

De verspreiding van COVID-19 begrijpen: de rol van AI

Robert Kooij 2 November 2020



Who am I ?





1988 - 1993



1994 - 1996



Royal Dutch Telecom

1997 - 2003



2005 - ...

2003 - 2018



2018 - 2020



Do you know who this is?



- Prof. Neil Ferguson
 - Mathematical epidemiologist at Imperial College London
 - Scientific Advisor to UK Government



Imperial College COVID-19 Response Team

Report 9: Impact of non-pharmaceutical interventions (NPIs) to reduce COVID-19 mortality and healthcare demand

Neil M Ferguson, Daniel Laydon, Gemma Nedjati-Gilani, Natsuko Imai, Kylie Ainslie, Marc Baguelin, Sangeeta Bhatia, Adhiratha Boonyasiri, Zulma Cucunubá, Gina Cuomo-Dannenburg, Amy Dighe, Ilaria Dorigatti, Han Fu, Katy Gaythorpe, Will Green, Arran Hamlet, Wes Hinsley, Lucy C Okell, Sabine van Elsland, Hayley Thompson, Robert Verity, Erik Volz, Haowei Wang, Yuanrong Wang, Patrick GT Walker, Caroline Walters, Peter Winskill, Charles Whittaker, Christl A Donnelly, Steven Riley, Azra C Ghani.

Prediction: based upon mathematical models

No action taken



500.000 deaths in UK!



MATHEMATICS

exponential growth R₀

moving averages

flattening the curve

the R-number



Importance of understanding spread

- Intervention policy
- Healthcare allocation
- Economic impact
- Mental well-being



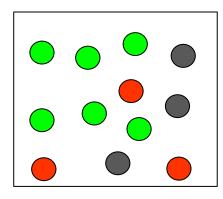
Two types of approaches

- Approach 1: Compartmental models
 - Population divided into categories

• Equation based

$$\frac{dS}{dt} = -\beta SI$$
$$\frac{dI}{dt} = \beta SI - \gamma I$$
$$\frac{dR}{dt} = \gamma I$$

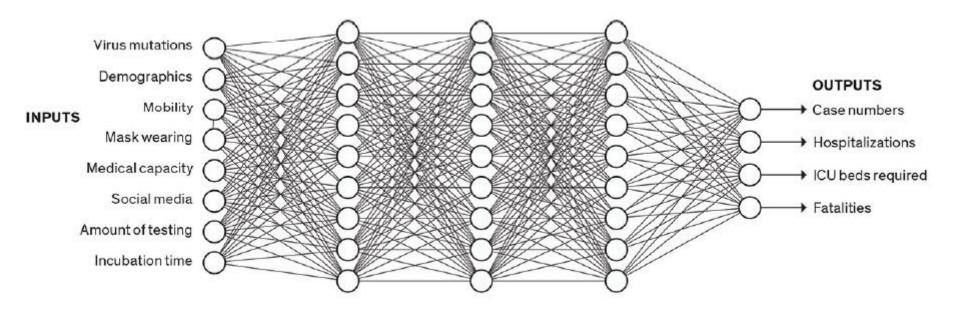
• Agent based





Two types of approaches

• Approach 2: AI-driven models





Overview

- Compartmental models: Equation Based
- Compartmental models: Agent Based
- AI-driven models
- Comparing the approaches
- Equation based approach on networks
- TU Delft COVID-19 Digital Campus
- Wrap-up





• SIR - model

- No births or immigration
- Curing leads to immunity
- Fixed infection rate per day: β
- Fixed recovery time: $1/\gamma$
- People are well-mixed

