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

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Abstract

To be updated.

Author summary

To be updated.

Introduction

To be updated

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 considerd a $RW(1)$ model, but the model overfits the data (See the supplemntary appendix). Similarly, we considerd 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,

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

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,

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

where α is the intercept term as before and τ_u is the precison 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 precison 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 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 computer code that we used for our analyses is avaliable at https://github.com/belayb/COVIDincidenceSA/tree/master/COVIDincidenceSA.

Results

Figure 1 presents the daily number of reported COVID-19 cases from 12 March 2020 to 27 February 2021. Similar to elsewhere in the world, South Africa pass through a two-wave pandemic. The pandemic's 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 new 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.

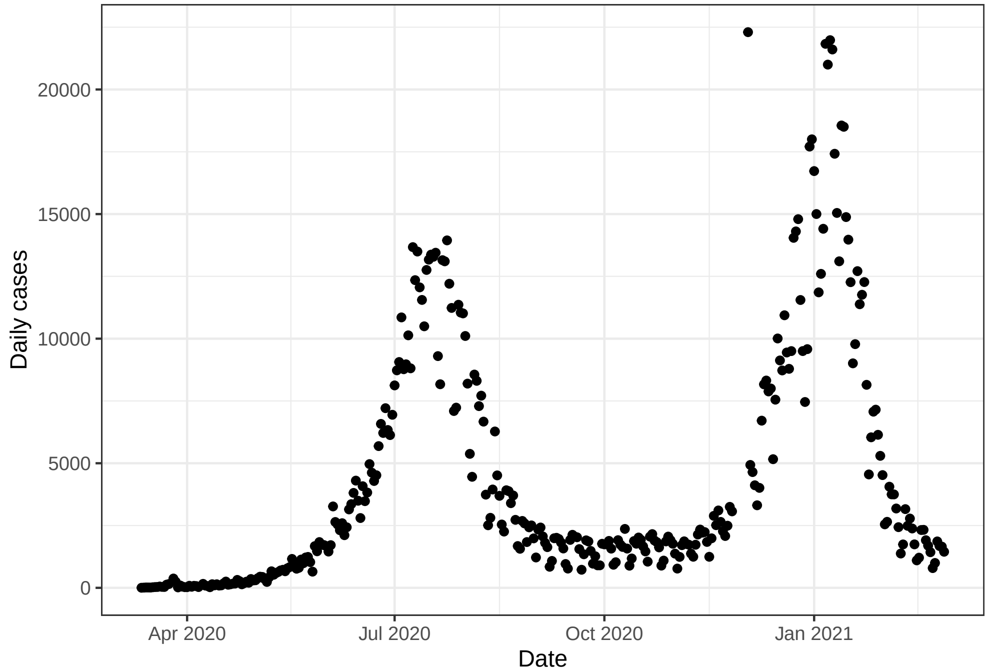

Fig 1. Daily number of COVID-19 cases in South Africa from 12/03/2020-27/02/2021.

