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On the necessity of proper quarantine without lock down for 2019-nCoV in the absence of vaccine

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ABSTRACT

Presently the world is passing through a critical phase due to the prevalence of the Novel Corona virus, 2019-nCoV or COVID-19, which has been declared a pandemic by WHO. The virus transmits via droplets of saliva or discharge from the nose when an infected person coughs or sneezes. Due to the absence of vaccine, to prevent the disease, social distancing and proper quarantine of infected populations are needed. Non-resident citizens coming from several countries need to be quarantined for 14 days prior to their entrance. The same is to be applied for inter-state movements within a country. The purpose of this article is to propose mathematical models, based on quarantine with no lock down, that describe the dynamics of transmission and spread of the disease thereby proposing an effective preventive measure in the absence of vaccine.

Introduction

Viruses are not new on Earth. Since the evolution of first living cells they existed as most numerous biological entity in almost ecosystem having infected all life forms besides human beings. The history of human race was disrupted several times by terrible impacts of viral infections. During last one year the entire world is facing challenges posed by novel corona virus 2019, commonly known as COVID-19 and the battle is not over yet. A new strain of COVID-19 has become very dangerous through its exceptional infectious qualities [3]. Reportedly, China country office of WHO, for the first time on December 31st, 2019, come to know about cases of pneumonia of unknown aetiology which was detected in Wuhan city, Hubei Province of China. Up to January 3rd, 2020, a total of 44 cases of pneumonia with unknown cases were reported. Subsequently on January 7th, 2020, it was identified that the pathogenic agent behind the cases of pneumonia was corona virus of a new strain. On January 13th, 2020, the first imported case of novel corona virus (2019-nCoV) was reported by the Ministry of Public Health, Thailand [9,14,21]. With the passage of time the whole world have been hit by the rapid spread of COVID-19. As reported by WHO [22] a total of 8385440 confirmed cases of COVID-19 positive and total of 450686 deaths have taken place by 19th June, 2020. The following table 1 shows a list of top 12 countries with respect to transmission of COVID-19.

Way back in 1918–19 the world had seen a pandemic of similar extent when the human civilization was attacked by H1N1 influenza. Since there was no vaccine available, different governments took different measures to contain the transmission of the virus in their countries. These pharmaceutical interventions (NPIs) included closing schools, churches, bars and other places of social gathering. The places where these interventions were implemented early were successful in reducing the number of cases while a lower mortality rate were experienced in place where interventions remained in place. But with the lifting of controls the transmission renounced once again [10].

Now we, the whole human civilization over the globe, experience a very similar kind of situation in combating COVID-19 with so called non-pharmaceutical interventions which aims at reducing contact rates in the population to arrest the transmission of the virus [11,24]. Such measures include the reduction of social contact in work places, schools

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and other public domains. For quantitative estimates of the impact of these measures in reducing morbidity, infection rate, excess mortality, proper mathematical model of virus transmission are required and these can contribute significantly in the public health planning.

To combat against COVID-19 the measures that included closing of international borders, shutting school, colleges and workplaces containing large gatherings (“Great lock down”, the phrase coined by the IMF), has invited huge impact on global economy which resulted in people to lose their jobs and businesses being disrupted. It’s truly a global crisis.

At this point of time mathematics has emerged to be an invaluable weapon to combat against COVID-19. Mathematical models allow public health officials to conduct virtual experiments thereby evaluating the efficacious of control strategies. By studying the transmission dynamics [15,19] a systematic quarantine strategy can be taken up.

Epidemic outbreaks evolve in geographic territories with considerable variability in spatial footings. This spatial inconstancy is significant in understanding the dominance of public health policies and interventions in regulating these epidemics. A major trouble in provoking models to narrate spatial inconstancy in epidemics is accounting for the movement of people in spatial contexts. Multiple approaches to generate such statement have been exhibited, including individual based models, network models, stochastic models, and partial differential equations models. For the geographic spread [4] and control of COVID-19, we consider two epidemic models of partial differential equations [8,17] corresponding to the epidemic 2019-nCoV, one is due to proper quarantine of infected population and other is due to no quarantine. To include and quantify spatial effect we consider these models as diffusion models [1,5,15] for the geographic spread of the epidemic. The problem we interested in including a number of E class and I class in a uniform population together with homogeneous initial susceptible volume \( S_0 \) and calculating geotemporal propagation of the malady. Nomenclatures used in the epidemic models of the pandemic disease COVID-19 is given in the Table 2.

Basic terminologies:

**Susceptible population** \( S(x, t) \): A susceptible population \( S(x, t) \) in an epidemiology is a population at \( x \in \Omega \subset \mathbb{R}^d \) in time \( t \), in which an infectious disease is not present but each individuals of this population is at risk of gaining infection by the disease in forward time.

**Latent infected population** \( E(x, t) \): A latent infected population \( E(x, t) \) in an epidemiology is a population at \( x \in \Omega \subset \mathbb{R}^d \) in time \( t \), in which an infectious disease (COVID-19 in our case) is present without symptoms. In forward time they may belong to the infected population with symptoms or may become susceptible. They have ability to transmit the disease.

**Infected population** \( I(x, t) \): An infected population \( I(x, t) \) in an epidemiology is a population at \( x \in \Omega \subset \mathbb{R}^d \) in time \( t \), in which an infectious disease (COVID-19 in our case) is present with symptoms. In forward time they have full ability to transmit the disease through migrant population transmission or by local individual transmission or community transmission.

**Removed population** \( R(x, t) \): A removed population \( R(x, t) \) in an epidemiology is a population at \( x \in \Omega \subset \mathbb{R}^d \) in time \( t \), whose members are recovered from the infection of the infectious disease or died due to the infectious disease (COVID-19 in our case).

**Basic reproduction rate** \( R_0 \): For an infectious disease (COVID-19 in our case) the basic reproductive number is the number of secondary infections delivered by a single infected individual in whole susceptible population. This quantity indicates the initial growth rate for the infected class and the potential for a large-scale epidemic. It is one of the touchstones of epidemiology.

**Herd immunity**: The immunization of an individual not only protects that individual but also indirectly protects others against the possibility of disease transmission from the immunized individual. If a sufficient fraction of a population is immunized, then an epidemic may be averted altogether. The protection of an entire population via the immunity of a fraction of the population is called herd immunity.

**Epidemic models**

In this section we develop two non-linear epidemic models [12], one is due to proper quarantine of infected population and other is due to no quarantine.