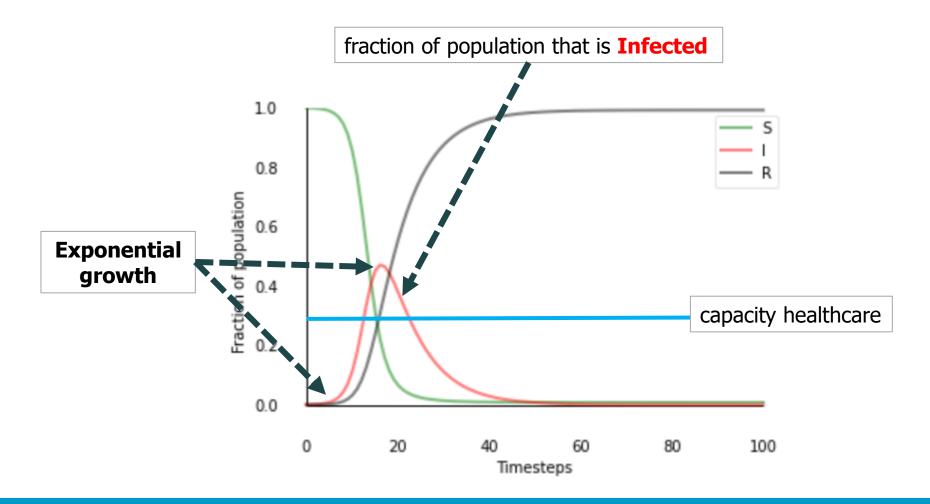
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- S: fraction of population that is **Susceptible**
- I: fraction of population that is **Infected**
- R: fraction of population that is **Removed**

$$\frac{dS}{dt} = -\beta SI$$
$$\frac{dI}{dt} = -\gamma I + \beta SI$$
$$\frac{dR}{dt} = \gamma I$$

W.O. Kermack and A.G. McKendrick, 1927, A contribution to the mathematical theory of epidemics, Proc. R. Soc. Lond. A, 115, 700–721.

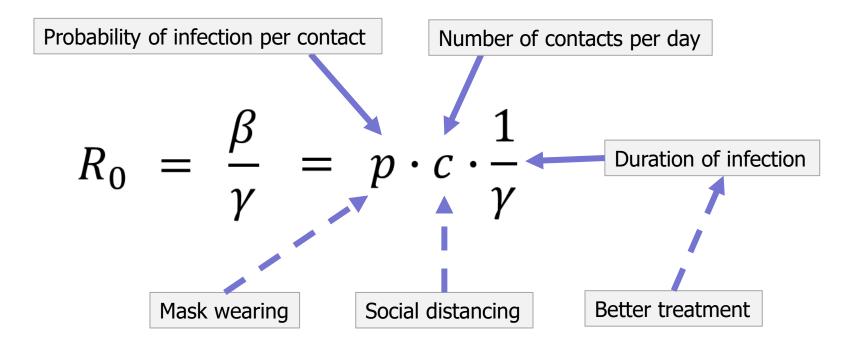




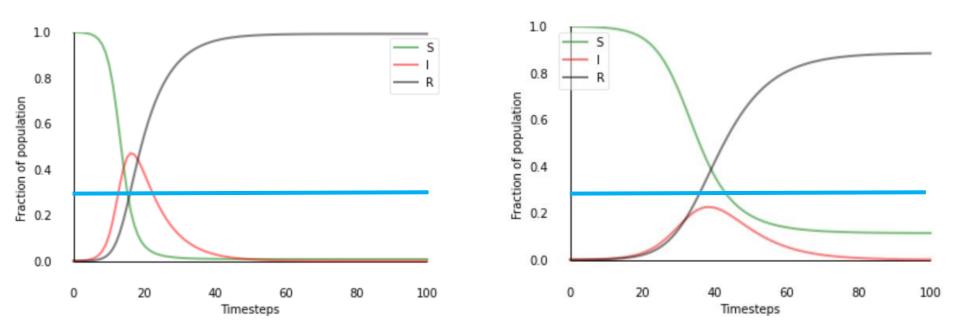


- Growth governed by R₀: **basic reproduction number**
- R₀: # of new infections due to first infection
 - $R_0 > 1$ exponential growth # of infections
 - $R_0 = 1 \#$ of infections stays constant
 - $R_0 < 1 \#$ of infections decreases fast to 0
- Flattening the curve: lowering R₀



