



Scenario-driven forecasting: Lessons Learned from Modeling the COVID-19 Pandemic

Men is beginnen forecasten: scenario denken
= populair geworden door 2 grote events in de jaren 70
1. De oliecrisis

Kristof Decock & prof. dr. Bart Van Looy



0. Introduction

Major differences between foresight and forecasting (Cuhls, 2003)

Foresight (Scenario Thinking)

Basic points, needs, research questions are still open and looked for as part of the foresight process

More **qualitative** than quantitative

Looks for '**information**' about the future for priority-setting

Brings people together for **discussions** about the future and for networking, makes use of the distributed intelligence

Criteria for assessment and preparation for **decisions**

Communication about the future as an objective

Long-, medium- and short-term orientation with implications for today

Finds out if there is **consensus** on themes

'Experts' and other participants, very dependent on **opinions**

Forecasting

Basis points, topics and research questions have to be clarified in advance

More **quantitative** than qualitative

Questions what the future in the selected area might look like

More result-oriented, can also be performed by **individual** people or in single studies (depends on methodology)

Not necessarily assessments, different **options** and choices or the preparation for decisions

Describes future **options**, results more important than the communication aspects

Long-, medium- and short-term orientation as well as the path into the future are the major points

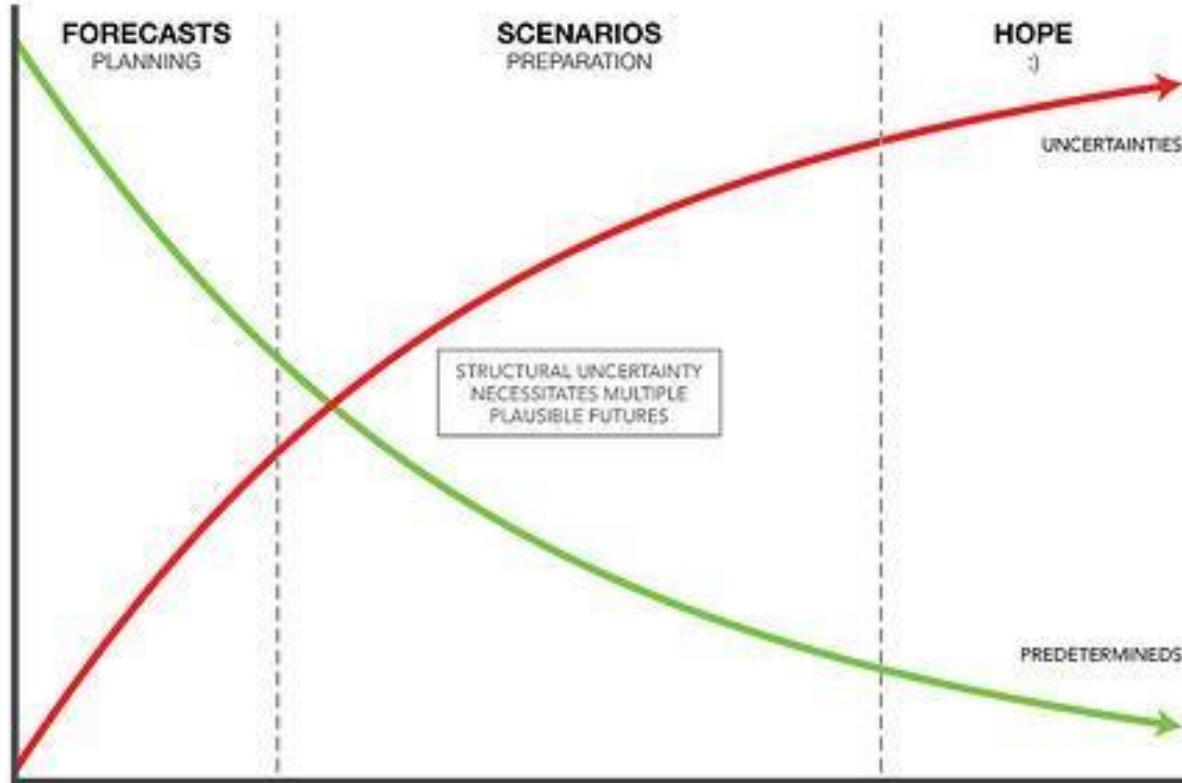
No information about consensus necessary

Mainly 'expert' and/or strict methodologies, less dependent on opinions

Uncertainty rises with time

FORECASTS VS. SCENARIOS

SCAFFOLDING INTO THE UNCERTAIN FUTURE



Hier kom ik bij het ding van daarnet
 > x-as: toename onzekerheid. Als het weinig onzeker is kun je goed voorspellen
 bv apple: heeft een concreet getal voor ogen hoeveel ze bij de christmas sales gaan verkopen
 > Ga je naar meer onzekerheid gaat het moeilijker zijn om kwantitatief juist te zijn

Source: Based on Van der Heijden, K. (1997). Scenarios: The Art of Strategic Conversation.
<https://medium.com/larimer/stories-about-the-future-c9e3bf6b1f52>

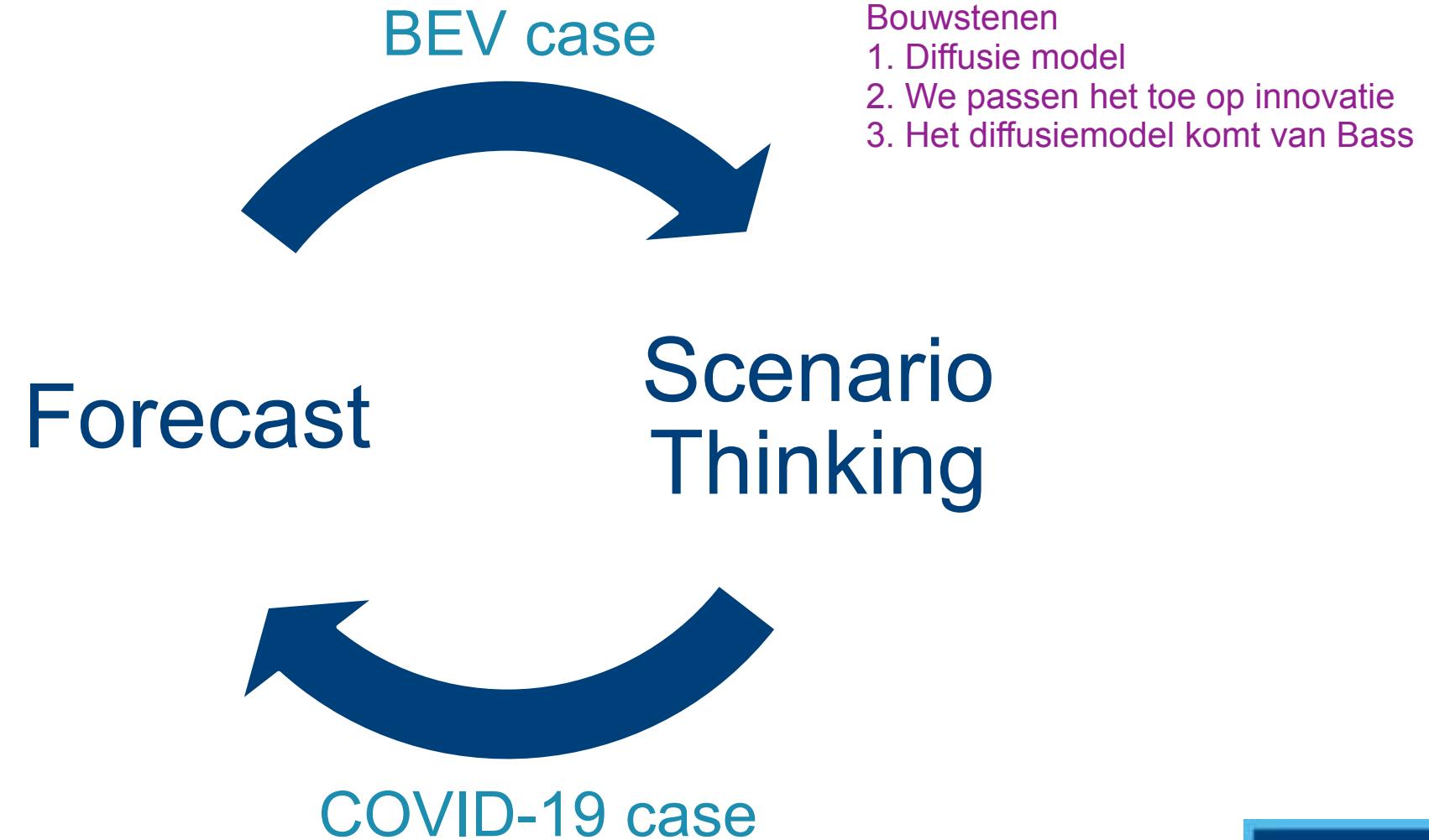


Can (quantitative) forecasting models inform (qualitative) scenario development efforts? And vice versa?

Literatuur is opgesplitst tussen forecasting en foresighting

- > je kunt het niet kwantificeren, je moet het kwalitatief maken. De mensen die beweren dat het wel kunnen zijn arrogante mensen die denken dat de wereld van hun is
- > mensen die denken dat ze dat wel kunnen

Can both disciplines inform each other?



1. Diffusion of an innovation: the Bass Model

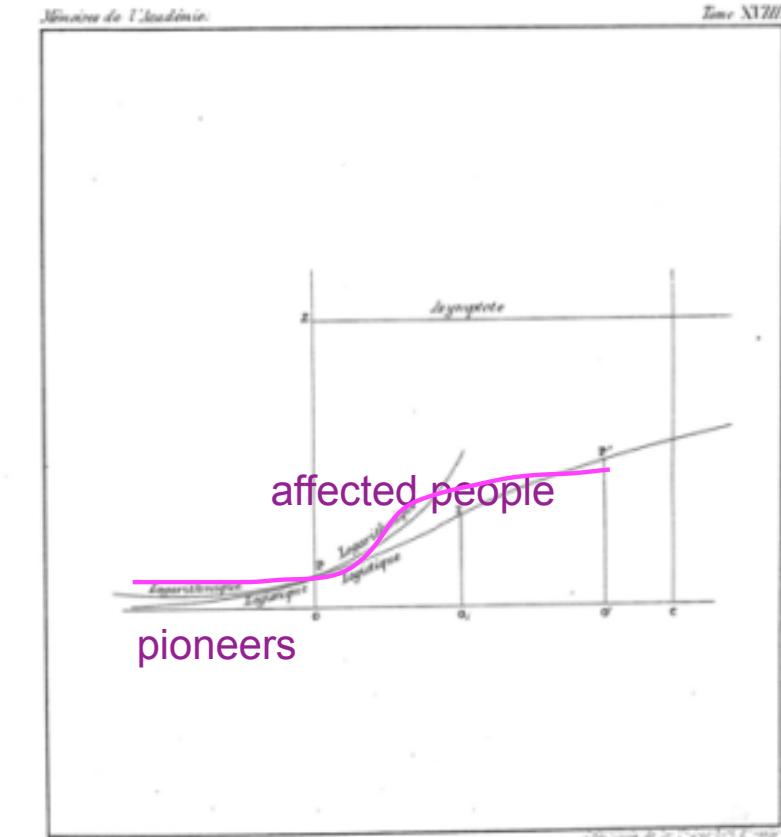
Als een nieuw product op de markt komt volgt dat een S-curve

> begin = traag: er zijn maar een # mensen geïnteresseerd in het product

> zodra de eerste persoon bv met een tesla rond rijdt rijden wij eens graag mee (en zijn onder de indruk) > aansteken mensen

Introduction

- Since the seminal work of Verhulst on population growth, the logistic equation has been introduced by adjusting the exponential growth model with a maximum value for the population (Verhulst, P.-F. (1838). *Note on the law of population growth*).
- Numerous studies in a variety of management disciplines suggest that growth patterns of cumulative sales of new products (over time) result in a S-shape curve (Bass, 1969; Mahajan et. al, 1990; Chandrasekaran & Tellis 2007).
- The logistic function could be used to model life cycles.



Original image of a logistic curve, contrasted with a logarithmic curve.
Source: Verhulst, P.-F. (1845): *Recherches mathématiques sur la loi d'accroissement de la population*.

The Bass Model (Bass, 1969)

Het model beschrijft een S curve

- The Bass model is one of the best-known quantitative models in marketing/innovation studies, describing the diffusion of an innovation (e.g. Mahajan et. al, 1990; Chandrasekaran & Tellis 2007, Massiani & Gohs, 2015)
- The model has been widely adopted by modelers for a number of reasons, e.g., its simplicity and 'good' predictive ability (for an overview, see e.g. Parker, 1994; Chandrasekaran & Tellis 2007).
- The Bass model disentangles innovators from imitators regarding the initial adoption of an innovation (e.g. : Bass, 1969; Lekvall and Wahlbin, 1973; Mahajan et al., 1990; Bass, Krishnan, and Jain, 1994).
- Similarity in the underlying growth dynamics to be modelled:
 - Innovation: adoption followed by imitation
 - Epidemics: infection followed by contamination

The Bass Model (Bass, 1969)

In its discrete form, the Bass model can be written as:

Hoeveel mensen gaan op welk moment in de tijd een elektrische auto kopen

$$N_t = N_{t-1} + p(m - N_{t-1}) + q \frac{N_{t-1}}{m} (m - N_{t-1})$$

Als je zo'n S curve wilt schatten, wanneer wil je die schatten?

> in het aller begin bv e-staps: wordt dat een ding?

- als belegger wil je weten of het een take-off gaat hebben

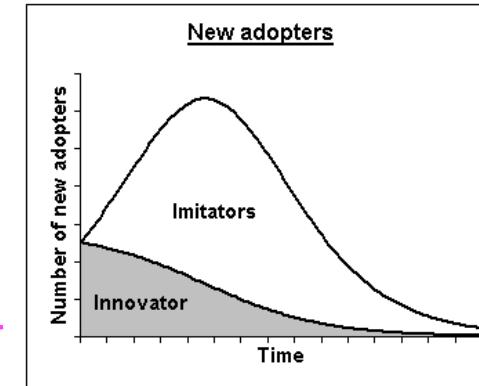
- dan zijn er heel veel onderzoekers die dit willen weten en die vroeg willen intekenen

Where:

- N_t = the number of adopters at time t (later: ICU occupation or total number of deceased),
- m = the potential market (later: the ultimate number of ICU occupation / deceased),
- p = coefficient of innovation (later: the infection parameter) and,
- q = coefficient of imitation (later: the intensity of contamination),
- $(m - N_{t-1})$ = non-adopters at the beginning of period t,
- (N_{t-1} / m) = the fraction who have already adopted.

In het aller begin: je ziet een # puntjes en dat lijkt omhoog te gaan, maar wat is omhoog gaan?

> dat weet je pas wanneer je in het buigpunt zit (maar dan is het te laat om te investeren)



Parameter estimation techniques

Waar we gaan eindigen met de electrische auto, gaan we pas weten zodra we dus aan het buigpunt zitten (waarbij de groei begint te vertragen)

- Notwithstanding sales data for only three observations would be sufficient to theoretically estimate the three parameters p, q and m (as illustrated in Bass, 1969, p. 224), this procedure may be hazardous because of the chaotic nature of early adoption data (Modis & Debecker, 1992).
- Scholars developed several approaches (e.g., OLS, MLE, NLS) to delineate a unique set of parameters (i.e. the one ‘optimal’ point estimate for m, p, q); however all methods have their limitations, drawbacks and biases and hence are not useful to generate forecasts during the growth phase of the diffusion curve (e.g., Heeler and Hustad, 1980; Schmittlein and Mahajan, 1982; Srinivasan and Mason, 1986; Mahajan et al., 1990; Venkatesan & Kumar, 2002; Chandrasekaran & Tellis, 2007).
- Therefore, Mahajan et. al (1990, p.9) conclude that:

“ [...] parameter estimation for diffusion models is primarily of historical interest; by the time sufficient observations have developed for reliable estimation, it is too late to use the estimates for forecasting purposes.”

Results

- Thus, ex-ante ‘forecasting’ the diffusion of new products (or novel technologies) by means of quantitative models is problematic for a number of reasons:
 - Diffusion models yield insights retrospectively.
 - Model estimations result in unstable and rudimentary predictions when only few ‘trend’ data are available (i.e. during the early stages of diffusion).
 - Parameter estimations are typically *not* used when the data does only contain pre-takeoff sales (Venkatesan et al., 2004).
- As such, scholars and practitioners alike tend to favor foresight efforts of a more qualitative nature (e.g. scenario development) to project foreseeable futures and their constituents.

