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## Intervention Analysis of Daily South African Rand/Nigerian Naira Exchange Rates

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**ABSTRACT:** Time series plot of a realization of daily exchange rates of South African Rand and Nigerian Naira from April 2017 to December, 2017 shows the occurrence of an intervention on 4th August, 2017. This research work has an aim of proposing an intervention model to explain the impact of this intervention believed to be due to the economic recession in Nigeria. Pre-intervention series is observed to be stationary by the Augmented Dickey Fuller Test. Following the shown autocorrelation structure of the series, an adequate subset ARMA (12, 2) model is fitted to it. On the basis of this model forecasts are made for the post-intervention period. Difference between these forecasts and their corresponding actual observations are modeled to obtain the intervention transfer function and the desired overall intervention model. Management of these exchange rates may be made on the basis of this model.

**Keywords:** South Africa, Nigeria, Exchange Rate, ARIMA Modeling, Interrupted Time Series, Forecasting.

### INTRODUCTION

The legal tender of South Africa is the Rand which has an acronym ZAR (Rand). On the other hand, Naira is the Nigerian currency and is denoted by NGN (for Nigerian Naira). An investigation of the daily exchange rates of South Africa and Nigeria from April 2017 to December, 2017, shows an unforeseen jump in the amount of NGN per ZAR on August 4<sup>th</sup>, 2017. In finance, an exchange rate is the value of one country's currency in relation to another currency. Exchange rate between the two currencies are the basis for international trade between the two nations and may be used as proxy for relative performance of their economies. The aim of this work is to propose an intervention model for the exchange rate between South African Rand and Nigerian Naira.

The intervention situation in the ZAR/NGN exchange is believed to be due to the current economic recession in Nigeria. The approach to the intervention model of the exchange rate, shall be the Autoregressive Integrated Moving Average (ARIMA) approach which was introduced by Box and Tiao (1975) [1]. This approach is well tested and efficaciously applied by many scholars. For instance, Masukawa et al. (2014) studied the impact of the introduction of a rotavirus vaccine on rates of hospitalization of children less than 5 years old for acute diarrhea [2]. Valadkhani and Layton (2004) examined the effect of goods and services tax on inflation in Australia. The observed a transitory effect [3]. Etuk et al. (2017) has fitted an intervention model of the Euro/British pound exchange rate occasioned by BREXIT [4]. Udoudo and Etuk (2018) conducted an intervention study on daily exchange rate of Thailand Thai-Bath/Nigerian Naira, still due to the current economic recession in Nigeria [5]. Ebhuoma et al., (2017) studied the positive effect of the re-introduction of dichlorodiphenyltrichloroethane in the lowering of malaria incidence using ARIA intervention analysis [6]. Michael et al. (2004) studied the impact of illicit drug supply



reduction on health and social outcomes: the heroin shortage in the Australian Capital Territory. They observed that a sustainable decline in the supply of heroin, as measured by indicators such as drug purity, is related to changes in drug-related health indicator such as ambulance callouts to heroin overdoses [7].

## MATERIALS AND METHOD

### Data

The data used in this work are of secondary sources. The data analyzed in this work are daily ZAR/NGN exchange rates from 8<sup>th</sup> April, 2017 to 26<sup>th</sup> December, 2017 from the website [www.exchangerates.org.uk/ZAR-NGN-exchange-rate-history.html](http://www.exchangerates.org.uk/ZAR-NGN-exchange-rate-history.html). They are read as the amounts of NGN per ZAR. The used data is listed in the appendix.

### Intervention Modeling

Let  $X_t$  be a time series encountering an intervention at time  $t = T$ . Box and Tiao (1975) proposed that the pre-intervention part of the series be modeled by ARIMA techniques. That is, for  $t < T$ , suppose that the  $ARIMA(p, d, q)$  model.

$$\nabla^d X_t = \alpha_1 \nabla^d X_{t-1} + \alpha_2 \nabla^d X_{t-2} + \dots + \alpha_p \nabla^d X_{t-p} + \varepsilon_t + \beta_1 \varepsilon_{t-1} + \beta_2 \varepsilon_{t-2} + \beta_p \varepsilon_{t-p} \quad (1)$$