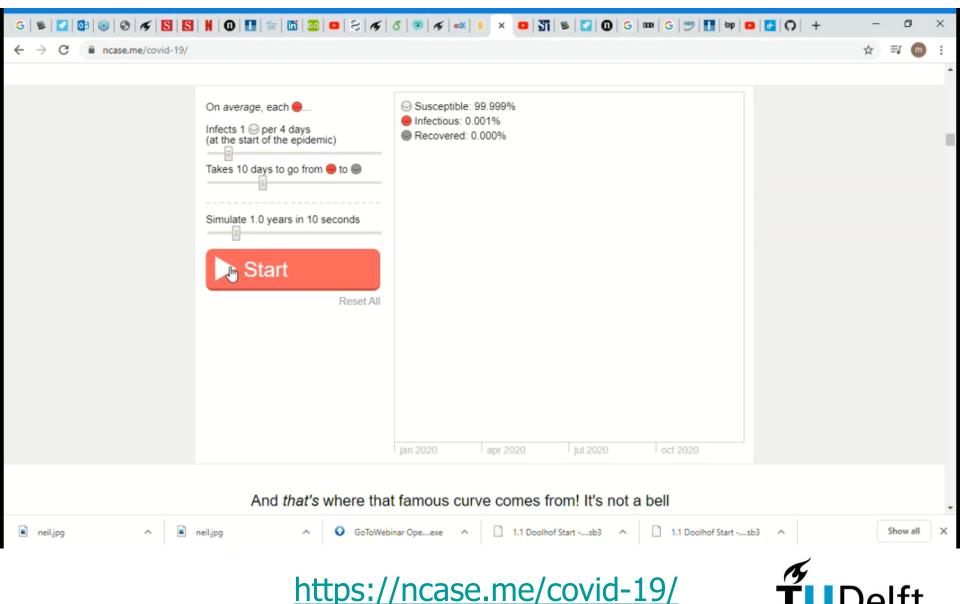






• Reduce number of contacts per day by 50%

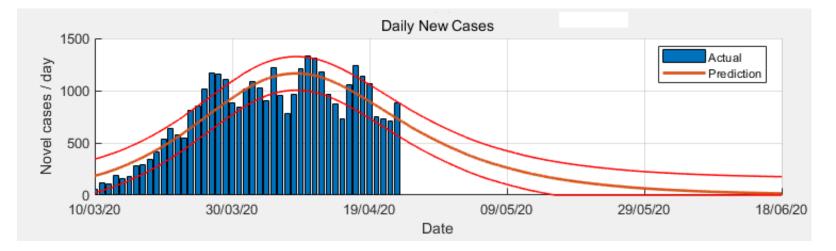




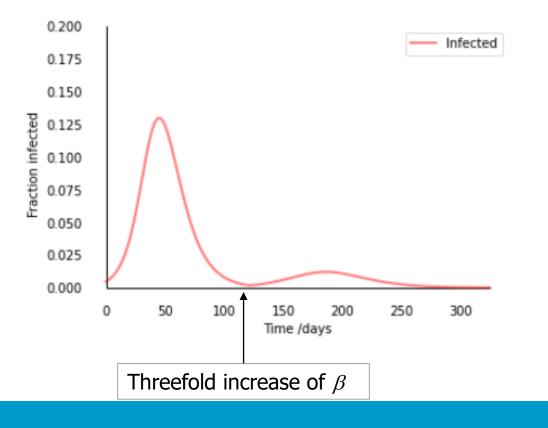
Predictive Monitoring of COVID-19

Jianxi Luo, Singapore University of Technology and Design, May 2020

- Use data on daily number of infections
- Estimate β and γ by using AI



• Infection rate β increases after relaxation interventions

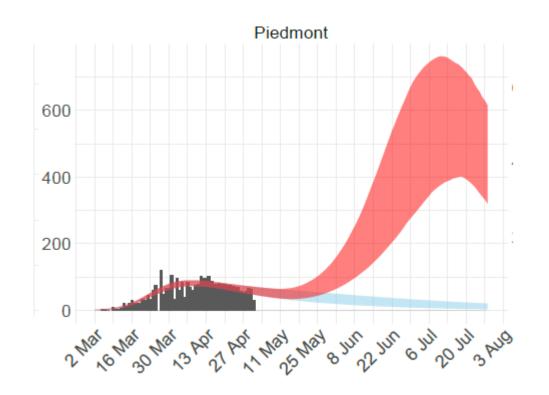






4th May Imperial College COVID-19 Response Team

Report 20: Using mobility to estimate the transmission intensity of COVID-19 in Italy: A subnational analysis with future scenarios



