

Short-term forecasting of total Number of reported COVID-19 cases in South Africa - A Bayesian temporal modeling approach

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Abstract

To be updated.

Author summary

To be updated.

Introduction

In this paper we present (1) South Africa's COVID trajectory to the first 100,000 (22 June 2020) cases and (2) fit a series of non-linear growth models, calibrated to COVID-19 cumulative number of reported case data from 5 March 2020 to 22 June 2020. The models are used to produce short term predictions of the number of reported cases expected for a period of 30 days ahead. These forecasts are generated at the national level

Methods

Data

We downloaded data from Coronavirus COVID-19 (2019-nCoV) Data Repository for South Africa maintained by Data Science for Social Impact research group at the University of Pretoria [ref]. The data repository captures the daily number of new cases, number of tests, number of deaths and recoveries. Our primary outcome of interest was the daily number of newly diagnosed COVID-19 cases and the unit of time used in modelling was a day. We used the daily case reports from March 12, 2020, until February 27, 2021, in our analysis.

Statistical analysis

We considered two widely used temporal models to model the daily number of newly diagnosed COVID-19 cases. We let $Y(t)$ denote the daily number of newly diagnosed COVID-19 cases at time t and $\mu(t)$ represent the expected number of cases at time t . We considered a Negative binomial distribution for $Y(t)$ to account for possible overdispersion. That is, $Y(t) \sim NB(\mu(t), \delta)$, where δ is the overdispersion parameter. We considered two temporal models to capture the trend over time: a random walk of order two ($RW(2)$) and an autoregressive model of order one ($AR(1)$) [1]. We also

considered a $RW(1)$ model, but the model overfits the data (See the supplementary appendix). Similarly, we considered an AR model order $p = 2$ but the result is similar to $AR1$ model and we prefer the simpler $AR1$ model.

The $AR(1)$ model [1] is given by,

$$\begin{aligned} Y(t) &\sim NB(\mu(t), \delta) \quad t = 1, \dots, n, \\ \log(\mu(t)) &= \alpha + u_t, \\ u_1 &\sim N(0, \tau_u(1 - \rho^2)^{-1}), \\ u_t &= \rho u_{t-1} + \epsilon_t, \quad t = 2, \dots, n, \\ \epsilon_t &\sim N(0, \tau_\epsilon), \end{aligned}$$

where, α is an intercept, ρ a temporal correlation term (with $|\rho| < 1$) and ϵ_t is a Gaussian error term with zero mean and precision τ_ϵ .

Similarly, the $RW(2)$ model [1] is given by,

$$\begin{aligned} Y(t) &\sim NB(\mu(t), \delta) \quad t = 1, \dots, n, \\ \log(\mu(t)) &= \alpha + u_t, \\ u_t - 2u_{t+1} + u_{t+2} &\sim N(0, \tau_u), \quad t = 2, \dots, n, \end{aligned}$$

where α is the intercept term as before and τ_u is the precision parameter.

The two models were fitted within the Bayesian framework using *inla* [2]. To complete the specification of both models, we assume the following priors. For the $AR(1)$ model, we denote $\theta_1 = \log(\tau_u(1 - \rho^2))$ where $\Gamma(10, 100)$ prior is specified for θ_1 , and we denote $\theta_2 = \log \frac{1+\rho}{1-\rho}$ and assume a $N(0, 0.15)$ prior for θ . Similarly, we represent the precision parameter of $RW(1)$, τ_u , as $\theta = \log(\tau_u)$ and assume a $\Gamma(10, 100)$ prior for θ .

To assess the models' accuracy in predicting COVID-19 cases, we present the forecast period's actual observed values and the predicted values. Additionally, the model fits were evaluated by using DIC (Deviance information criteria).

The R-code that we used for our analyses is available at <https://github.com/belayb/COVIDincidenceSA/tree/master/COVIDincidenceSA>.

