Appendix

Using early detection data to estimate the date of emergence of an epidemic outbreak

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Contents

1	Rea	nalysis of Roberts et al. (2021)	4			
2	Running our model on simulated data					
3	Sup	plementary tables	7			
4	Alpl	ha SARS-CoV-2 variant in the UK	11			
	4.1	Alpha sequences reported in the UK	11			
	4.2	GISAID data on Alpha cases in the UK	12			
	4.3	Cumulative cases	13			
	4.4	Using the first Alpha case only	14			
5	COV	/ID-19 in Wuhan	15			
	5.1	Early cases of COVID-19 reported in Wuhan	15			
	5.2	Cumulative cases	16			
	5.3	Using other COVID-19 datasets	17			
6	Pseu	udo-algorithm	19			
Re	eferences 21					

List of supplementary figures

А	Reanalysis of Roberts et al. (2021) [1] with updated datasets	4
В	Distribution of the difference (in days) between the estimated date of the first infec- tion and the actual date of the first infection in simulated data	5
С	Estimated dates of first infection, for each of the 100 simulated datasets	6
D	Alpha sequences in the UK	11
Е	Cumulative cases of Alpha variant in the UK	13
F	Estimates of the emergence date using information on the first case only	14
G	Early COVID-19 cases in Wuhan	15
Н	Cumulative cases of COVID-19 in Wuhan	16
Ι	Estimates of the emergence date using truncated or outdated case datasets	17

List of supplementary tables

А	Model calibration	7
В	Estimates for the emergence date	8
С	Sensitivity analyses	9
D	Impact of using different datasets	18

1 Reanalysis of Roberts et al. (2021)

We used the same methodology as Roberts et al. (2021) [1] and ran it on updated datasets to assess the consequences of changes in the input case data. We found that the most recent datasets led to much later estimated dates of the first infection (see Figure A).

Figure A: Reanalysis of Roberts et al. (2021) [1] with updated datasets. The original analysis was done with Huang et al.'s dataset [2]. We re-ran the analysis on updated case datasets, using the same N = 10 number of case-days as in the original analysis.



2 Running our model on simulated data

Here, we ran our model on simulated case datasets and quantified the error of our estimates versus the *observed* date of emergence (observed in the simulation scenario and thus known). We generated 100 simulated datasets and further ran our model, with 1000 replicates for each combination, on subsets of different sizes; that is, we considered the case data up to the day of occurrence of the N^{th} case, for increasing values of $N = \sim 1$, $N = \sim 10$ and $N = \sim 1000$. Overall, the estimated date of first infection is slightly earlier that the real date of first infection (Figure B).

Figure B: Distribution of the difference (in days) between the estimated date of the first infection and the actual date of the first infection in simulated data, for different values of *N*, the number of cases considered for the evaluation. The vertical lines represent the median (full line) and the mean (dashed line) of the distributions. The different simulated datasets are gathered together.



We also explored whether the real date of first infection was within the 95% IPR for each simulation. When combining estimations with N = 1 and N = 1000, all simulations had the real first infection date in the 95% IPR. With individual values of N, the first date of infection could occasionally be outside of the 95% IPR, the estimation being mostly at an earlier date than the simulated data's date (Figure C).

Figure C: Estimated dates of first infection, for each of the 100 **simulated datasets**. The dates of first infection in the simulated datasets are set at 0. The squares show the dates at which there were N = 1, 10, 1000 cases in the simulated dataset, while the diamond, cross and bars show the mean, median and 95% interval of estimated dates of the first infection corresponding to each N, using the simulated dataset as source data. A perfect estimation lands on 0.



3 Supplementary tables

Table A: Model calibration. Our results are obtained from a set of 5 000 simulations selected by the model calibration. Here, we summarize some metrics obtained by calibrating our model for the two applications, and show the observed data for comparison. The values and ranges of estimates correspond to medians and 95% interpercentile ranges.

	Alpha		COVID-19	
Metric	Observed	Estimated	Observed	Estimated
Date of 1st case	Sep 20	Sep 21 (Aug 24–Oct 4)	Dec 10	Dec 8 (Nov 13–Dec 18)
Delay from 1^{st} to N^{th} case (days)	53	51 (38–79)	41	42 (32–67)
Total cases at day of N^{th} case	406	432 (407–465)	3072	~3 360 (3 080–3 690)
Proportion of accepted simulations	-	$\sim 36\%$	-	$\sim 14\%$

Table B: Estimates for the emergence date. Estimated dates of emergence. Median values and central 95% interpercentile ranges (values between 2.5th and the 97.5th percentiles) are shown. Abbreviations: tMRCA= time of most recent ancestor.

