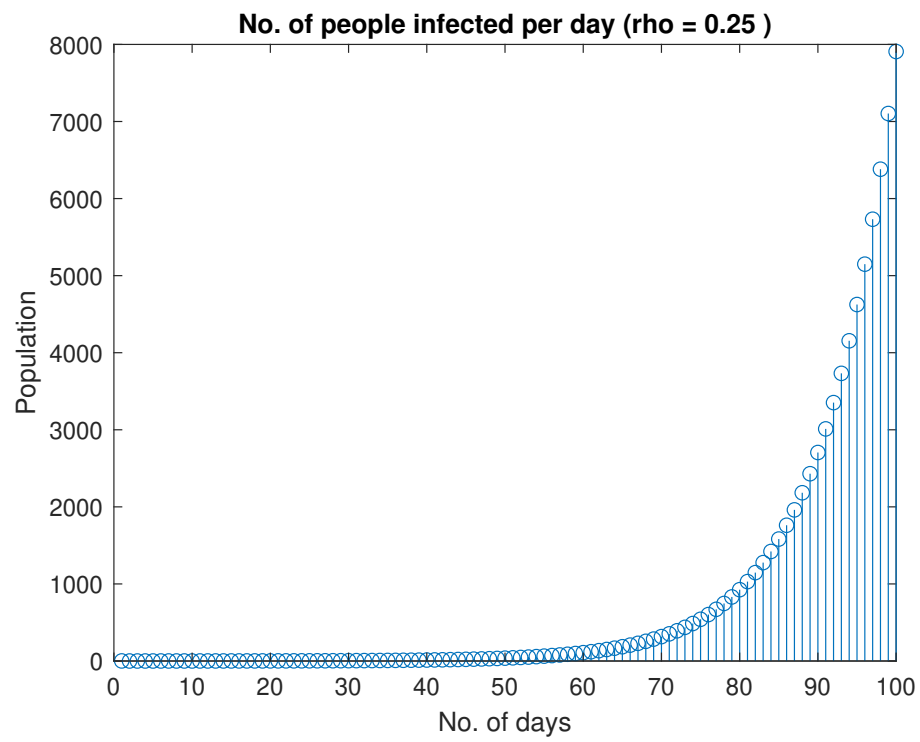
**(Solution)** Here,  $\rho$  is like a measure showing how social distancing impacts the number of people affected by the coronavirus. When  $\rho$  is high, it indicates good social distancing, and if it's low, it means poor social distancing. For instance, when  $\rho = 1$ , it signifies 100 percent effective social distancing, resulting in only one case on Day 1. This aligns with the outcome obtained from plotting the data.

Now, if we consider  $\rho = 1$ , here's what the graph looks like:

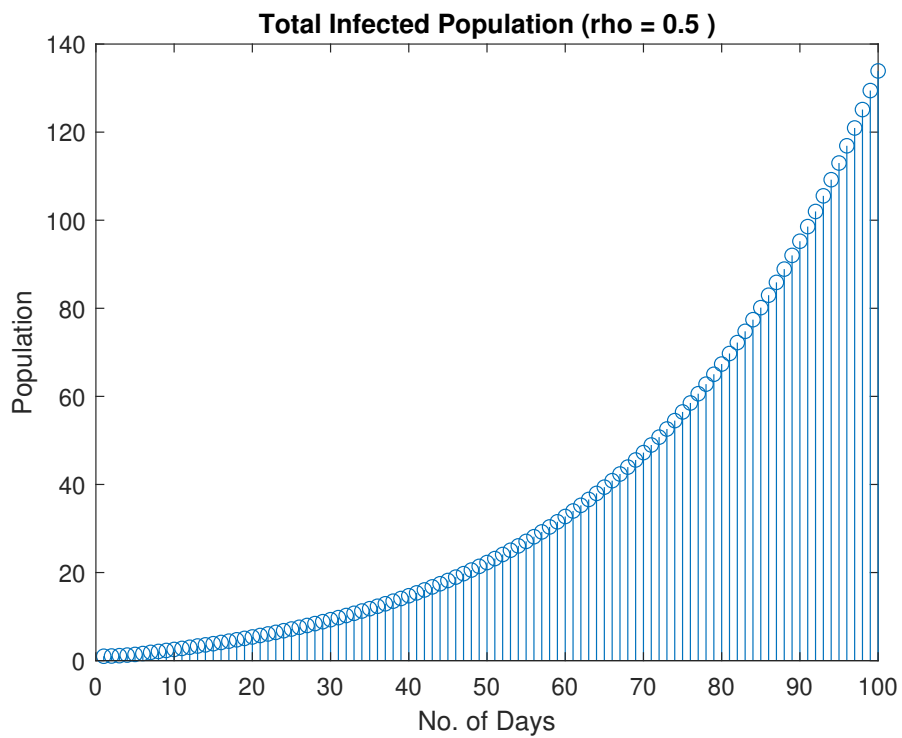
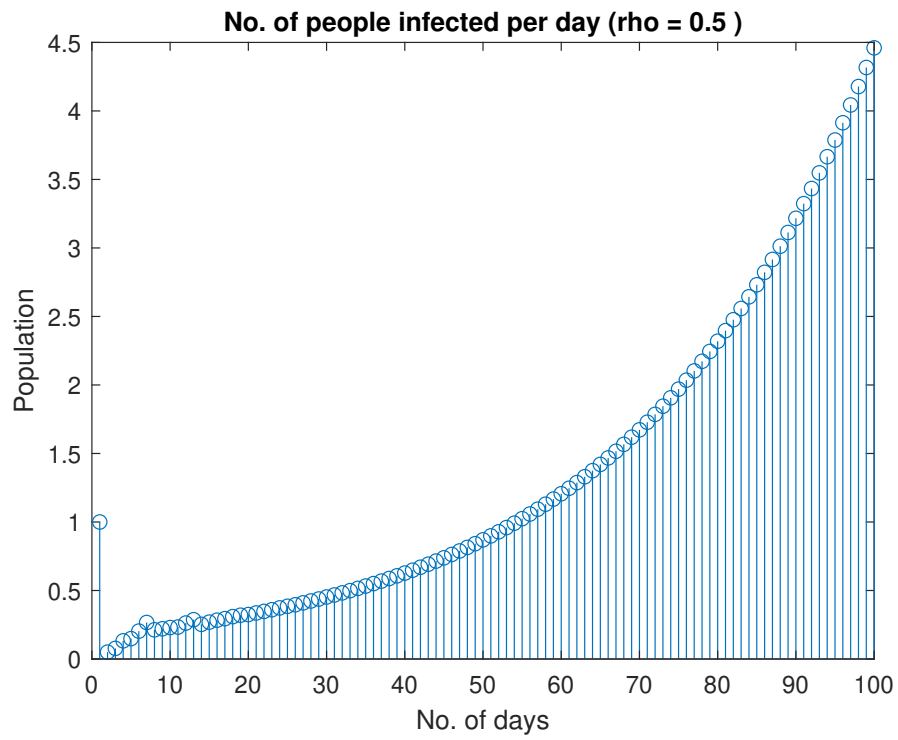


