



## SARIMA MODELLING OF DAILY LABORATORY CONFIRMED CASES OF CORONAVIRUS IN NIGERIA

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**ABSTRACT:** This study is an attempt to model daily confirmed cases of coronavirus in Nigeria. A time plot of the series shows an upward trend with some seasonality. It is tested for unit test and is shown to be non-stationary. Its difference shows evidence of stationarity. The correlogram of the difference shows significant spikes at the partial autocorrelation function at lags 1 and 12 and at its autocorrelation function at lags 1 and 13, with the lag 13 spike surrounded by spikes of comparable lengths in the same direction. This suggests an autoregressive fit of lags 1 and 12 and a moving average fit of lags 1, 13 and 14. A fit of the model shows that only the moving average lags are significant. A more specific SARIMA(0, 1, 1) $\times$ (0, 0, 1)<sub>13</sub> model is fitted to the series. This shows that the series may be regarded as a SARIMA(0, 1, 1) $\times$ (0, 0, 1)<sub>13</sub> case.

**Key Words:** Covid-19 pandemic, SARIMA modelling

### INTRODUCTION

Coronavirus disease has attained a worldwide status as a pandemic. The prime case recorded in Nigeria appeared on February 27, 2020 and was confirmed at the Virology Laboratory of the Lagos State University Teaching Hospital. He is from Italy and arrived Nigeria from Milan, Italy on a business trip. He was being managed at the Infectious Disease Hospital, Yaba, Lagos. The Federal Ministry of Health announced the first confirmed case in Nigeria on the next day, 28<sup>th</sup> February 2020. (Ehanire, 2020). The occurrence of a phenomenon like this is often an opportunity for researchers to model its incidence and in the case of a medical condition like this to proffer a curative solution to it. As published in Thisday newspaper, Gumel (2020) underscored this point saying that a researcher after studying the components of this phenomenon can model it using mathematical tools. Voice of Nigeria announces that Professor Maurice Iwu has claimed to have discovered a cure for it Ukoh(2020).

The approach of seasonal autoregressive integrated moving average modelling is to be adopted in this work. Proposed by Box and Jenkins (1976) it has been widely successfully applied to model seasonal time series. To mention a few, look at Etuk (2013), Mwanaga et al. (2017) and Adams and Bamanga (2020). Here it is our intention to see the daily occurrence of this disease in Nigeria as a time series and model it accordingly. Section 2 dwells on Materials and Methods, section 3 on Results and Discussion and section 4 on the Conclusion.

### MATERIALS AND METHODS

Data: The data used for this study are 64 values of daily cumulative laboratory confirmed cases of coronavirus-19 recorded by Nigerian Centre of Disease Control with website <http://covid19.ncdc.gov.ng/>. They are displayed in the appendix.

### Seasonal Autoregressive Integrated Moving Average Modelling

It is our intention to study this covid-19 phenomenon as a daily time series. It is hoped that it shall be modelled as a seasonal autoregressive integrated moving average model.

A time series  $X_1, X_2, \dots, X_n$  is said to follow an autoregressive moving average model if

$$X_t = \alpha_1 X_{t-1} + \alpha_2 X_{t-2} + \dots + \alpha_p X_{t-p} + \beta_1 \varepsilon_{t-1} + \beta_2 \varepsilon_{t-2} + \dots + \beta_q \varepsilon_{t-q} \quad (1)$$

where  $\{\varepsilon_t\}$  is a white noise process, the  $\alpha$ 's and  $\beta$ 's are constants chosen such that the model (1) is both stationary and invertible. Model (1) is denoted as an ARMA( $p, q$ ). It may be written as an ARMA( $p, q$ ). It may be written as

$$A(L)X_t = B(L)\varepsilon_t \quad (2)$$

where  $A(L) = 1 - \alpha_1 L - \alpha_2 L^2 - \dots - \alpha_p L^p$  and  $B(L) = 1 + \beta_1 L + \beta_2 L^2 + \dots + \beta_q L^q$  and  $L^k X_t = X_{t-k}$ .