(where  $\nabla X_t = X_t - X_{t-1}$ ) is fitted. Model (1) may be put as

$$\Phi(L)(1-L)X_t = \Theta(L)\varepsilon_t \quad (2)$$

Where  $L^k X_t = X_{t-k}$ ,  $L^k \varepsilon_t = \varepsilon_{t-k}$ ,  $\Phi(L) = 1 - \alpha_1 L - \alpha_2 L^2 - \dots - \alpha_p L^p$  is the autoregressive (AR) operator and  $\Theta(L) = 1 + \beta_1 L + \beta_2 L^2 + \dots + \beta_q L^q$  is the moving average (MA) operator. The  $\alpha$ 's and  $\beta$ 's are chosen such that the zeros of  $\Theta(L) = 0$  are outside of the unit circle for model stationarity and the zeros of  $\Phi(L) = 0$  are outside of the unit circle for model invertibility.

From (2), the noise part of the intervention model is

$$V_t = \frac{\Theta(L)\varepsilon_t}{\Theta(L)(1-L)^d} \quad (3)$$

On the basis of the model forecasts are obtained for the post-intervention part of the time series. Suppose these are  $F_t, t \geq T$ . Then for  $t \geq T$

$$Z_t = X_t - F_t = \frac{c(1) * (1 - c(2)^{t-T+1})}{(1 - c(2))} \quad (4)$$

(The Pennsylvania State University, 2016 [8]).

This is the transfer function of the intervention model. The model is then obtained by combining (3) and (4) to have

$$V_t = \frac{\Theta(L)\varepsilon_t}{\Theta(L)(1-L)^d} + I_t Z_t \quad (5)$$



Where  $I_t$  is an indicator variable that  $I_t = 1$  is the post-intervention period and zero otherwise. Secondly still on the basis of the model, forecast are obtained for the post-intervention part of the time series. Suppose these are  $F_t, t \geq T$ . Then for  $t \geq T$

$$Z_t = X_t - F_t = c(1) + c(2) * (t - T + 1) + c(3) * (t - T + 1)^2 \quad (6)$$

This is the transfer function of the intervention model. The model is given by combining (3) and (6) to have

$$Y_t = \frac{\Theta(L)\epsilon_t}{\Theta(L)(1-L)^d} + I_t Z_t \quad (7)$$

Where  $I_t$  is an indicator variable that  $I_t = 1$  is the post-intervention period and zero otherwise. In practice the model (2) is fitted first by the determination of the orders,  $p$ ,  $d$  and  $q$ . The differencing order is determined sequentially starting from 0 if the series is stationary. If not, with  $d = 1$ , the series is tested for stationary. If non-stationary,  $d = 2$ . Stationary may be tested with the Augmented Dickey Fuller (ADF) unit root test procedure. The autoregressive (AR) order may be determined by the lap at which the partial autocorrelation function (PACF) cuts off. The moving average (MA) order may be estimated as the lap at which the autocorrelation function (ACF) cuts off. Estimation of  $\alpha$ 's and  $\beta$ 's may be done by the method of least squares.

Computer Package: Eviews 10 was used to do all computations in this work.

## RESULTS AND DISCUSSION

The time plot of the realization of the time series used in this work is shown in figure 1. After three spikes, there is a sudden sharp increase on 4th August 2017 after which there is no fall in the series. This is the point of intervention. Prior to this point the exchange rates, apart from three spikes the exchange rates point 33 and 56 exhibit a fairly flat trend (see figure 2). They are adjudged stationary by the Augmented Dickey Fuller Test (See Table 1). Their correlogram of figure 3 shows evidence of seasonality of  $MA(2)$  and  $AR(2)$ . This inform the fitting of an  $ARMA(2, 2)$  model estimated in table (2) as:

$$X_t = 0.664041X_{t-2} - 0.401345\epsilon_{t-2} + \epsilon_t$$

The autocorrelation structure of its residuals shown in Figure 4 looks like that of white noise, an indication of model adequacy. On its basis the noise component of the model is

$$V_t = \frac{(1 - 0.401345^2)}{(1 - 0.664041L^2)} \epsilon_t$$

The estimate in Table (2)  $\alpha_2 = 0.664041$  and  $\beta_2 = -0.401345$  are highly statistically significant.