Mobility held constant Increased mobility: 40% return to pre-lockdown level





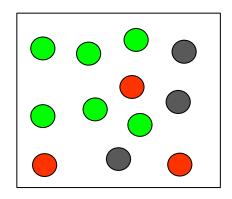
?

- Parameters are not constant
- Availability and quality of data
- Undetected cases
- Infection delay
- Duration immunity
- Seasonal variation
- Heterogeneous mixing
- Human factors





- Each individual (agent)
 - Moves around
 - Is in one of the **S-I-R** states



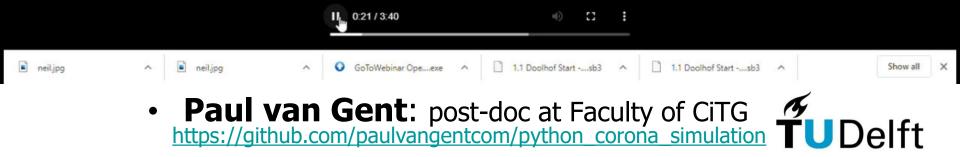


← → C ① Not secure | paulvangent.com/covid/Covid_Compilation_reinfection.mp4

Baseline simulation

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2000 people 3% infection risk when near infected person baseline mortality: 2% at-risk age: 55+ critical risk age: 75+ healthcare capacity: 300 beds



Modelling transmission and control of the COVID-19 pandemic in Australia

Sheryl L. Chang¹, Nathan Harding¹, Cameron Zachreson¹, Oliver M. Cliff¹, and Mikhail Prokopenko^{1,2,*}

¹ Centre for Complex Systems, Faculty of Engineering, University of Sydney, Sydney, NSW 2006, Australia

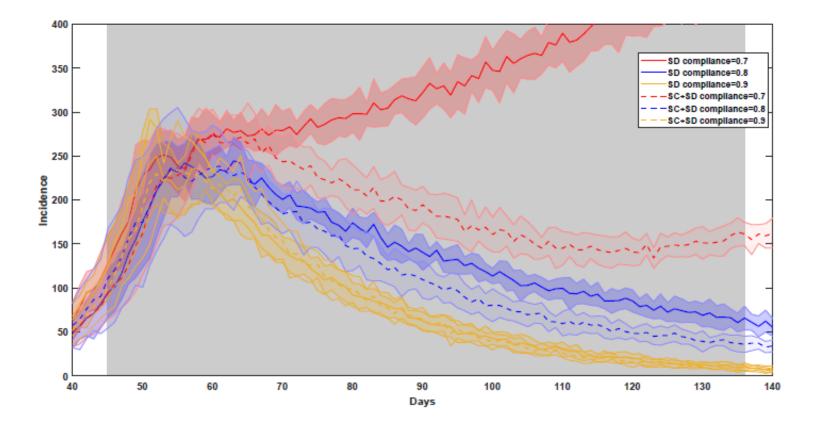
² Marie Bashir Institute for Infectious Diseases and Biosecurity, University of Sydney, Westmead, NSW 2145, Australia

Corresponding author: mikhail.prokopenko@sydney.edu.au (ORCID: 0000-0002-4215-0344)