Hardly is a time series  $\{X_t\}$  stationary. In that case Box and Jenkins (1976) proposed that differencing to an order  $d$  could render it stationary. Then if the difference of  $\{X_t\}$ ,  $\nabla^i(X_t)$ ,  $i = 1, 2, \dots, d-1$ , is non-stationary and the  $d^{\text{th}}$  difference is stationary then a replacement of  $X_t$  by  $\nabla^d(X_t)$ , in (1) or (2) yields an autoregressive integrated moving average model of order ( $p, d, q$ ) in  $X_t$  denoted by an ARIMA( $p, d, q$ ) model. Then  $\nabla = 1 - L$ .

If seasonal periodicity  $s$  is observed to happen in the series, assuming that there is a seasonal trend of period  $D$ , then the model becomes

$$\Phi(L^s)A(L)X_t = \square(L^s)B(L)\varepsilon_t \quad (3)$$

where  $\Phi(L^s) = 1 - \phi_1 L^s - \phi_2 L^{2s} - \dots - \phi_P L^{Ps}$  and  $\square(L^s) = 1 + \theta_1 L^s + \theta_2 L^{2s} + \dots + \theta_Q L^{Qs}$ . Model (3) is expressed as a seasonal autoregressive integrated moving average model of order ( $p, d, q$ ) $\times$ ( $P, D, Q$ ), denoted by SARIMA( $p, d, q$ ) $\times$ ( $P, D, Q$ ). An indication of autoregressive order  $p$  is a significant spike at lag  $p$  on the partial autocorrelation function and an indication of a moving average order  $q$  is a significant spike on the partial autocorrelation function at lag  $q$ .

Computer Software: The software used for this work is eviews 10. It uses the least square approach too estimation of model parameters.

## RESULTS AND DISCUSSION

The difference of the data yields daily laboratory confirmed cases of coronavirus in Nigeria. Its time plot is in figure 1. This shows a positive trend and some seasonality. The Augmented Dickey Fuller test of stationarity of Table 1 adjudges it as non-stationary, in which case differencing has to be done on it.

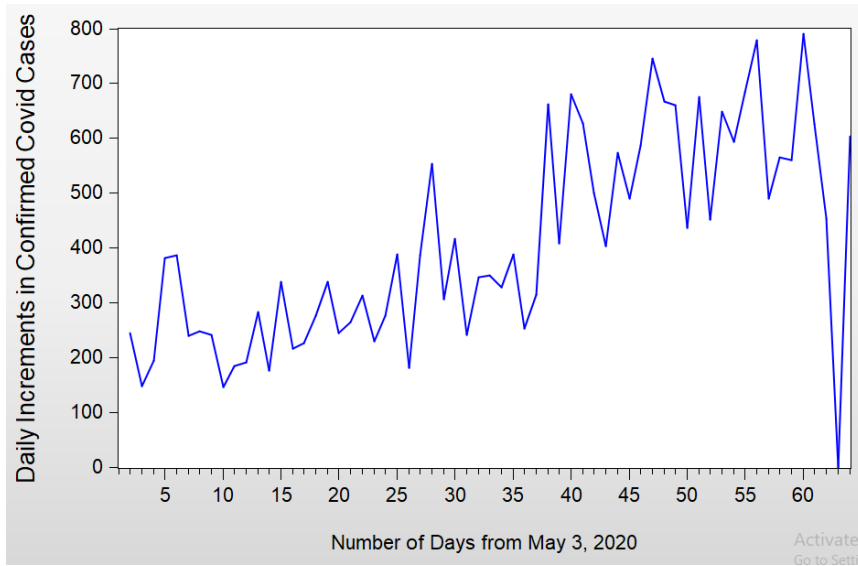


Figure 1: Time Plot of number of daily confirmed cases of coronavirus in Nigeria.