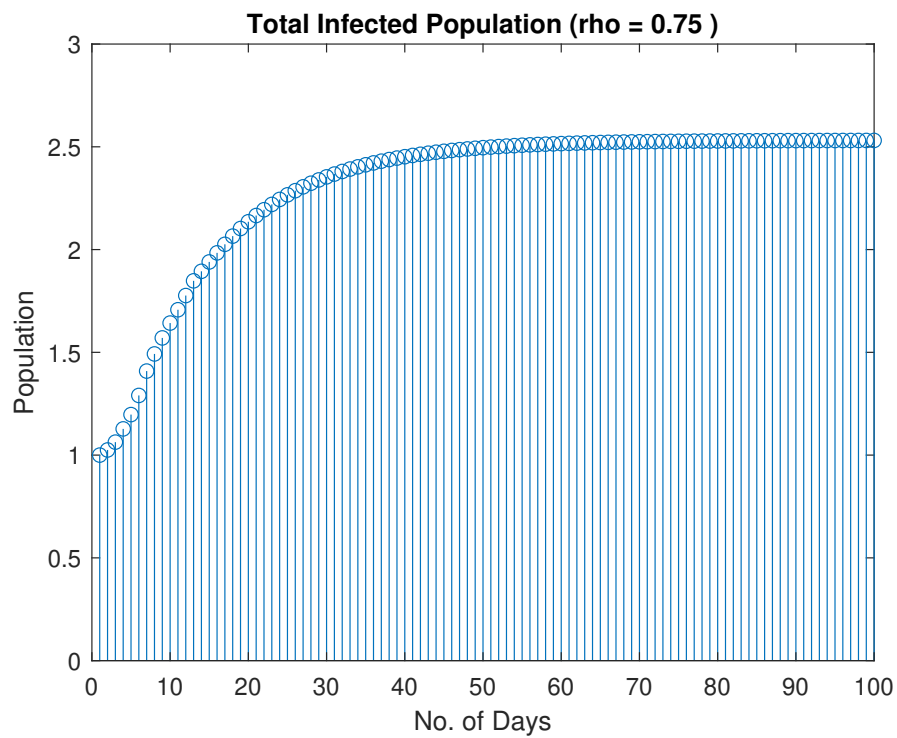
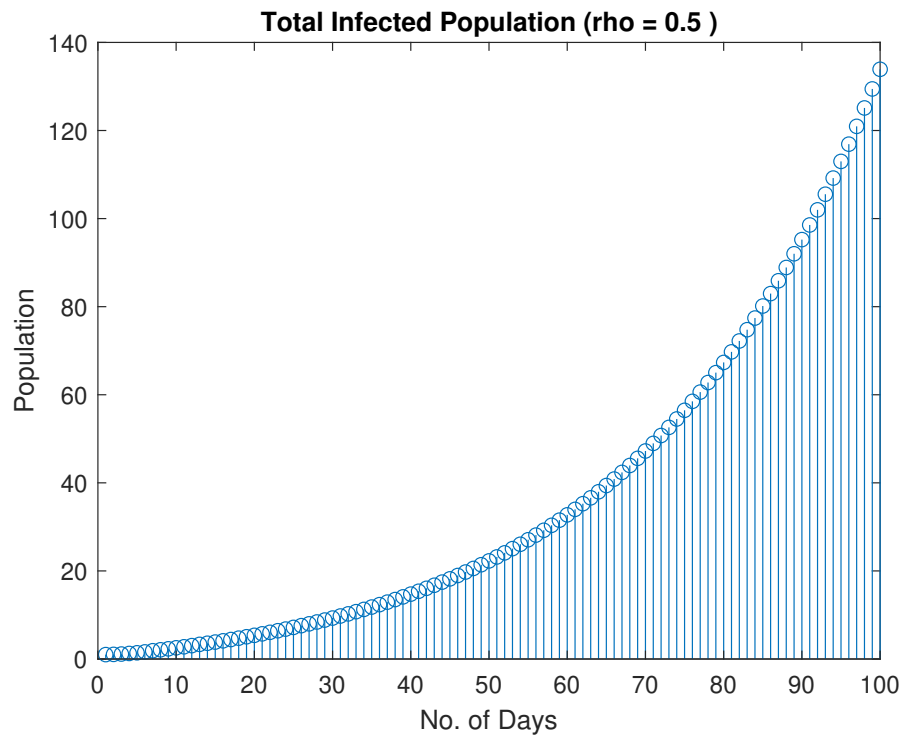
(2. Plot the new daily infections for the first  $n = 100$  days by implementing the filter given in Equation 3.1 with the Kronecker delta as the input for  $\rho = 0.25, 0.50, 0.75$  Use an integrator filter to obtain the total number of infections for  $n = 100$  days)

**(Solution)** For  $\rho = 0.25, \rho = 0.5$  and  $\rho = 0.75$ , we get the following plot for number of people infected per day and cumulative infections