Results

The daily number of reported COVID-19 cases from 12 March 2020 to 27 February 2021 is presented in Figure 1. Similar to elsewhere in the world, South Africa pass through a two-wave pandemic. The pandemic first peak was on 07 July 2020, where up to 13944 new COVID-19 cases reported, followed by a second peak in January 2021, where more than 21,000 daily cases reported. Figure 2 presents the cumulative number of reported COVID-19 cases and tests performed. To date, 8,838,937 tests have been conducted, and a total of 1,500,677 cases reported.

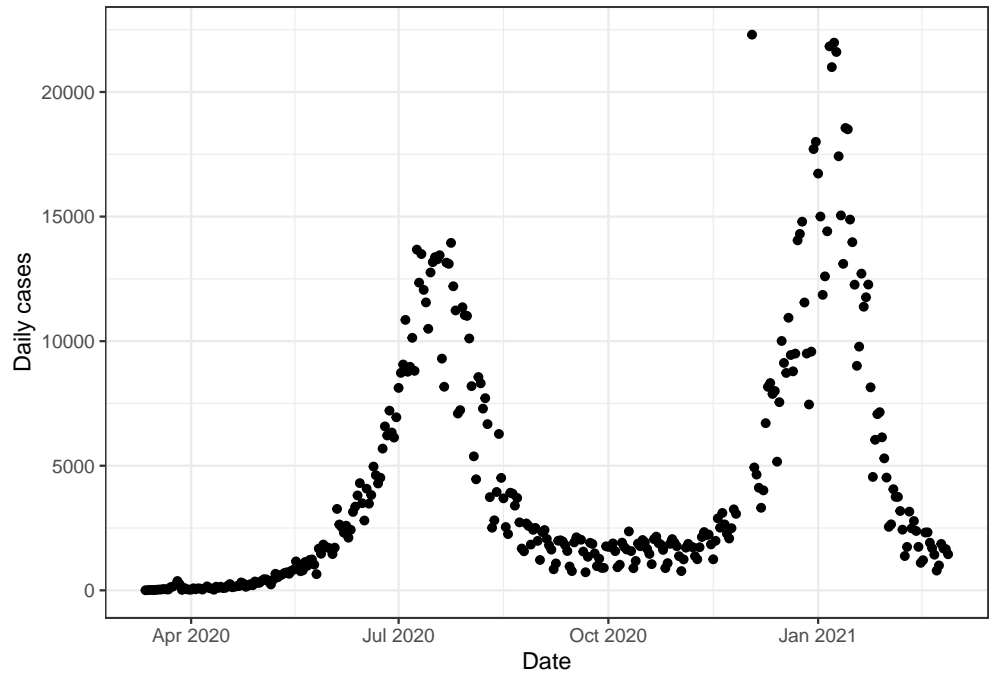


Fig 1. Daily number of COVID-19 cases in South Africa from 12/03/2020-27/02/2021.