Table 1: Unit Root Test for confirmed cases

Null Hypothesis: DCOVID has a unit root  
 Exogenous: Constant  
 Lag Length: 1 (Automatic - based on SIC, maxlag=10)

	t-Statistic	Prob.*
<b>Augmented Dickey-Fuller test statistic</b>	<b>-2.519660</b>	<b>0.1159</b>
Test critical values:		
1% level	-3.542097	
5% level	-2.910019	
10% level	-2.592645	

\*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation  
 Dependent Variable: D(DCOVID)  
 Method: Least Squares  
 Date: 07/05/20 Time: 06:15  
 Sample (adjusted): 4 64  
 Included observations: 61 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
DCOVID(-1)	-0.280798	0.111443	-2.519660	0.0145
D(DCOVID(-1))	-0.400361	0.135526	-2.954133	0.0045
C	119.8430	49.07743	2.441916	0.0177
R-squared	0.320722	Mean dependent var		7.459016
Adjusted R-squared	0.297299	S.D. dependent var		171.7062
S.E. of regression	143.9366	Akaike info criterion		12.82455
Sum squared resid	1201629.	Schwarz criterion		12.92837
Log likelihood	-388.1489	Hannan-Quinn criter.		12.86524
F-statistic	13.69242	Durbin-Watson stat		2.035145
Prob(F-statistic)	0.000013			

The time plot of differences of confirmed cases in Figure 2 shows a horizontal trend and some seasonality. The Augmented Dickey Fuller test on the series in Table 2 shows that they are stationary.

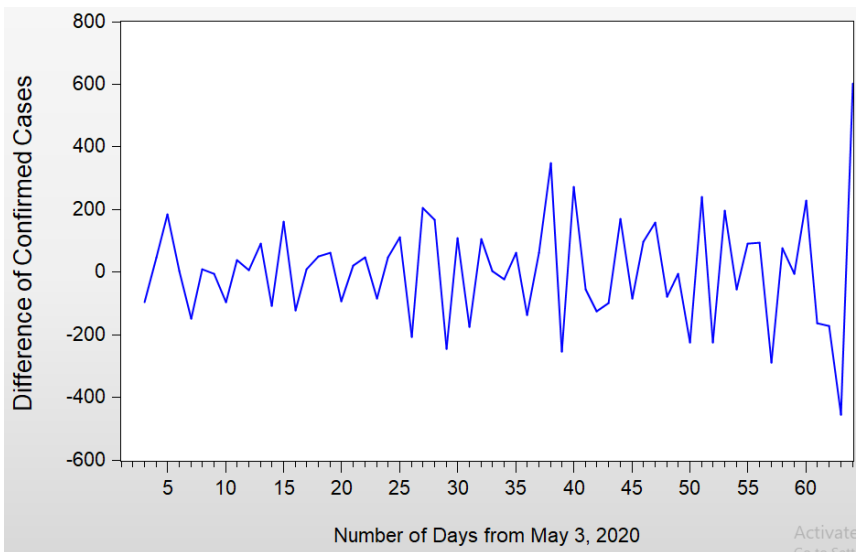


Figure 2: Difference of confirmed cases of coronavirus

Table 2: Unit Root Test on Difference of confirmed cases



Null Hypothesis: DDCOVID has a unit root  
 Exogenous: Constant  
 Lag Length: 0 (Automatic - based on SIC, maxlag=10)

	t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic	-12.31078	0.0000
Test critical values:		
1% level	-3.542097	
5% level	-2.910019	
10% level	-2.592645	