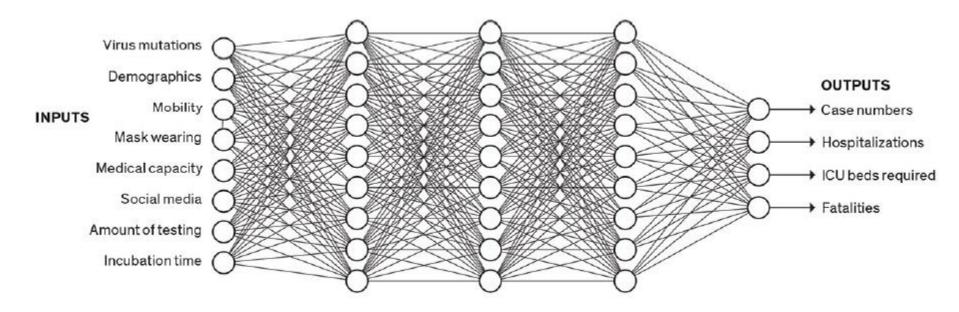
- Virtual Australia with 24 million agents
 - Demography
 - Mobility
 - Disease



- Requires massive computation power
- Hard to calibrate
- Once the model is calibrated
 - Relate interventions to changes in infection rates
 - Assess impact of several interventions









394,088 views | Feb 5, 2020, 12:14pm EST

AI Predicts Coronavirus Could Infect 2.5 Billion And Kill 53 Million. Doctors Say That's Not Credible, And Here's Why





Received August 21, 2020, accepted August 25, 2020, date of publication August 28, 2020, date of current version September 14, 2020. Digital Object Identifier 10.1109/ACCESS.2020.3019989

DeepCOVIDNet: An Interpretable Deep Learning Model for Predictive Surveillance of COVID-19 Using Heterogeneous Features and Their Interactions

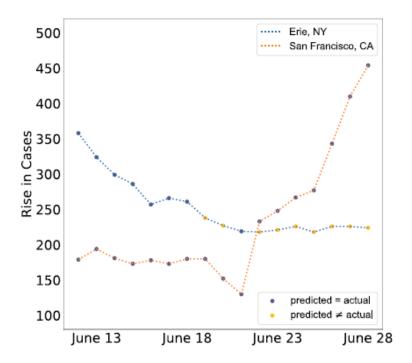
ANKIT RAMCHANDANI^{®1}, CHAO FAN^{®2}, AND ALI MOSTAFAVI^{®2}

¹Department of Computer Science and Engineering, Texas A&M University, College Station, TX 77840, USA ²Zachry Department of Civil and Environmental Engineering, Texas A&M University, College Station, TX 77840, USA



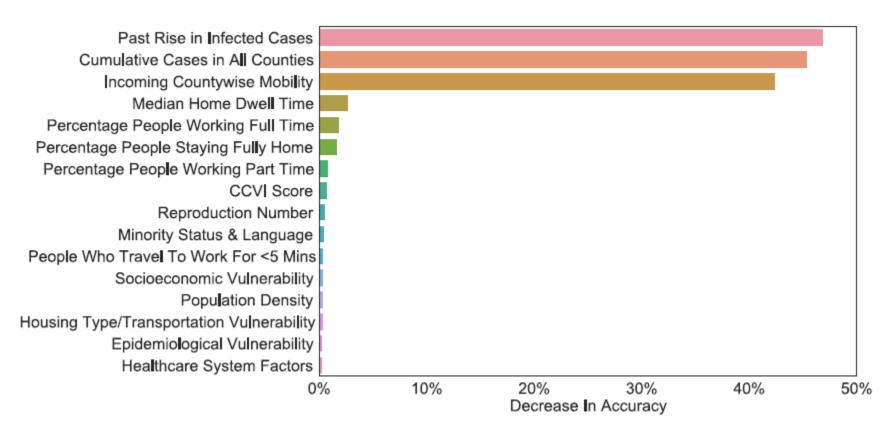
- Obtain data
- Selecting features
 - Population attributes
 - Population activities
 - Mobility
 - Disease spread attributes
- Build Deep Learning model (LSTM)
- Evaluate performance of model





• Predictive performance: 7 days





• Identification most influential features



- Compartment Models
 - Relatively simple tool for long-term forecasts
 - Parameter estimation
- Agent Based Models
 - Lot of work to build
 - Tricky to calibrate
- AI-driven models
 - Short-term predictions
 - Unreliable when conditions change