2. Bass revisited: illustrated with the BEV-case

Insights from:

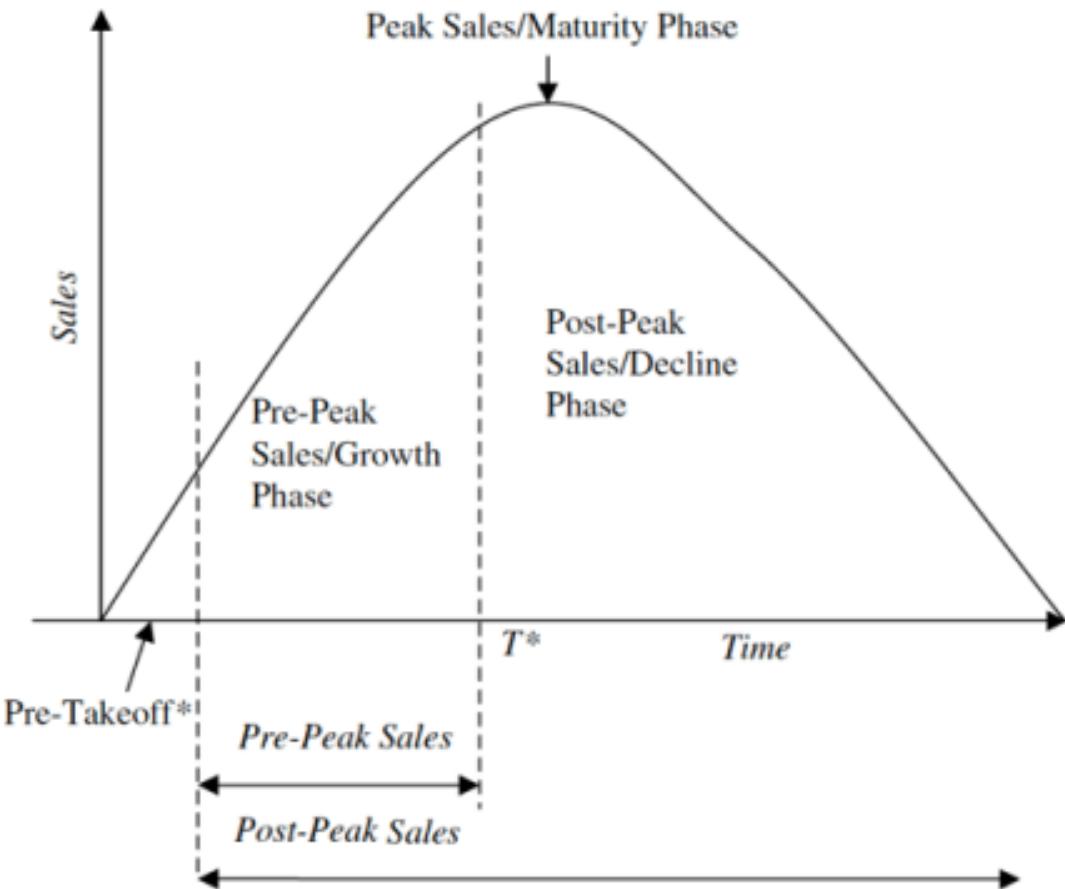
Decock, K., Debackere, K. and Van Looy, B. (2020). Bass Re-visited:
Quantifying Multi-Finality. *MSI working paper*, n° 2003.

Research question

Can (quantitative) forecasting models inform (qualitative) scenario development efforts in situations where only relatively short time series are available (i.e. during the pre-takeoff phase)?

Helpt het nu bij het nemen van beslissingen in hoge onzekere situaties?

Figure 1 Sample Diffusion Path of a New Product Innovation



*Typically not used in diffusion model estimation.

Source: Venkatesan, R., Krishnan, T. and Kumar, V. (2004)

Novel Heuristics connecting Scenario Thinking with Forecasting Logics (and assuming multi-finality)

- Starting from the Bass model, we develop an exhaustive, three-dimensional search space to assess a wide range of values for each parameter looking for local optima instead of one global optimum. (Scenario Thinking)
- The grid consists of 500.000 scenarios for each region (EU and US), i.e. 10 (m) x 250 (p) x 200 (q), with ranges consistent with parameters from literature:
 - $10\% \leq m \leq 100\%$
 - $0,00001 \leq p \leq 0,00250$
 - $0,01 \leq q \leq 2,00$
- A loss function is defined, based on the fit with actual diffusion patterns; the goodness-of-fit for each scenario is calculated as $R^2 = 1 - \frac{SS\ Error}{SS\ Observed}$
- Models with an $R^2 > 0,99$ were retrieved and labelled as the “*more likely* scenarios”, i.e. are best in explaining what we currently already observe in terms of market adoption (Connecting with Forecasting).

We gaan niet meer op zoek naar de S, maar gaan er vanuit dat het alle kanten kan uitgaan (the principle of multifinality), we stellen ons dus open voor een veelheid aan scenario's

Multi-Finality

- Buckley (1967): “Similar initial conditions may lead to dis-similar end-states” (Sociology and Modern System Theory, Prentice Hall).
- *Multifinality* holds that similar initial conditions may lead to dissimilar outcomes (Feiring & Lewis, 1987).

Data: Key figures about European and U.S. fleet

	Europe	United States
Number of cars (2016, mio)	259,7	113
New annual registrations (2016, mio)	15	6,3
Average age	11	11,6
Renewal period (in years)	17	18
Market share (2017) of BEV	0,12%	0,35%

(Sources: European Automobile Manufacturers Association & European Environment Agency; statista.com)

Data: Cumulative sales data on BEV in Europe and U.S.

Er waren dus maar 7 datapunten

> begin in 2011: door oprichting Tesla

Year	Stock BEV in Europe	Stock BEV in U.S.
2011	8.493	10.060
2012	22.479	24.710
2013	46.654	72.404
2014	84.509	135.820
2015	141.265	206.864
2016	205.581	293.595
2017	302.724	398.066

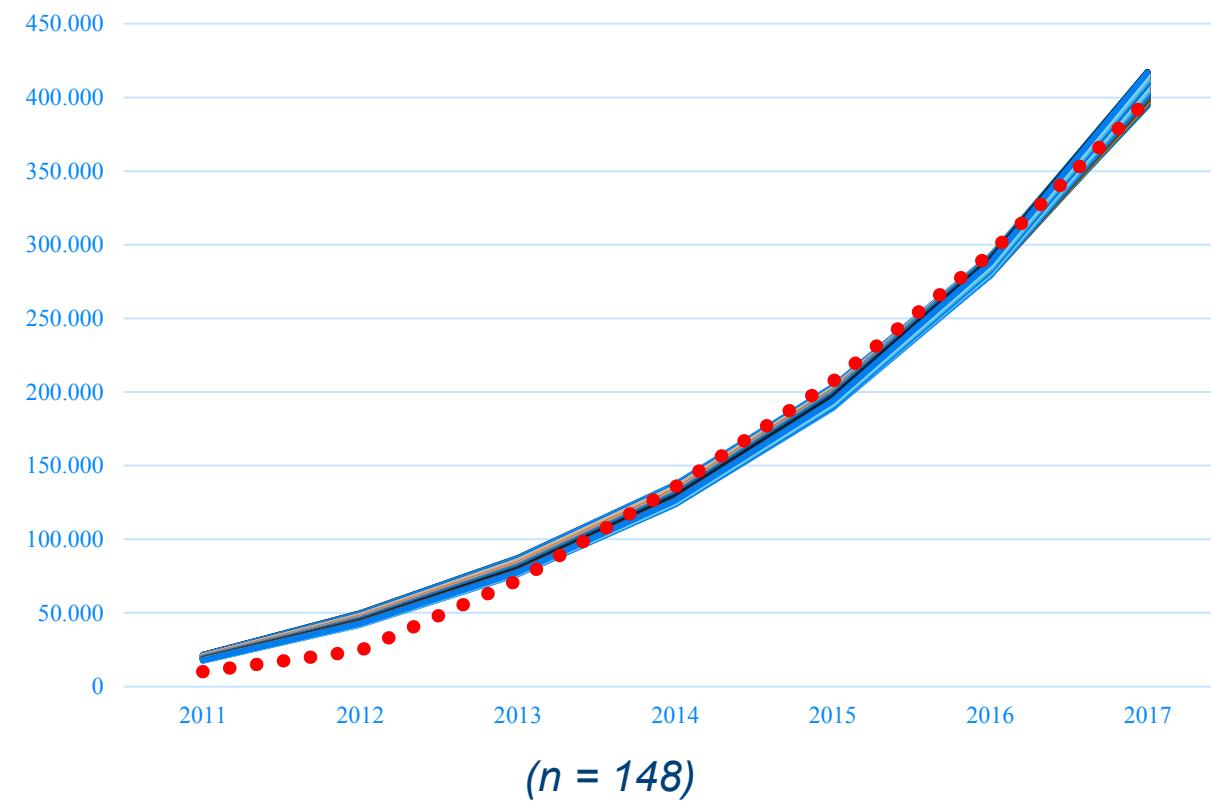
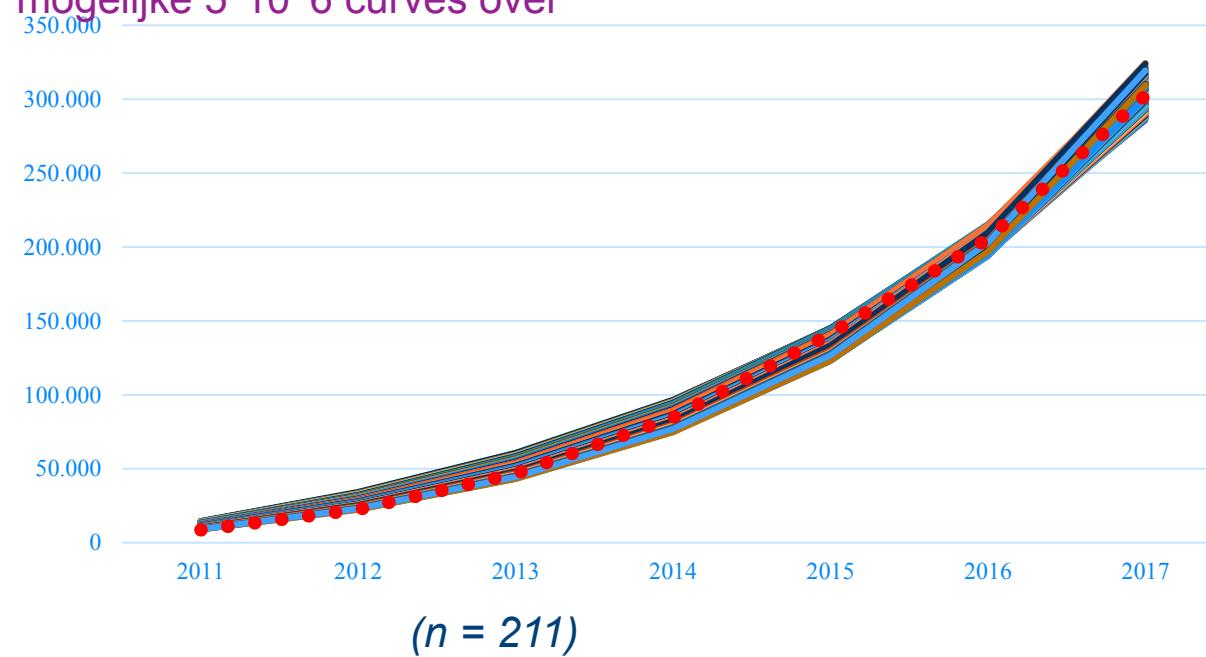
(Sources: European Environment Agency; U.S. Department of Energy – Energy Efficiency and Renewable Energy, hybridcars.com)

Goodness-of-fit of the *more likely* scenarios ($R^2 > ,9965$) in Europe (left) and U.S. (right)

Hiermee bereken je heel veel curves

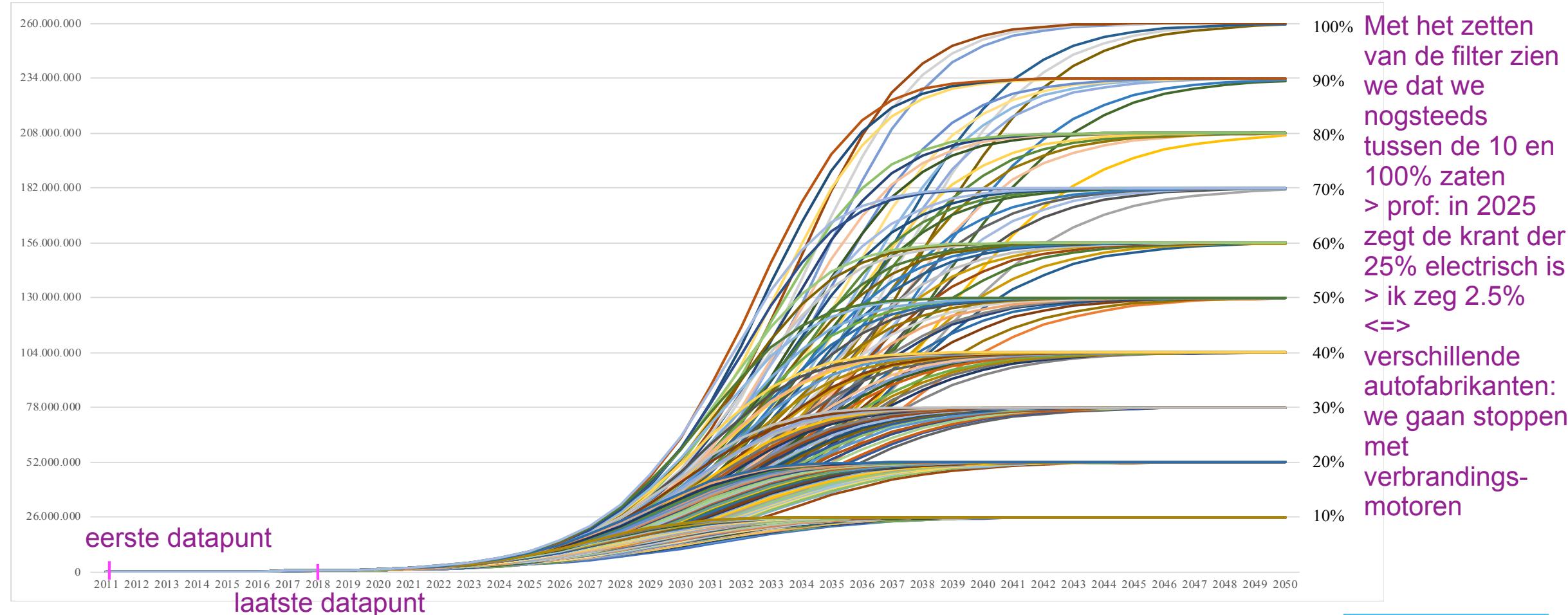
> je gaat alleen die curves bekijken die voor 90 of 99% fitten met wat we zeker weten

> goede nieuws: er bijven nog maar 150 van alle berekende mogelijke $5 \cdot 10^6$ curves over



More likely scenarios ($R^2 > ,9965$) for Battery EV in Europe (n = 211)

Als je de 200 curves dan bekijkt dan krijg je



Als we hetzelfde voen voor BEV voor BE als met covid:

Descriptive Statistics BEV in Europe, incl. 2010 ($R^2 > ,99$; n= 321)

Year	<u>Market Share</u>			<u>New BEV entrants</u>					
	Min	Max	Average	Min	%	Max	%	Average	%
2018	0,15%	0,21%	0,18%	108.093,90	0,72%	215.541,19	1,44%	<u>152.699,95</u>	1,02%
2019	0,21%	0,35%	0,26%	145.416,81	0,97%	351.241,89	2,34%	226.631,29	1,51%
2020	0,28%	0,56%	0,39%	195.378,54	1,30%	577.036,93	3,85%	336.297,46	2,24%
2021	0,38%	0,93%	0,58%	262.057,74	1,75%	945.305,52	6,30%	498.177,50	3,32%
2022	0,52%	1,52%	0,87%	348.128,24	2,32%	1.541.410,04	10,28%	734.992,48	4,90%
2023	0,70%	2,48%	1,28%	458.779,77	3,06%	2.494.249,59	16,63%	1.076.210,61	7,17%
2024	0,93%	4,01%	1,88%	599.618,20	4,00%	3.985.767,67	26,57%	1.556.098,14	10,37%
2025	1,23%	6,41%	2,73%	775.148,73	5,17%	6.240.048,53	41,60%	2.206.728,15	14,71%
2030	4,13%	44,72%	12,62%	693.547,67	4,62%	32.323.147,47	215,49%	7.211.956,36	48,08%
2035	8,14%	94,54%	26,63%	4.290,21	0,03%	29.145.225,56	194,30%	5.979.862,00	39,87%
2040	9,72%	99,92%	32,79%	22,51	0,00%	20.369.162,15	135,79%	1.667.145,40	11,11%
2045	9,97%	100,00%	33,95%	0,12	0,00%	6.580.460,67	43,87%	210.859,55	1,41%

New BEV entrants, Actuals 2018: 150.003

Market share distribution of the *more likely* scenarios (i.e. $R^2 > ,99$)

Market share	<u>Europe</u>		<u>United States</u>	
	Count	%	Count	%
10%	234	34%	646	35%
20%	117	17%	320	17%
30%	79	11%	211	11%
40%	58	8%	157	8%
50%	47	7%	128	7%
60%	40	6%	104	6%
70%	34	5%	91	5%
80%	29	4%	78	4%
90%	27	4%	67	4%
100%	24	3%	62	3%
Total	689	100%	1.864	100%
Parameter	Min	Max	Min	Max
p value	0,00003	0,00067	0,00012	0,00241
q value	0,28	0,62	0,23	0,51

Conclusions

- Rather than searching for the ultimate set of parameter estimates, the proposed heuristics focus on revealing a set of potential outcomes, building on relatively short time series (before growth phase).
- The heuristics implied allow to assess and quantify the multi-finality of novel trajectories, enabling to consider multiple pathways towards local optima, ex-ante and simultaneously.
- It becomes feasible to assess the likelihood as well as time horizon of different scenarios, and thus offers the potential to assess and model uncertainty (Knight, 1920) in a quantitative manner; however, not in a deterministic way.
- The presence of multiple trajectories towards the future, directs our attention to consider explicitly the antecedents required for different scenarios to unfold.
- As such, forecasting becomes a valuable, informative complement of scenario development exercises rather than an inferior or even neglected substitute.

From BEV towards COVID-19

Nederlandse onderzoekers voorspellen evolutie corona met modellen uit 1845 van Belgische wiskundige

Door modellen voor de groei van populaties los te laten op wereldwijde coronadata zijn Eindhovense datawetenschappers in staat om het te verwachten aantal nieuwe besmettingen en doden als gevolg van het SARS-CoV-2 virus te berekenen voor de nabije toekomst. Voor zeven landen - China, Zuid-Korea, Iran, Italië, Frankrijk, de VS en Nederland - is een nauwkeurige voorspelling mogelijk van één tot drie dagen vooruit. Voor sommige landen kunnen ze ook een schatting maken van het maximum aantal bevestigde infecties.