\*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation  
 Dependent Variable: D(DDCOVID)  
 Method: Least Squares  
 Date: 07/05/20 Time: 06:18  
 Sample (adjusted): 4 64  
 Included observations: 61 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
DDCOVID(-1)	-1.554586	0.126278	-12.31078	0.0000
C	5.240674	19.25306	0.272200	0.7864
R-squared	0.719788	Mean dependent var		11.45902
Adjusted R-squared	0.715039	S.D. dependent var		281.5934
S.E. of regression	150.3194	Akaike info criterion		12.89564
Sum squared resid	1333160.	Schwarz criterion		12.96485
Log likelihood	-391.3170	Hannan-Quinn criter.		12.92276
F-statistic	151.5552	Durbin-Watson stat		2.118366
Prob(F-statistic)	0.000000			

The correlogram of the difference of the confirmed cases in Figure 3 below shows a positive spike at lags 1 and 12 for the partial autocorrelation function (PACF) and significant spikes at lags 1 and 13 in the autocorrelation function (ACF), the spike at lag 13 unrounded with comparative spikes, in direction and length, at lags 12 and 14, an indication of seasonality of lag 13. This has led to the hypothesis of an ARIMA(12, 1, 14) with the autoregressive lags 1 and 12 and the moving average lags 1, 13 and 14. A summary of the model fit in Table 3 shows that the only significant parameters are at the moving average lags of 1, 13 and 14, which is an indication of a seasonality of period 13.

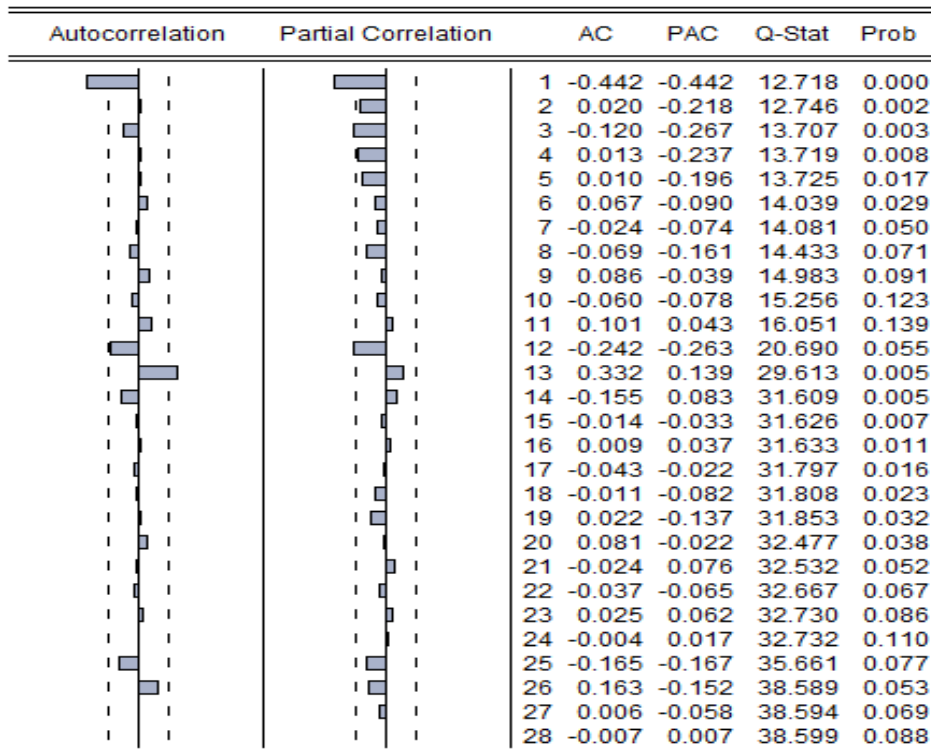


Figure 3: Correlogram of difference of confirmed cases

Table 3: Estimation of ARIMA model for the difference of confirmed cases

Dependent Variable: DDCOVID  
 Method: Least Squares  
 Date: 07/05/20 Time: 05:51  
 Sample: 3 64  
 Included observations: 62  
 Convergence achieved after 46 iterations  
 Coefficient covariance computed using outer product of gradients

