PREDICTING THE ULTIMATE OUTCOME OF THE COVID-19 OUTBREAK IN ITALY

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ABSTRACT

During the COVID-19 outbreak, it is essential to monitor the effectiveness of measures taken by governments on the course of the epidemic. Here we show that there is already a sufficient amount of data collected in Italy to predict the outcome of the process. We show that using the proper metric, the data from Hubei Province and Italy has striking similarity, which enables us to calculate the expected number of confirmed cases and the number of deaths by the end of the process. Our predictions will improve as new data points are generated day by day, which can help to make further public decisions. The method is based on the data analysis of logistic growth equations describing the process on the macroscopic level. At the time of writing, the number of fatalities in Italy is expected to be 6000, and the crisis ends before April 15, 2020.

1 Introduction

The present study is motivated by the paper [1] that appeared in The Lancet on 13 March 2020, where the authors are calling for trend analysis. At the beginning of the outbreak, Italian data followed closely the exponential growth trend observed in Hubei Province, China. The question is if Italian measures can break the exponential trend? Can they turn the process to a logistic growth with a final saturation value like in China?

Exponential growth turns into logistic growth when the number of susceptible subjects is limited. If the epidemic is uncontrolled, this limit is defined by the herd immunity, and the vast majority of the population gets infected. Another possibility is when physical contacts in the social network are so severely restricted that the system remains below the epidemic threshold [2] so that only finite-sized infected pockets can develop from infected seeds. This second scenario is in China, where it led to the termination of the epidemic in Hubei Province and only small infected pockets in other parts of China.

Exponential growth is described by the differential equation

$$\frac{dN(t)}{dt} = \lambda N(t),\tag{1}$$

where λ is the rate of growth. If we apply this for the cumulative number of cases N_n on a given day n, then the rate of growth can be calculated from the relative growth of the cumulative number of cases

$$\frac{N_n - N_{n-1}}{N_n} \approx \lambda. {2}$$

When the exponential growth has some limit $N_n < N_\infty$, the growth rate depends on the number of cases $\lambda(N)$ and decreases as we approach the limit. Growth should stop when we reach the limit $\lambda(N_\infty) = 0$. We can expand the growth rate near this point and get

$$\lambda(N) \approx \lambda'(N_{\infty})(N - N_{\infty}) + \frac{1}{2}\lambda''(N_{\infty})(N - N_{\infty})^2 + \dots, \tag{3}$$

where the first derivative is negative $\lambda'(N_\infty) < 0$. Introducing $\lambda_0 = -\lambda'(N_\infty)N_\infty$ and $b = \lambda''(N_\infty)N_\infty^2/2$ we can write $\lambda(N) = \lambda_0(1-N/N_\infty) + b(1-N/N_\infty)^2 + \dots$ If $b \ll \lambda_0$, the second and higher order terms are negligable and we can determine λ_0 and N_∞ from a linear fit

$$\lambda(N) \approx \lambda_0 \left(1 - \frac{N}{N_\infty} \right),$$
 (4)

which leads to the logistic growth equation

$$\frac{dN(t)}{dt} = \lambda_0 N(t) \cdot \left(1 - \frac{N(t)}{N_{\infty}} \right). \tag{5}$$

The solution of the logistic equation is the logistic curve

$$N(t) = \frac{N_{\infty}}{1 + \left[\frac{N_{\infty}}{N_0} - 1\right]e^{-\lambda_0 t}}.$$
 (6)

We can define the end time of the epidemic when the number of cases reaches the fraction 1-p of total cases. The solution of the equation gives and

$$(1-p)N_{\infty} = \frac{N_{\infty}}{1 + \left[\frac{N_{\infty}}{N_0} - 1\right]e^{-\lambda_0 t}},\tag{7}$$

where $p \ll 1$ is a very small fraction. The solution for very small p values is then given by

$$t = \frac{1}{\lambda_0} \log \left[\frac{N_\infty}{N_0} - 1 \right] - \frac{1}{\lambda_0} \log p \tag{8}$$

Next, we show that this is an excellent approximation to the actual data, which we obtained from the JHU site[3].