Method Date of emergence Earliest infection Study This study Alpha Population dynamics, Aug 21 (Jul 23-Sep 5), 2020 Jun 13, 2020 based on N observations. Population dynamics, Aug 19 (Jul 19–Sep 9), 2020 Jun 2, 2020 [3] updated* based on the first observation. Phylodynamics estimating Aug 28 (Aug 14-Sep 9), 2020 Aug 6, 2020 [4] tMRCA. COVID-19 Population dynamics, Nov 28 (Nov 2–Dec 9), 2019 Sep 26, 2019 This study based on N observations. Phylodynamics model Nov 18 (Oct 16-Dec 6), 2019 Sept 13, 2019 [5] coupling transmission and viral genome evolution.

* Simulations run using the model presented in [3], but with updated parameters matching ours.

Table C: Sensitivity analyses. Estimated dates of emergence, obtained by varying the key epidemiological parameters (reproduction number, over-spreading and detection) as well as the parameters regarding the model calibration (i.e., the tolerances for simulation selection; see details in the Methods section of the main text). Median values and 95% interpercentile ranges (values between 2.5th and the 97.5th percentiles) are shown. Baseline values are marked in boldface.

	R	к	$p_{ m detect}$	$\boldsymbol{\theta}_{\tau}$	$\boldsymbol{\delta}_{\mathbf{tol}}^{Y}$	Date of emergence
Alpha	1.7	0.57	0.0105	12	0.3	Aug 8 (Jul 6–Aug 26), 2020
	1.9	0.57	0.0105	12	0.3	Aug 21 (Jul 23–Sep 5), 2020
	2.1	0.57	0.0105	12	0.3	Aug 29 (Aug 3–Sep 12), 2020
	1.9	0.35	0.0105	12	0.3	Aug 23 (Jul 21–Sep 8), 2020
	1.9	0.57	0.0105	12	0.3	Aug 21 (Jul 23–Sep 5), 2020
	1.9	0.75	0.0105	12	0.3	Aug 20 (Jul 18–Sep 3), 2020
	1.9	0.57	0.0050	12	0.3	Aug 14 (Jul 18–Aug 29), 2020
	1.9	0.57	0.0105	12	0.3	Aug 21 (Jul 23–Sep 5), 2020
	1.9	0.57	0.0250	12	0.3	Aug 28 (Jul 31–Sep 11), 2020
	1.9	0.57	0.0105	7	0.3	Aug 23 (Jul 27–Sep 6), 2020
	1.9	0.57	0.0105	12	0.3	Aug 21 (Jul 23–Sep 5), 2020
	1.9	0.57	0.0105	15	0.3	Aug 19 (Jul 23–Sep 2), 2020
	1.9	0.57	0.0105	12	0.3	Aug 21 (Jul 23–Sep 5), 2020
	1.9	0.57	0.0105	12	0.5	Aug 20 (Jul 24–Sep 4), 2020
COVID-19	2.0	0.10	0.1500	6	0.3	Nov 17 (Oct 20–Dec 2), 2019
	2.5	0.10	0.1500	6	0.3	Nov 28 (Nov 2-Dec 9), 2019
	3.5	0.10	0.1500	6	0.3	Dec 5 (Nov 14–Dec 10), 2019
	2.5	0.05	0.1500	6	0.3	Nov 30 (Nov 7–Dec 10), 2019
	2.5	0.10	0.1500	6	0.3	Nov 28 (Nov 2-Dec 9), 2019
	2.5	0.25	0.1500	6	0.3	Nov 25 (Oct 29–Dec 5), 2019
	2.5	0.10	0.1000	6	0.3	Nov 26 (Oct 31–Dec 8), 2019
	2.5	0.10	0.1500	6	0.3	Nov 28 (Nov 2-Dec 9), 2019
	2.5	0.10	0.2500	6	0.3	Nov 30 (Nov 4–Dec 10), 2019
	2.5	0.10	0.1500	3	0.3	Dec 1 (Nov 5–Dec 10), 2019
	2.5	0.10	0.1500	6	0.3	Nov 28 (Nov 2-Dec 9), 2019
	2.5	0.10	0.1500	12	0.3	Nov 23 (Oct 28–Dec 5), 2019
	2.5	0.10	0.1500	6	0.1	Nov 28 (Nov 4–Dec 10), 2019
	2.5	0.10	0.1500	6	0.3	Nov 28 (Nov 2-Dec 9), 2019
	2.5	0.10	0.1500	6	0.5	Nov 28 (Nov 2–Dec 10), 2019