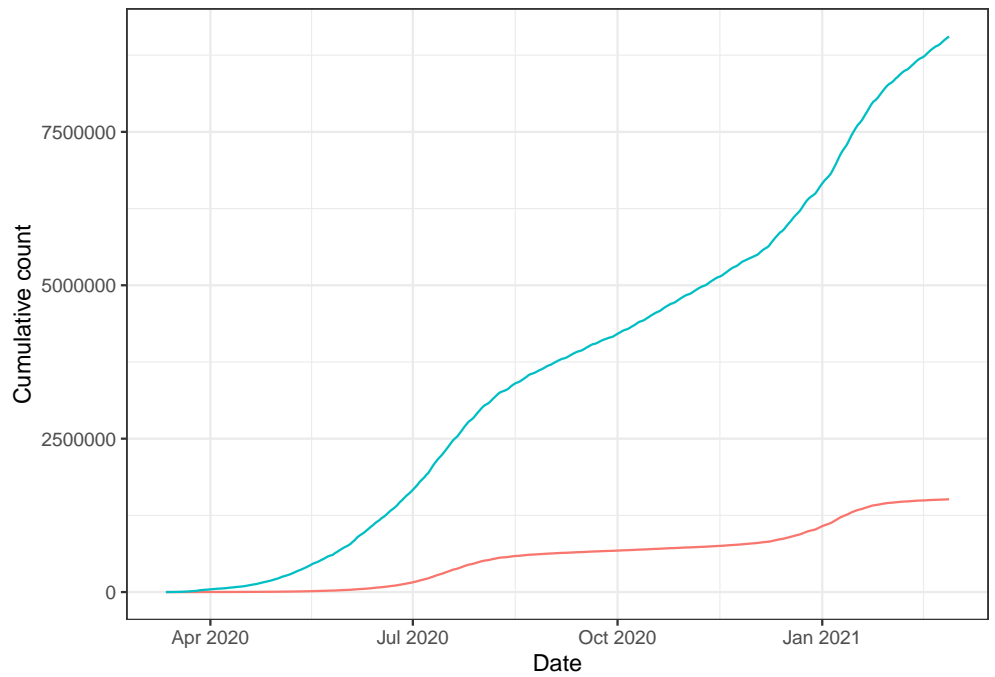


Fig 2. The cummulative number of COVID-19 cases and Cummulative number of tests in South Africa from 12/03/2020-27/02/2021. Red-line denote the number of cases and blue-line denotes the number of tests.

Short-term prediction of the total number of reported COVID-19 cases

We fit the four models described in the previous section to the daily reported new COVID-19 cases. The parameter estimates for each of the fitted models are presented in Supplementary Table 1. Our main interest is to estimate the rate of change of daily confirmed COVID-19 cases and produce a short term forecast for number of reported cases.

As depicted in Figure 3, two models fitted to the data appear to fit the observed data (within the estimation period) well with a narrow confidence interval obtained for the $RW(2)$ model. The two models provide similar predictions over the 7-day ahead period.

Table 1. Accuracy metrics of forecasting for AR1, AR2, RW1, and RW2 models.

Days ahead point forecasts	Mean Absolute Error			
	RW1	RW2	AR1	AR2
One day	449.2918	3482.401	1201.786	1056.635
Two day	1308.8048	3575.291	1871.106	1716.237
Three day	2351.2483	3758.596	2777.020	2636.543
Four day	3327.8508	3890.326	3620.173	3460.322
Five day	4227.1623	3951.276	4348.963	4164.685
Six day	4982.3087	3909.428	4893.645	4679.546
Seven day	5821.8832	3959.609	5479.801	5230.147
	Mean Absolute Percentage Error			
	RW1	RW2	AR1	AR2
One day	0.0003005	0.0023281	0.0008036	0.0007065
Two day	0.0008742	0.0023875	0.0012498	0.0011463
Three day	0.0015690	0.0025074	0.0018531	0.0017593
Four day	0.0022186	0.0025927	0.0024135	0.0023069
Five day	0.0028151	0.0026302	0.0028962	0.0027734
Six day	0.0033140	0.0025989	0.0032550	0.0031125
Seven day	0.0038681	0.0026291	0.0036407	0.0034748
	Information Criteria			
	DIC	WAIC		
DIC	5059.06	5447.40	5278.87	5284.00
WAIC	5123.56	5460.81	5286.90	5294.74