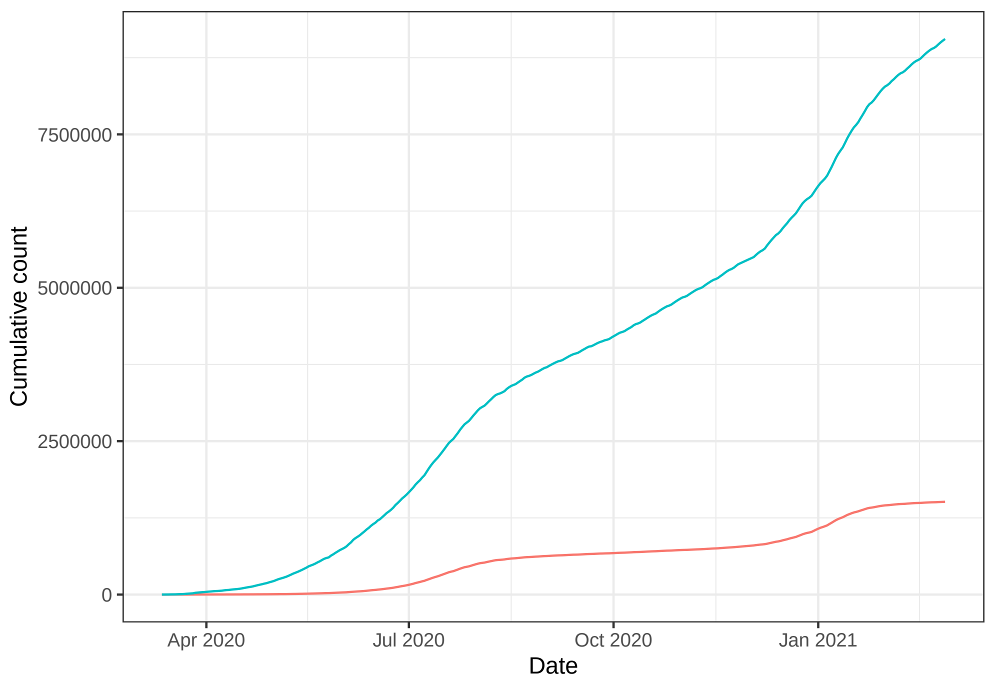

Fig 2. The cummulative number of COVID-19 cases and Cummulative number of tests in South Africa from 12/03/2020-27/02/2021. Red-line denote the number of cases and blue-line denotes the number of tests.

