Investigating Malaysia’s preparedness towards the second wave of COVID-19 via SIR modelling

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Abstract:

**Purpose:** This work simulates the infectious trend of COVID-19 in order to understand, predict and estimate the severity of the disease and to gauge Malaysia’s existing preparations and policies on the issue. The simulation is used to predict the number of cumulative infectious individuals’ within the society, which then serves as a measure to investigate the readiness of Malaysia in battling against the second wave of the outbreak.

**Methods:** Confirmed positive infectious cases (as reported by Ministry of Health, Malaysia (MOH)) from Jan 25, 2020 to March 31, 2020 were used to infer the severity of the COVID-19 infectious trend in Malaysia. We simulated the infectious count for the same duration and extended the trajectory for another two and four week to assess the predictive capability of the Susceptible-Infectious-Recovered (SIR) model. We also utilized the transmission rate $\beta$ to predict the cumulative number of infectious individuals.

**Results:** Our predicted cumulative number of infectious individuals’ tallies with the preparations undertaken by the MOH and the simulation does provide an indication of the severity of COVID-19 disease outbreak in Malaysia.

**Conclusion:** The SIR model that we obtained was not far off from the actual count and its simulated decline is in correlation with the introduction of Malaysian control measures such as the Movement Control Order (MCO), social distancing and increased hygienic awareness.

**Keywords:** COVID-19 trend; COVID-19 readiness; epidemic models; predictive analytics;

**Introduction**

The first incidence of coronavirus disease 2019 (COVID-19) was reported in mainland China, in the city of Wuhan on Dec 31, 2019. Ever since then, the geographical spread of the epidemic was swift and beyond control, hence turning it into a pandemic.\(^1\) Beginning on Feb 26, 2020, the number of COVID-19 cases increased rapidly worldwide when compared to inside China. Massive substantial outbreaks were observed in the USA, Italy and Iran. COVID-19 is caused by the severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2)\(^2\) that has resulted in more than 130,000 deaths within a short span of time. The death toll of COVID-19 is extremely greater
than that of the other infectious diseases caused by viruses such as SARS, MERS, Ebola, and H1N1. At the onset of COVID-19, patients usually show symptoms connected to viral pneumonia, usually fever, cough, sore throat myalgia and fatigue. The incubation period of COVID-19 is generally between two to 14 days or longer and usually takes five days as the average. The virus can spread from an individual to another through respiratory droplets and close contact.

The first incidence in Malaysia was reported on Jan 25, 2020 through three cases of imported COVID-19 involving three tourists from China who had entered Malaysia via Johor from Singapore on Jan 23, 2020. According to the MOH, the first wave of the outbreak commenced on the first incidence day until Feb 16, 2020. During these 23 days period, only 22 confirmed cases were recorded with eight recovered patients and zero deaths, with minimal outbreak control measures exercised. Moreover, there were no new cases recorded for the following ten days. However, on Feb 27, 2020 new incidence involving two cases marked the beginning of the second wave of the outbreak in Malaysia. The first 18 days of the second wave did not see tremendous increase in the number of new cases, however, on March 15, 2020, the numbers had a sudden rise up to 190 individuals from only 35 on the previous day. This alarming rate was due to the identification of large clusters of susceptible individuals that were in contact with infectious individual(s). Following that, the first death incidence involving two cases occurred on March 17, 2020. The sudden hike and drastic spread of the outbreak changed the Malaysian context of the outbreak from being in-control to perilous state and triggered the implementation of the Movement Control Order (MCO) on March 18, 2020. Since the first rise on March 15, 2020, the inclining trend of new cases each day continued with an average of 150 cases per day and was hit with a major spike of 235 cases on March 26, 2020. On top of that, the number of deaths started spiraling upwards, with more than 60% of deaths were patients over the age of 60 years which included those with underlying conditions such as hypertension and diabetes (as of March 31, 2020), in line with the findings from Guan and colleagues’ clinical progress study. The Malaysian scenario of COVID-19 outbreak from Jan 25, 2020 to Mar 31, 2020 can be visualized as in Figure 1.
In such a perturbing situation, assessing the infectious trend of COVID-19 in Malaysia is crucial to measure the pandemic's severity, as evidenced by the enormous growth from three cases to 2,766 cases within only 67 days (March 31, 2020). Estimation of infectious count over time is able to provide a better understanding of the current epidemiological situation in Malaysia and can provide insights into the measurable effect of undertaken outbreak control measures. Analysis providing such estimations enables predictions of potential future growth, can assist in risk estimation of regional countries, and planning of alternative interventions or increasing the intensity of existing interventions.