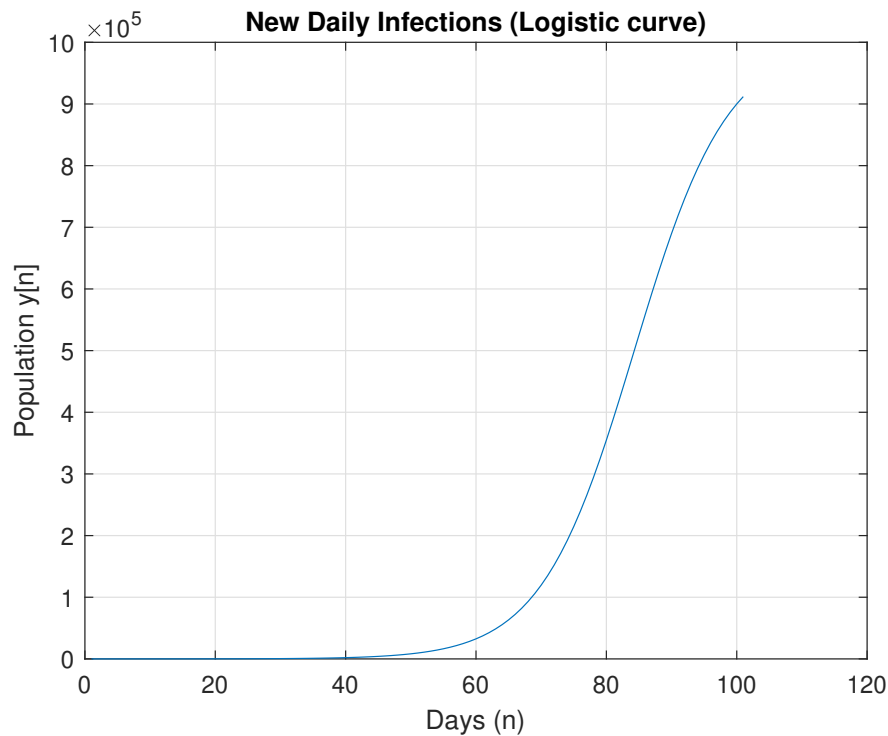
The total number of infected people for  $\rho = 0.25, \rho = 0.5$  and  $\rho = 0.75$  after of 100 days is  $7.7745e+04$ , 133.8987 and 2.5310 respectively.

**Saturation and Towards Normality** The cumulative number of infections,  $x(n)$ ,

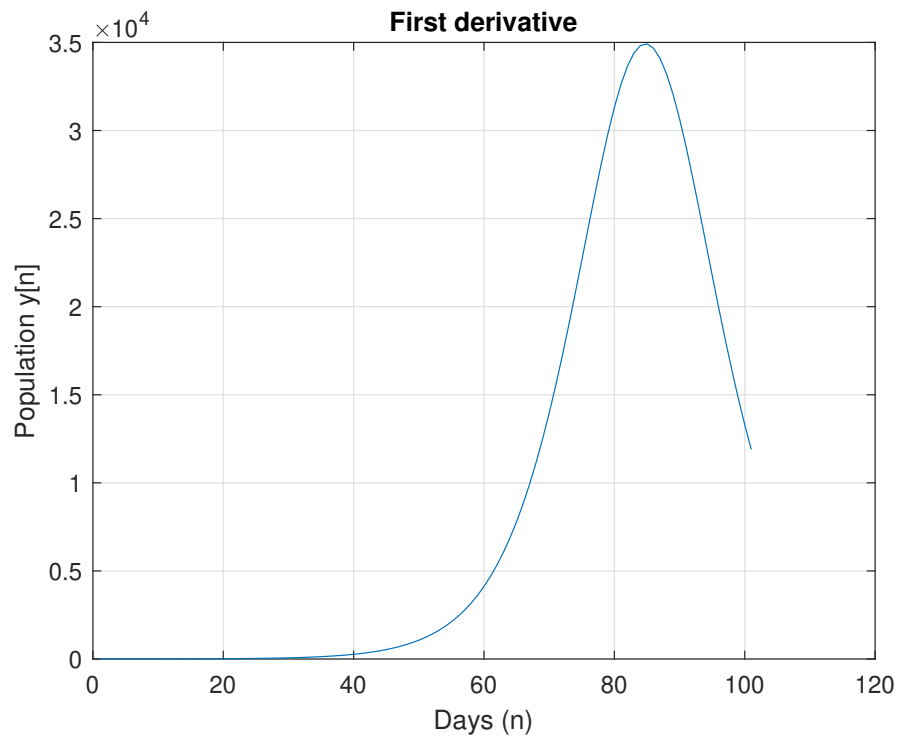
$$x(n) = \frac{K}{1 + [K(R_0 - 1) + R_0]R_0^{-(n+1)}} - \frac{1}{R_0 - 1} \quad (0.10)$$