2 Deaths

First, we can do a post-fact analysis of the data from Hubei Province. We have the confirmed cases and deaths data, which both show logistic growth. Both of them have some subjective elements, and never the less the deaths data seems to be more objective. In Fig.1, we show the daily growth rate of death cases. Since the reporting of death cases is uneven, deaths are sometimes reported a day later. Therefore we calculate the daily rate from the average of two consecutive days $\lambda_n = 0.5 \cdot (N_n - N_{n-2})/N_n$. The first data point is January 24, 2020, three days after the lockdown of Wuhan. The daily growth rate is fluctuating around 25-30%, which seems to be the "natural" rate of growth in the absence of public actions. Then 14 days after the lockdown at about 500 deaths, the daily growth rate drops to about 10%. Then a linear trend sets in with some fluctuations due to some errors in the reporting. The slope of the linear fit is $\lambda_0/N_\infty = 4.6 \pm 0.4 \cdot 10^{-5}$, $\lambda_0 \approx 0.146$, and $N_\infty \approx 3158$.

We repeat the same analysis for the data from Italy and show the result in Fig.2. The Italian data starts on February 25, 2020. The beginning of the data fluctuates around 20-30% daily growth. The lockdown of Lombardy starts on March 9, at about 463 deaths, and the linear trend can be observed starting from this date. The slope of the linear trend is $\lambda_0/N_\infty = 4.05 \pm 1.13 \cdot 10^{-5}, \, \lambda_0 \approx 0.243, \, {\rm and} \, N_\infty \approx 5996.$

Remarkably, the slope of the curve is the same in Hubei and Italy within the error of fitting. It seems that the lockdown has a similar effect in both cases. The daily rate of growth of deaths appears to be falling 4% for each 1000 new deaths in Italy and the daily growth rate will be reduced from roughly 24% to zero after 6000 deaths. In Hubei the initial daily growth rate of 12% fell to zero after about 3000 deaths. It seems that this is a universal consequence of lockdown, which needs further explanation.

3 Length of the outbreak

Using the fitted parameters, we attempt to calculate the expected end of the outbreak in Italy. First, we validate the model on the Hubei data. In Fig.1, the linear regime starts when the number of deaths is $N_0=549$ on February 5. Today, 40 days later, on March 17, the number of deaths is 3111 in Hubei, which is 98.5% of the expected deaths $N_{\infty}=3158$ or p=0.015. Using these numbers (8) yields t=39.44 days, which is a very precise estimate for the length of the process. In Italy, the linear regime starts on March 9, and $N_0=463$. Using the same p=0.015 and the number of ultimate deaths $N_{\infty}=5996$ (8) yields t=27.49 days that is April 6. The p=0.001 milestone will be reached around April 15, and the lockdown can probably be partially lifted.

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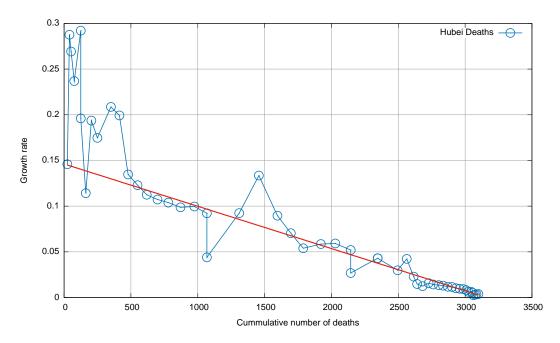


Figure 1: Daily growth rate of deaths as a function of the cumulative number of deaths in Hubei Province, China.

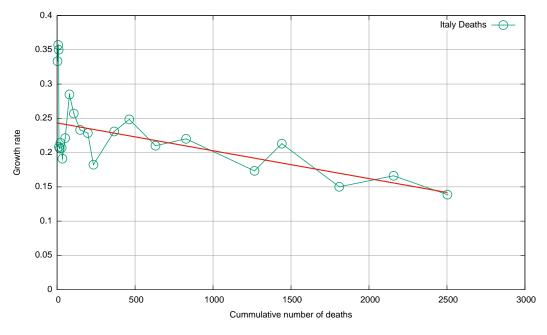


Figure 2: Daily growth rate of deaths as a function of the cumulative number of deaths in Italy.

References

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- [2] Pastor-Satorras, Romualdo and Vespignani, Alessandro. Epidemic Spreading in Scale-Free Networks, Phys. Rev. Lett. 86, 14, 3200–3203 (2001) https://link.aps.org/doi/10.1103/PhysRevLett.86.3200
- [3] https://coronavirus.jhu.edu/