On the basis of these estimate, forecast have been made for the post intervention period. The observed/forecast is modeled using equation (4) and obtained from Table (3),  $c(1) = 8.490537$  and  $c(2) = 0.681409$ . Clearly we see that  $c(1)$  and  $c(2)$  are statistically significant indicating that the model is adequate for forecasting.

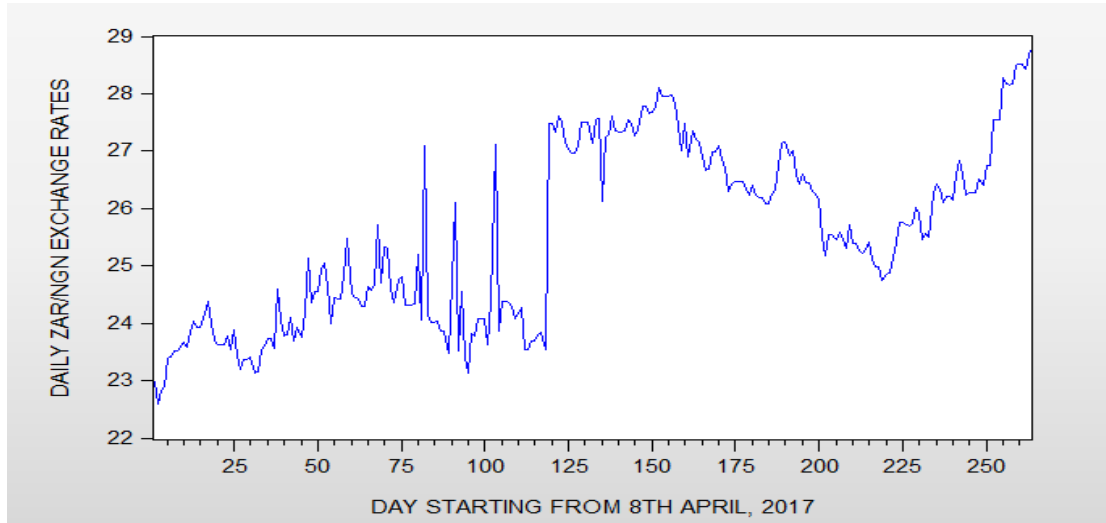


Fig.1: Time plot of daily ZAR/NGN Exchange rate

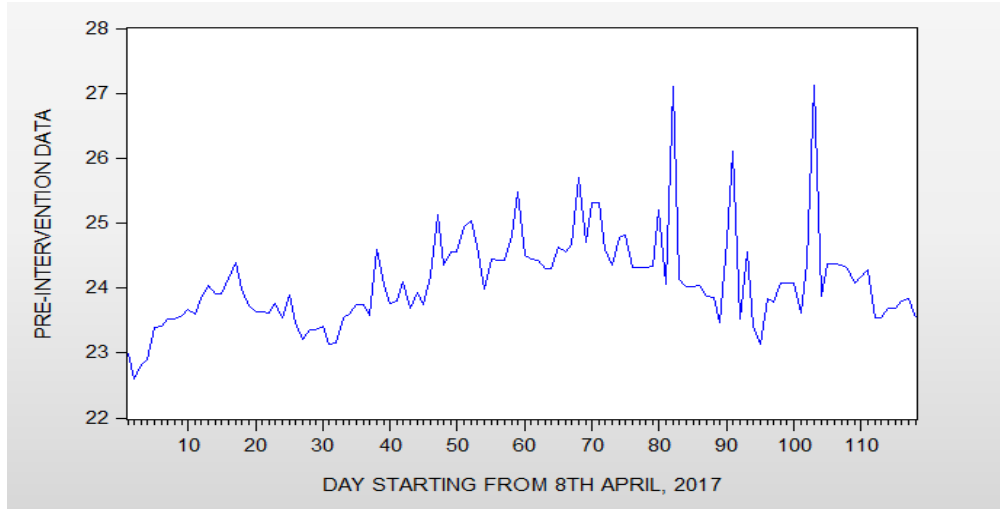


Fig. 2: Time plot of the pre-intervention model



Table 1: Stationarity Test for Pre-intervention Data

Null Hypothesis: SERIES01 has a unit root				
Exogenous: Constant				
Lag Length: 1 (Automatic - based on SIC, maxlag=12)				
			t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic			-4.453163	0.0004
Test critical values:				
	1% level		-3.487550	
	5% level		-2.886509	
	10% level		-2.580163	
*MacKinnon (1996) one-sided p-values.				
Augmented Dickey-Fuller Test Equation				