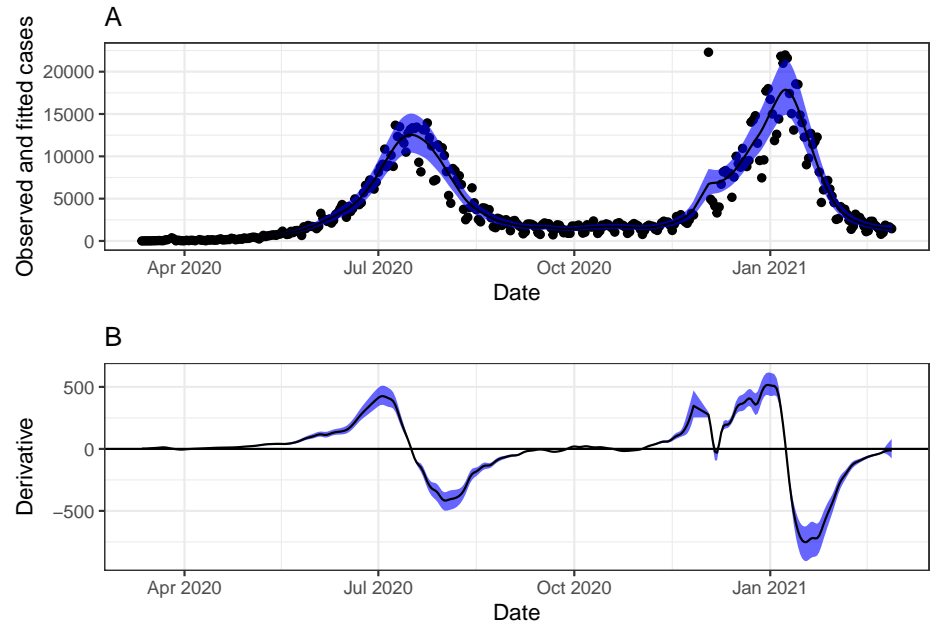


Fig 3. RW2 model for the daily confirmed COVID-19 cases in South Africa 12/03/2020-27/02/2021. Panel A-fitted and observed data. The black dots are the observed number of daily cases, the black solid line the fitted curve, and the blue shaded area is the 95% credible interval. Panel B-first-order derivative of the fitted curve along with the 95% credible interval.

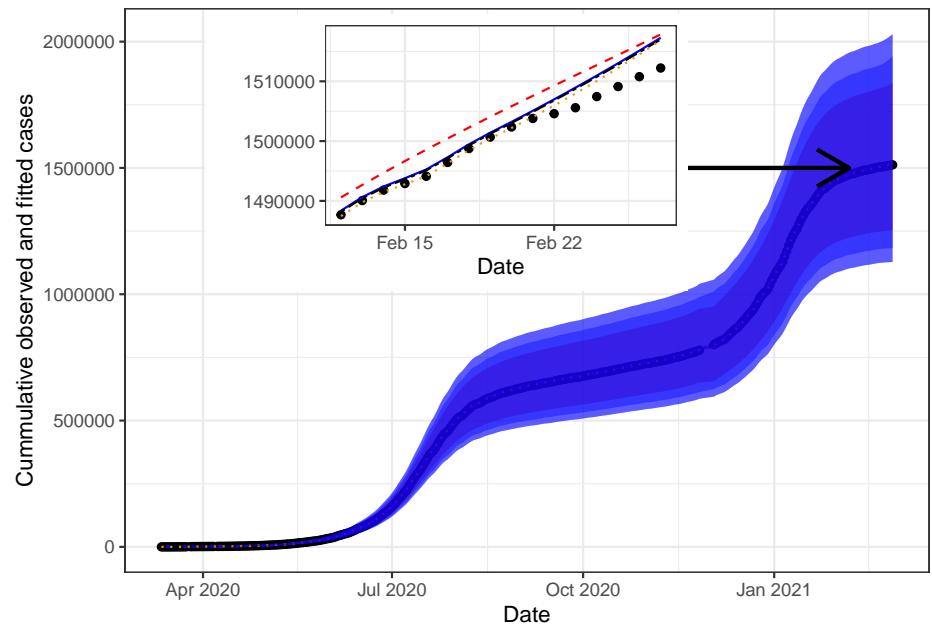


Fig 4. Predicted cumulative COVID-19 cases in South Africa under the RW(1) and AR(1) model. Estimation period 12/03/2020-20/02/2021. The black dots are the observed cumulative cases. The red dashed lines are for RW(2) and the blue line for AR(1). The shaded bands are the prediction intervals.

