

Discussion about *COVID-19: data, modeling, and communication* Alessandro Vespignani Northeastern University

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Statistical methods and models for complex data
800 years of research to understand a complex world

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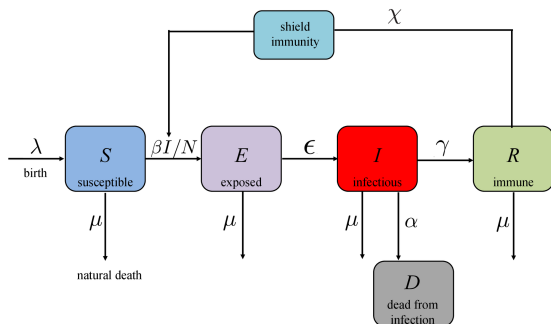
21 Settembre 2022 - Padova

Comment 1 - The theoretical model

1. Get the experts talking

Fact: At the beginning of the pandemic phase, since the availability of public data, a large field of experts produced analysis and interpretation using their own skills.

Comment: However only a limited set of experts used appropriate models for epidemiological diffusion, starting from compartmental models to more complex methods.



Comment 1 - The theoretical model

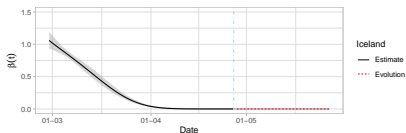
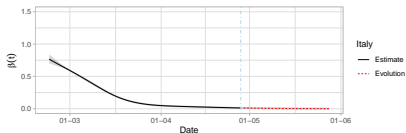
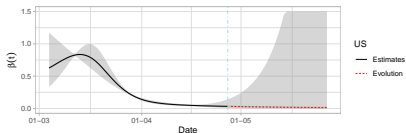
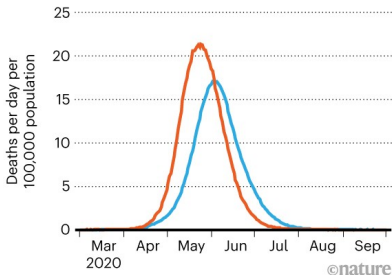
Simulations [1] often produce scenarios far from what happens in the real world (but this is however fine).

Most of the future contagion estimates was presented without any measure of uncertainty (i.e. interval confidence; and this is not fine).

SIMULATION SHOCK

A model by Imperial College London in mid-March predicted a total of more than 500,000 UK deaths from COVID-19, and more than 2.2 million in the United States if no action was taken to stop the virus spreading in those countries.

— United Kingdom — United States



Comment 2 - GIGO [Garbage in - Garbage out]

2. *Your results is good as your data*

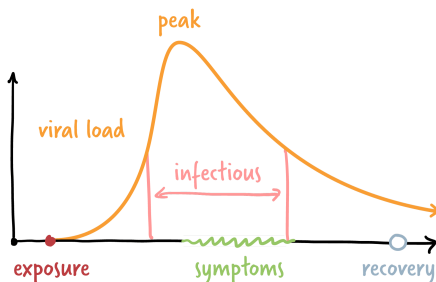
Fact: The availability of public data (number of infected, deaths, etc... with temporal and spatial profile) allows easy use of them for data analysis.

Comment: All that glitters is not gold. Issues about the quality and reliability of public data affect the results by producing biased estimations and bad predictions.



Comment 2 - GIGO [Garbage in - Garbage out]

For example, the daily numbers of occurrences related to COVID-19 are time-lagged (with respect to the date of diagnosis). Without account for that, results are biased [2].



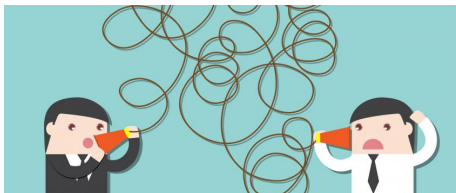
In fact, the main biases are due to problems in data collection and management [3, 5]. The reliability of future scenarios is also affected by prediction length [4].

Comment 3 - The communication

3. The problem of communication is the illusion that it takes place.

Fact: The communication about COVID-19 is often misunderstood.

Comment: Most part of people is confused by the COVID-19 numbers and statistics.



Comment 3 - The communication

There are some pieces of evidence:

- COVID-19 news have always a negative tone [6] and it implies a certain difficult to be followed;
- Authorities and agencies sometimes failed to communicate COVID-19 risks and decisions [7];
- The mathematical COVID-19 language (rates, percentages, trends, ...) is hard to be understood for someone [5].

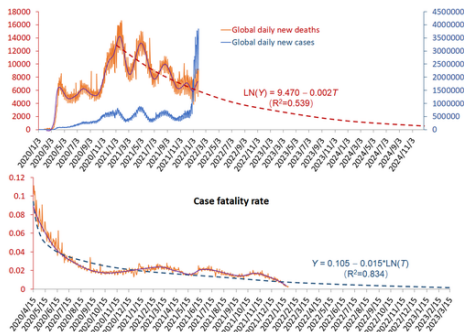


Comment 4 - What's the next?

4. History teaches, but has no pupils.

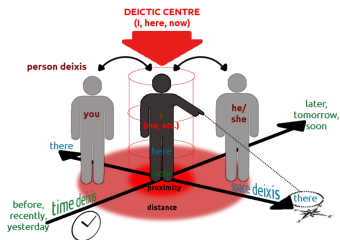
Fact: One week ago, the WHO declares that we are close to the end of the pandemic.

Comment: The lessons learned could help in designing effective public responses for constraining future waves of COVID-19 worldwide.



Comment 4 - What's the next?

- COVID-19 provides a modern example of why the classic mantra of “person, place, and time” in epidemiology is essential. [5]



- Both the global and national public health communities must remain in constant dialogue with governments.
- As for other sectors (i.e. nuclear energy) States and WHO should promote international networks and protocols for data sharing and analysis.

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