- Epidemiological models (e.g. Bayesian SEIR modeling) require knowledge on the underlying epidemiological parameters to yield accurate predictions. These parameters are unknown during the early phases of the emerging virus (i.e. before the inflection points).
- To what extent could diffusion models - widely known and used within the innovation (and technology) discipline - bear relevance for modeling and forecasting pandemic phenomena like the COVID-19 virus :
 - Innovation followed by imitation
 - Infection followed by contamination



FACULTY OF ECONOMICS
AND BUSINESS



Modeling the COVID-19 Pandemic

Kristof Decock, *Faculty of Economics & Business (MSI), KULeuven & Flanders Business School*

Jorge Ricardo Nova Blanco, *Rega Institute & Institute for the Future, KULeuven*

Michela Bergamini, *Faculty of Economics & Business (MSI), KULeuven*

Prof. Koenraad Debackere, *Faculty of Economics & Business (MSI), ECOOM KULeuven*

Sien Luyten, *Flanders Business School*

Xiaoyan Song, *Faculty of Economics & Business, ECOOM KULeuven*

Prof. Anne-Mieke Vandamme, *Rega Institute & Institute for the Future, KULeuven*

Prof. Bart Van Looy, *Faculty of Economics & Business (MSI), KULeuven, ECOOM & Flanders Business School*

3. Italian case

Insights from:

Decock, K., Bergamini, M., Debackere, K., Lupi, E., Vandamme A.-M. and Van Looy, B. (2020). Predicting when peaks will occur, *ex ante*: Insights from the COVID-19 Pandemic in Italy and Belgium. *Under review*

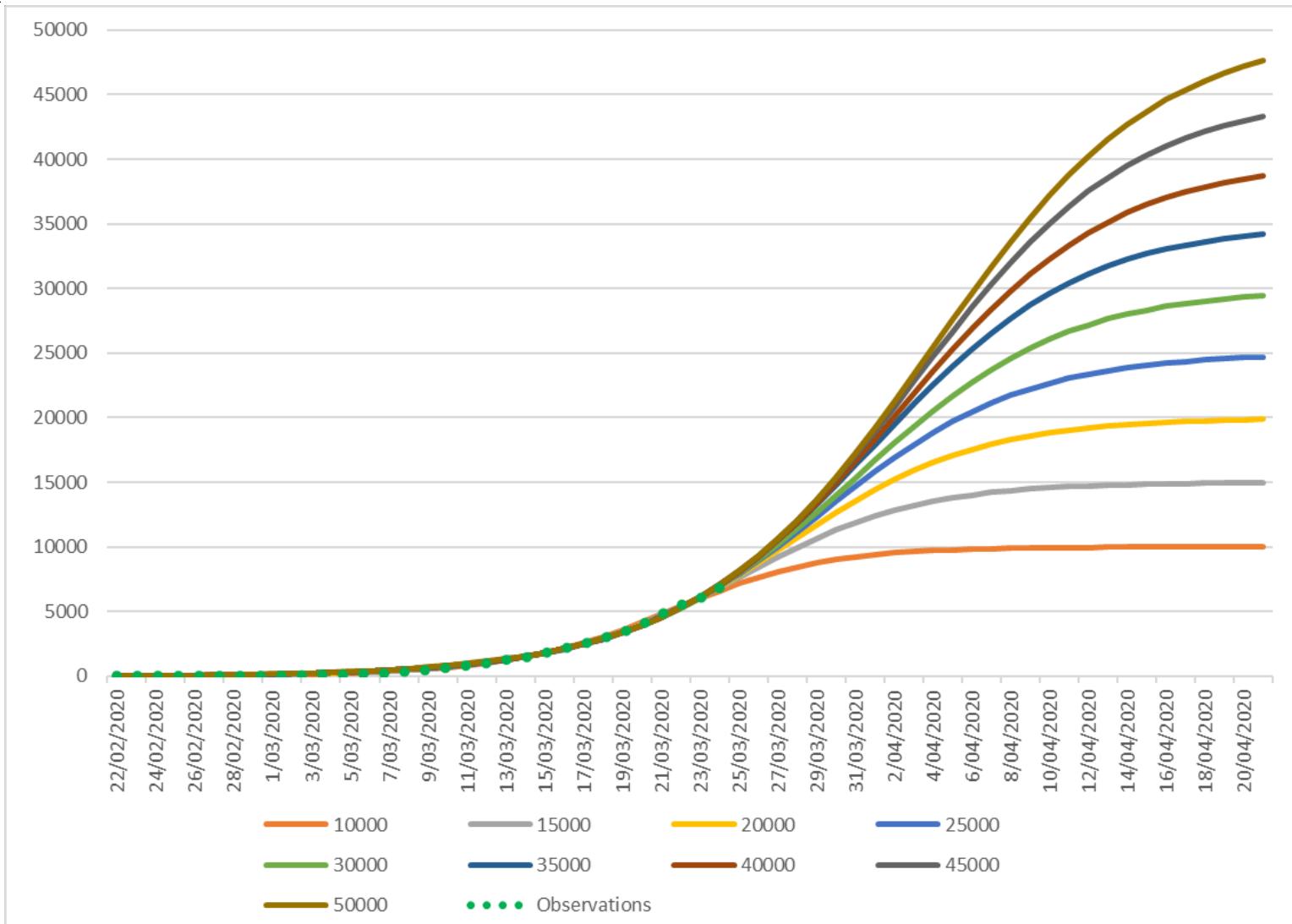
Introduction

- On Friday 20/03, we started apply the heuristics developed to model the diffusion of the Electrical Car (Decock, Debackere & Van Looy, 2020) towards the COVID-19 pandemic.
- We focused our efforts on deceased on the one hand, and ICU occupation levels on the other hand.
- In this PPT, we report major findings for Italy, pertaining to the number of cumulative deceased.
- We report models with different levels of granulation (in terms of end states) as this allows to assess the consistencies of the patterns emerging.
- Source data: www.protezionecivile.gov.it

First round of models

- The reported COVID-19 deceased for Italy exceeded 10 on 25/02.
- We created first models for Italy, on 25/03, based on data starting from 22/02, running till 24/03 (32 days).
- End-states ranged from 10 to 50.000 (cumulative deceased) with steps of 5.000.
- The heuristics at that moment no longer include an end state totaling 5.000 cumulative deceased.
- As the following graph shows, 9 different end states (ranging between 10.000 & 50.000) are at that moment plausible.

Date	N. deceased (cumulative)
22/02/2020	2
23/02/2020	2
24/02/2020	6
25/02/2020	11
26/02/2020	12
27/02/2020	17
28/02/2020	21
29/02/2020	29
1/03/2020	35
2/03/2020	52
3/03/2020	80
4/03/2020	107
5/03/2020	148
6/03/2020	197
7/03/2020	233
8/03/2020	366
9/03/2020	464
10/03/2020	631
11/03/2020	827
12/03/2020	1016
13/03/2020	1268
14/03/2020	1441
15/03/2020	1811
16/03/2020	2158
17/03/2020	2505
18/03/2020	2978
19/03/2020	3407
20/03/2020	4032
21/03/2020	4825
22/03/2020	5476
23/03/2020	6072
24/03/2020	6820
25/03/2020	7503
26/03/2020	8165
27/03/2020	9134
28/03/2020	10023
29/03/2020	10779
30/03/2020	11591
31/03/2020	12428
1/04/2020	13155
2/04/2020	13915
3/04/2020	14681
4/04/2020	15362
5/04/2020	15887
6/04/2020	16523



First round of models

- When one adds to this graph the reported numbers for the time period 25/03 till 06/04 (red dots – figure on the next slide), it seems that Italy is walking a line towards an end-state situated between 10.000 & 20.000 deceased, with actual levels much closer to the 20.000 curve than the 10.000.
- According to these graphs, the peak for the 10 K curve would imply 619 deceased on 21/03; the peak for the 20 K curve would imply 984 deceased on 28/03.
- The peak for Italy (so far) was Friday 27/03 with 969 reported deceased.
- On that evening, we reached out to Sant'Anna to signal our work and to indicate that our models suggested a peak soon (based on the 18K-19K deceased scenarios, which signaled peak levels on Friday 27/03 and the weekend).

Da: Bart Van Looy <bart.vanlooy@kuleuven.be>

Data: 27/03/20 19:09 (GMT+01:00)

A: Arianna Menciassi <arianna.menciassi@santannapisa.it>, sabina.nuti@santannapisa.it

Cc: Andrea Piccaluga <andrea.piccaluga@santannapisa.it>, alberto.diminin@santannapisa.it

Oggetto: Italy COV19

Hello Sabina, Arianna,

I'm reaching out related to the COV19 situation.

Since one week, we (Kristof, I and a small team) are modelling the diffusion (deceased & ICU) of COV19 for BE, NL, IT, ES, FR,...based on diffusion models we develop in the framework of the PhD of Kristof Decock and which also have been applied as part of the EIB project.

We (including Koen Debackere who is following the process closely) are currently convinced that our approach does a better job than what is available/used.

We are already advising/informing the Belgian authorities on capacity planning etc.

As for Italy, we also have estimates on how things will evolve during the next 10 days. I would like to discuss these estimates so you can assess whether this could be useful/hopeful for Italy.

I'm now going out for a run, but could skype this evening, around 21.00 or so. If someone else at Sant'Anna would be better placed to have a look at this, let me know.

Best wishes,

Talk soon,

Bart

From: sabina.nuti <sabina.nuti@santannapisa.it>

Sent: vrijdag 27 maart 2020 22:41

To: Bart Van Looy <bart.vanlooy@kuleuven.be>; Arianna Menciassi <arianna.menciassi@santannapisa.it>

Cc: Andrea Piccaluga <andrea.piccaluga@santannapisa.it>; alberto.diminin@santannapisa.it

Subject: RE: Italy COV19

Dear Bart, we have a large group of researchers working on this issue . The Italian contest has large differences among the italian regions that have to be considered. We are working with our government and we will be happy to collaborate with you. Best Sabina

Inviato da smartphone Samsung Galaxy.

From: Bart Van Looy

Sent: zaterdag 28 maart 2020 2:32

To: 'sabina.nuti' <sabina.nuti@santannapisa.it>; Arianna Menciassi <arianna.menciassi@santannapisa.it>

Cc: Andrea Piccaluga <andrea.piccaluga@santannapisa.it>; alberto.diminin@santannapisa.it; Kristof Decock <kristof.decock@kuleuven.be>

Subject: RE: Italy COV19

Hello Sabina, all,

Please find included the xls in which we model Italy. Only deceased.

We examine several scenarios of diffusion (250.000) and then select the most plausible ones.

This approach – for the details see paper in attach – yielded a prediction for the reported casualties on 27.03 of 931 while the actual number was 919. We have the same level of accuracy for Spain, Belgium,...

We started to develop/apply the models 8 days ago and the results for Italy have been consistent/accurate for 7 days in a row now. If you extend this forecast, Italy will have similar numbers today (Saturday) and tomorrow, and then the decline will set in

The end state which we predict for Italy will be around 18.000 casualties. While this might sound as a lot and still 9.000 to go, we also think that it is important to signal that after a very heavy weekend (with again casualties around 900 for Saturday and Sunday), it will become better from next week onwards (Monday).

Let me/us know if this is useful and if this can be helpful to cope with the current situation in Italy.

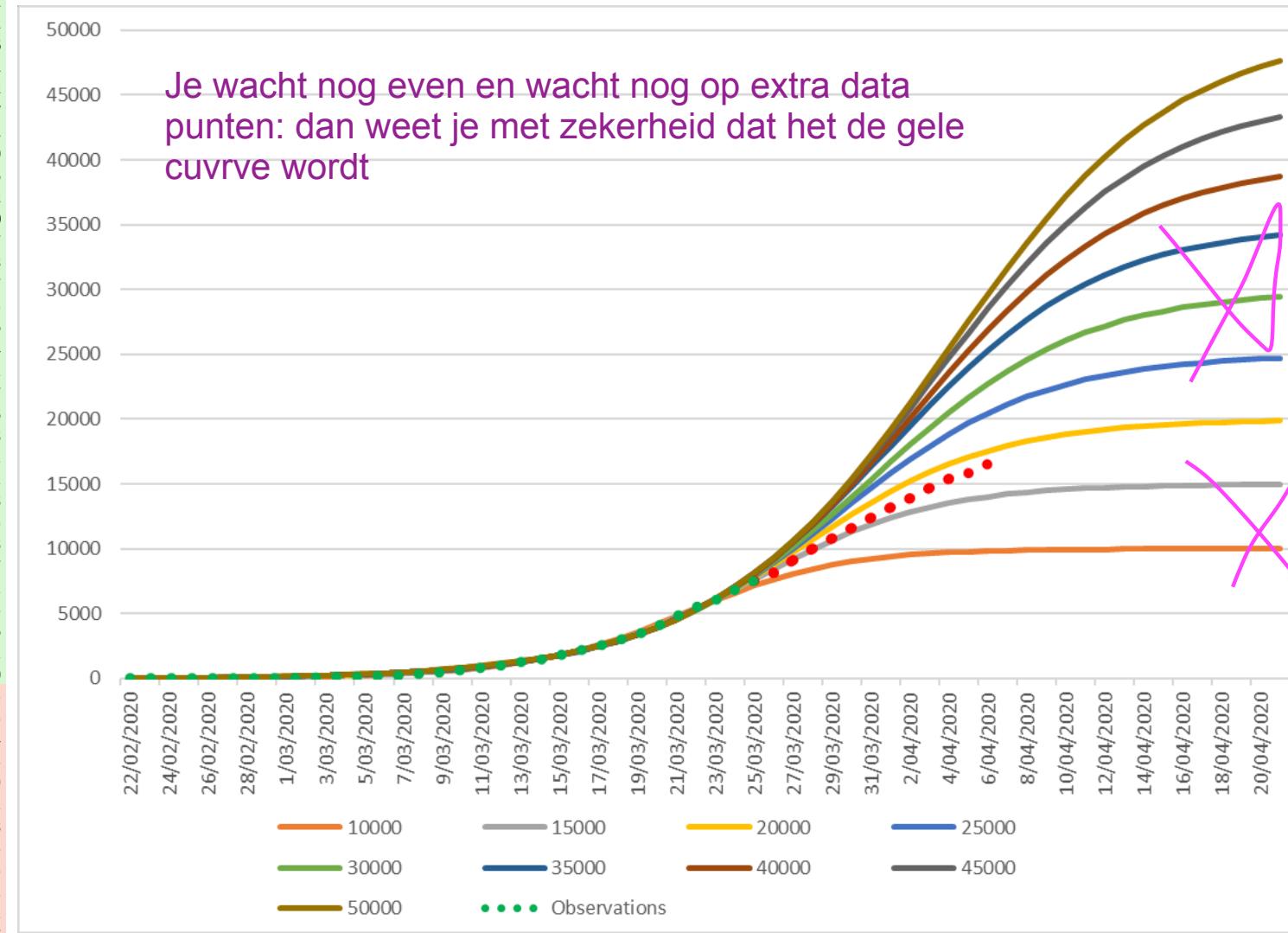
Similar models for Belgium are now being used here to 'manage' the situation, including the ICU capacity.

We can call tomorrow around lunch time, i.e. after we have incorporated the most recent data for BE in our forecasts (the numbers are released around 11.00).