Table 2. Short-term predictions of total number of reported cases at the national level under the rw2 model. Estimation period 12/03/2020-20/02/2021

Model	Date	Total	Prediction	Prediction Interval	Total - Prediction
RW1	2021-02-21	1503796	1504037	(1179288.4-1906832.48)	-241.2192
	2021-02-22	1504588	1505956	(1180039.06-1910889.23)	-1367.7046
	2021-02-23	1505586	1507956	(1180669.91-1915712.14)	-2370.0667
	2021-02-24	1507448	1510042	(1181213.73-1921302.31)	-2593.7729
	2021-02-25	1509124	1512217	(1181690.53-1927674.02)	-3092.5479
	2021-02-26	1510778	1514484	(1182113.69-1934850.96)	-3706.3254
	2021-02-27	1512225	1516849	(1182492.77-1942862.03)	-4624.2376
RW2	2021-02-21	1503796	1507571	(1248627.38-1814985.53)	-3774.603
	2021-02-22	1504588	1509293	(1249674.84-1817725.91)	-4704.805
	2021-02-23	1505586	1511002	(1250614.44-1820669.35)	-5415.664
	2021-02-24	1507448	1512704	(1251450.03-1823858.23)	-5255.522
	2021-02-25	1509124	1514405	(1252187.39-1827339.44)	-5281.407
	2021-02-26	1510778	1516115	(1252833.56-1831166.58)	-5337.172
	2021-02-27	1512225	1517842	(1253396.63-1835400.97)	-5616.650
AR1	2021-02-21	1503796	1505071	(1124486.84-1998035.3)	-1274.995
	2021-02-22	1504588	1506982	(1125286.9-2001911.8)	-2393.764
	2021-02-23	1505586	1508942	(1125972.18-2006363.99)	-3356.174
	2021-02-24	1507448	1510953	(1126572.19-2011372.04)	-3504.598
	2021-02-25	1509124	1513014	(1127105.33-2016924.89)	-3889.609
	2021-02-26	1510778	1515126	(1127584.21-2023016.99)	-4347.854
	2021-02-27	1512225	1517290	(1128017.98-2029646.04)	-5065.020
AR2	2021-02-21	1503796	1504897	(1123980.48-1998091.99)	-1101.411
	2021-02-22	1504588	1506802	(1124788.11-2001923.68)	-2213.521
	2021-02-23	1505586	1508752	(1125480.87-2006309.88)	-3165.515
	2021-02-24	1507448	1510747	(1126088.3-2011227)	-3299.344
	2021-02-25	1509124	1512789	(1126628.81-2016661.08)	-3665.250
	2021-02-26	1510778	1514878	(1127115.05-2022603.28)	-4099.507
	2021-02-27	1512225	1517012	(1127556.18-2029047.89)	-4787.420