**Hypothesis**

For the non-linear epidemic model [12] the whole population \( N \) is considered to be constant. Due to diffusion, we consider the spread of the infection within the population as a function of time and space both. Let \( \Omega \subset \mathbb{R}^d \), be a bounded domain. Suppose the disease is such that the population can be separated into four distinct classes: the susceptible population, \( S(x, t) \), who can grab the disease; the infected population, \( I(x, t) \), who have the disease and can emit it; a class in which the disease is latent, \( E(x, t) \), who also can transmit the disease; and the removed population, \( R(x, t) \), namely, those who are recovered, immune or isolated until recovered or dies out; at the location \( x \in \Omega \) and at time \( t \). Then

\[
S + E + I + R = N, \tag{1}
\]

Also suppose that \( \lambda_1 \) is the rate at which the interaction between the \( S \) class and the \( E \) class occur, \( \lambda_2 \) is the rate at which the interaction between the \( S \) class and the \( I \) class occur, \( \lambda_3 \) is the rate for which the interaction between \( S \) class and \( E \) class belong to the \( E \) class, \( \lambda_4 \) is the rate at which the interaction between the \( I \) class and the \( E \) class occur, \( \lambda_5 \) be

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### Table 1

| Country Name          | Total confirmed cases | Total confirmed new cases | Total deaths | Total new deaths |
|-----------------------|-----------------------|---------------------------|--------------|------------------|
| United States of America | 2149166              | 23139                     | 117472       | 770              |
| Brazil                | 955377               | 32188                     | 46510        | 1269             |
| Russian Federation    | 569063               | 7972                      | 7841         | 181              |
| India                 | 380532               | 13586                     | 12573        | 336              |
| The United Kingdom    | 300473               | 1218                      | 42288        | 135              |
| Spain                 | 245268               | 585                       | 27136        | 0                |
| Peru                  | 240908               | 3752                      | 7257         | 201              |
| Italy                 | 238159               | 331                       | 34514        | 66               |
| Chile                 | 225103               | 4475                      | 3841         | 226              |
| Iran                  | 197647               | 2596                      | 9272         | 87               |
| Germany               | 187764               | 0                         | 8856         | 0                |
| Turkey                | 184031               | 1304                      | 4882         | 21               |

### Table 2

| Symbols | Descriptions |
|---------|--------------|
| \( S(x, t) \) | Susceptible population |
| \( E(x, t) \) | A population in which the disease is latent |
| \( I(x, t) \) | Infected population |
| \( R(x, t) \) | Removed population |
| \( \lambda_1 \) | Rate of contact between \( S \) and \( E \) class |
| \( \lambda_2 \) | Rate of contact between \( S \) and \( I \) class |
| \( \lambda_3 \) | A fraction of \( \lambda_1 \) that belong to \( E \) class |
| \( \lambda_4 \) | Rate of contact between \( E \) and \( I \) class |
| \( \lambda_5 \) | A fraction of \( \lambda_1 \) that belong to \( I \) class |
| \( r \) | Rate at which both \( E \) and \( I \) classes release their individuals from their respective classes |
| \( D \) | Diffusion coefficient |
the remaining fraction of $\lambda_1$ for which the interaction between $S$ class and $E$ class belong to the $I$ class, that is $\lambda_1 + \lambda_3 = \lambda_1$. $\gamma$ be the rate at which both $E$ class and $I$ class release the individuals from their respective classes (in this case we choose the same rate for both $E$ and $I$ class) and $D$ be the diffusion coefficient for all the population.

Model for no quarantine case

If there are no quarantine for infected populations, no social distancing and not considering for citizens of several countries that needed to be quarantined for 14 days earlier to entering their topical countries or topical state who comes outside their own country or own state respectively, the population of all $S$, $E$ and $I$ classes are interact with each other. For which as time goes, the total populations has high probability to become latent infected as well as infected and the corresponding model is followed by the flow chart Fig. 1, given by:

$$\frac{\partial S}{\partial t} = - \left( \lambda_1 SE + \lambda_2 SI \right) + D \frac{\partial^2 S}{\partial x^2} \tag{2}$$

$$\frac{\partial E}{\partial t} = \lambda_2 SE - \lambda_1 IE - \gamma E + D \frac{\partial^2 E}{\partial x^2} \tag{3}$$

$$\frac{\partial I}{\partial t} = \lambda_1 SI + \lambda_1 IE + \lambda_2 SE - \gamma I + D \frac{\partial^2 I}{\partial x^2} \tag{4}$$

$$\frac{\partial R}{\partial t} = \gamma \left( I + E \right) \tag{5}$$

It is easy to check that the system (2)–(4) has infinitely many solutions and are non-negative and uniformly bounded. First we represent a non-negativity lemma, which can be found in any standard book.

**Lemma 1.** Let $u \in C^{2,1}(\Omega \times [0, t]) \cap C^{2,1}(\Omega \times [0, t])$ with

$$u_t - D \Delta u = a(x, t)u; \quad x \in \Omega, 0 \leq t \leq \tau$$

$$\frac{\partial u}{\partial t} \geq 0; \quad x \in \partial \Omega, 0 \leq t \leq \tau$$

$$u(x, 0) \geq 0; \quad x \in \Omega$$

where $a(x, t) \in C^{2,1}(\Omega \times [0, t])$ and $\frac{\partial u}{\partial n}$ is the outward normal derivative of $u$ at $\partial \Omega$. Then $u(x, t) \geq 0$ on $\Omega \times [0, \tau]$. Moreover $u(x, t) > 0$ or $u(x, t) = 0$ in $\Omega \times [0, \tau]$.

**Proposition 1.** For a non-negative initial condition, the system (2)–(4) possesses a non-negative solution.

![Flow of COVID-19 transmission for 'no quarantine case'.](image)

**Proof.** Let $(S, E, I)$ be a solution of the system (2)–(4) in $\Omega \times [0, T_{max}]$. Then for every $t \in (0, T_{max})$ and from the system (2)–(4)

$$S_t - D \Delta S = - (\lambda_1 E + \lambda_2 I) S$$

$$E_t - D \Delta E = (\lambda_1 S - \lambda_1 I - \gamma) E$$

$$I_t - D \Delta I = \lambda_2 S + \lambda_2 E + \gamma I - \gamma I \frac{\partial (\lambda_2 S + \lambda_2 E - \gamma I)}{\partial x}$$

where $\gamma \in \Omega$, $0 < t < \tau$. Also $\lambda_1 E + \lambda_2 I, \lambda_2 S - \lambda_2 E - \gamma$ and $\lambda_2 S + \lambda_2 E - \gamma$ are bounded in $\Omega \times [0, \tau]$.

Thus from Lemma 1, as $t$ is arbitrary in $(0, T_{max})$, we must have $S(x, t) \geq 0, E(x, t) \geq 0$ and $I(x, t) \geq 0$ in $\Omega \times [0, \tau]$. Hence the proof. $\square$

**Proposition 2.** For a non-trivial and non-negative initial value let $(S, E, I) \in \left[ C^{2,1}(\Omega \times [0, T_{max}) \cap C^{2,1}(\Omega \times [0, T_{max}) \right]$ be a solution of the system (2)–(4).

Then $T_{max} = \infty$ and

$$0 < S(x, t) + E(x, t) + I(x, t) \leq \max \{|S_0(x) + E_0(x) + I_0(x)|, |S_0(x) + E_0(x) + I_0(x)|, |S_0(x) + E_0(x) + I_0(x)| + |S_0(x) + E_0(x) + I_0(x)|\}$$

where $\gamma$ denotes the differentiation with respect to $x$ and $N$ is given by the relation (1).