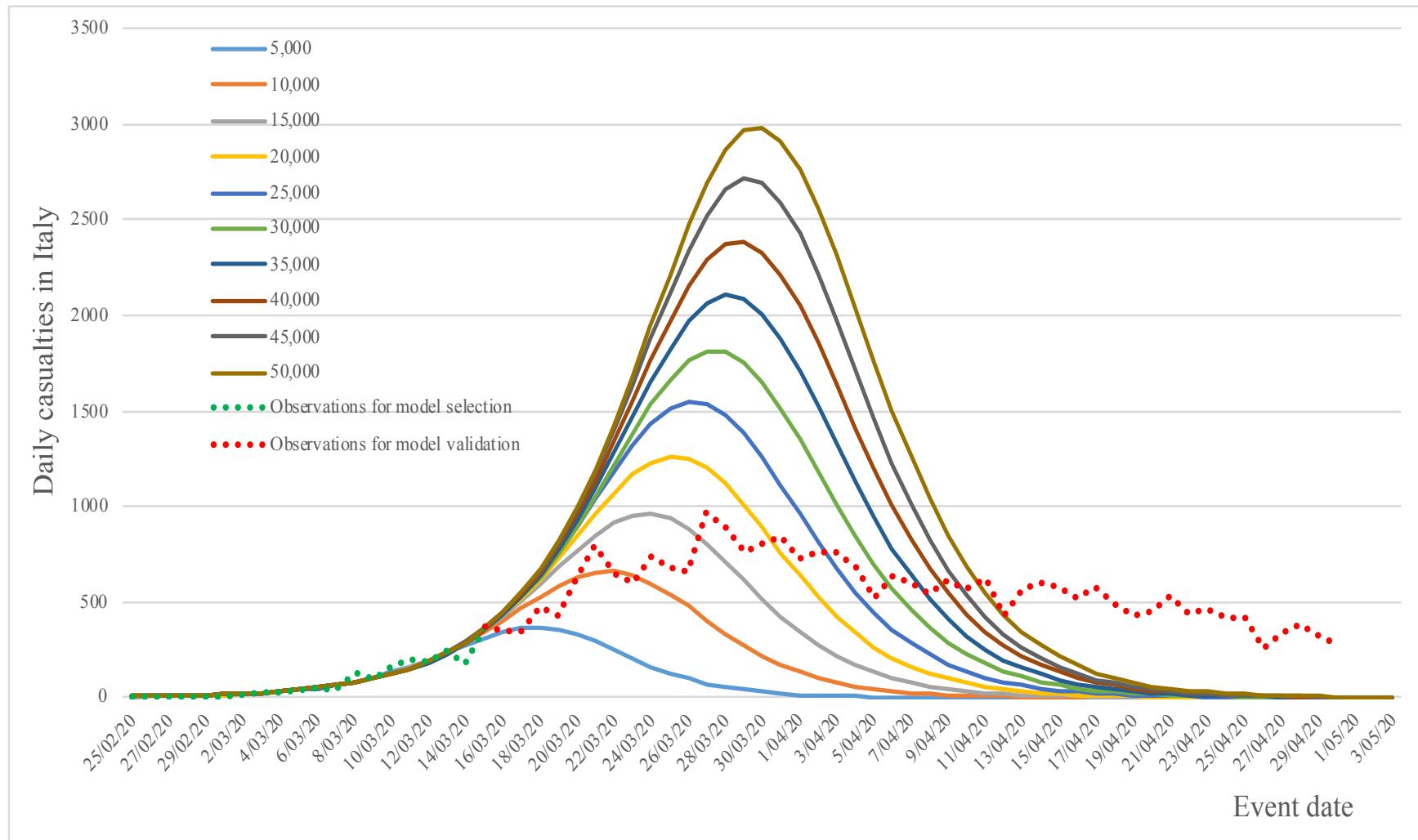
Take care, talk soon,

Bart

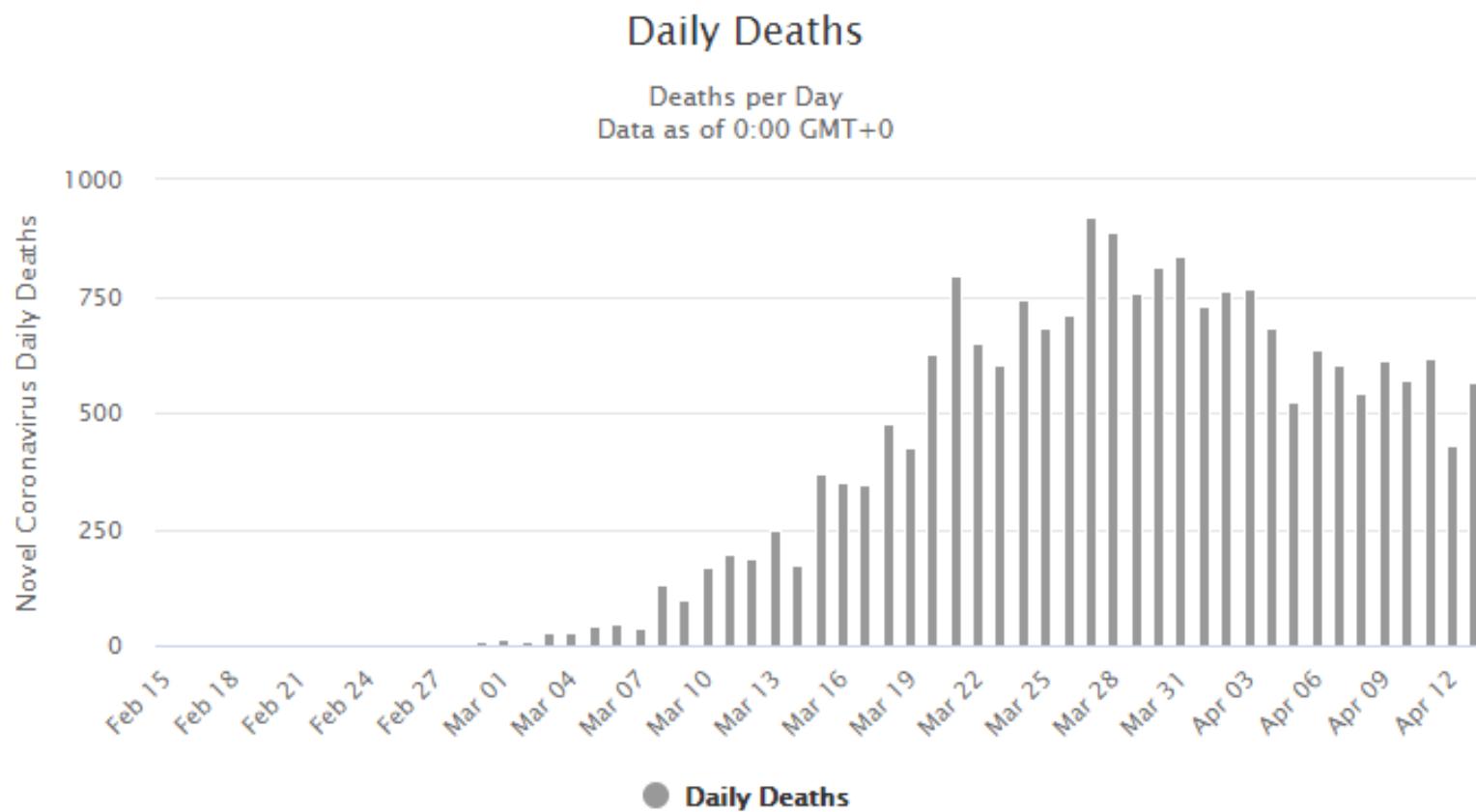
Date	N. deceased (cumulative)
22/02/2020	2
23/02/2020	2
24/02/2020	6
25/02/2020	11
26/02/2020	12
27/02/2020	17
28/02/2020	21
29/02/2020	29
1/03/2020	35
2/03/2020	52
3/03/2020	80
4/03/2020	107
5/03/2020	148
6/03/2020	197
7/03/2020	233
8/03/2020	366
9/03/2020	464
10/03/2020	631
11/03/2020	827
12/03/2020	1016
13/03/2020	1268
14/03/2020	1441
15/03/2020	1811
16/03/2020	2158
17/03/2020	2505
18/03/2020	2978
19/03/2020	3407
20/03/2020	4032
21/03/2020	4825
22/03/2020	5476
23/03/2020	6072
24/03/2020	6820
25/03/2020	7503
26/03/2020	8165
27/03/2020	9134
28/03/2020	10023
29/03/2020	10779
30/03/2020	11591
31/03/2020	12428
1/04/2020	13155
2/04/2020	13915
3/04/2020	14681
4/04/2020	15362
5/04/2020	15887
6/04/2020	16523



Italy as a whole, model selection based on 24/2-15/3



Daily New Deaths in Italy

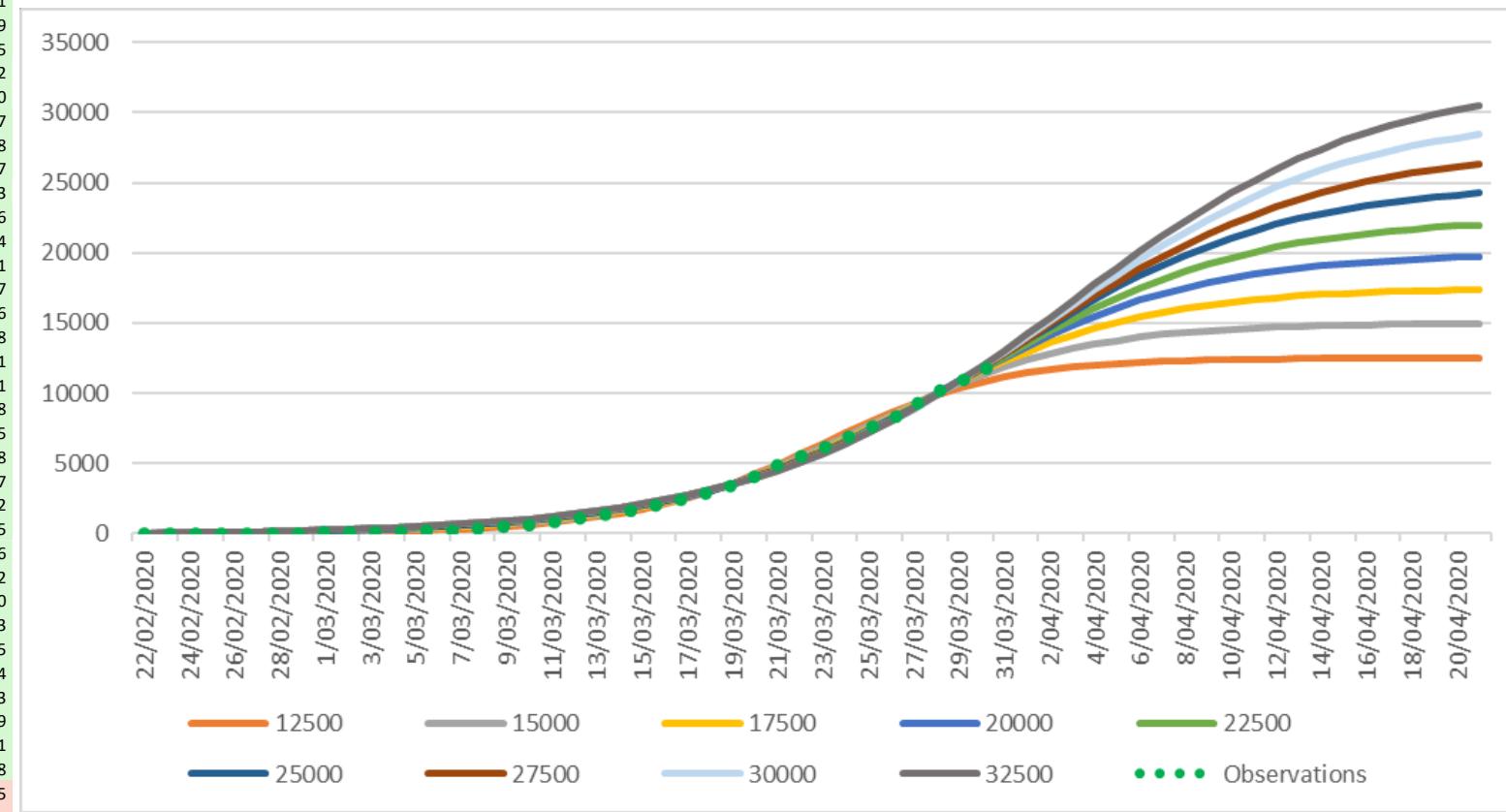


<https://www.worldometers.info/coronavirus/country/italy/>

Second round of models

- We updated the models by including data until 31/03.
- In total, the number of days used for calculating the models and selecting the more plausible ones, amounts now to 39 days.
- End-states ranged from 10.000 to 32.500 (cumulative deceased) with steps of 2.500.
- The heuristics at that moment no longer include end-state totaling 10.000 cumulative deceased.
- As the following graph shows, 9 different end-states (ranging 12.500 to 32.500) are at that moment plausible.

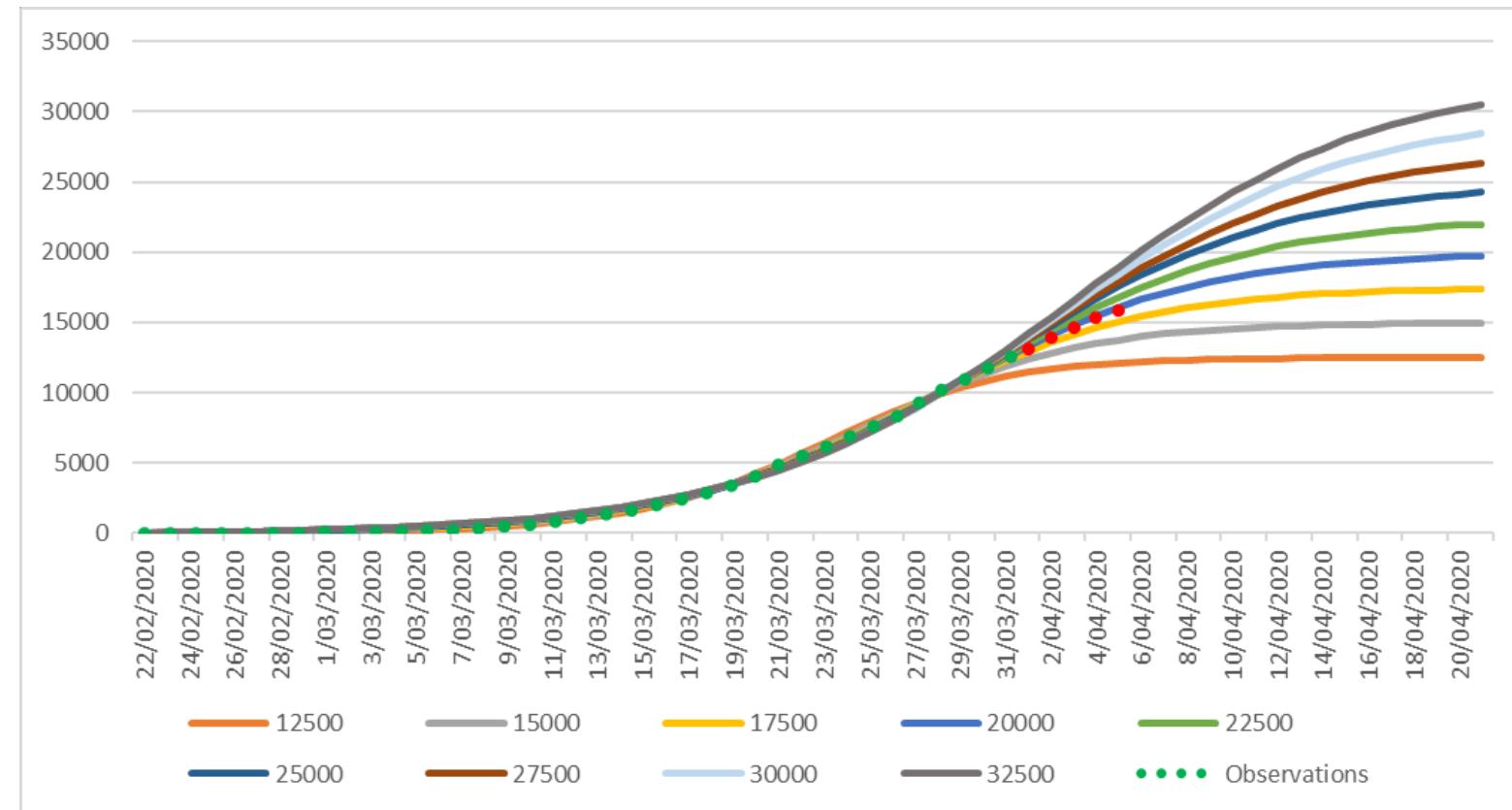
Date	N. deceased (cumulative)
22/02/2020	2
23/02/2020	2
24/02/2020	6
25/02/2020	11
26/02/2020	12
27/02/2020	17
28/02/2020	21
29/02/2020	29
1/03/2020	35
2/03/2020	52
3/03/2020	80
4/03/2020	107
5/03/2020	148
6/03/2020	197
7/03/2020	233
8/03/2020	366
9/03/2020	464
10/03/2020	631
11/03/2020	827
12/03/2020	1016
13/03/2020	1268
14/03/2020	1441
15/03/2020	1811
16/03/2020	2158
17/03/2020	2505
18/03/2020	2978
19/03/2020	3407
20/03/2020	4032
21/03/2020	4825
22/03/2020	5476
23/03/2020	6072
24/03/2020	6820
25/03/2020	7503
26/03/2020	8165
27/03/2020	9134
28/03/2020	10023
29/03/2020	10779
30/03/2020	11591
31/03/2020	12428
1/04/2020	13155
2/04/2020	13915
3/04/2020	14681
4/04/2020	15362
5/04/2020	15887
6/04/2020	16523



Second round of models

- When one adds to this graph the reported numbers for the time period till 06/04 (red dots), it seems that Italy is walking a line towards an end-state situated between 18.000 and 20.000 deceased, with actual levels closer to the 20.000 curve than the 18.000.
- This range is in line with the observed patterns and their implications which build on data till 24/03
- Given the peak levels reported before, this update confirms that decline of the curve is manifesting itself in a consistent manner.
- The final graph on the next slide depicts the predictions derived from different end-states together with the actual observations on a *daily* base.