Dependent Variable: D(SERIES01)				
Method: Least Squares				
Date: 12/03/18 Time: 12:20				
Sample (adjusted): 3 118				
Included observations: 116 after adjustments				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
SERIES01(-1)	-0.412232	0.092571	-4.453163	0.0000
D(SERIES01(-1))	-0.285462	0.088279	-3.233632	0.0016
C	9.953103	2.233143	4.456994	0.0000
R-squared	0.351856	Mean dependent var		0.008145
Adjusted R-squared	0.340384	S.D. dependent var		0.749080
S.E. of regression	0.608378	Akaike info criterion		1.869482
Sum squared resid	41.82402	Schwarz criterion		1.940695
Log likelihood	-105.4299	Hannan-Quinn criter.		1.898390
F-statistic	30.67199	Durbin-Watson stat		2.116784
Prob(F-statistic)	0.000000			

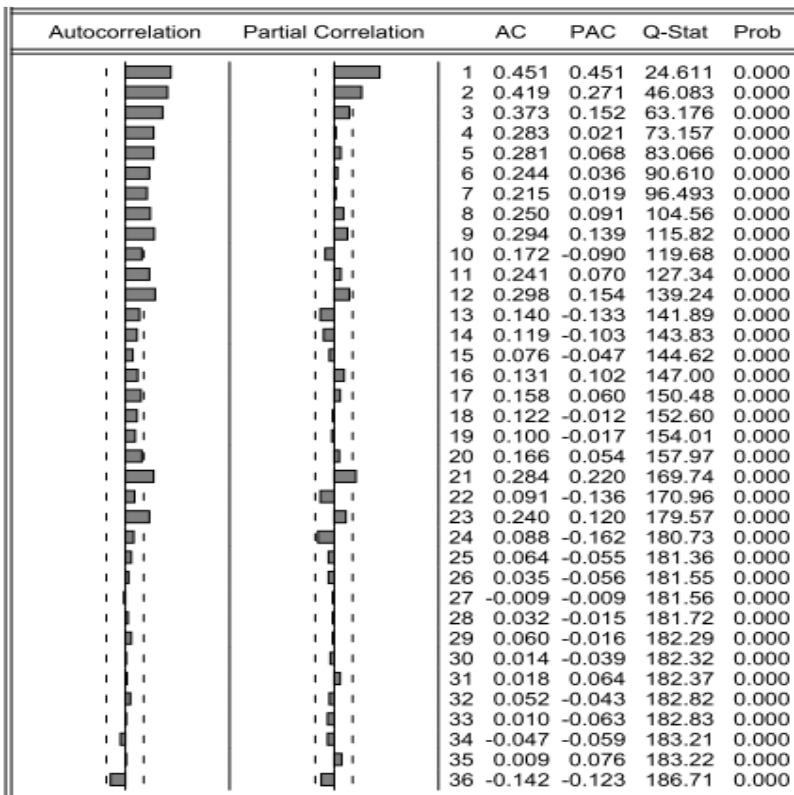


Fig 3: Correlogram of the Pre-intervention Data



Table 2: Estimate of the Pre-intervention model showing that the AR (2) MA (2) are the only significant components of the model.