**Proof.** First we show that all of $S(x, t), E(x, t)$ and $I(x, t)$ are bounded in $\Omega \times (0, T_{max})$. Let $U(x, t) = S(x, t) + E(x, t) + I(x, t)$, As,

$$0 < U(x, 0) \leq \max \{|U_0(x)|, |U_0(x)| + |U_0(x)| + |U_0(x)|\}$$

and

$$\gamma (S + E + I) - D \Delta (S + E + I) = - \gamma (S + E + I) \leq N \gamma (S + E + I)$$

we must have $0 < U(x, 0) \leq \gamma (S + E + I)$ in $\Omega \times (0, T_{max})$, where

$$w(t) = \left[ N + (\{U_0(x)\} + |U_0(x)| + \{U_0(x)\} + |U_0(x)|) \right] e^{-\gamma t}$$

for $t \in [0, \infty)$ is the solution of the ODE

$\frac{dw(t)}{dt} = N - \gamma w(t)$

$w(0) = \max \{|U_0(x)|, |U_0(x)| + |U_0(x)|\}$

Thus we have $\max \{U_0(x), U_0(x), U_0(x)\} e^{-\gamma t}$ for $t \in [0, \infty)$.

Therefore $0 < S(x, 0) + E(x, 0) + I(x, 0) \leq \gamma (S + E + I)$.

$s_{max} \left\{ |S_0(x) + E_0(x) + I_0(x)|, |S_0(x) + E_0(x) + I_0(x)|, |S_0(x) + E_0(x) + I_0(x)| \right\} + |S_0(x) + E_0(x) + I_0(x)| + |S_0(x) + E_0(x) + I_0(x)| + |S_0(x) + E_0(x) + I_0(x)|$,

Hence the proof.

**R_0 and herd immunity for ‘no quarantine case’:** For the reproductive rate of COVID-19 [23,25], in the model (2)–(5), considering initially $E_0 = I_0$ the per capita increase of $E$ and $I$ class is given by

$$\frac{1}{E + I} \frac{d(E + I)}{dt} = \frac{(\lambda_1 + \lambda_3) S_0}{E + I} - \gamma$$

Which gives us the basic reproductive rate

$$R_0 = \frac{(\lambda_1 + \lambda_3) S_0}{\gamma}$$

If $R_0 > 1$, then every infected member of the population will emit the disease to leastwise one other member during the infectious epoch, and the model argue that the disease will propagate within the population. If not, then the disease is desired to fall through before overreaching a substantive fraction of total population. Therefore $R_0 = 1$ is a critical epidemiological grade. In other terms, pathogens with elevated equilibrium of contagion and subordinate rescue and mortality rates will gesture an ideal threat. The reciprocal of the removal rate is the average time interval during which an individual from both $E$ and $I$ class remain contagious, given by $\frac{1}{\gamma}$.

The expression for $R_0$ can be rearranged to find the minimum size of a susceptible population, necessary for an epidemic to occur. Assuming that $R_0 = 1$, the threshold condition is given by

$$S_0 = \frac{(\lambda_1 + \lambda_3)}{\gamma}.$$
A pathogen will go extinct if the size of the susceptible population is less than this threshold ($S < S_*$). If the population size is above this threshold, then we can rewrite the basic reproductive rate as

$$ R_0 = \frac{S_0}{S_*} $$

Immunization reduces the size of the $S$ class and thus leads to a smaller basic reproductive rate of the pathogen. In particular, immunizing a fraction $p$ of a population reduces $R_0$ to

$$ R'_0 = \frac{(1 - p)S_0}{S_*} \left(1 - \frac{R_0}{R_0}ight). $$

Immunization will successfully eradicate the disease if it causes the basic reproductive rate to drop below one. Thus the critical immunization rate $p_c$ is

$$ p_c = 1 - \frac{1}{R_0}. $$

Expansion of this model has been utilized to anticipate the necessity of minimum coverage to urge some other tangible diseases to mitigation. For example, measles and whooping cough two of the most contagious diseases are thought to require 90–95% coverage, chicken pox and mumps 85–90% coverage, polo and scarlet fever 82–97% coverage, and smallpox 70–80% coverage [2].

Re-parametrisation

For non-dimensionalisation, re-scaling dependent variables $S, E, I, R$ by $S_0$, the initial susceptible population and independent variables as $x, t$ by $x_0$ and $t_0$ respectively, let $S = S_0S, E = S_0E, I = S_0I, R = S_0R, t = t_0t, $ where $S_0, t_0$ and $x_0$ are the characteristic units used to scale the above variables. Putting these in the system (2)–(5) we get

$$ \frac{dS}{dt} = -\left(\lambda t_S S E + \lambda i s S S I\right) + D_S \frac{\partial^2 S}{\partial x^2}, $$

$$ \frac{dE}{dt} = \lambda t_S S S E - \rho S I E + D_E \frac{\partial^2 E}{\partial x^2}, $$

$$ \frac{dI}{dt} = \lambda i S S S I + \lambda i S I S E + \lambda i S S S S E - \rho S I E + D_E \frac{\partial^2 I}{\partial x^2}, $$

$$ \frac{dR}{dt} = \rho S I + D_E \frac{\partial^2 R}{\partial x^2}. $$

that is,

$$ \frac{dS}{dt} = -\left(\lambda t_S S E + \lambda i S S I\right) + D_S \frac{\partial^2 S}{\partial x^2}, $$

$$ \frac{dE}{dt} = \lambda t_S S S E - \rho S I E + D_E \frac{\partial^2 E}{\partial x^2}, $$

$$ \frac{dI}{dt} = \lambda i S S S I + \lambda i S I S E + \lambda i S S S S E - \rho S I E + D_E \frac{\partial^2 I}{\partial x^2}, $$

$$ \frac{dR}{dt} = \rho S I + D_E \frac{\partial^2 R}{\partial x^2}. $$

With $S'(x, 0) = 1, E'(x, 0) = \frac{E_0}{I_0} I'(x, 0) = \frac{R'}{E_0}, R'(x, 0) = 0$ as initial conditions, where $E_0$ and $I_0$ are respectively the initial population of $E$ and $I$ class.

Letting $r_0 = 1$ and $D \frac{t_0}{x_0} = 1$, we get $t_0 = \frac{1}{t_0}$ and $x_0 = \left(\frac{L}{t_0}\right)^2$. Thus we end up with the scaled model

$$ \frac{dS'}{dt'} = -a S' E' - \left(R_0 - a\right) S' I' + \frac{\partial^2 S'}{\partial x'^2}, $$

$$ \frac{dE'}{dt'} = b S' E' - \mu I' E' - \frac{\partial^2 E'}{\partial x'^2}, $$

$$ \frac{dI'}{dt'} = \left(R_0 - a\right) S' I' + \mu I' E' + \left(a - b\right) S' E' - \frac{\partial^2 I'}{\partial x'^2}. $$

$$ \frac{dR'}{dt'} = I' + E'. $$

with $S'(x, 0) = 1, E'(x, 0) = \frac{E_0}{I_0}, I'(x, 0) = \frac{R'}{E_0}, R'(x, 0) = 0$ and four dimensionless numbers $a, b, \mu$ and $R_0$ as $a = \frac{t_S S_0}{t_0}, b = \frac{t_i S_0}{t_0}, \mu = \frac{t_S S_0}{t_0}$ and $R_0 = \frac{\lambda t_S S_0 S_0}{\rho t_0}$ respectively.