Date	N. deceased (cumulative)
22/02/2020	2
23/02/2020	2
24/02/2020	6
25/02/2020	11
26/02/2020	12
27/02/2020	17
28/02/2020	21
29/02/2020	29
1/03/2020	35
2/03/2020	52
3/03/2020	80
4/03/2020	107
5/03/2020	148
6/03/2020	197
7/03/2020	233
8/03/2020	366
9/03/2020	464
10/03/2020	631
11/03/2020	827
12/03/2020	1016
13/03/2020	1268
14/03/2020	1441
15/03/2020	1811
16/03/2020	2158
17/03/2020	2505
18/03/2020	2978
19/03/2020	3407
20/03/2020	4032
21/03/2020	4825
22/03/2020	5476
23/03/2020	6072
24/03/2020	6820
25/03/2020	7503
26/03/2020	8165
27/03/2020	9134
28/03/2020	10023
29/03/2020	10779
30/03/2020	11591
31/03/2020	12428
1/04/2020	13155
2/04/2020	13915
3/04/2020	14681
4/04/2020	15362
5/04/2020	15887
6/04/2020	16523

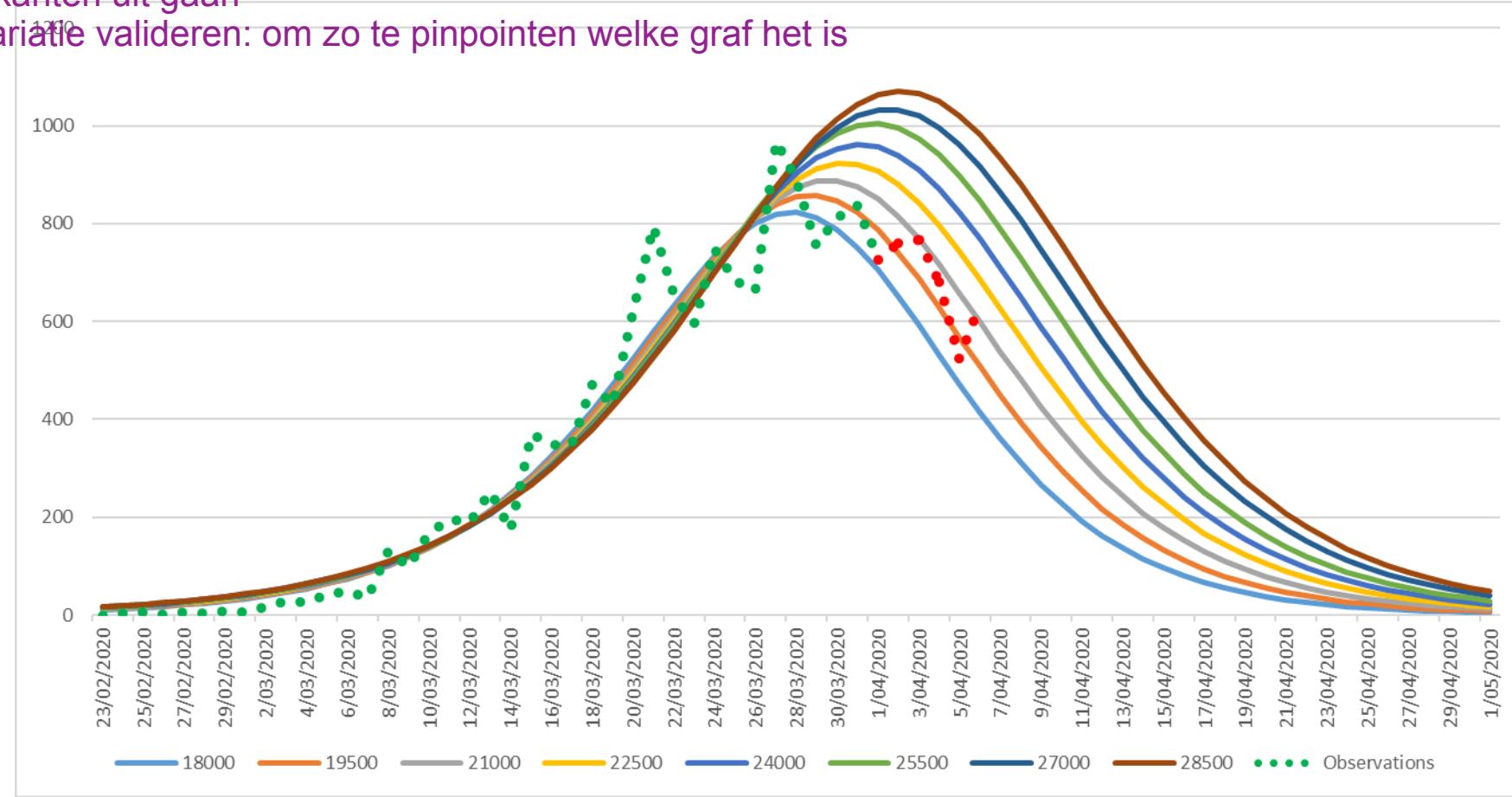


Italy – Update 07/04

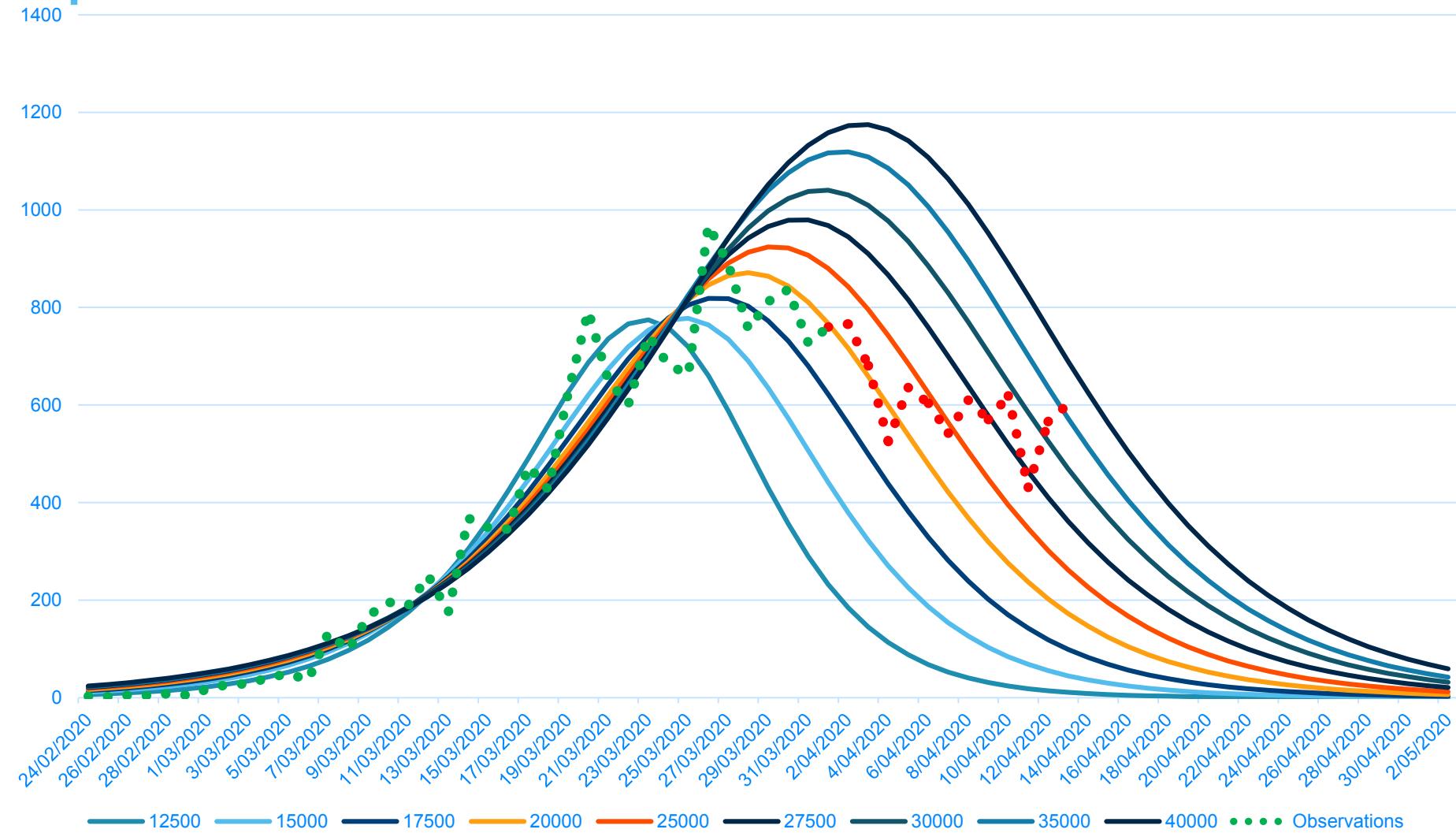
Je gaat naar de toekomst denken vanuit scenario logica

> het kan alle kanten uit gaan

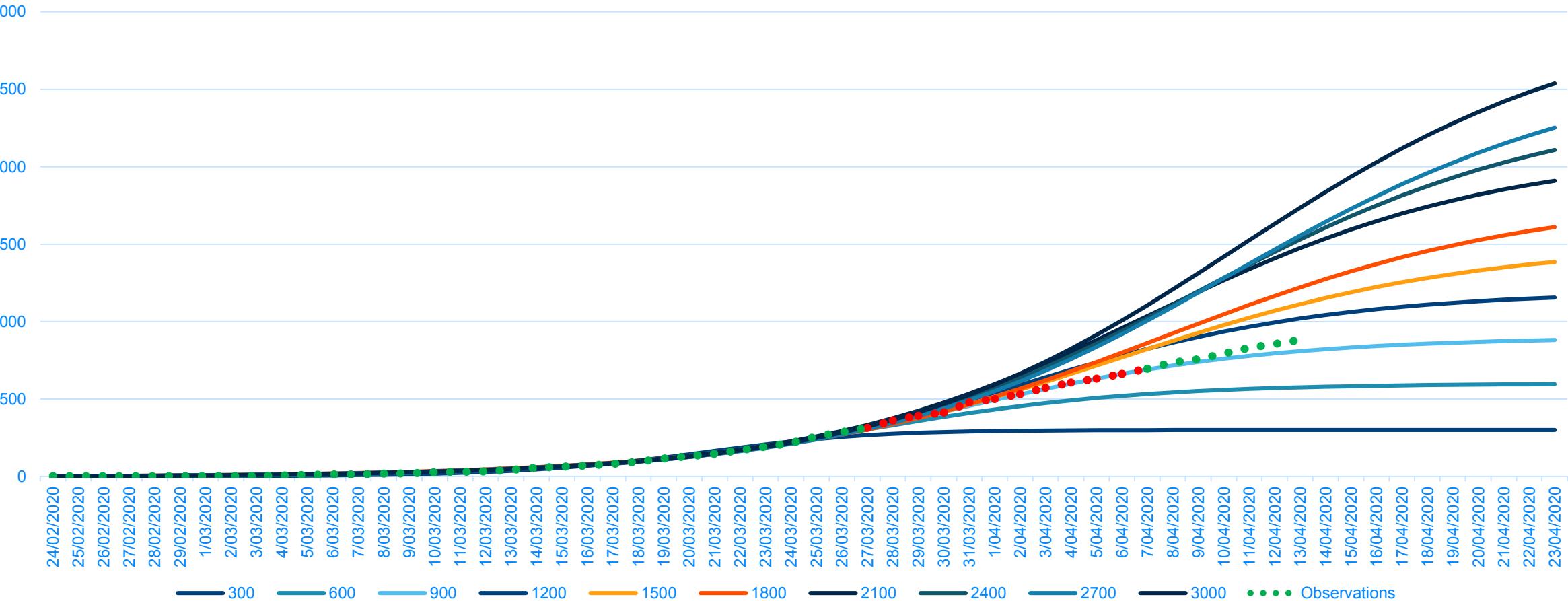
> en dan de variatie valideren: om zo te pinpointen welke graf het is