(1. With  $R_0 = 1.15$  and  $K = 106$ , plot the total number of infections in the first-order model versus the logistic evolution for a population of one million for  $n = 100$  days. Does the first order model follow the logistic hypothesis?)

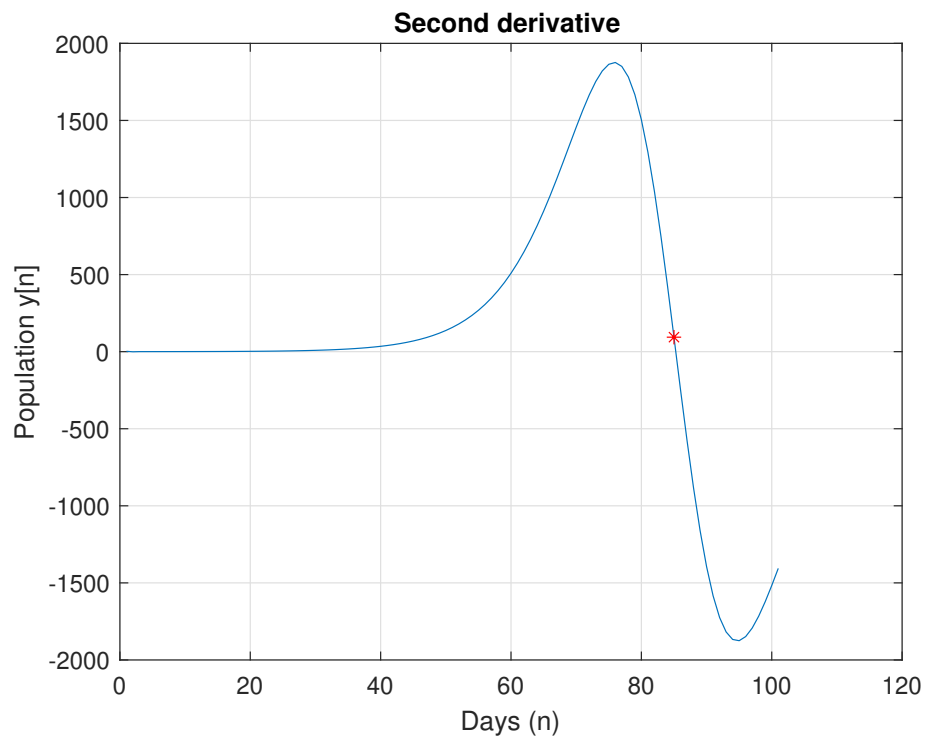
**(Solution)** Applying logistic evolution to the parameters mentioned earlier, the outcomes are as follows:



(2. Find the point of inflection using the global maximum of the first derivative and the zero-crossing of the second derivative) The first order derivative is plotted as follows,



The second order derivative when plotted is as follows



By analyzing the results obtained from these calculations, I determined that the global maximum is  $3.4918 \times 10^4$ , and it occurs at  $n = 85$ . Additionally, the zero-crossing point is also equal to this value.

## **A Code Repositories**

<https://github.com/VenugopalRadhakrishnan/DSP-Lab>.