The parameters $\lambda_1, \ldots, \lambda_5, \gamma$ and $D$ in the dimensional model have been reduced to four dimensionless quantities $R_0, a, b$ and $\mu$.

Method of solution

In this model we investigate the local spread of an epidemic wave [13] of infection into a uniform susceptible population. We want to designate conditions for the existence of such travelling wave, its speed of propagation and, when it exists.

Looking for the one dimensional travelling wave solutions, let

$$ S'(x', t') = S' \left(x', t'\right), E'(x', t') = E' \left(x', t'\right), I'(x', t') = I' \left(x', t'\right), R' \left(x', t'\right) = R' \left(x', t'\right), $$

where the wave speed $c$, have to determine. The above consideration will give us a travelling wave of constant shape in the direction of positive $x'$-axis. Substituting the above consideration into the system (6)–(9) we get the system of equations as

$$ -c \frac{dS'}{dx'} = -a S' E' - \left(R_0 - a\right) S' I', $$

$$ -c \frac{dE'}{dx'} = b S' E' - \mu I' E' - E', $$

$$ -c \frac{dI'}{dx'} = \left(R_0 - a\right) S' I' + \mu I' E' + \left(a - b\right) S' E' - I', $$

$$ -c \frac{dR'}{dx'} = I' + E'. $$

Which can be represented as

$$ S'' + cS' - a S' E' - \left(R_0 - a\right) S' I' = 0, $$

$$ E'' + cE' + b S' E' - \mu I' E' - E' = 0, $$

$$ I'' + cI' + \left(R_0 - a\right) S' I' + \mu I' E' + \left(a - b\right) S' E' - I' = 0, $$

$$ cR' + \left(I' + E'\right) = 0. $$

(13)

where prime represents differentiation w.r.t. $z$. The above system consists of finding range of parameters considered above, for which there exists a solution with positive wave speed $c$ and non-negative $S', I'$ such that

$$ E'(\infty) = 0, \quad S'(\infty) = 0, \quad S'(-\infty) = 1, \quad S'(\infty) = 0, $$

$$ 0 < I'(\infty) < 1. $$

The conditions on $E'$ and $I'$ imply a pulse wave of infective population which propagates into the uninfected population. As time goes,
the equations (11) and (12) changes by linearising, with \( S' - E', E' - I' \) and \( I' \to 1 \) to get
\[
S'' + cS'' - R_0S' \approx 0, \\
E'' + cE'' + (b - \mu - 1)E' \approx 0, \\
I'' + cI'' + (R_0 + \mu - b - 1)I' \approx 0.
\]

(14) \( \quad \) \( \quad \) \( \quad \) \( \quad \)

(15) \( \quad \) \( \quad \) \( \quad \) \( \quad \)

(16) \( \quad \) \( \quad \) \( \quad \) \( \quad \)

Phase plane analysis: From equation (15) a typical wave front solution is where \( E' \) towards one side, say, as \( z \to -\infty \), is at one steady state and as \( z \to \infty \) it is at the other. In that case we have to determine the value or values of \( c \), for which the equation (15) has a non-negative solution \( E' \) which satisfies.
\[
\lim_{z \to -\infty} E'(z) = 1, \lim_{z \to \infty} E'(z) = 0.
\]

In \((E',U)\) phase plane.
\( E' = U, U = -cU - (b - \mu - 1)E' \) we have the phase plane trajectories as the solution of
\[
\frac{du}{dz} = \frac{-cU - (b - \mu - 1)E'}{U}
\]

which has one singular point for \((E', U)\) as \((0, 0)\). Corresponding to this singular point we define the matrix
\[
A = \begin{pmatrix}
-c & -(b - \mu - 1) \\
1 & 0
\end{pmatrix}
\]

whose eigenvalues are
\[
\frac{1}{2} \left[-c \pm \sqrt{c^2 - 4(b - \mu - 1)}\right]
\]

\( \Rightarrow \) stable node, if \( c > 4(b - \mu - 1) \)

stable spiral, if \( c < 4(b - \mu - 1) \) if \( c > c_{\text{min}} = 2\sqrt{b - \mu - 1} \)

the singular point \((0, 0)\) is a stable node. The case when \( c = c_{\text{min}} \) gives us a degenerate node. If \( c^2 < 4(b - \mu - 1) \), it is a stable spiral; i.e., \( E' \) oscillates in a neighbourhood of the origin. By continuity disputes, in dimensional terms the range of \( c \) must satisfy.
\[
c > c_{\text{min}} = 2\sqrt{R_0 + \mu - b - 1} = 2\sqrt{\frac{\lambda_1 + \lambda_2 + \lambda_3}{\lambda_3} R_0 - 1},
\]

\( \mu > b + 1 - R_0 \).

No wave solution exists for \( \mu < b + 1 - R_0 \). So this condition is necessary for \( I' \) class to propagate an epidemic wave.

There are a typical travelling wave solution when \( c \leq 2\sqrt{\sqrt{b - \mu - 1}} \). As \( I' < 0 \) for some \( z \), they are physically unrealistic, because in that case \( I' \) spirals around the origin. In this seance \( E' \to 0 \) at the leading edge with decreasing oscillation around \( I' = 0 \).

Analysis of analytical solution

Now solutions of equations (15) and (16) are respectively given by
\[
E'(z) \propto e^{\sqrt{-c \pm \sqrt{c^2 - 4(b - \mu - 1)}} z} \frac{1}{2},
\]

(17) \( \quad \) \( \quad \) \( \quad \) \( \quad \)

and
\[
I'(z) \propto e^{\sqrt{-c \pm \sqrt{c^2 - 4(R_0 + \mu - b - 1)}} z} \frac{1}{2} \]

(18) \( \quad \) \( \quad \) \( \quad \) \( \quad \)

As we required \( E'(z) = I'(z) \) and \( \Gamma'(z) \to 1 \) with \( E'(z) > 0 \) and \( \Gamma'(z) > 0 \), these solutions can not oscillate about \( E' = I' \) and \( I' = 1 \) respectively; otherwise \( E'(z) < 0 \) and \( \Gamma'(z) < 0 \) for some \( z \). Then from relations (17) and (18), the travelling wave speed \( c \) and \( \mu \) must satisfy
\[
c > c_{\text{min}} = \max \left\{ 2\sqrt{b - \mu - 1}, 2\sqrt{R_0 + \mu - b - 1} \right\}
\]

\( = \max \left\{ 2\sqrt{\frac{\lambda_1 S_0}{\lambda_3} R_0 - 1}, 2\sqrt{\frac{\lambda_1 + \lambda_2}{\lambda_3} S_0 - \frac{\lambda_3 S_0}{\lambda_3} - 1} \right\} \)

where \( b + 1 - R_0 < \mu < b - 1 \). This maximum represents whether due to the pandemic COVID-19, the population in which the disease is in latent state or the infected population increase more from the starting of pandemic or not.