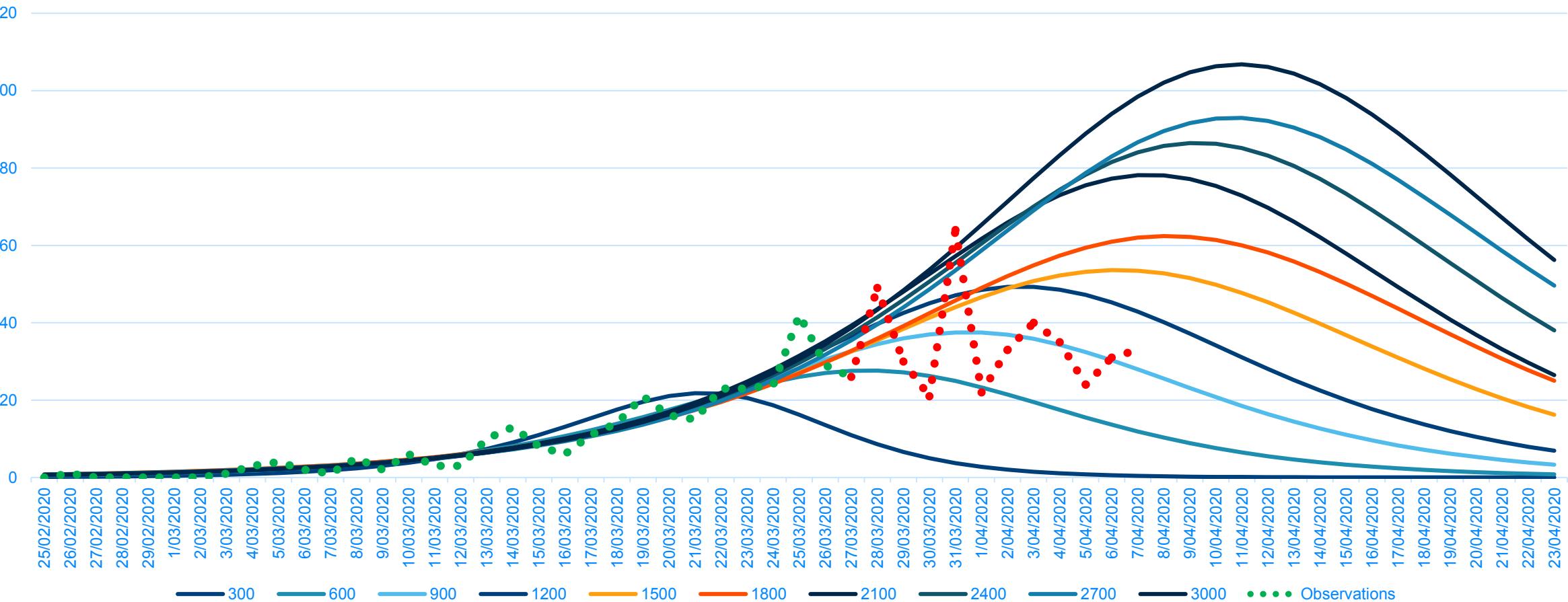
Italy Update 14/04



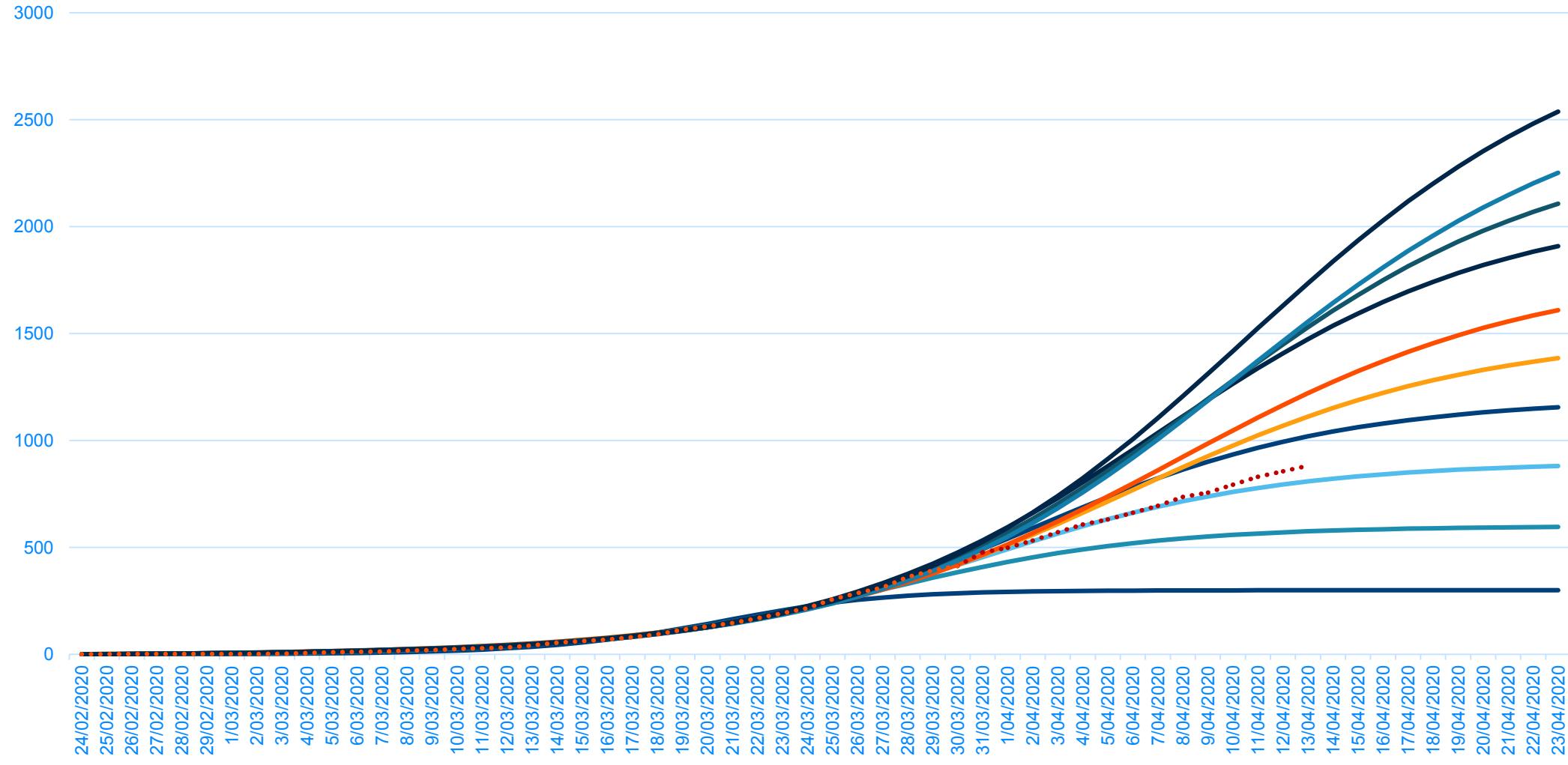
Veneto



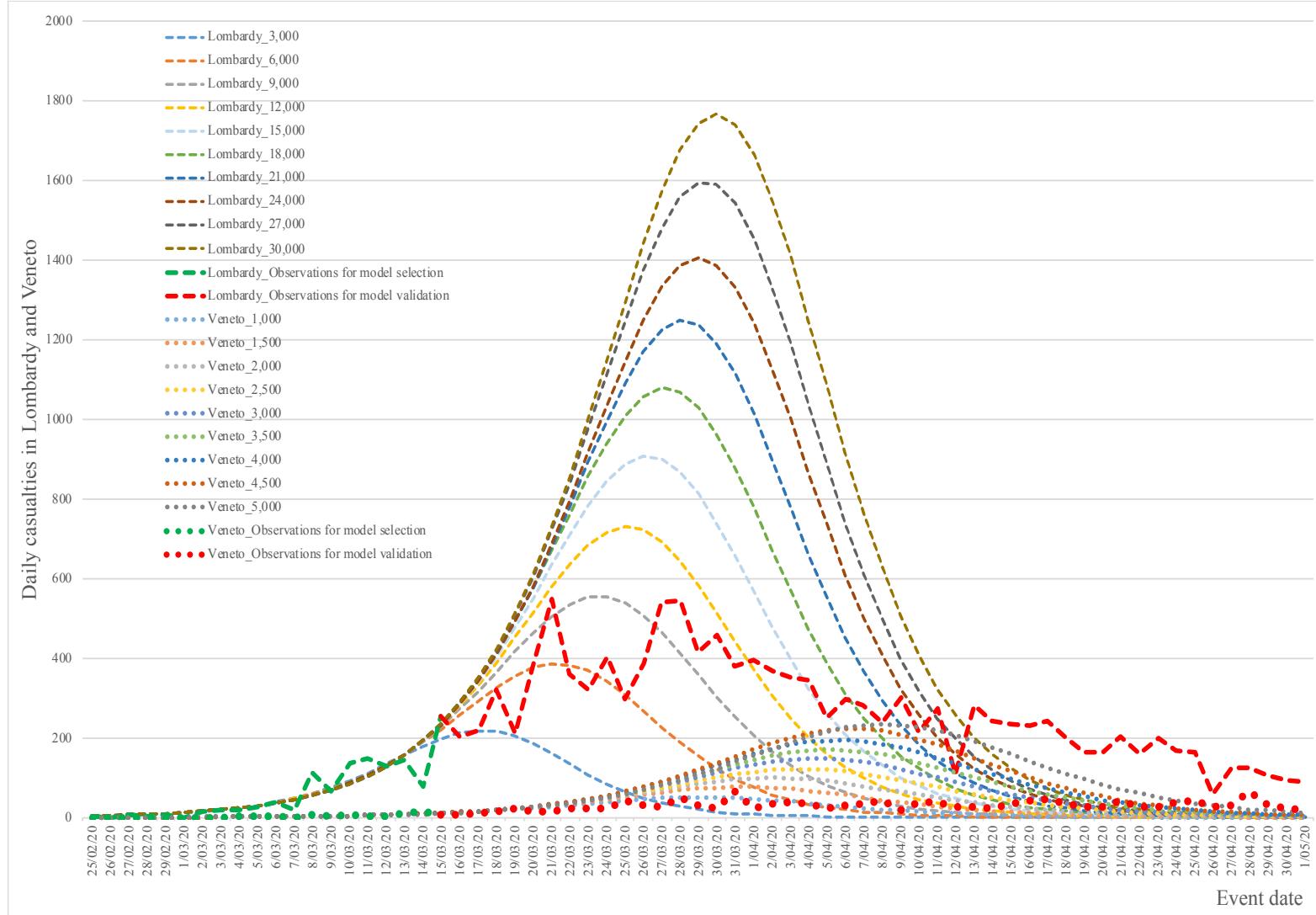
Veneto



Veneto – Update 14/04



Lombardy versus Veneto



4. ICU occupation in Belgium

Insights from:

- Decock, K., Debackere, K., Vandamme A.-M. and Van Looy, B. (2020). Scenario-driven Forecasting: Lessons Learned from Modeling the COVID-19 Pandemic. *ISSI Newsletter*, Vol. 16, Iss. 1., pp. 2-6
- Decock, K., Debackere, K., Vandamme A.-M. and Van Looy, B. (2020). Scenario-driven Forecasting: Modeling peaks and paths. Insights from the COVID-19 Pandemic in Belgium. *Under review*.

The Belgium Case: Data

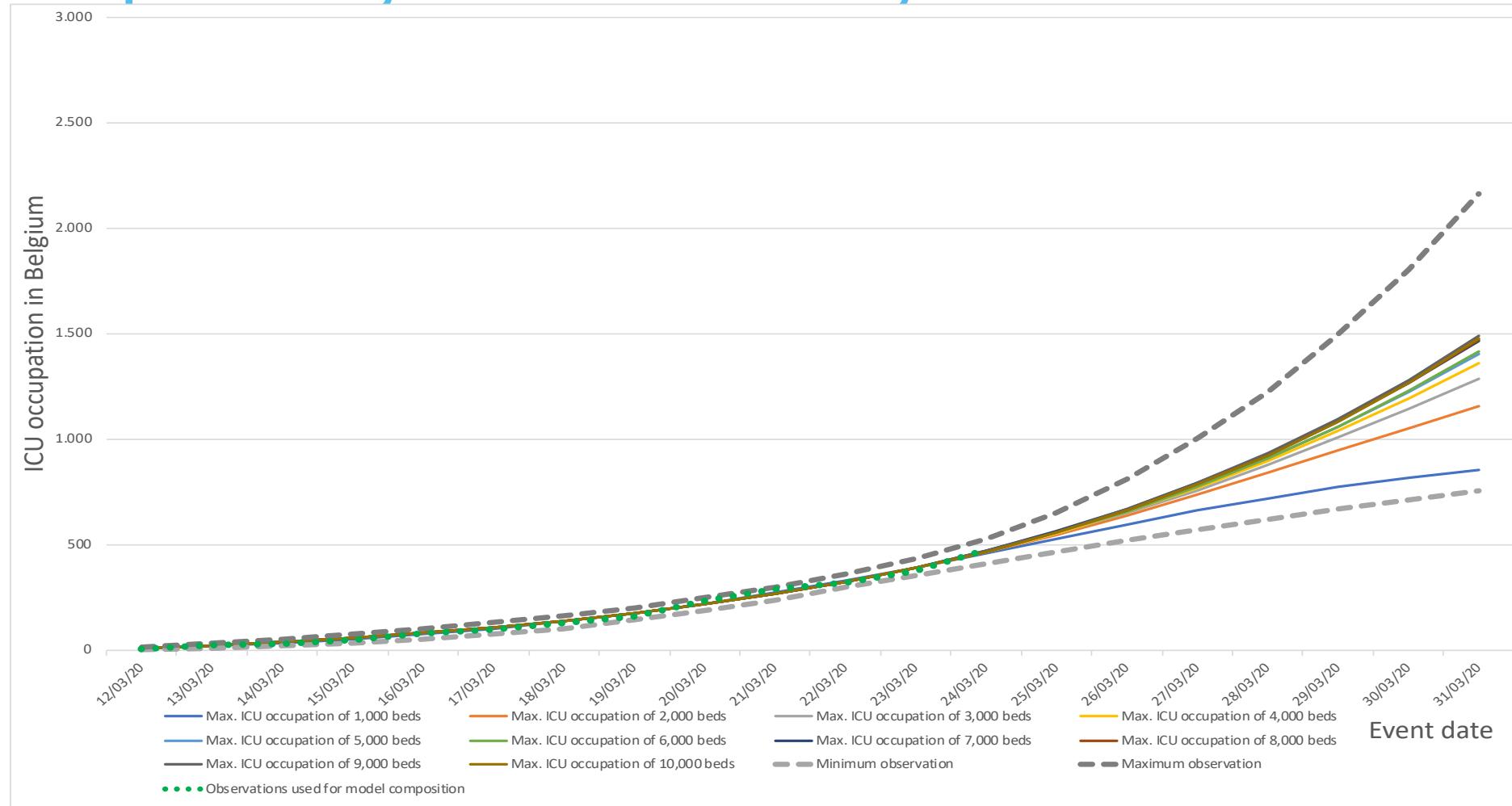
- 03/02/2020: First confirmed case
- 10/03/2020: First COVID-19 deceased registered
- Initial ICU capacity: 1.650 beds in total
- Gradual increase towards approx. 2.300 ICU beds exclusively for COVID-19 patients
- Will this be enough for the coming weeks/months?

COVID-19 ICU occupation between 12/03 - 24/03/2020

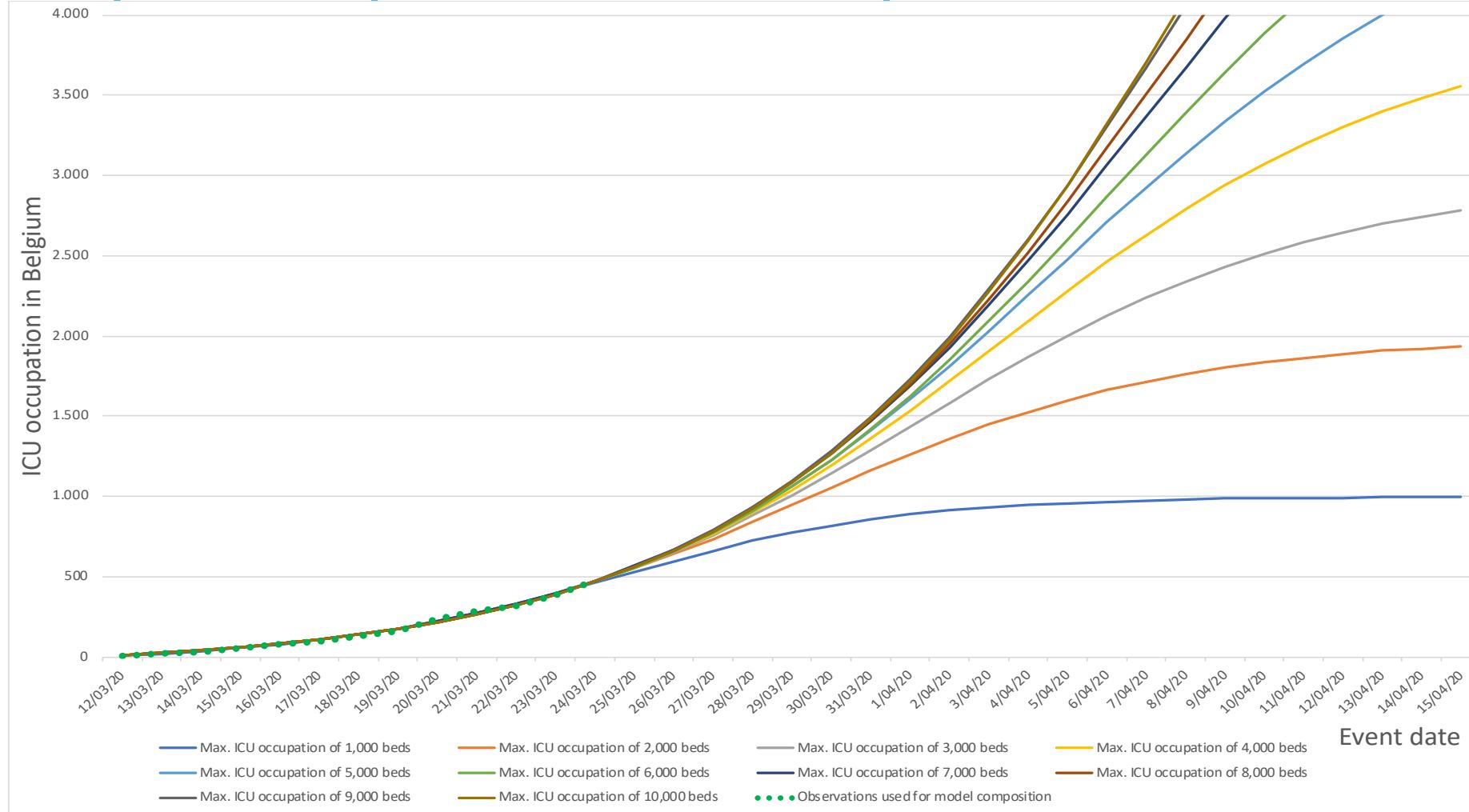
Date	ICU occupation	Date	ICU occupation	Date	ICU occupation
12/03/20	5	17/03/20	100	22/03/20	322
13/03/20	24	18/03/20	130	23/03/20	381
14/03/20	33	19/03/20	164	24/03/20	474
15/03/20	53	20/03/20	238		
16/03/20	79	21/03/20	290		

Source: Sciensano

Stylized pathways of more likely scenarios – zoom-in



Stylized pathways of more likely scenarios – zoom-out



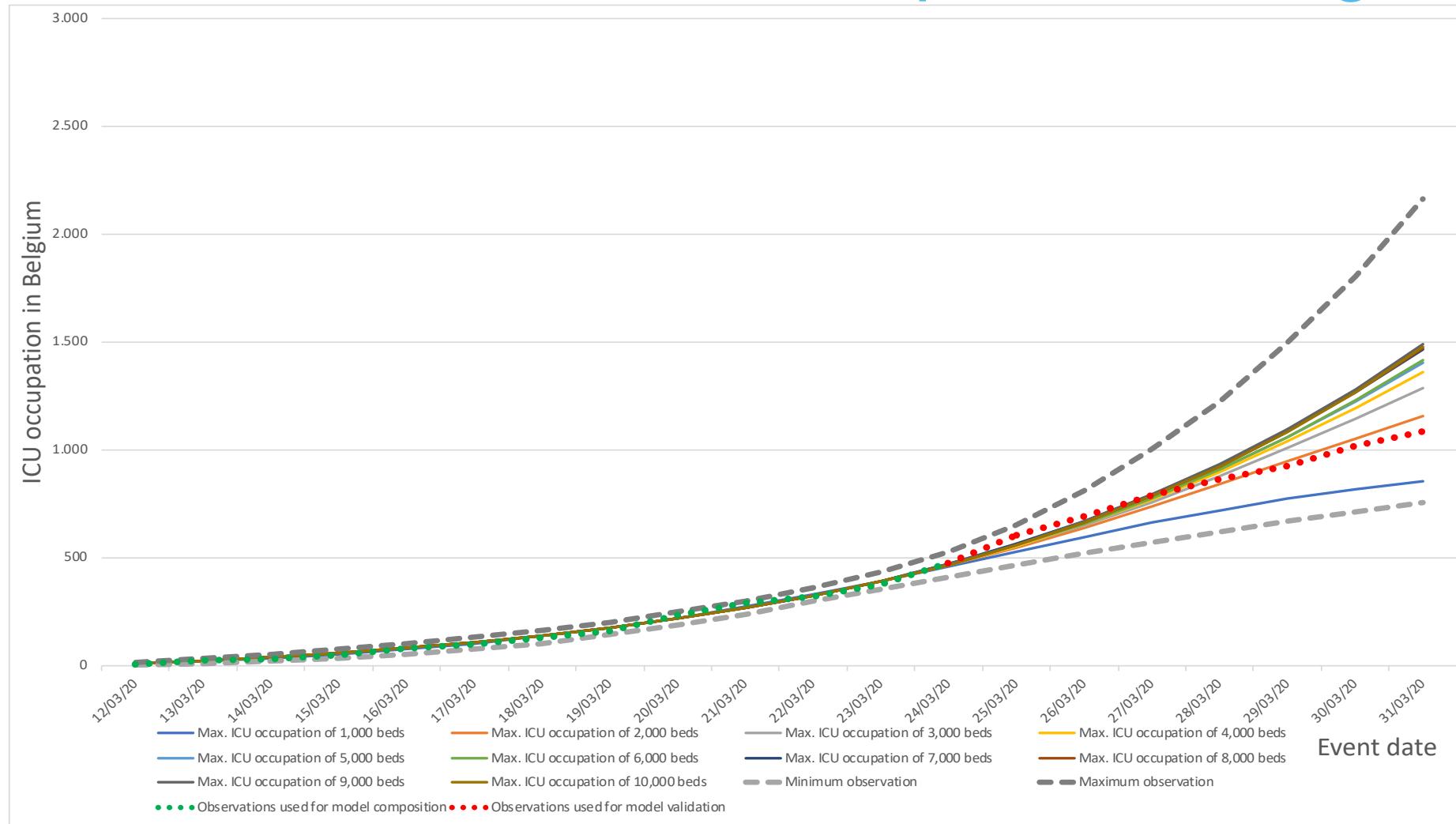
Forecasting the ICU occupation between 25-31/03/2020

Max. ICU capacity	Average ICU occupation for the corresponding pathways						
	25/03/20	26/03/20	27/03/20	28/03/20	29/03/20	30/03/20	31/03/20
1,000	530.41	598.30	662.88	722.03	774.37	819.26	856.77
2,000	549.63	640.24	737.86	840.94	947.44	1,054.95	1,160.93
3,000	554.59	651.96	760.39	879.53	1,008.49	1,145.73	1,289.10
4,000	556.01	656.58	770.32	897.71	1,038.78	1,192.99	1,359.08
5,000	560.06	662.81	780.00	912.65	1,061.44	1,226.54	1,407.53
6,000	557.22	659.75	777.21	910.96	1,062.18	1,231.69	1,419.82
7,000	565.75	671.08	792.29	931.05	1,088.92	1,267.17	1,466.65
8,000	560.28	665.33	786.64	926.13	1,085.62	1,266.81	1,471.03
9,000	562.46	668.74	791.80	933.75	1,096.72	1,282.72	1,493.55
10,000	559.37	664.17	785.49	925.47	1,086.32	1,270.25	1,479.31
Overall minimum	467.52	520.39	571.98	621.99	669.57	713.99	754.72
Overall maximum	655.25	812.99	1,002.77	1,229.04	1,495.87	1,806.41	2,162.24
Overall average	545.50	631.60	723.48	820.43	921.78	1,026.99	1,135.52
Average of the 10 end states	555.58	653.90	764.49	888.02	1,025.03	1,175.81	1,340.38