Figure 1 Visualization of first 67 days of COVID-19 outbreak in Malaysia.

Nevertheless, performing such analyses, especially in real time is often difficult due to constraints such as delay in symptom appearance resulting from the incubation period and delay in confirmation of positive cases resulting from limitations in testing and detection capacity. Mathematical modelling of infectious diseases can help to overcome the constraints caused by the delays and uncertainty. The most common modelling approach to simulate the probable outbreak trajectory and severity of an infectious disease outbreak is the Susceptible-Infectious-Recovered (SIR) model. As anticipated, several studies have widely applied the SIR model.
and its extensions such as the Susceptible-Exposed-Infectious-Recovered (SEIR) model\textsuperscript{24-27} and
the Susceptible-Exposed-Infectious-Hospitalized-Recovered (SEIHR)\textsuperscript{28} to the current COVID-19
outbreak at global and regional levels.

Until now, there has been scarce information in understanding the changing severity and
transmission dynamics of COVID-19 in South East Asia, particularly in Malaysia which is severely
hit by the outbreak in the region. Using the SIR model, we simulated the infectious trend of
COVID-19 in Malaysia to estimate the COVID-19 transmission pattern for a period of 67 days.
The simulation is used to obtain an overall picture of COVID-19’s potential severity in Malaysia.
Furthermore, we predicted the cumulative number of infectious individuals in order to assess the
preparations undertaken by the MOH. That is, whether the MOH has prepared the minimal
number of beds or not.

\textbf{Methods}

\textbf{Data sources and collection}

In this modelling study, we extracted the daily number of confirmed positive (infectious) cases
from the official daily statistics of COVID-19 provided on the MOH web portal.\textsuperscript{29} The extracted
records were then collated as time-series data, which begins from day zero of the COVID-19
outbreak in Malaysia (Jan 25, 2020). In order to avoid any possibility of biasness in using a single
data source, we validated our figures with Kini News Lab COVID-19 tracker\textsuperscript{30}, a local website that
provides real time data and information on COVID-19. These data are collected through daily
press conference statements by the Director General of Health, Malaysia, where patients’ data
are not identifiable and remain anonymous. Hence, ethical approval is not required.

\textbf{Infectious trend simulation of COVID-19 in Malaysia}

In the modelling procedure, we divided the population into three compartments, as follows:
susceptible $S(t)$ (number of not yet infectious and disease free individuals at time $t$), infectious $I(t)$
(number of confirmed or isolated individuals at time $t$), and recovered $R(t)$ (no longer infectious
individuals at time $t$. We used the standard SIR epidemic model (Figure 2) to simulate the infectious severity of COVID-19 in Malaysia beginning from the first day of the outbreak. This model is widely used as a first approach to analyze virus spreading and is reasonably predictive for infectious diseases which are transmitted from human to human, and where recovery confers lasting resistance, such as measles, mumps and rubella.\(^{21}\)

![Figure 2 SIR state diagram.](image)

The standard SIR model assumes no births or deaths i.e. a fixed population, $N=S(t)+I(t)+R(t)$. The primary components of this model are the parameters $\beta$: transmission rate which controls the rate of spread and $\gamma$: recovery rate. If the average duration of recovery is denoted $D$, then the recovery rate is given by $\gamma=1/D$, since an individual experiences one recovery in $D$ days. Apart from these parameters, another important measure in epidemiology is the basic reproductive number $R_0$, which estimates the speed at which a disease is capable of spreading in a specific population.\(^{31}\) The variable $R_0$ also indicates the number of secondary infections stemming directly from the first case in a susceptible population. When $R_0>1$, one infected individual will on average infect $>1$ person in total. When $R_0=1$, we are right at the threshold between an epidemic and not. Finally, when $R_0<1$, one infected individual will on average infect $<1$ person in total. Thus, it is the target to have mechanisms to achieve $R_0<1$. As disclosed by Dr. Noor Hisham Abdullah, Director General of Health, Malaysia on April 10, 2020, Malaysia is approaching $R_0=1$.\(^{32}\) This significant improvement is due to among others – the Movement Control Order (MCO), better social distancing etiquettes and hygienic practices. The dynamics of the COVID-19 transmission can be described using the following nonlinear ordinary differential equations (ODEs) as shown below:
The differential equations were numerically solved with R software environment (version 3.6.3), with fourth order Runge–Kutta (RK4) method via the package deSolve (version 1.28). Data and code are available upon request.