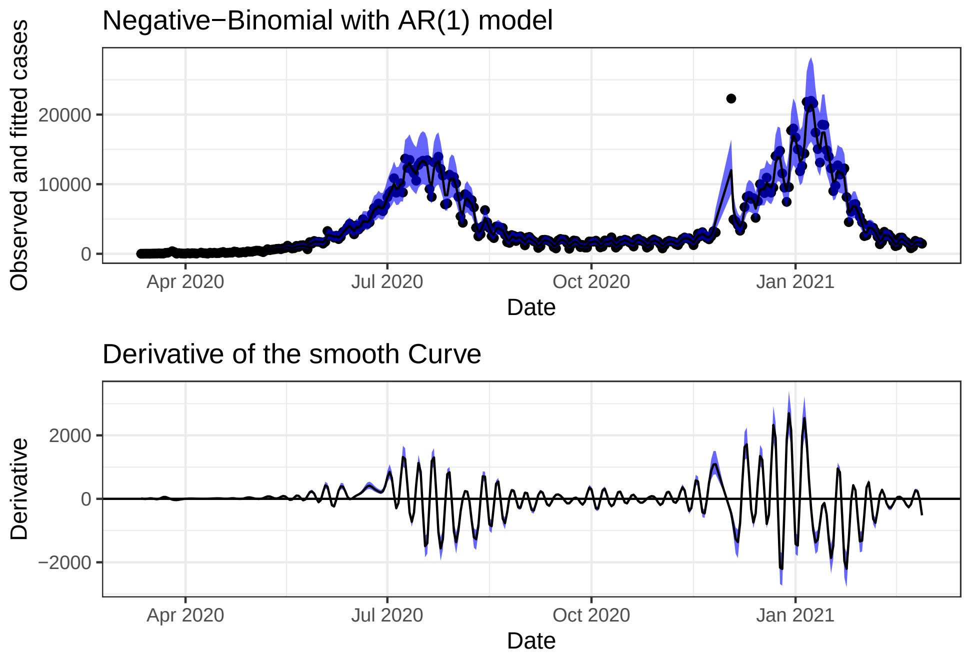

Fig 3. Fitted and observed data AR(1) model

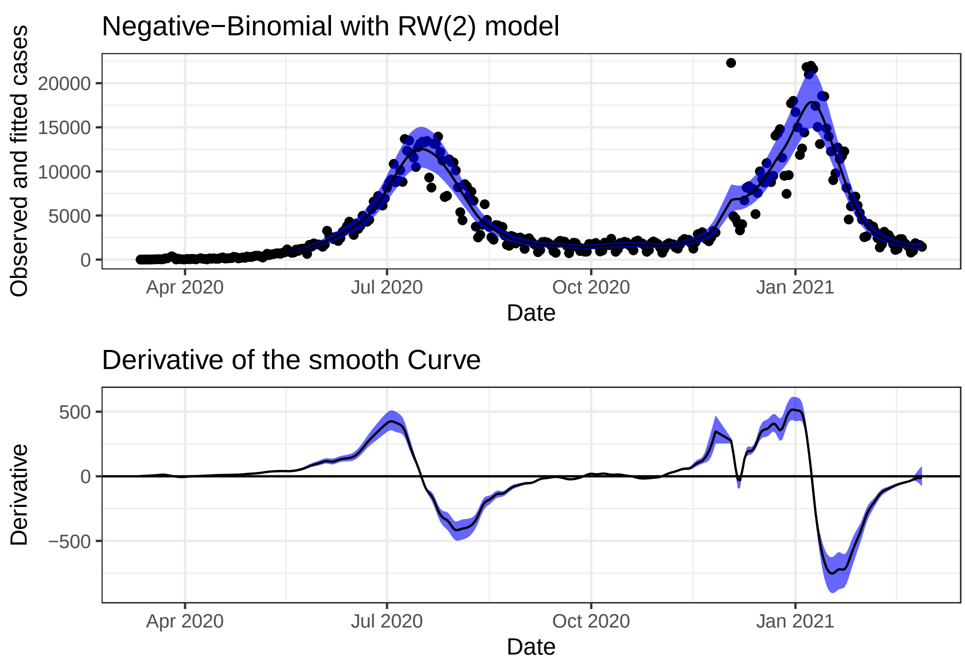

Fig 4. Fitted and observed data RW(2) model

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

We fit the two models described in the previous section to the daily reported new COVID-19 cases. The parameter estimates for the two models are presented in Supplementary Table 1. 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 provides similar predictions over the 7-day ahead period.

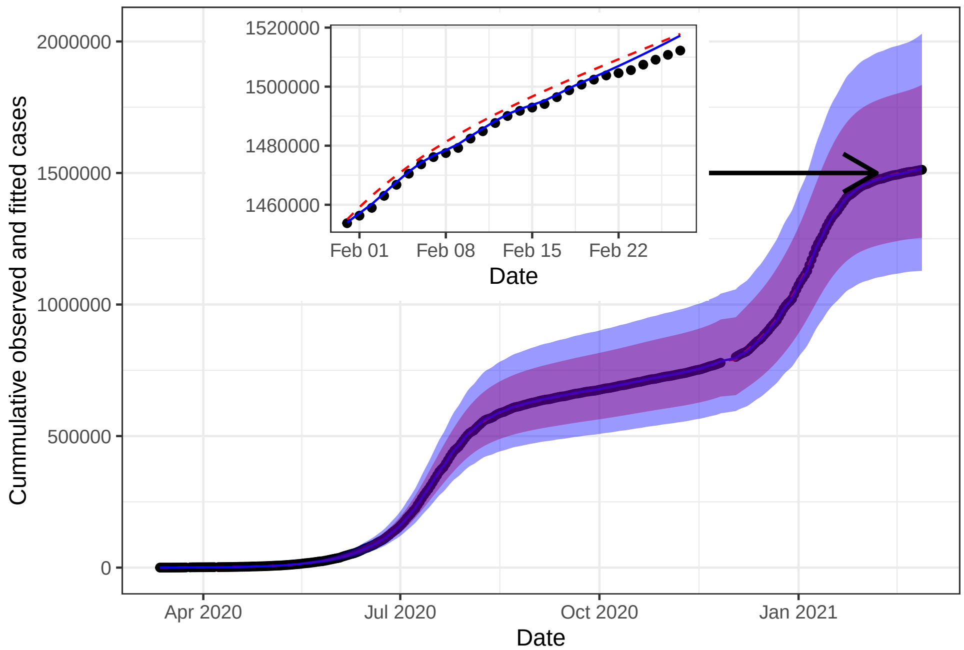

Fig 5. Predicted cummulative COVID-19 cases in South Africa under the RW(1) and AR(1) model. Estimation period 12/03/2020-20/02/2021. The black dots are the observed cummulative cases. The red dashed lines are for RW(2) and the blue line for AR(1).The shaded bands are the prediction intervals.

Table 1. 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

Date	Total	Prediction	Prediction Interval	Total - Prediction
2021-02-21	1503796	1507571	(1248627.89-1814985.06)	-3774.615
2021-02-22	1504588	1509293	(1249675.34-1817725.43)	-4704.813
2021-02-23	1505586	1511002	(1250614.94-1820668.85)	-5415.668
2021-02-24	1507448	1512704	(1251450.53-1823857.7)	-5255.520
2021-02-25	1509124	1514405	(1252187.89-1827338.87)	-5281.398
2021-02-26	1510778	1516115	(1252834.07-1831165.97)	-5337.154
2021-02-27	1512225	1517842	(1253397.14-1835400.29)	-5616.621

Table 2. Short-term predictions of total number of reported cases at the national level under the AR1 model. Estimation period 12/03/2020-20/02/2021

Date	Total	Prediction	Prediction Interval	Total - Prediction
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

Table 3. 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 4. 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