Forecasting the ICU occupation between 25-31/03/2020

Max. ICU capacity	Average ICU occupation for the corresponding pathways						
	25/03/20	26/03/20	27/03/20	28/03/20	29/03/20	30/03/20	31/03/20
1,000	530.41	598.30	662.88	722.03	774.37	819.26	856.77
2,000	549.63	640.24	737.86	840.94	947.44	1,054.95	1,160.93
3,000	554.59	651.96	760.39	879.53	1,008.49	1,145.73	1,289.10
4,000	556.01	656.58	770.32	897.71	1,038.78	1,192.99	1,359.08
5,000	560.06	662.81	780.00	912.65	1,061.44	1,226.54	1,407.53
6,000	557.22	659.75	777.21	910.96	1,062.18	1,231.69	1,419.82
7,000	565.75	671.08	792.29	931.05	1,088.92	1,267.17	1,466.65
8,000	560.28	665.33	786.64	926.13	1,085.62	1,266.81	1,471.03
9,000	562.46	668.74	791.80	933.75	1,096.72	1,282.72	1,493.55
10,000	559.37	664.17	785.49	925.47	1,086.32	1,270.25	1,479.31
Overall minimum	467.52	520.39	571.98	621.99	669.57	713.99	754.72
Overall maximum	655.25	812.99	1,002.77	1,229.04	1,495.87	1,806.41	2,162.24
Overall average	545.50	631.60	723.48	820.43	921.78	1,026.99	1,135.52
Average of the 10 end states	555.58	653.90	764.49	888.02	1,025.03	1,175.81	1,340.38
Actuals	605	690	789	867	927	1,021	1,088

Short term evolution of ICU occupation in Belgium

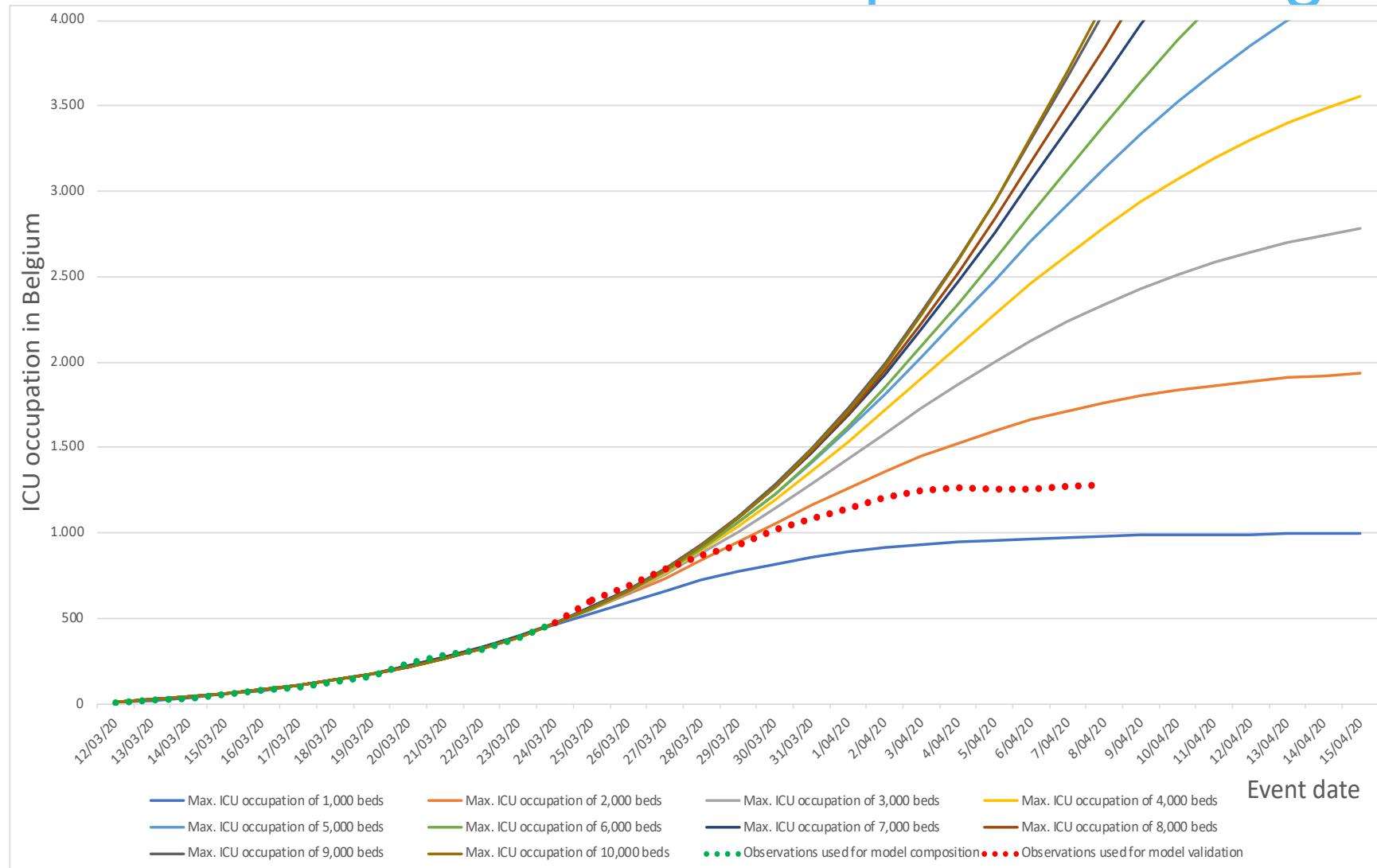


COVID-19 ICU occupation between 25/03 - 08/04/2020

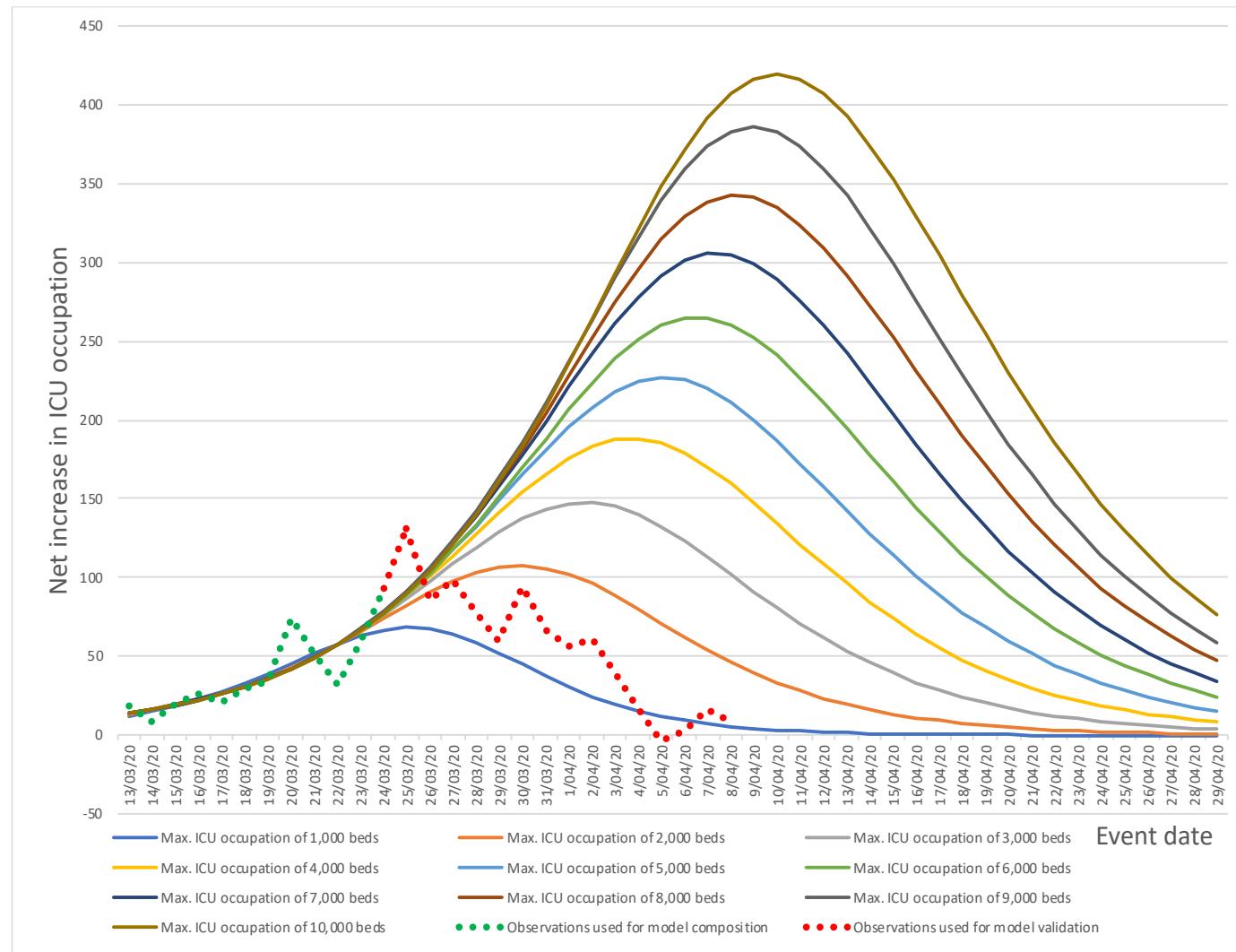
Date	ICU occupation	Date	ICU occupation	Date	ICU occupation
25/03/20	605	30/03/20	1,021	4/04/20	1,261
26/03/20	690	31/03/20	1,088	5/04/20	1,257
27/03/20	789	1/04/20	1,144	6/04/20	1,260
28/03/20	867	2/04/20	1,205	7/04/20	1,276
29/03/20	927	3/04/20	1,245	8/04/20	1,285

Source: Sciensano

Long term evolution of ICU occupation in Belgium



Daily net increase in ICU occupation



Conclusions

- Epidemiological models (e.g. Bayesian SEIR modelling) suffer from accuracy since the basic epidemiological parameters are unknown during the early phases of the emerging virus.
- At such critical moments during an unfolding crisis, scenario-driven forecasting serves as a valid complement for other modelling efforts.
- This dynamic blending has the potential to inform policy makers with accurate predictions in the short term and signal plausible end-states in the scenario portfolio when additional observations become available (requiring temporal distance).
- Our combination of forecast and foresight hence does not signal yet another case in evidence-based policy, but it rather illustrates the good governance of evidence as advocated by Parkhurst (2016).

5. Deceased in Belgium

Insights from:

Decock, K., Bergamini, M., Debackere, K., Lupi, E., Vandamme A.-M. and Van Looy, B. (2020). Predicting when peaks will occur, *ex ante*: Insights from the COVID-19 Pandemic in Italy and Belgium. *Under review*

“Forced quarantaine” in the elderly care centers

Vlaamse rusthuizen vragen verbod op bezoek in verband met coronacrisis

11/03/20 om 11:40 Bijgewerkt om 11:39

Knack

DINSDAG 24 MAART 2020 - DS AVOND



CORONA: DE RUSTHUIZEN

‘Sommigen willen geen levensrekkend
verblijf in het ziekenhuis’

**11 MAART 2020; 18U: PERSBERICHT:
WOONZORGCENTRA MOETEN DEUREN
SLUITEN VOOR ALLE EXTERNE BEZOEKERS**

Deze avond werd een [nieuw persbericht](#) ontvangen vanuit het kabinet van minister Beke.

Vanaf donderdag 12 maart 2020 moeten woonzorgcentra de deuren sluiten voor alle externe bezoekers.



‘Rusthuizen zijn een tikkende tijdbom’

27/03/2020 om 19:46 door [Veerle Beel](#)

KU LEUVEN

Challenges – data quality / registering issues

- Before 31/03: 1 data point, related to the total deceased of the last 24 hours
- On 31/03: reporting of 2 data points, i.e.:
 - 98 as the daily number of deceased (of the last 24 hours)
 - 94 of deceased people since March 11 (retro-active corrections)
- Since then, additional retro-active corrections reported, e.g. on:
 - 02/04: 42 cases
 - 05/04: 43 cases
 - 06/04: 29 cases
- On 07/04: 272 cases related to elderly care homes were allocated from 01/04 till 04/04
- Also on 10/04: another 206 ‘old’ cases allocated between 18/03 and 06/04.

From 10/04 onwards... “data issues” go public

The
Economist

Daily chart

Tracking covid-19 excess deaths across countries

Official covid-19 death tolls still under-count the true number of fatalities

APR 16TH 2020

UK | England | N. Ireland | Scotland | Wales | Politics

BBC NEWS

Coronavirus: Older people being 'airbrushed' out of virus figures

© 14 April 2020

NIEWS CORONASTATISTIEK

deVolkskrant

Huisartsen gaan zelf niet-geteste coronapatiënten bijhouden: beter beeld van virus in Nederland

Nederlandse huisartsen slaan de handen ineen om de coronastatistieken te verbeteren. Ze gaan de ernstig zieke patiënten en overledenen in kaart brengen, die niet op corona zijn getest maar van wie vrijwel zeker vaststaat dat ze besmet zijn of waren met het virus. Gezondheidsdienst RIVM is blij met het initiatief.

Arnout le Clercq 13 april 2020, 17:13

We gaan niet afwijken van het principe dat we zowel bevestigde als vermoedelijke gevallen tellen. Want dat is de standaardpraktijk.

Deel op

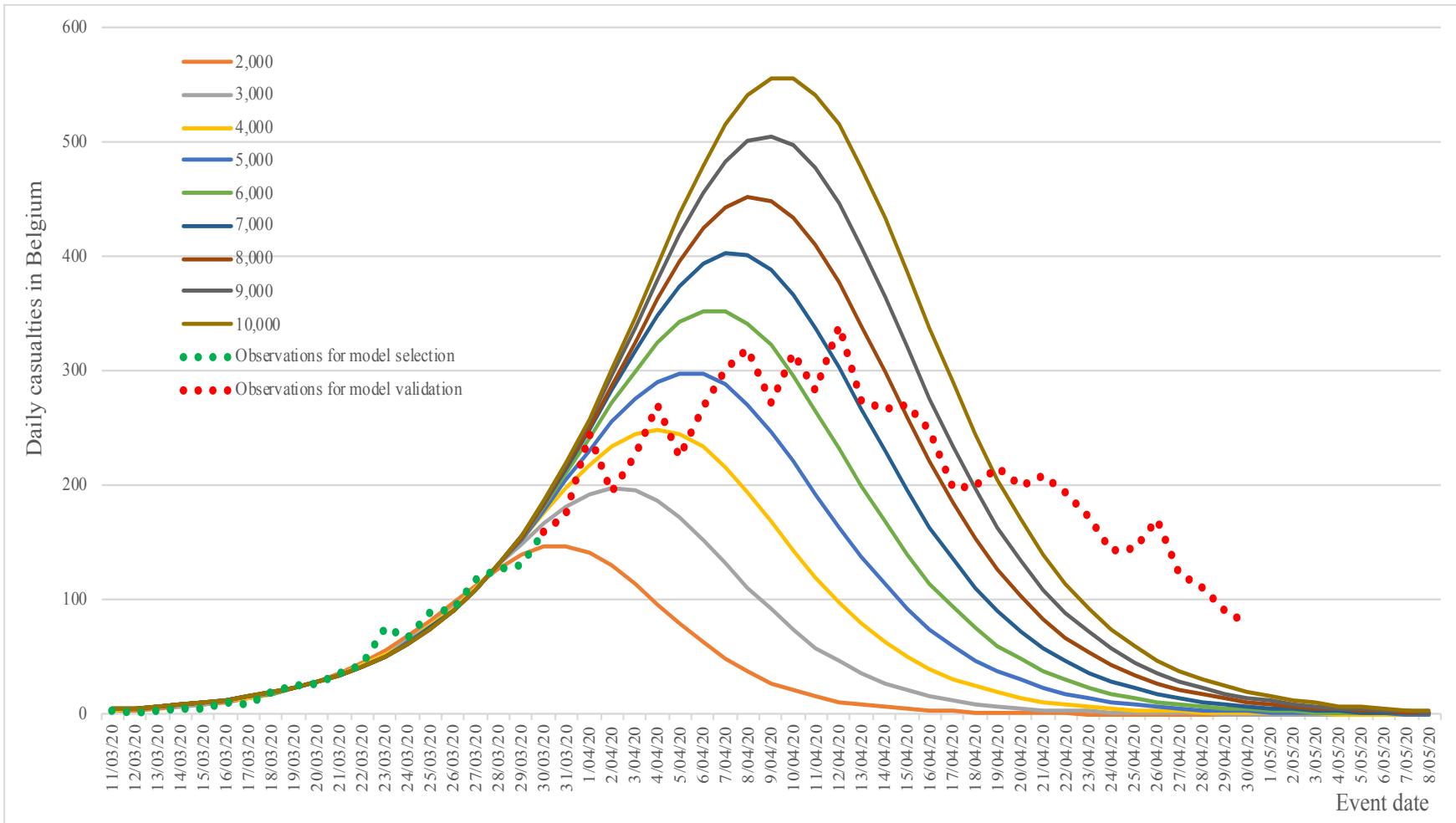
STEVEN VAN GUCHT
VIROOOG VOLKSGEZONDHEID

DE
TIJD

16/04/2020

KU LEUVEN

Daily number of casualties

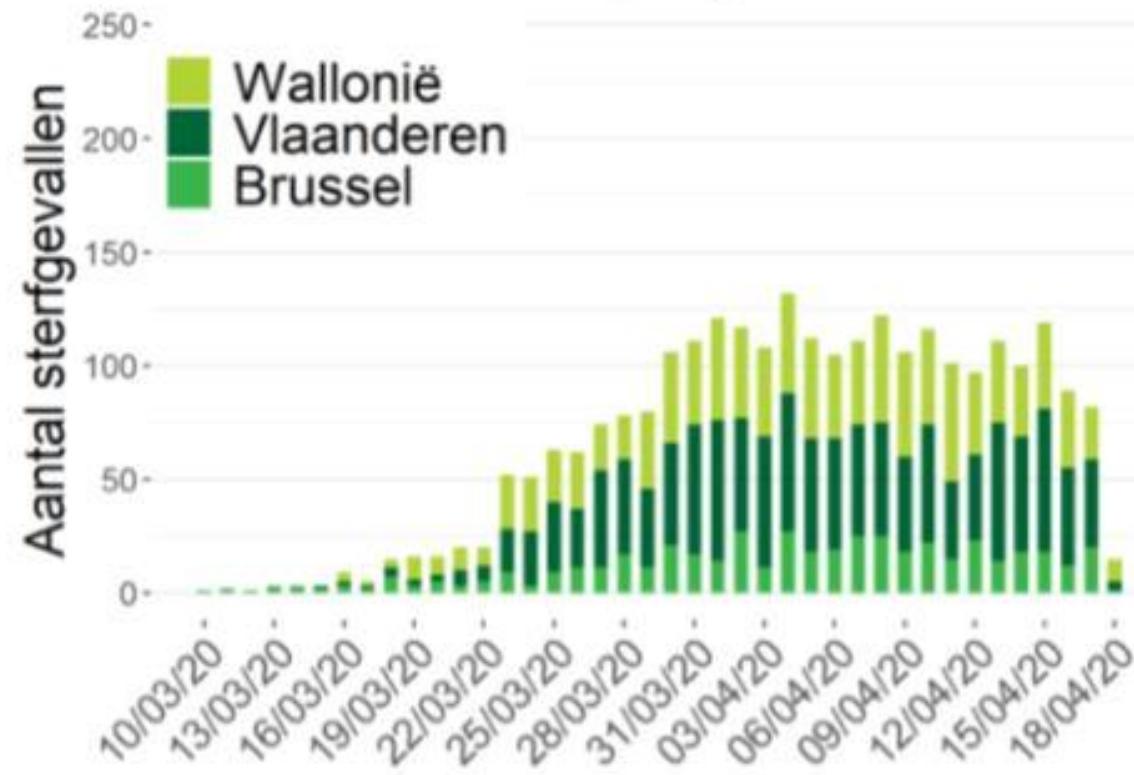


Not one curve, but at least two curves to be modeled...