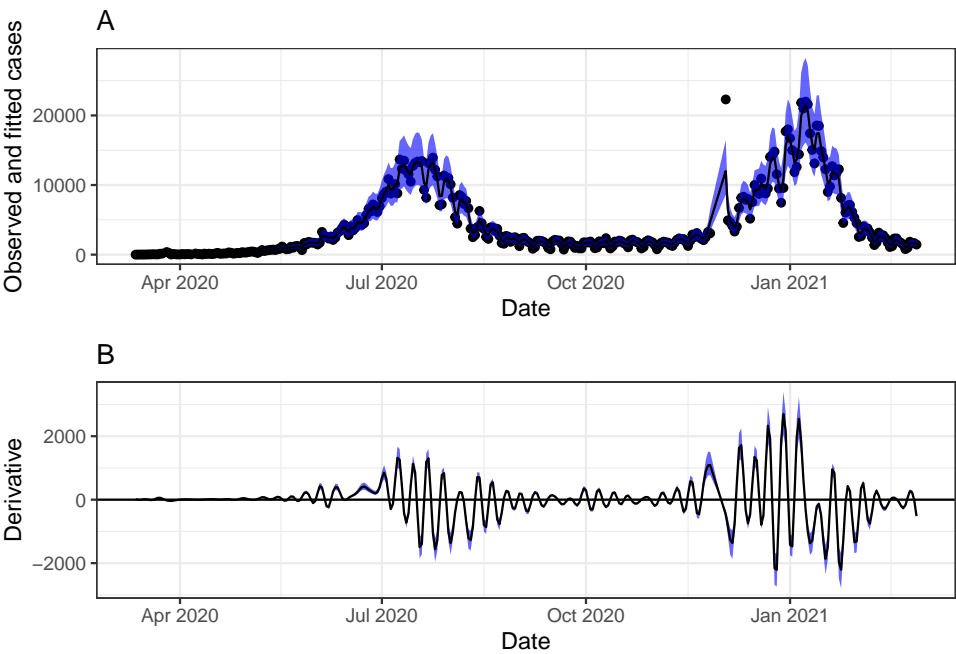


Fig S1. Fitted and observed data AR(1) model

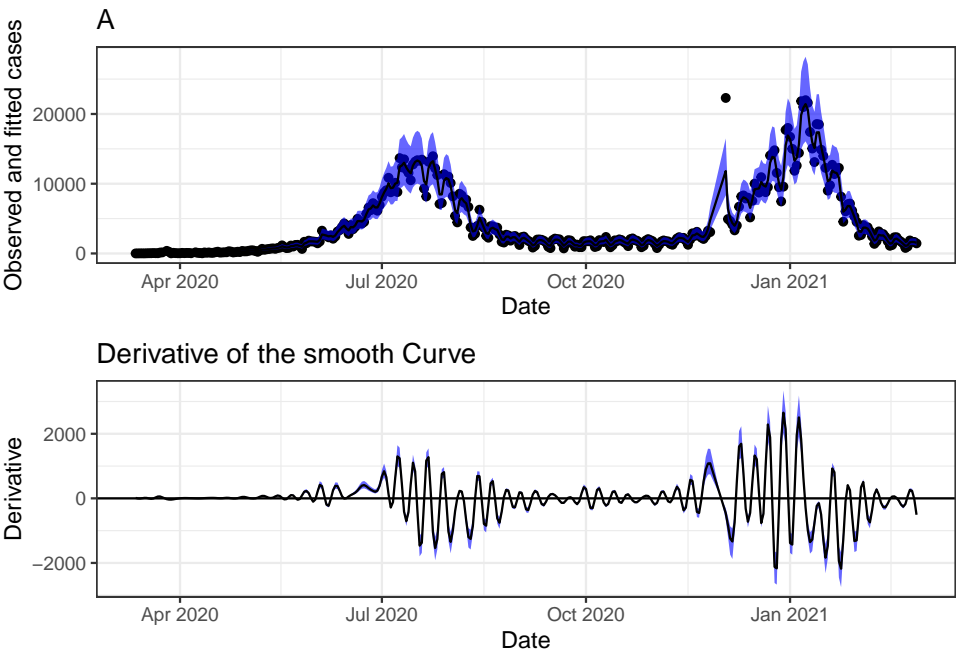


Fig S2. Fitted and observed data AR(2) model

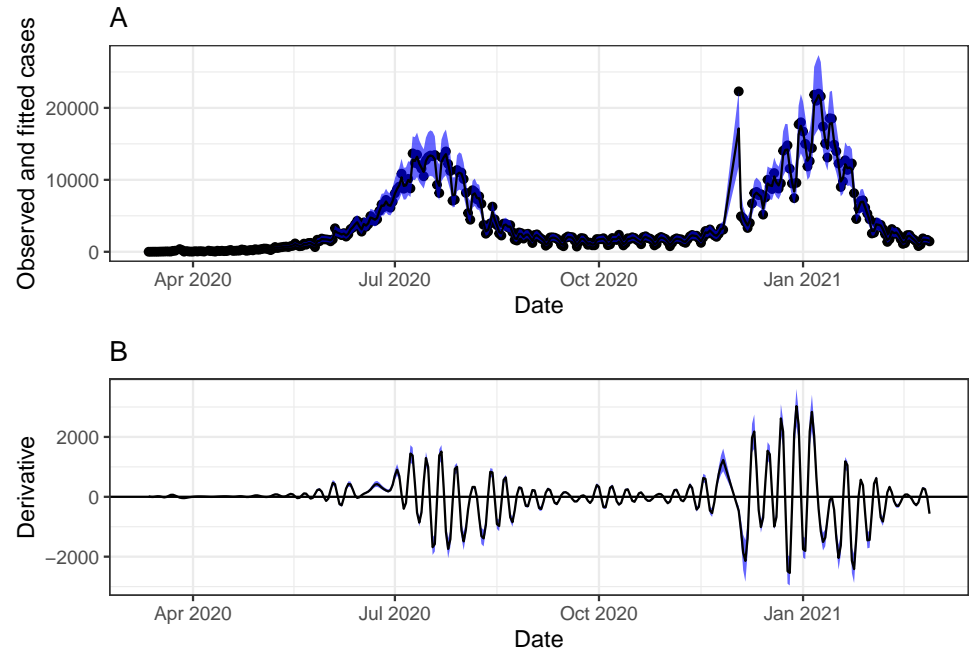


Fig S3. Fitted and observed data RW(1) model

Table S1. Parameter estimates AR1 model

	mean	sd	0.025quant	0.975quant
(Intercept)	6.109	2.441	0.145	10.569
Size	28.955	8.389	16.805	49.397
Precision for time	0.158	0.112	0.023	0.438
Rho for time	0.995	0.004	0.985	0.999

Table S2. Parameter estimates RW2 model

	mean	sd	0.025quant	0.975quant