Table 5. Information Creteria for AR1 and RW2 models

	DIC	WAIC
AR1	5278.871	5286.902
RW1	5447.398	5460.804

Appendix

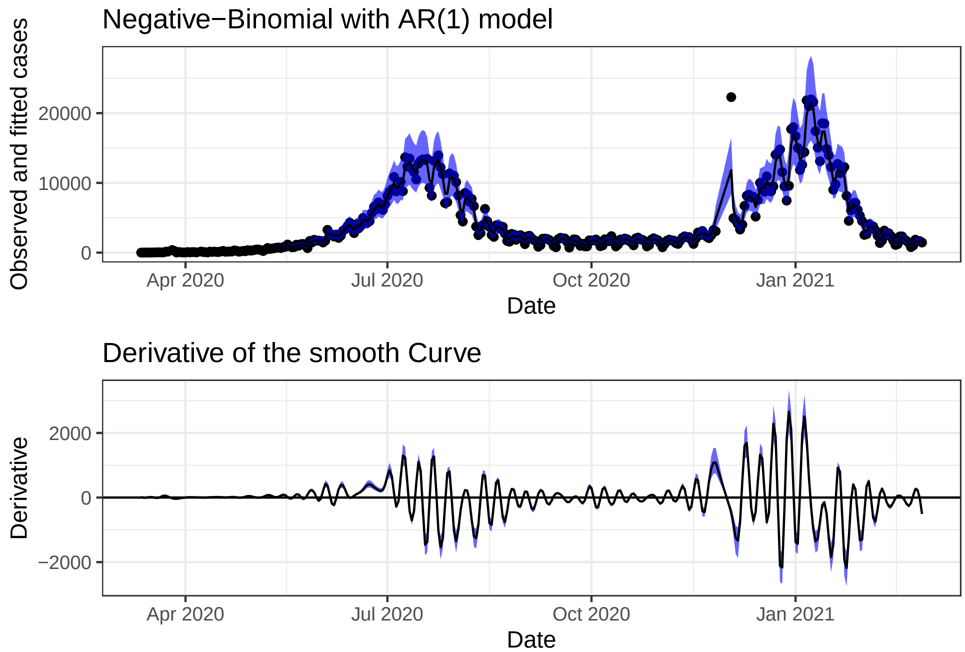

Fig S1. Fitted and observed data AR(2) model

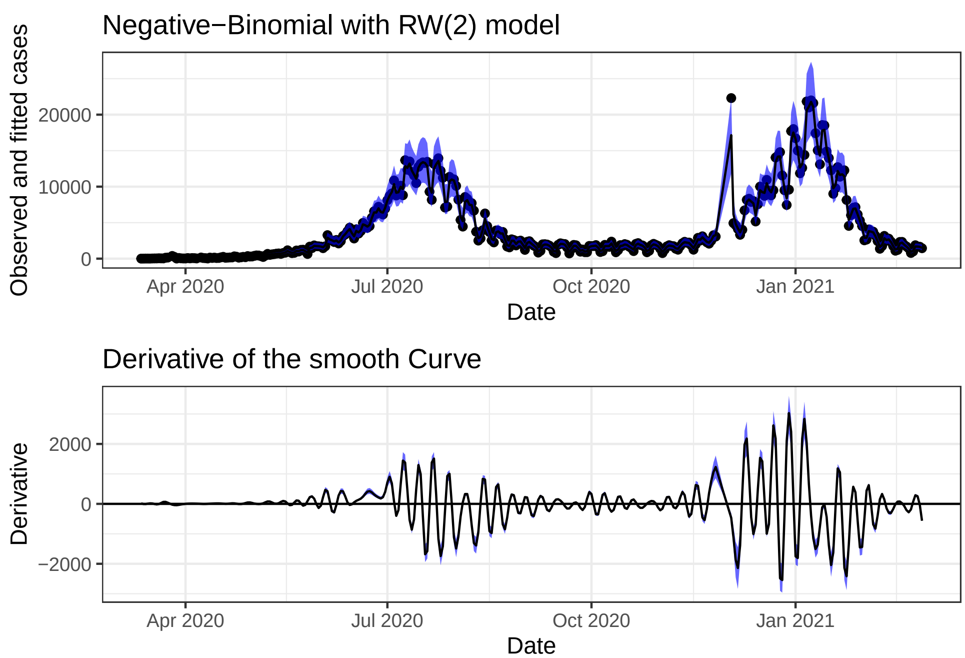

Fig S2. Fitted and observed data RW(1) model

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.