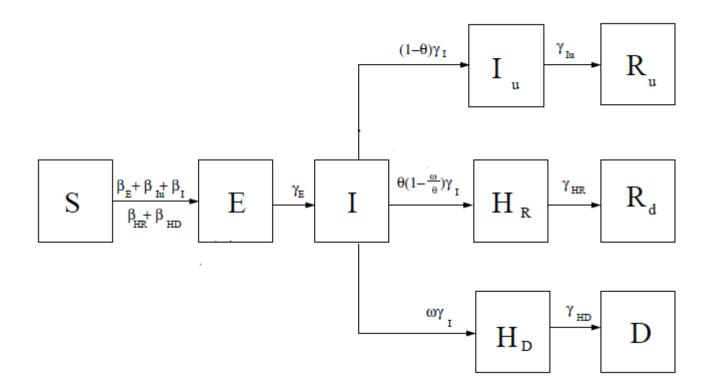


- Hybrid models
- More compartments
 - Undetected cases
 - Hospitalization
 - Incubation delay
- Time dependent control measures
- Parameters estimated through AI

Commun Nonlinear Sci Numer Simul. 2020 Sep; 88: 105303.

Mathematical modeling of the spread of the coronavirus disease 2019 (COVID-19) taking into account the undetected infections. The case of China <u>B. Ivorra, ^a M.R. Ferrández, ^b M. Vela-Pérez, ^a and A.M. Ramos^{a,*}</u>

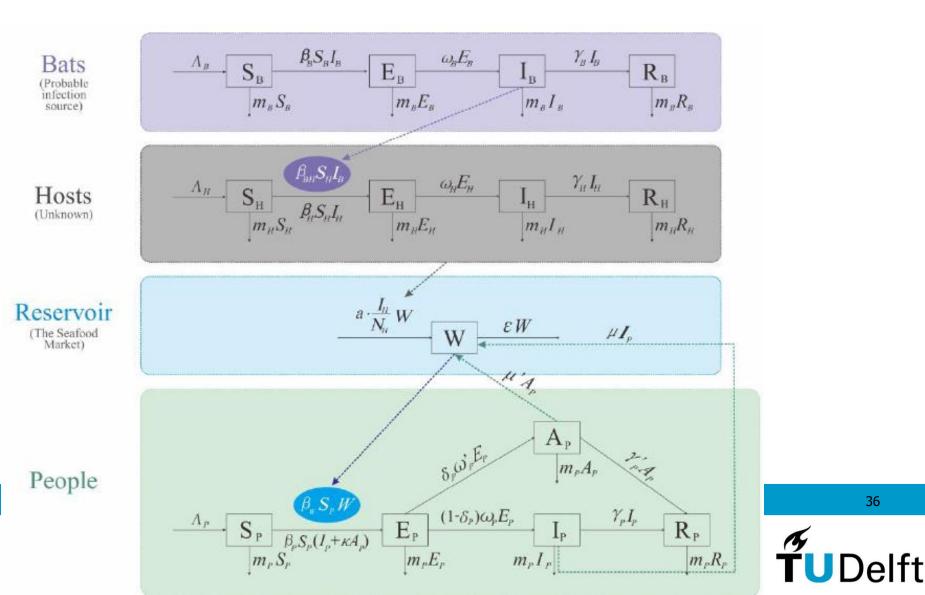
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$$\begin{split} \frac{\mathrm{d}S}{\mathrm{d}t}(t) &= -\frac{S(t)}{N} \left(m_E(t)\beta_E E(t) + m_I(t)\beta_I I(t) + m_{I_u}(t)\beta_{I_u}(\theta(t))I_u(t) \right) \\ &- \frac{S(t)}{N} \left(m_{H_R}(t)\beta_{H_R}(t)H_R(t) + m_{H_D}(t)\beta_{H_D}(t)H_D(t) \right), \\ \frac{\mathrm{d}E}{\mathrm{d}t}(t) &= \frac{S(t)}{N} \left(m_E(t)\beta_E E(t) + m_I(t)\beta_I I(t) + m_{I_u}(t)\beta_{I_u}(\theta(t))I_u(t) \right) \\ &+ \frac{S(t)}{N} \left(m_{H_R}(t)\beta_{H_R}(t)H_R(t) + m_{H_D}(t)\beta_{H_D}(t)H_D(t) \right) - \gamma_E E(t) + \tau_1(t) - \tau_2(t), \\ \frac{\mathrm{d}I}{\mathrm{d}t}(t) &= \gamma_E E(t) - \gamma_I(t)I(t), \\ \frac{\mathrm{d}I_u}{\mathrm{d}t}(t) &= (1 - \theta(t))\gamma_I(t)I(t) - \gamma_{I_u}(t)I_u(t), \\ \frac{\mathrm{d}H_R}{\mathrm{d}t}(t) &= \theta(t) \left(1 - \frac{\omega(t)}{\theta(t)} \right) \gamma_I(t)I(t) - \gamma_{H_R}(t)H_R(t), \\ \frac{\mathrm{d}H_D}{\mathrm{d}t}(t) &= \gamma_{H_R}(t)H_R(t), \\ \frac{\mathrm{d}R_d}{\mathrm{d}t}(t) &= \gamma_{H_R}(t)H_R(t), \\ \frac{\mathrm{d}R_d}{\mathrm{d}t}(t) &= \gamma_{H_D}(t)H_D(t). \end{split}$$