4 Alpha SARS-CoV-2 variant in the UK

4.1 Alpha sequences reported in the UK

Figure D: Alpha sequences in the UK. We apply our model to 406 sequenced samples (blue) collected up to November 11 (light-blue highlight) and reported by November 30, 2020 (date of submission to GISAID [6]). We do not consider samples collected between November 12 and November 30 to avoid the effects of reporting delays in the last days. For comparison, here we present the sequences reported after November 30, and up to December 15 (white), corresponding to a total of 455 sequenced samples.



4.2 GISAID data on Alpha cases in the UK

SUPPLEMENTAL TABLE

Data Availability

GISAID Identifier: EPI_SET_230104xg 10.55876/gis8.230104xg doi:

All genome sequences and associated metadata in this dataset are published in GISAID's EpiCoV database. To view the contributors of each individual sequence with details such as accession number, Virus name, Collection date, Originating Lab and Submitting Lab and the list of Authors, visit 10.55876/gis8.230104xg

Data Snapshot

- EPI_SET_230104xg is composed of 409 individual genome sequences.
 The collection dates range from 2020-09-20 to 2020-11-11;
- Data were collected in 1 countries and territories;
- All sequences in this dataset are compared relative to hCoV-19/Wuhan/WIV04/2019 (WIV04), the official reference sequence employed by GISAID (EPI_ISL_402124). Learn more at https://gisaid.org/WIV04.

4.3 Cumulative cases

Figure E: Cumulative cases of Alpha variant in the UK. Accepted simulations (gray) and observed data (black).



Date of sample collection (2020)

4.4 Using the first Alpha case only

Figure F: Estimates of the emergence date using information on the first case only. Distributions of the emergence date for the Alpha variant in the UK, estimated using data on the first observed Alpha case (i.e., September 20, 2020) only, using our model (violet, upper middle) and the model previously developed by (Czuppon et al., 2021) as published (yellow, bottom) and updating the parameters to match ours (red, lower middle; cf. the Methods section of the main text). We also plot the results from running our model on data of samples collected up to November 11, 2020 (i.e., our main results; blue, top), for comparison.



Date of emergence (2019)

5 COVID-19 in Wuhan

5.1 Early cases of COVID-19 reported in Wuhan

Figure G: Early COVID-19 cases in Wuhan. We apply our model to 3072 COVID-19 reported cases with symptoms onset by January 19, 2020 (blue). On January 20, the first public declaration of human-to-human transmission of the virus was made. Soon after that, the lockdown intervention was deployed nationally, along with testing, which explains the change in epidemic dynamics observed after that date (white).



5.2 Cumulative cases

Figure H: Cumulative cases of COVID-19 in Wuhan. Accepted simulations (gray) and observed data (black).



Date of symptoms onset (2019--2020)

5.3 Using other COVID-19 datasets

Shorter dataset. Here, we analyze the impact of applying our model to the dataset of COVID-19 cases (cf. Figure S2) with symptoms onset before December 31, 2019, the day of the first public declaration of the Wuhan cluster [7]. Based on the 169 reported cases of COVID-19 with symptoms onset before December 31, we estimate a date of emergence by November 24 (October 29–December 04), 2019, and not earlier than September 26, 2019.

Anterior dataset. In addition, we apply our model to case data published by the WHO in 2020 [8], which were later corrected [9]. The WHO data sets used in our model are adapted from the data sets available in a GitHub repository that summarizes published epidemic curves of early COVID-19 cases [10]. Results are summarized in Table D.

Figure I: Estimates of the emergence date using truncated or outdated case datasets.

Distributions of the emergence date for the COVID-19 epidemic in Wuhan, estimated using data on cases with symptoms onset by December 31, 2019 (violet, middle) and by January 19, 2020 (i.e., our main results; blue, top), as well as an outdated, later corrected dataset (red, bottom). The first cases reported in the datasets are depicted by dashed lines colored accordingly.