Thus the wave speed for the pandemic COVID-19 in total population is given by
\[
||c||_{2} \geq 2 \left[ \sqrt{b - \mu - 1}^2 + \sqrt{R_0 + \mu - b - 1}^2 \right] = 2\sqrt{R_0 - 2}.
\]

(19) \( \quad \) \( \quad \) \( \quad \) \( \quad \)

In dimensional term this is given by
\[
||c||_{2} \geq 2 \left[ \frac{\lambda_1 S_0}{\lambda_3} R_0 - \frac{2}{2} \right]
\]

we expect such travelling waves derived from fully non-linear system of equations, will evolve into a travelling waveform with the minimum wave speed given by the equation (19), except in exceptional conditions. The wave velocity for COVID-19, \( V \) say, in dimensional term is then given as
\[ V = (\gamma D)^2 |c| \partial |c|, \]

where
\[ \|c\| = 2\sqrt{R_0 - 2}. \]

The travelling wave solution for susceptible population \( S' \) cannot exhibit a local maximum, since \( S' = 0 \) there and the equation for \( S' \) shows that \( S'' = \alpha S'E' + (R_0 - \alpha)S' > 0 \), which indicates a local minimum. So \( S'(z) \) is a monotone increasing function of \( z \). By linearising the equation (10) for \( S' \) as \( z \to \infty \), where \( S' = 0 + s \), with \( s \) small, we get
\[ s' + cs = 0 \]

which gives \( S'(z) \to 0 \) as \( z \to \infty \).

**Model for proper quarantine case**

Due to proper quarantine of infective population and considering for citizens of several countries that needed to be quarantined for 14 days earlier to entering their topical countries or state, who comes outside their own country or same state respectively and treating them as infective, though the interaction between several classes of people, considered above, is reduced but not properly [18]. Because of daily life survival, some of the susceptible population comes in contact with some of the infective, though the interaction between several classes of people, mainly the travelled waves of COVID-19 transmission for ‘proper quarantine case’ will be a solution of the system (20)–(21). Then \( T_{\text{max}} = \infty \) and

\[ 0 < S(x, t) + E(x, t) \leq \max \{ \|S_0\|, \|E_0\| \} \]

\[ + E_0(x) | |_{\infty} + \|S(x, t) + E(x, t)\|_{\infty} + \|S_0(x) + E_0(x)\|_{\infty}, N \]

where prime denote differentiation with respect to \( x \).

**Proof.** The proof is similar as Proposition (2). \( \Box \)

\( R_0 \) for proper quarantine case: Due to ‘proper quarantine’ [6], the strict isolation [7] of infected individuals and considering for citizens of several countries that required to be quarantined for 14 days prior to entering their native countries or native state, who comes outside their own own country or own state respectively, mainly the \( E \) class is dominant to spread the pandemic COVID-19. From the equation (21), the per capita increase of \( E \) class is given by

\[ \frac{1}{E} \frac{\partial E}{\partial t} = \lambda S - \gamma. \]

Which gives us the basic reproductive rate

\[ R_0 = \frac{\lambda S_0}{\gamma}. \]

If \( R_0 > 1 \), then every infected member of the population will emit the disease to leastwise one other member during the infectious epoch, and the model argue that the disease will propagate within the population. If not, then the disease is desired to fall through before overreaching a substantive fraction of total population. Therefore \( R_0 = 1 \) is a critical epidemiological grade. In other terms, pathogens with elevated equilibriums of contagion and subordinate rescue and mortality rates will gesture an ideal threat. The reciprocal of the removal rate is the average time interval during which an individual from both \( E \) and \( I \) class remain contagious, given by \( \gamma \).

**Re-parametrisation**

For non-dimensionalisation, rescaling dependent variables \( S, E, I, R \) by \( S_0 \), the initial susceptible population and independent variables as \( x, t \) by \( x_0 \) and \( t_0 \) respectively, let \( S = S_0 S', E = S_0 E', I = S_0 I', R = S_0 R', t = t_0 t' \) and \( x = x_0 x' \), where \( S_0, t_0 \) and \( x_0 \) are the characteristic units used to scale the above variables. Putting these in the system (20)–(23) we get

\[ \frac{\partial S'}{\partial t} = -\lambda S_0 S' S_0 E' + R_0 \frac{\partial S'}{\partial x}, \]

\[ \frac{\partial E'}{\partial t} = \lambda S_0 S' S_0 E' - \gamma S_0 E' + \frac{\partial E'}{\partial x}, \]

\[ \frac{\partial I'}{\partial t} = \lambda S_0 S' S_0 E' - \gamma S_0 I' \]

\[ \frac{\partial R'}{\partial t} = \gamma S_0 \left( I' + E' \right) \]

that is,

\[ \frac{\partial S'}{\partial t} = -\lambda S_0 S' S_0 E' + \frac{\partial S'}{\partial x}, \]

\[ \frac{\partial E'}{\partial t} = \lambda S_0 S' S_0 E' - \gamma S_0 E' + \frac{\partial E'}{\partial x}, \]

\[ \frac{\partial I'}{\partial t} = \lambda S_0 S' S_0 E' - \gamma S_0 I' \]

\[ \frac{\partial R'}{\partial t} = \gamma S_0 \left( I' + E' \right) \]

With \( S' \left( x, 0 \right) = 1, E' \left( x, 0 \right) = \frac{\varepsilon}{\gamma}, I' \left( x, 0 \right) = \frac{\varepsilon}{\gamma}, R' \left( x, 0 \right) = 0 \) as

\[ \frac{\partial S'}{\partial x}, \]

\[ \frac{\partial E'}{\partial x}, \]

\[ \frac{\partial I'}{\partial x}, \]

\[ \frac{\partial R'}{\partial x}, \]

\[ \frac{\partial |c|}{\partial x}, \]

\[ \frac{\partial |c|}{\partial x}. \]
initial conditions; where $E_0$ and $I_0$ are respectively the initial population of $E$ and $I$ class.

Letting $y_0 = 1$ and $D \frac{y}{\partial z} = 1$, we get $t_0 = \frac{1}{r}$ and $x_0 = \left( \frac{r}{2} \right)^{\frac{1}{2}}$. Thus we end up with the scaled model

$$\frac{dS}{dt} = -\mu S + \frac{c^2 S}{c^2 + I},$$

$$\frac{dI}{dt} = R_0 S E - E + \frac{c^2 E}{c^2 + I},$$

$$\frac{dR}{dt} = \left( \mu - R_0 \right) S E - I'.$$

Division of equation (28) by $S$ and integration over $(0, z)$ gives

$$S'(z) = \exp \left( \frac{\mu}{c} \int_0^z E(s) ds \right)$$

and

$$S'(z) = \exp \left( \frac{\mu}{c} \int_0^z E(s) ds \right) \neq 0 \text{ Now differentiating equation (29) with respect to } z \text{ we get}$$

$$\frac{d^2 E}{dz^2} = \frac{R_0 E}{c} \left( \frac{dE}{dz} \right) + S \frac{dE}{dz} + \frac{1}{c} \frac{dE}{dz}.$$

If there exists $z^* \in [0, \infty)$ such that $\left( \frac{dE}{dz} \right)^* = 0$ then from the above equation we get

$$\frac{d^2 E}{dz^2} = \frac{R_0 E}{c} \left( \frac{dE}{dz} \right)^* < 0.$$

Thus if $\left( \frac{dE}{dz} \right)^* = 0$ for some $z^* \in [0, \infty)$, then $E(z)$ is concave downwards.