Results

**Actual and Simulated Current Infectious Trend**

The SIR model is initialised (at time $t=0$) with the initial conditions $S(0)=S_0=999$ (99.9% from total population), and $I(0)=I_0=1$ (0.1% from total population). We obtained the basic reproductive number $R_0$ for this study using the average $R_0 = 2.44$ estimated using stochastic methods in two previous studies.\textsuperscript{24,25} It is consistent with the range estimated by WHO\textsuperscript{11}, and Ying and colleagues.\textsuperscript{34} The average number of days of recovery is assumed to be $D=11$ days based on the first recovered case in Malaysia. It follows that the recovery rate, $\gamma=1/11 = 0.09$. Next, using $R_0=\beta/\gamma$,\textsuperscript{31} we derive the value of the transmission rate, $\beta = 2.44(0.09) = 0.22$. Finally, with the values of $S_0$, $I_0$, and $R_0$, the differential equations were solved to obtain the values of each compartments at each time point (days) beginning from day zero (25 January 2020) to day 67 (31 March 2020).

It can be seen that in the initial stage (day zero to day 15, Figure 3), the simulated counts were approximately close to actual counts. However, after day 20, the simulation had an upwards trend with the peak value of 224 infectious individuals on day 56 (21 March 2020). The simulation trend started declining after day 58. On the other hand, the actual infectious counts only started increasing drastically after day 50 and there were three major spikes between days 50 and 61. The actual peak value was 235 infectious individuals on day 61 (26 March 2020).
Figure 3 Actual versus simulated infectious trend.

The standard SIR model was able to approximately mimic the actual trend and predict the actual spikes. The discrepancy in SIR simulation between day 20 to day 50 is due to the nature of the actual counts (discrete values) when there is a sudden spike in the number of confirmed cases. Note that, the simulated peak was also approximately close to that of the actual peak. All three simulated compartments of the SIR model for the same period of time are shown in Figure 4. The simulated susceptible trend declines after day 30 and further down after day 60. Whereas, the simulated recovered trend inclines after day 30 while the infectious trend steadily approaching downward trend.
Figure 4 Simulated SIR compartments.

Trajectory Projection Simulation of COVID-19 in Malaysia

The predictive capability of the SIR model in the previous section is fairly good and thus, we used the same initial conditions and parameters’ values to simulate the trajectory projection of COVID-19 in Malaysia. Two trajectories were produced: 80 days form day zero (until April 13, 2020) and 110 days from day zero (until May 13, 2020). Numbers of days for the trajectories were chosen arbitrarily to determine the approximate number of days needed for the infectious trend to completely decline since the first day of COVID-19’s establishment in Malaysia.

It can be observed from Figure 5 that the simulated maximum count is 224 infectious individuals in a day. Thereafter, the simulated figures dropped to below 200 infectious individuals in a day and further to below 110 infectious individuals at the end of the trajectory on April 13, 2020. For the second trajectory (Figure 6), which is the extended trajectory of 30 days from the previous one, the infectious count trend declines steadily by exhibiting a downward trend from day 80 (April 13, 2020) and reaches the lowest point with less than 20 cases in a day on May 13, 2020, which indicates that the severity of COVID-19 in Malaysia may reduce by mid-May, 2020. Actual infectious count reported on April 13, 2020 was 134 new infectious individuals. In between April 14 until April 23, 2020, actual data showed a downward trend. The actual
infectious counts reported were 170, 85, 110, 69, 54, 84, 36, 57, 50 and 71 new infectious individuals respectively within that time period. Except for the spike on April 14, 2020, the preceding counts maintained below the generated SIR curve. As such, evidence from actual data on infectious individuals has a strong correlation to the fact that our SIR curve is based on $\beta = 0.22$. This implies a strong assumption that infection from an infectious individual to a susceptible individual in Malaysia is at a conservative rate. That is, at the high end, an infectious individual in Malaysia will transmit the disease to a susceptible individual every four days, while at the low end in five days.