Date of emergence (2019)

Table D: Impact of using different datasets. Results obtained using different COVID-19 cases datasets. The estimated time elapsed between the first infection to the N^{th} observed case yields the estimated date of outbreak emergence. In addition, we estimate the epidemic size at the date of detection of the N^{th} case. The proportions of detected infections are retrieved for comparison with the input epidemic parameters (cf. Table 2 of the main text). Median and 95% interpercentile ranges (i.e., values between 2.5th and the 97.5th percentiles) are shown, unless stated otherwise.

		Case dataset	
	Pekar et al. 2022 ^{<i>a</i>}	Pekar et al. 2022	WHO 2020
Observations			
Date of N^{th} case	Jan 19, 2020	Dec 31, 2019	Dec 31, 2019
Date of first case	Dec 10	Dec 10	Dec 2
Number of reported cases	3 072	169	202
Simulations			
Emergence date (2019)	Nov 28 (Nov 2-Dec 09)	Nov 24 (Oct 29–Dec 6)	Nov 22 (Oct 28-Dec 2)
Date of earliest infection	Sep 26, 2019	Sep 24, 2019	Sep 23, 2019
Number of cases	~3 360 (3 080–3 690)	180 (169–210)	220 (202–250)
Epidemic size ^b	~63 400 (57 300–69 900)	~3 500 (2 680–4 240)	$\sim 4160\;(3280 - 5000)$
Proportion of detected infections	5.31% (5.06%-5.57%)	5.3 (4.52-6.66)	5.32 (4.55-6.49)

^{*a*} Dataset used in our baseline analyses.

^{*b*} At day of infection of the N^{th} case.

6 Pseudo-algorithm

The code was built in Julia and is available at a public Github repository: https://github.com/sjijon/estimate-emergence-from-data.

Algor	rithm A Estimate the delay between the first infection and the $N^{{f th}}$ ca	ase.
Abbre	eviations: dist. = distribution.	
Requ	tire: Time-series of the N first observed cases, $\{d_i^{obs}\}_{i=1,\dots,K'}$	
	Context-specific epidemiological parameters (Distributions)	
	Lower bound for the number of infections (I_{\min})	
	Upper bound for the running time (t_{max})	
	Number of simulations (num_sims)	
	Error tolerance (δ_{tol})	
Ensu	re: Time elapsed between first infection and N^{th} observed case	
	The simulated underlying epidemic (SimEpi)	
	The simulated cases (SimCases)	
1: p	procedure Read Data	Returns ObsCases
2:	Read available data on cases	
3:	Retrieve the number of observed cases, N, and their dates of occure	ence.
4: e i	nd procedure	
5: W	vhile sim_num≤ num_sims do	
6:	procedure Transmission model	⊳ Returns SimEpi
7:	Initialize infections $I(t_0) = 1$	
8:	while $t < t_{\max} \operatorname{do}$	
9:	Draw the number of secondary infections	• Negative binomial dist.
10:	if offspring was generated then	
11:	Compute the time of those secondary infections, $\{t_i\}_{i=1,2,3}$	⊳ Gamma dist.
12:	Update number of infected at time of infection, $I(t)$	
13:	end if	
14:	Increase time step	
15:	end while	
16:	Compute daily number of infections, $I(d_k)$, $k = 1, 2,$	
17:	end procedure	
	⊳…Con	tinues on the next page.

Algorithm A (continuation)

18:	if $I \ge I_{\min}$ then	
19:	procedure Detection model	▹ Returns SimCases
20:	k=1	
21:	for $d_k, k = 1, 2, \dots$ do	
22:	Draw the daily number of detected infections among $I(d_k)$	⊳ Binomial dist.
23:	Compute detection time for each individual, $\{\tau_i\}_{i=1,\dots,M}$	⊳ Gamma dist.
24:	Increase the daily number of cases, $Y(d_i)$	
25:	if N^{th} case detected then	
26:	Keep only the detections occurring on the same day as N-th case	
27:	Exit loop	
28:	end if	
29:	k=k+1	
30:	end for	
31:	end procedure	
32:	end if	
33:	Identify the epidemic size at the day where the N^{th} case was infected	
34:	if $N_{\text{cases}} \ge N$ then	
35:	procedure Compare SimCases and ObsCases	
36:	Condition C1: $d_1 \ge t_1$	
37:	Condition C2: $ Y^{obs}(d_k) - Y^{sim}(d_k) \le \delta'_{tol} N$, $\forall k = K, K-1, K-2,$	
38:	if Conditions C1–C2 hold then	
39:	$sim_num = +1$	▷ Select simulation
40:	end if	
41:	end procedure	
42:	end if	
43: e i	nd while	

20

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