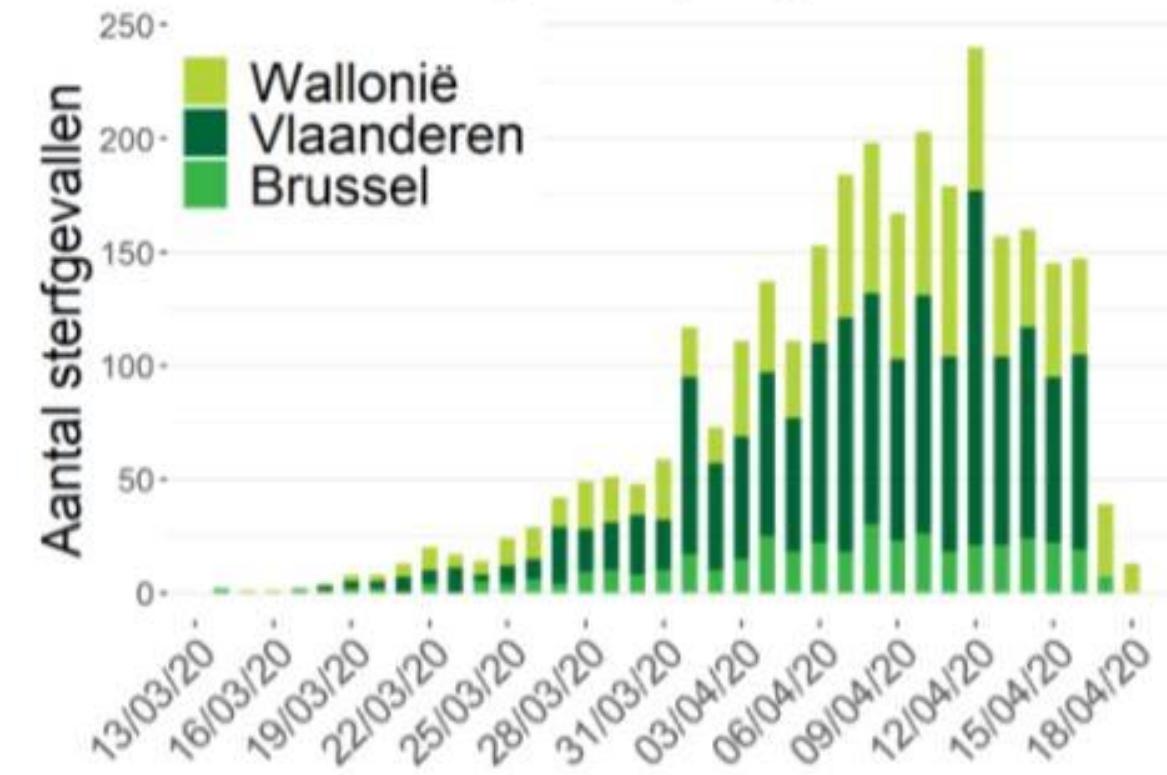
- Underlying (growth) dynamics are different between deceased in the hospitals and deceased outside the hospitals (e.g. elderly care homes)
 - Started later, no direct imports from abroad;
 - Higher contamination due to the location-specific circumstances;
 - Higher death ratio (age, other type (lack) of medical support / equipment, ...);
 - Other age-related issues (more “pre-existing conditions” like diabetes, cardiac problem...);
 - Real cause of death (i.e. hospitals) VS Presumed cause of death (i.e. care homes; only 2,8% of the deaths has been COVID-19 confirmed by a lab test);
 - Different stages of the underlying sub-epidemics (affecting the R-value).
- As such, the two curves need to be modeled separately, as the sum of two S-curves results not always an new S-curve... (other peaks to be reached).

Disentangled curves?!

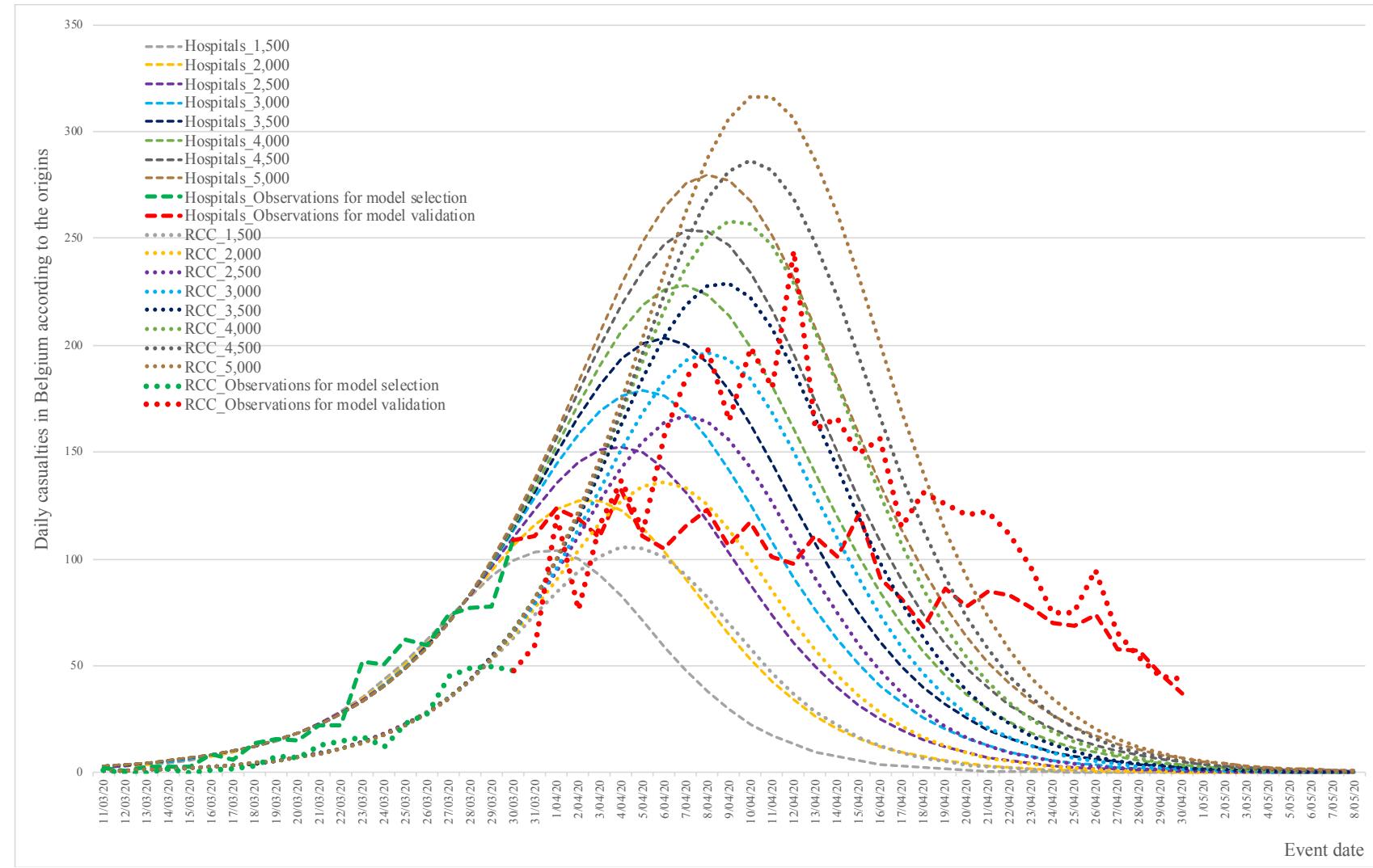
Evolutie van het aantal COVID-19 sterfgevallen in ziekenhuizen per gewest



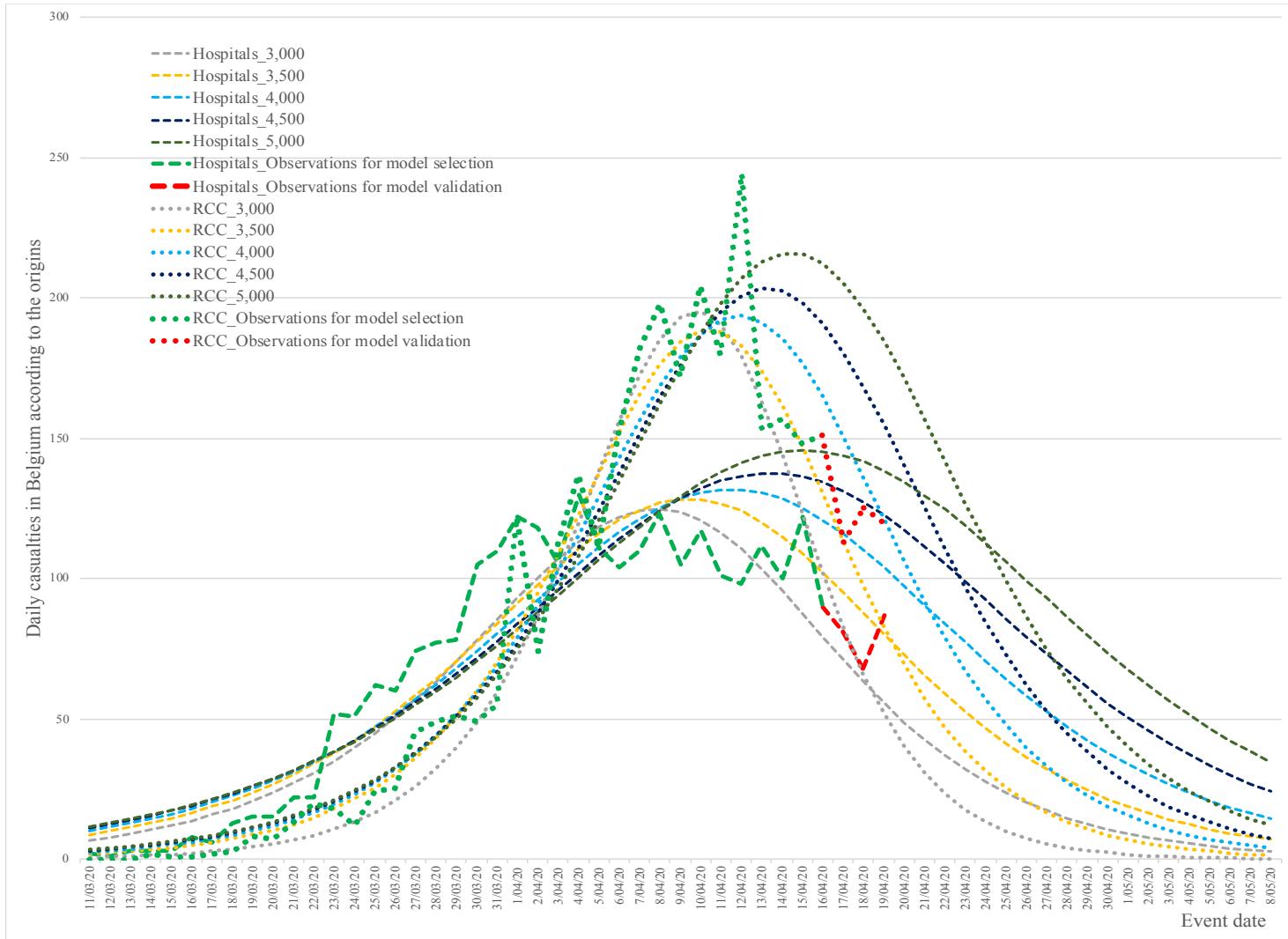
Evolutie van het aantal COVID-19 sterfgevallen in woonzorgcentra per gewest



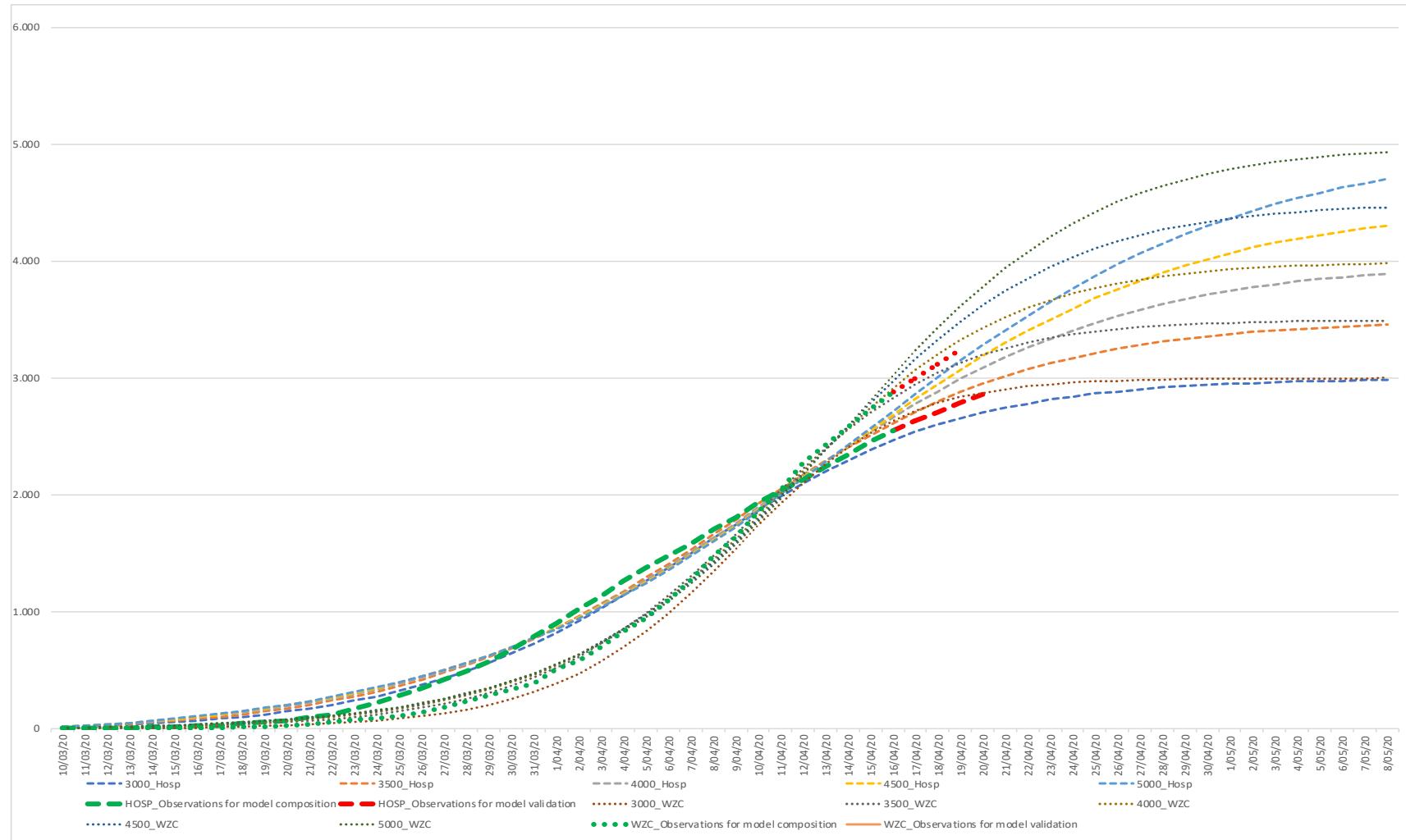
Hospitals versus elderly care homes – DAILY



Model updates – 2 weeks later



Hospitals versus elderly care homes – CUMUL



Key takeaways

1. Scenario-driven forecasting: 1 + 1. Becomes 3?
2. Predictive ability of the outcomes depends critically on reliable / complete / consistent data (incl. separate time series)!
3. Short term accuracy: high (ex ante) – contributing to decision making under ‘Knightian’ uncertainty
4. Importance of understanding the context: impact on the underlying growth dynamics
5. Dis-aggregation (including the reverse logic)
6. ‘Peaks <> End States (flattening of the curves, cohorts, data...)