![Simulated projection of 80 days trajectory.](image)

**Figure 5** Simulated projection of 80 days trajectory.
Discussion

Based on the SIR model’s simulation, we extend our discussion to predict cumulative positive infectious cases of COVID-19 in Malaysia. This discussion is subject to further analysis, since it is not conclusive in nature. However, we would like to highlight at this point of our argument that discussions provided here are based on definitions within the literature. We will base our discussion on $\beta = 0.22$ which implies a one-to-one transmission after four days. This assumption is conservative.

We hypothesize that after an individual becomes infected, thus becoming an infectious individual, he will only display symptom(s) after 14 days in order to be confirmed positive with the COVID-19 disease and will be hospitalized thereafter. As such, after being hospitalized he will not be able to be in contact with other susceptible individuals. We also hypothesize that during the 14 days period, an infectious individual will be able to spread the disease to another individual in a four days interval.

We assume that first three infectious individuals in Malaysia (confirmed on Jan 25, 2020) were still at large within society at that point in time. With a four days transmission rate, the next three individuals to be infectious would be on Jan 29, 2020 and the next six individuals to be

**Figure 6** Simulated projection of 110 days trajectory.
infectious would be on Feb 2, 2020. On Feb 8, 2020, the first cohort of three individuals would display symptoms and be hospitalized. On Feb 12 2020, the second cohort consisting of three individuals to be infectious on Jan 29, 2020 would display symptoms and be hospitalized. On Feb 16 2020, the third cohort consisting of six individuals to be infectious on Feb 2, 2020 will show symptoms and will be hospitalized. Prior to anybody from the first to the third cohort to display symptoms and hospitalized, the fourth cohort of twelve individuals to be infectious will occur on Feb 6, 2020. This cycle repeats itself ad infinitum until certain measures are able to halt the process. Figure 7a–7d depicts this process.

Figure 7a Predicted Cumulative between Jan 25, 2020 and Feb 12, 2020.
Figure 7b Predicted Cumulative between Feb 12, 2020 and Feb 26, 2020.

Figure 7c Predicted Cumulative between March 1, 2020 and March 15, 2020.
Figure 7d Predicted Cumulative between March 21, 2020 and April 4, 2020.

Remark: The color code in Figure 7a–7d is used to identify the 14 day period of each infectious cohort beginning from the first day the cohort becomes infectious. For example, in Figure 7a the color purple-blue for the first cohort of infectious individuals on Jan 25, 2020: it is predicted that there are three infectious individuals within the society, where three of them will show symptoms and be hospitalized after a 14 day period on Feb 8, 2020. Another example in Figure 7c and 7d, is the color yellow for the tenth cohort of the infectious individuals on March 7, 2020: it is predicted that there will be 879 infectious individuals within society, where 408 of them would display symptoms and be hospitalized after a 14 days period on March 21, 2020.

We note here that the actual cumulative positive cases (where we assume they are hospitalized and not in contact with other susceptible individuals) reported on March 29, 2020 is 2,470 individuals. Meanwhile, the predicted cumulative number of infectious individuals to be hospitalized on March 29, 2020 is 3,063. Furthermore, we have predicted that there are around 19,926 infectious individuals still roaming around in society on March 29, 2020.

Although the prediction depicted in Figure 7a–7d is of exponential growth of infectious individuals, we also take note of the following:
i. The above ‘avalanche’ effect is under the assumption that no remedial action has been taken to halt interaction (except for hospitalizing the infectious ones – which translates into no longer being in contact with susceptible individuals).

ii. The above ‘avalanche’ effect is under the assumption that the best fit SIR curve upon the actual infectious count (discrete data) produces the transmission rate, $\beta = 0.22$, that is translated into the idea of one infectious individual will transmit the disease to another individual within a four day interval.

The following figures present an account of our predicted values when compared to actual values reported by the MOH Malaysia on a daily basis. In both the figures below (Figure 8 and 9), the simulated cumulative is based on Figure 7a–7d.