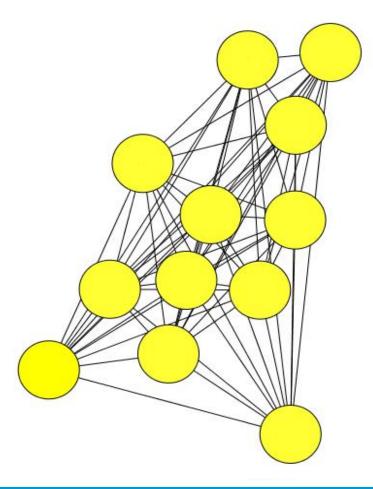


4th May Imperial College COVID-19 Response Team

Report 20: Using mobility to estimate the transmission intensity of COVID-19 in Italy: A subnational analysis with future scenarios

occur, the number of deaths are read is likely to be considerably lower in both scenarios. It should be noted that in our model we do not account for cross-region movement, which, given increased mobility, is likely to increase infections and subsequently deaths, in regions not experiencing major epidemics.





https://www.nas.ewi.tudelft.nl/index.php/coronavirus

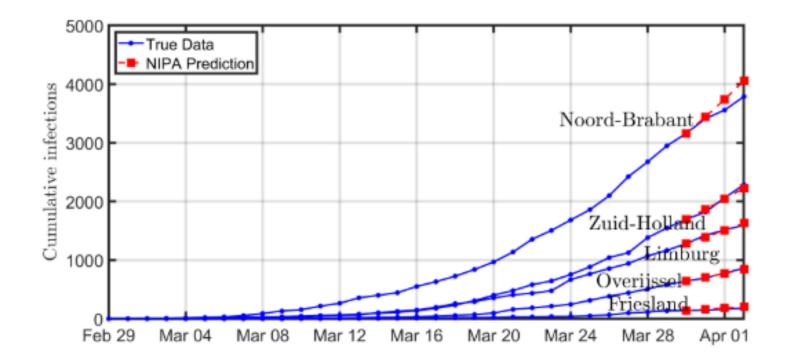


• NIPA = Network Inference-based Prediction Algorithm

For every province estimate

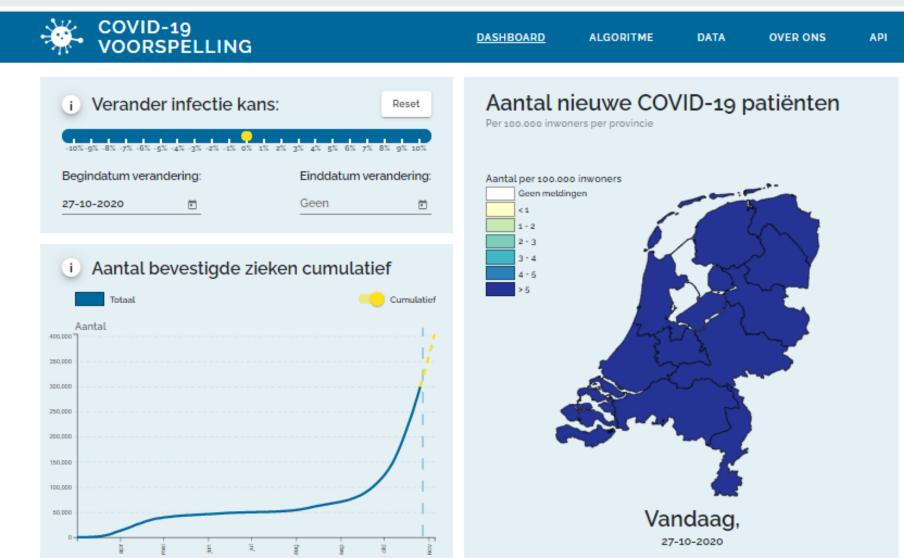
 Self-infection and curing rate
 Infection to other provinces
 156 parameters!
 Estimated using AI







networkdatascience.ewi.tudelft.nl



TU Delft – COVID-19 Digital Campus

Home

Mobility

Wellbeing

Contact Networks