Again from (29)

$$\frac{dE}{dz} = \frac{1}{c} \left( -R_0 S(z) + 1 \right) E(z).$$

Then $E'(z)$ increases at $z = 0$ if $-R_0 + 1 > 0$, i.e. if $R_0 < 1$ and decreases at $z = 0$ if $-R_0 + 1 < 0$, i.e. if $R_0 > 1$.

Therefore $E'(z)$ has at most one peak. Also with the convergence of $E'(z)$ to 0, our claim on $E(z)$ follows. Hence the proof.

Now the system (28)–(31) can be represented as

$$S' + cS^* - \mu S E^* = 0,$$

$$E' + cE' + R_0 S E^* - E^* = 0,$$

$$I' + cI' + (\mu - R_0) S E^* - I' = 0,$$

$$cR' + (I' + E') = 0.$$

where prime represents differentiation w.r.t. $z$. The above system consists of finding range of parameters considered above, for which there exists a solution with positive wave speed $c$ and non-negative $S', I'$ such that

$$E'(-\infty) = E'(+\infty) = 0, I'(-\infty) = I'(+\infty) = 0.$$  

The conditions on $E'$ and $I'$ imply a pulse wave of infective population which propagates into the uninfected population. As time goes, the system (33)–(34) changes by linearising, with $S' \rightarrow 1, E' \rightarrow 0$ and $I' \rightarrow 0$ to get

$$E' + cE' + (R_0 - 1) E' \approx 0,$$

$$I' + cI' - I' \approx 0.$$

Phase plane analysis: From equations (36) a typical wave front solution is where $E'$ at one side, say, as $z \to -\infty$, is at one steady state and as $z \to +\infty$ it is at the other. So here we have to determine the value or values of $c$, for which the equation (36) has a non-negative solution $E'$ which satisfies.

$$\lim_{z \to -\infty} E'(z) = 0; \lim_{z \to +\infty} E'(z) = 1.$$
In \((E', U)\) phase plane.

\[ E' = U, U' = -cU - (R_0 - 1)E \]

we have the phase plane trajectories as the solution of

\[ \frac{dz}{dt} = -cU - (R_0 - 1)E \]

which has one singular point for \((E', U)\) as \((0, 0)\). Corresponding to this singular point we define the matrix

\[ A = \begin{pmatrix} -c & -(R_0 - 1) \\ 1 & 0 \end{pmatrix} \]

whose eigenvalues are

\[ \frac{1}{2} \left[ -c \pm \sqrt{c^2 - 4(R_0 - 1)} \right] \]

stable node, if \( c^2 > 4(R_0 - 1) \)

stable spiral, if \( c^2 < 4(R_0 - 1) \)

If \( c > c_{\text{min}} = 2\sqrt{R_0 - 1} \) then the singular point \((0, 0)\) is a stable node. The case when \( c = c_{\text{min}} \) gives us a degenerate node. If \( c^2 < 4(R_0 - 1) \), it is a stable spiral; that is, in a neighbourhood of the origin, \( E' \) oscillates. By continuity disputes, in dimensional terms the range of \( c \) must satisfy

\[ c > c_{\text{min}} = 2\sqrt{R_0 - 1} = 2\sqrt{\frac{\lambda S_0}{\gamma}} - 1, R_0 > 1. \]

No wave solution exists for \( R_0 < 1 \). So this condition is necessary for the spread of epidemic wave for \( E' \) class.

There are a typical travelling wave solution when \( \epsilon \in 2\sqrt{R_0 - 1} \). As \( E' < 0 \) for some \( z \), they are physically unrealistic, because in that case \( E' \) spirals around the origin. In this season \( E' \rightarrow 0 \) at the leading edge with decreasing oscillation around \( E' = 0 \).

From equations (37) a typical wave front solution is where \( I' \) at one side, say, as \( z \rightarrow -\infty \), is at one steady state and as \( z \rightarrow -\infty \) it is at the other. In that case we have to determine the value or values of \( c \) for which the equation (37) has a non-negative solution \( I' \) which satisfies

\[ \lim_{z \to -\infty} I'(z) = 0, \lim_{z \to -\infty} I'(z) = 1. \]

In \((I', V)\) phase plane.

\[ I' = V, V = -cV + \gamma \]

we have the phase plane trajectories as solution of

\[ \frac{dz}{dt} = -cV + \gamma \]

which has one singular point for \((I', V)\) as \((0, 0)\). Corresponding to this singular point we define the matrix

\[ A = \begin{pmatrix} -c & 1 \\ 1 & 0 \end{pmatrix} \]

whose eigenvalues are

\[ \frac{1}{2} \left[ -c \pm \sqrt{c^2 + 4} \right] \Rightarrow \text{saddle point.} \]

**Analysis of analytical solution**

Now solutions of the system (36)–(37) are respectively given by

\[ E'(z) \sim \exp \left[ \frac{-c \pm (c^2 - 4(R_0 - 1))}{2} z \right] \]

and

\[ I'(z) \sim \exp \left[ \frac{-c \pm (c^2 + 4)}{2} z \right] \]

As we required \( E'(z) \rightarrow 0 \) and \( I'(z) \rightarrow 0 \) with \( E'(z) > 0 \) and \( I'(z) > 0 \), these solutions can not oscillate about \( E' = 0 \) and \( I' = 0 \) respectively; otherwise \( E'(z) < 0 \) and \( I'(z) < 0 \) for some \( z \). Then from the relations (38) and (39), the wave speed \( c \) and \( R_0 \) must satisfy

\[ c > c_{\text{min}} = 2\sqrt{R_0 - 1} = 2\sqrt{\frac{\lambda S_0}{\gamma}} - 1, R_0 > 1 \]

and the threshold condition in dimensional terms is given by

\[ R_0 = \frac{\lambda S_0}{\gamma} > 1 \]

we expect such travelling waves derived from the fully non-linear system of equations will evolve into a travelling waveform with the minimum wave speed \( c = 2(R_0 - 1)^{\frac{1}{2}} \), except in exceptional conditions. The wave velocity for COVID-19, \( V \), say, in dimensional term is then given by

\[ V = (\gamma D)^{\frac{1}{2}} = 2(\gamma D)^{\frac{1}{2}} \left[ \frac{\lambda S_0}{\gamma} - 1 \right]^{\frac{1}{2}} \]

The travelling wave solution for susceptible population \( S' \) cannot have a local maximum, since \( S' = 0 \) there and the equation for \( S' \) shows that \( S'^{\prime} = \mu E' S' > 0 \), which indicate a local minimum. So \( S'(z) \) is a monotone increasing function of \( z \). By linearising the equation (32) for \( S' \) as \( z \rightarrow -\infty \), where \( S' = 1 - s, \) with \( s \) small, we get

\[ s'' + cs' - \mu E' = 0 \]

with which \( E'(z) \) from (38),

\[ S'(z) \sim 1 - O \left( \exp \left[ \frac{-c \pm (c^2 - 4(R_0 - 1))}{2} z \right] \right) \]

and so, as \( z \rightarrow -\infty \), \( S'(z) \rightarrow 1 \) exponentially.