(Intercept)	7.603	0.017	7.570	7.637
Size	10.422	0.911	8.734	12.319
Precision for time	0.036	0.014	0.016	0.070

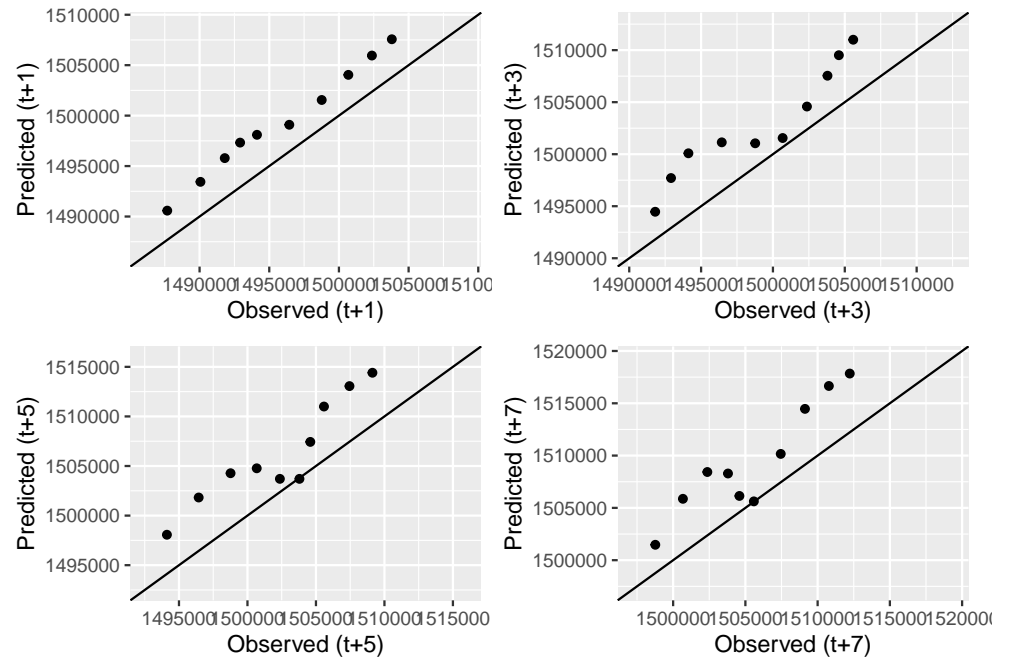


Fig S4. Predicted vs observed cummulative COVID-19 cases in South Africa under the RW(2) model. Base estimation period 12/03/2020-10/02/2021. The estimation period was expanded by 1 day until 20/02/2021

References

1. Gómez-Rubio V. Bayesian inference with inla. CRC Press; 2020.
2. Martins TG, Simpson D, Lindgren F, Rue H. Bayesian computing with inla: New features. Computational Statistics & Data Analysis. Elsevier; 2013;67: 68–83.