Variable	Coefficient	Std. Error	t-Statistic	Prob.
AR(1)	0.039049	0.200023	0.195224	0.8459
AR(12)	-0.050128	0.173767	-0.288480	0.7740
MA(1)	-0.744949	0.154957	-4.807473	0.0000
MA(13)	0.348918	0.185932	1.876595	0.0658
MA(14)	-0.368165	0.150161	-2.451806	0.0174
SIGMASQ	15833.89	2767.697	5.720963	0.0000

R-squared	0.448331	Mean dependent var	5.790323
Adjusted R-squared	0.399075	S.D. dependent var	170.7991
S.E. of regression	132.4023	Akaike info criterion	12.74843
Sum squared resid	981701.3	Schwarz criterion	12.95428
Log likelihood	-389.2014	Hannan-Quinn criter.	12.82925
Durbin-Watson stat	1.972650		

This is a SARIMA(0, 1,1)X(0, 0,1)<sub>13</sub> model. A more specific estimation of the above-mentioned model is done on Figure 4 to have the model



$$\nabla X_t = -0.7331\varepsilon_{t-1} + 0.3503\varepsilon_{t-13} - 0.3779\varepsilon_{t-14} + \varepsilon_t$$

which is a SARIMA(0, 1, 1) $\times$ (0, 0, 1)<sub>13</sub> model for the confirmed cases of coronavirus in Nigeria.

## CONCLUSION

The daily confirmed cases of coronavirus in Nigeria have been shown to show a SARIMA(0, 1, 1) $\times$ (0, 0, 1)<sub>13</sub> model. This model may be used to approximate its daily variation. Any study of the series may be based on this model.

Table 4: More specific Estimation of a SARIMA(0,1,1) $\times$ (0, 0, 1)<sub>13</sub> model for the confirmed cases

Dependent Variable: DDCOVID  
 Method: ARMA Maximum Likelihood (OPG - BHHH)  
 Date: 07/05/20 Time: 06:24  
 Sample: 3 64  
 Included observations: 62  
 Convergence achieved after 35 iterations  
 Coefficient covariance computed using outer product of gradients

Variable	Coefficient	Std. Error	t-Statistic	Prob.
MA(1)	-0.733143	0.088343	-8.298805	0.0000
MA(13)	0.350301	0.170952	2.049119	0.0450
MA(14)	-0.377918	0.138582	-2.727031	0.0084
SIGMASQ	15839.22	2763.845	5.730863	0.0000
R-squared	0.448146	Mean dependent var	5.790323	
Adjusted R-squared	0.419602	S.D. dependent var	170.7991	
S.E. of regression	130.1214	Akaike info criterion	12.68608	
Sum squared resid	982031.5	Schwarz criterion	12.82332	
Log likelihood	-389.2685	Hannan-Quinn criter.	12.73996	
Durbin-Watson stat	1.941012			
Inverted MA Roots	.92	.88+.29i	.88-.29i	.68-.64i
	.68+.64i	.31-.88i	.31+.88i	-.12-.93i
	-.12+.93i	-.54-.77i	-.54+.77i	-.83+.43i
	-.83-.43i	-.94		

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**APPENDIX**

Cumulative daily laboratory confirmed cases of Coronavirus in Nigeria (starting from 3 May 2020) read row wise

2558	2802	2950	3145	3526	3912	4151	4399	4641	4787	4971
5162	5445									
5621	5959	6175	6401	6677	7016	7261	7526	7839	8068	8344
8733	8915									
9302	9855	10162	10578	10819	11166	11516	11844	12233	12486	12801
13873										
14554	15181	15682	16085	16658	17148	17735	18480	19147	19808	20244
20919	21371									
22020	22614	23298	24077	24567	25133	25694	26484	27110	27564	28167

Source: NCDC Coronavirus COVID-19 Microsite <http://covid19.ncdc.gov.ng/>