**Results and discussion**

In this paper for the epidemic COVID-19, we have investigated both analytical and numerical solutions for both the models \( 'no quarantine case' \) and \( 'proper quarantine case' \). From our analysis of analytical solution for both the models it is observed that for \( 'no quarantine case' \) as time goes, both the susceptible and latent infected population tend to zero and the total population will become infected where as \( 'proper quarantine case' \) as time goes; the infected population decrease, latent infected population tend to zero and the total population will become susceptible.

For \( 'no quarantine case' \) we separately calculate the speed of spread of COVID-19 in both \( E \) and \( I \) class, which tells us that, as time goes, whether \( E \) or \( I \) class increase. After that we consider the Euclidean norm to calculate the wave velocity for the pandemic COVID-19 in total population, given by the relation (19). For \( 'proper quarantine case' \), due to proper quarantine for infected population, the speed of spread of COVID-19 only depend on \( E \) class. So the speed of spread of COVID-19 in \( E \) class is the wave velocity for the pandemic COVID-19 in total population, given by the relation (40).

For \( 'no quarantine case' \) the numerical simulation of travelling wave solution of the system (6)–(9) is done using Crank-Nicolson method. For the sake of convenience we truncate the time domain \([0, \infty) \) to \([0, 50] \) and the one dimensional spatial domain \([0, 1]\). With respect to this boundary condition of \( t' \) and \( x' \), the boundary conditions of \( S'(x', t') \), \( E'(x', t') \), \( I'(x', t') \) and \( R'(x', t') \) are considered as \( S'(x', 0) = S'(x', 50) = 1, S'(0, t') = 0; E'(x', 0) = 0; E'(x', 50) = 0; E'(0, t') = E'(1, t') = 0 and I'(x', 0) = I'(1, t') = 0, I'(0, t') = I'(1, t') = 1 and \( R'(x', 0) = 0 \) respectively. The 3D plots of \( S'(x', t'), E'(x', t'), I'(x', t') \) and \( R'(x', t') \) are shown in Figs. 3 and 4, respectively with respect to \( x' \) and \( t' \).
for the travelling wave solution. Based on above boundary conditions, numerical simulation shows that for the figure Fig. 3, the susceptible population \( S'(x', t') \) goes to zero in forward time and at any position \( x' \). From the figure Fig. 3 it is obvious that the latent infected population \( E'(x', t') \) first increase and after certain time it decrease and goes to zero in forward time at any position \( x' \). From the figure Fig. 4, it is observed that the infected population \( I'(x', t') \) increases and takes the value one in forward time at any position \( x' \). From the figure Fig. 4 we observe that

![Fig. 3](image)

**Fig. 3.** No quarantine case: Numerical simulation for Susceptible population \( S'(x', t') \) and Latent infected population \( E'(x', t') \) for the system of equations (6)-(9) when \( a = 8; b = 3; \mu = 1.8; ||c||_2 = 2.5 \) and \( R_0 = 2.5 \).

![Fig. 4](image)

**Fig. 4.** No quarantine case: Numerical simulation for Infected population \( I'(x', t') \) and Removed population \( R'(x', t') \) for the system of equations (6)-(9) when \( a = 8; b = 3; \mu = 1.8; ||c||_2 = 2.5 \) and \( R_0 = 2.5 \).

![Fig. 5](image)

**Fig. 5.** Proper quarantine case: Numerical simulation for Susceptible population \( S'(x', t') \) and Latent infected population \( E'(x', t') \) for the system of equations (24)-(27) when \( c = 5; \mu = 3 \) and \( R_0 = 2.5 \).
the graph of removed population $R'(x', t')$ is flat with respect to $t'$, because there are no diffusion of $R'$.

For ‘proper quarantine case’ the numerical simulation of travelling wave solution of the system (24) and (27) is done. Like ‘no quarantine case’ here we also truncate the time domain $[0, \infty)$ to $[0, 50]$ and the one dimensional spatial domain $\Omega$ to $[0, 1]$. With respect to this boundary condition of $t'$ and $x'$, the boundary condition of $S'(x', t'), E'(x', t'), I'(x', t')$ and $R'(x', t')$ are defined as by $S'(x', 0) = S'(x', 50) = 0, S'(0, t') = S'(1, t') = 0, E'(x', 0) = E'(x', 50) = E'(1, t') = 0, I'(x', 0) = I'(x', 50) = 1, I'(0, t') = I'(1, t') = 0$ and $R'(x', 0) = 0$. The 3D plots of $S'(x', t'), E'(x', t'), I'(x', t')$ and $R'(x', t')$ are shown in figures (5) and (6), respectively with respect to $x'$ and $t'$ for the travelling wave solution. Based on above boundary conditions, numerical simulation shows that for the Fig. 5, the susceptible population $S'(x', t')$ increases and goes to one in forward time and at any position $x'$. From the figure Fig. 5 it is obvious that the latent infected population $E'(x', t')$ first increase and after certain time it decrease and goes to zero in forward time at any position $x'$. From the figure Fig. 6 we observe that the infected population $I'(x', t')$ decrease and takes the value zero in forward time at any position $x'$. From the figure Fig. 6 we observe that the graph of removed population $R'(x', t')$ is increases in forward time and at any position $x'$.