Figure 8 Actual versus Predicted Cases (Jan 25 to March 31, 2020).
Figure 9 Actual versus Predicted Cases (March 1 to March 31, 2020).

Table 1 summarizes figure 8 and 9. One is able to observe from table 1 that the SIR model produces predicted daily values that are not far off from the actual daily values. As such, the SIR transmission rate, $\beta = 0.22$, which corresponds to a hypothesized scenario whereby an individual will infect another individual within a four days interval, should not be taken lightly. Furthermore, a one-to-one transmission on an interval of four days can be seen to be rather conservative.

Nevertheless, as shown in figure 7(a)–7(d), even at this rate, the exponential growth rate is visible.

Table 1 Difference between Actual and Predicted Cases on Selected Days.

<table>
<thead>
<tr>
<th>Date</th>
<th>March 15, 2020</th>
<th>March 21, 2020</th>
<th>March 25, 2020</th>
<th>March 29, 2020</th>
</tr>
</thead>
<tbody>
<tr>
<td>Actual Cumulative Infectious Cases</td>
<td>428</td>
<td>1183</td>
<td>1796</td>
<td>2470</td>
</tr>
<tr>
<td>Predicted Cumulative Infectious Cases</td>
<td>471</td>
<td>879</td>
<td>1641</td>
<td>3063</td>
</tr>
<tr>
<td>Difference</td>
<td>+43</td>
<td>-304</td>
<td>-155</td>
<td>+593</td>
</tr>
<tr>
<td>Actual Infectious Cases</td>
<td>190</td>
<td>153</td>
<td>172</td>
<td>150</td>
</tr>
<tr>
<td>Predicted Infectious Cases</td>
<td>205</td>
<td>224</td>
<td>217</td>
<td>198</td>
</tr>
<tr>
<td>Difference</td>
<td>+15</td>
<td>+71</td>
<td>+45</td>
<td>+48</td>
</tr>
</tbody>
</table>
Since the outbreak of COVID-19 in Malaysia, the MOH has been referring to the Malaysian Institute of Economic Research (MIER) for its data analysis related to the COVID-19 pandemic in Malaysia. MIER has been tracking the progression of the pandemic in Malaysia through its projection of new cases trajectory. The MIER projected trajectory predicted that burden on beds available and the burden on ICU beds will increase. Based on MIER's prediction, on March 29, 2020, the MOH made an official announcement on the preparation of around 19,200 new beds. However, the figure was nowhere mentioned in MIER's findings, as well as it was unclear on how the MOH came up with that specific figure.

The key finding of our SIR modelling is the prediction of around 19,926 infectious individuals still roaming around within the Malaysian society on March 29, 2020. Our prediction tallies with the preparations undertaken by the MOH. Hence, our prediction through SIR modelling verifies that the preparedness measure taken by the MOH as of 29 March 2020 is sufficient. In particular, with reference to the need of supporting the increased burden on beds available for potential infectious individuals. Furthermore, the MOH will soon be starting to utilize the 19,200 beds facility to treat undocumented migrants that are confirmed positive with COVID-19 lately.

**Conclusion**

Although the SIR model is a numerical simulation, the simulated numbers do provide us a high possible scenario in which the COVID-19 infectious cases can surge to. This gives us an overall picture of the infectious severity of COVID-19 in Malaysia. We showed that the severity dynamics of COVID-19 in Malaysia is rapidly changing and should be closely monitored. Hence, to this end, these trajectories could serve as a dependable means for the Malaysian government, businesses and citizens to plan and mitigate for such spike in infectious cases. We would like to reinstate that the mathematical epidemic modeling is beneficial in deciding the preparedness policies and strategies for flattening the COVID-19 curve, as well as for any such potential future outbreaks. This study is believed to serve as one of the initial efforts for investigating Malaysia’s policies with regards to readiness measures undertaken, especially in the preparation of sufficient number of beds for possible COVID-19 patients in Malaysia. In summary, we note that this study is
undertaken during Malaysian outbreak control measures such as the Movement Control Order (MCO), social distancing and increased hygienic awareness programs.

Acknowledgments

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Disclosure

The authors report no conflicts of interest in this work.

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