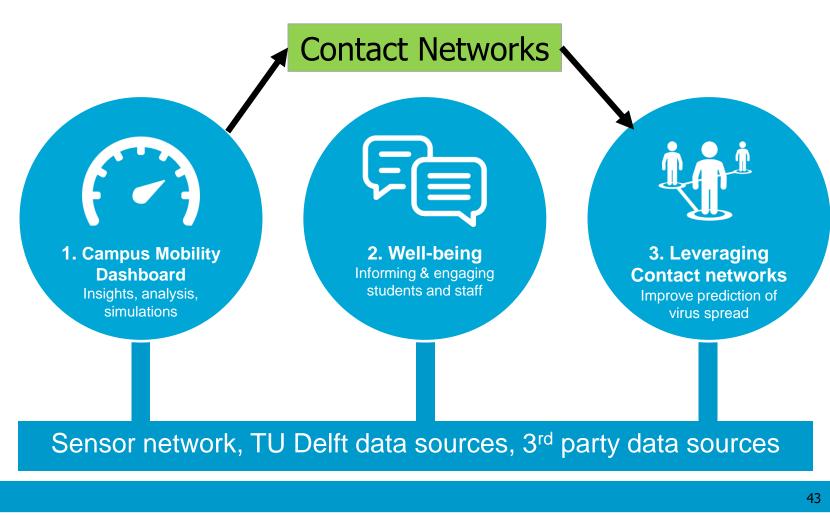
About us

COVID-19 Digital Campus

A living lab for digital technologies



TU Delft – COVID-19 Digital Campus





https://www.tudelft.nl/en/covid/

Wrap-up

- Importance of understanding COVID-19 spread
- Equation based approach
- Agent Based approach
- AI-driven approach
- Hybrid models
- COVID-19 Digital Campus



Wrap-up



Report 34: COVID-19 Infection Fatality Ratio: Estimates from

Seroprevalence

29 October 2020

Nicholas F Brazeau¹, Robert Verity¹, Sara Jenks², Han Fu¹, Charles Whittaker¹, Peter Winskill¹, Ilaria Dorigatti¹, Patrick Walker¹, Steven Riley¹, Ricardo P Schnekenberg³, Henrique Hoeltgebaum⁴, Thomas A Mellan¹, Swapnil Mishra¹, H Juliette T Unwin¹, Oliver J Watson¹, Zulma M Cucunubá¹, Marc Baguelin¹, Lilith Whittles¹, Samir Bhatt¹, Azra C Ghani¹, Neil M Ferguson¹, Lucy C Okell¹⁺.



Imperial College COVID-19 response team

Thanks for your attention!



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