For ‘no quarantine case’, we solve the system (10)–(13) numerically with boundary conditions $E'(-\infty) = E'(\infty) = 0, S'(-\infty) = S'(\infty) = 0, 0 \leq I'(-\infty) < I'(\infty) = 1$. The estimation of parameters are given by $a = 8, b = 3, \mu = 1.8, c = 2.5$ and $2.5$$\leq R_0 \leq 3.00$. Based on those boundary conditions and the estimation of parameters, a numerical simulation is done for all of $S', E', I'$ and $R'$ populations with respect to $z$ and the variations of Basic Reproduction number $R_0 = 2.50, R_0 = 2.75$ and $R_0 = 3.00$. In Fig. 7 we draw the graphs of $S'$ with respect to $z$ and the above stated variations of $R_0$. For all the variations of $R_0$ we observed that $S'$ tends to 0 as $z$ approaches to $\infty$. As for an infectious disease (COVID-19 in our case) the basic reproductive number is the number of secondary infections delivered by a single infected individual in whole susceptible population and this quantity indicates the initial growth rate of the infected population and the potential for a large-scale epidemic, it is also noticed that for the larger value of $R_0$ the susceptible population will become more infected and the rate of convergence of the susceptible population $S'$ to 0 becomes faster for a large population. In Fig. 8 we draw the graphs of $E'$ with respect to $z$ and for the variations of $R_0$. In this case for all the variations of $R_0$ we observed that the latent infected population $E'$ tends to 0 as $z$ approaches to $\infty$. In Fig. 9 we draw the graphs of $I'$ with respect to $z$ and the variations of $R_0$. For all the variations of $R_0$ we observed that $I'$ tends to 1 as $z$ approaches to $\infty$. It is also noticed that for the larger value of $R_0$ the volume of susceptible population $S'$ decreases, whereas the volume of infected population $I'$ increases. That is why the rate of convergence of the infected population $I'$ to 1 becomes faster for the larger value of $R_0$ for a larger population. In Fig. 10 the graphs of $R'$ with respect to $z$ and for the variations of $R_0$ are drawn. For all the above stated variations of $R_0$ the removed population $R'$ decreases as $z$ approaches to $\infty$. This discussion guarantee us that for vno quarantine case’ the total population will become infected after
certain time. That is why a ‘proper quarantine’ is essential, as a result we consider our next model as a ‘proper quarantine model’.

For ‘proper quarantine case’, we solve the system (28)–(31) numerically, with boundary conditions $E'(-\infty) = E'(\infty) = 0$, $I'(-\infty) = I'(\infty) = 0$. The estimation of parameters are given by $\mu = 3, c = 5$ and $2.40 < R_0 < 2.50$. Based on those boundary conditions and the estimation of parameters, a numerical simulation is done for all of $S$, $E'$, $I'$ and $R'$ populations with respect to $z$ and the variations of the Basic Reproduction Number $R_0$ as $R_0 = 2.40$, $R_0 = 2.45$ and $R_0 = 2.50$. In Fig. 11 we draw the graphs of $S$ with respect to $z$ and the variations of $R_0$. For all the above stated variations of $R_0$ we observed that $S$ increases as $z$ approaches to $\infty$. It is also noticed that for the larger value of $R_0$, the rate of increase of the susceptible population $S$ becomes slower. This happens because of ‘proper quarantine’ of infected individuals and proper isolation of latent infected population. In this case the number of secondary infection from a single primary infection is reduced and therefore the value of the Basic Reproduction Number $R_0$ is reduced. In Fig. 12 we draw the graphs of $E'$ with respect to $z$ and for the variations of $R_0$. In this case for all the variations of $R_0$ we observed that the latent infected population $E'$ tends to 0 as $z$ approaches to $\infty$. In Fig. 13 we draw the graphs of $I'$ with respect to $z$ and the variations of $R_0$. For all the variations of $R_0$ we observed that $I'$ decreases monotonically as $z$ approaches to $\infty$. It is also noticed that due to ‘proper quarantine’, for the smaller value of $R_0$, the rate of decrease of the infected population $I'$ becomes faster. In Fig. 14 the graphs of $R'$ with respect to $z$ and for the variations of $R_0$ are drawn. For all the above stated variations of $R_0$ the removed population $R'$ first decrease and after that it increases as $z$ approaches to $\infty$. This discussion guarantee us that for ‘proper quarantine case’ the total population will become susceptible after certain time. Thus in the absence of vaccine, proper quarantine of infected individuals and proper isolation of latent infected individuals are very essential together with social distancing and use of mask.

An elementary public health intention is to fetch disease from over an epidemic threshold grade to under threshold grade, thereby excluding a threat of a large scale epidemic. That can accomplish through interventions that either directly impact the infectiousness of the pathogen, modify patterns of interaction so that the pathogen cannot easily spread within the population, or immunize partitions of the population. We call those three forms of intervention as contact reducing, transmission reducing, and immunizing [16].

There are some important enlightenment of the threshold result (41) like, the critical population density $S_0 = \frac{1}{\gamma}$ for the existence of epidemic wave for ‘proper quarantine case’; the critical transmission coefficient $\lambda_3 = \frac{b}{c}$ to the $E$ class which, if not exceeded, obstruct the spread of the disease; the threshold mortality rate $R_0 = \frac{\lambda_3 S_0}{E}$ for $E$ class which, if not exceeded, obstruct the spread of the disease etc. All of these have adhesion for control strategies [20, 26]. If we can minimize the transmission measure $\lambda_3$ for COVID-19 to $E$ class, it may be feasible to overstep condition (41) and therefore again obstruct the spread of the disease. This can be done by social distancing, proper quarantine of infective population and considering for citizens of several countries that needed to be quarantined for 14 days earlier to entering their topical countries or topical state, who comes outside their own country or own state respectively. Finally with $R_0 > 1$ as the threshold criterion we notice that an accidental inflow of the susceptible population can increase $S_0$ above $S_i$ and hence commence an epidemic.
respectively. Later on we found the value of 
Fig. 14.

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dimensionalisation, rescaling dependent variables
infectious disease, which transmit through contact.
certain time. Not only for COVID-19, this model is also valid for any
quarantine case
population will become susceptible after certain time, where as for 'no
numerical method as well as analytical method we saw that the total
economy. Because, solving the model for 'proper quarantine case
vaccine. This strategy can also save them from the destroy of their
entering their native countries or native state, who comes outside their
several countries that required to be quarantined for 14 days prior to
isolation etc. for an infectious disease, proved that social distancing,
spread of the disease. But this strategy is not only hurting the economy of
many countries are adoption lock down as a possible way to prevent the
world are struggling to find ways to prevent the spread of COVID-19 and

Conclusions
In the absence of vaccine for COVID-19, governments across the
world are struggling to find ways to prevent the spread of COVID-19 and
countries are adoption lock down as a possible way to prevent the
spread of the disease. But this strategy is not only hurting the economy of
their respective countries, but also hurting the global economy.

In this article our investigation for the necessity of social distancing,
isoaltion etc. for an infectious disease, proved that social distancing,
proper quarantine of infective population and considering for citizens of
certain countries that required to be quarantined for 14 days prior to
entering their native countries or native state, who comes outside their
own country or own state respectively is one of the best possible way to
stop the spread of COVID-19 with out lock down, in the absence of vaccine.
This strategy can also save them from the destroy of their economy. Because, solving the model for 'proper quarantine case'
by numerical method as well as analytical method we saw that the total
population will become susceptible after certain time, where as for 'no
quarantine case', we saw that the susceptible population become 0 after
certain time. Not only for COVID-19, this model is also valid for any
infectious disease, which transmit through contact.

We also make the following important observations, for both the model 'no quarantine case' and 'proper quarantine case', the system of
diffusion equations possess non-negative solutions and the solution are
uniformly bounded. For both models we derive the wave velocity for the
epidemic, comparing them we get that the wave velocity for 'no quar-
teine case' is always greater than that of 'proper quarantine case'.

Declaration of Competing Interest
The authors declare that they have no known competing financial
interests or professional relationships that could have appeared to influence
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