Variable	Coefficient	Std. Error	t-Statistic	Prob.
Dependent Variable: SERIES01 Method: ARMA Maximum Likelihood (OPG - BHHH) Date: 12/03/18 Time: 12:36 Sample: 1 118 Included observations: 118 Convergence achieved after 388 iterations Coefficient covariance computed using outer product of gradients				
AR(1)	0.335933	0.234475	1.432706	0.1550
AR(2)	0.664041	0.326915	2.031236	0.0448
MA(1)	-0.034668	0.161103	-0.215192	0.8300
MA(2)	-0.401345	0.130968	-3.064454	0.0028
MA(3)	-0.005994	0.143699	-0.041715	0.9668
MA(4)	-0.140887	0.160434	-0.878161	0.3819
MA(5)	-0.067029	0.157434	-0.425760	0.6712
MA(6)	-0.019104	0.170422	-0.112100	0.9110
MA(7)	-0.084369	0.150683	-0.559911	0.5768
MA(8)	0.018290	0.122378	0.149452	0.8815
MA(9)	0.005159	0.078090	0.066068	0.9475
MA(10)	-0.119376	0.099013	-1.205653	0.2307
MA(11)	0.021779	0.161118	0.135171	0.8927
MA(12)	0.201793	0.117343	1.719683	0.0885
SIGMASQ	0.339598	0.036780	9.233293	0.0000
R-squared	0.335957	Mean dependent var	24.10502	
Adjusted R-squared	0.245699	S.D. dependent var	0.718179	
S.E. of regression	0.623742	Akaike info criterion	2.087148	
Sum squared resid	40.07258	Schwarz criterion	2.439354	
Log likelihood	-108.1418	Hannan-Quinn criter.	2.230154	
Durbin-Watson stat	2.039278			
Inverted AR Roots	1.00	-.66		
Inverted MA Roots	.89+.12i	.89-.12i	.64+.58i	.64-.58i
	.25+.84i	.25-.84i	-.25+.85i	-.25-.85i
	-.66+.53i	-.66-.53i	-.86-.19i	-.86+.19i

Table 3: Estimate of the Intervention transfer function

	Coefficient	Std. Error	t-Statistic	Prob.
Dependent Variable: Z Method: Least Squares (Gauss-Newton / Marquardt steps) Date: 12/04/18 Time: 11:25 Sample: 119 263 Included observations: 145 Convergence achieved after 10 iterations Coefficient covariance computed using outer product of gradients Z=C(1)*(1-C(2)^(T-118))/(1-C(2))				
C(1)	8.490537	0.315539	26.90804	0.0000
C(2)	0.681409	0.012018	56.69802	0.0000
R-squared	0.799833	Mean dependent var	26.24924	
Adjusted R-squared	0.798433	S.D. dependent var	2.204589	
S.E. of regression	0.989776	Akaike info criterion	2.831022	
Sum squared resid	140.0910	Schwarz criterion	2.872080	
Log likelihood	-203.2491	Hannan-Quinn criter.	2.847705	
Durbin-Watson stat	0.337881			

Fig 5: Graph of Intervention Transfer Function

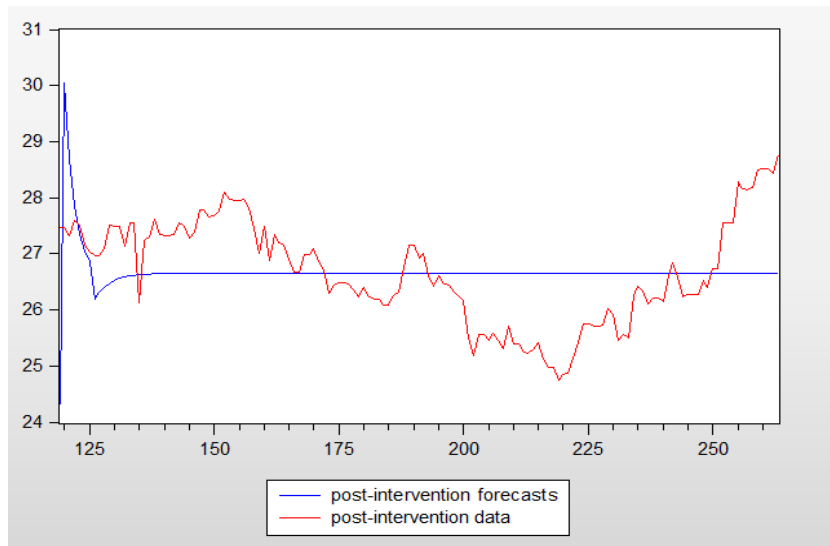


Fig.5: Post-intervention Observations and Intervention Forecast of Model 1

## CONCLUSION

From Fig 5 above, we observe that there is a close agreement between post-intervention observations and the forecast. Hence the intervention model (7) is adequate. The model explains the effect of the economic recession on the amount of Naira which is exchanged for a Rand. This is certainly going to assist the Nigerian Government as well as managers in the private sector to establish and maintain adequate intervention measures to remedy the situation for better trade relationship between Nigeria and South Africa.

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APPENDIX  
DATA



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April, 2017 (From 8<sup>th</sup> )

22.9825 22.5949 22.8233 22.8872 23.3938 23.4175 23.5195 23.5289 23.5787 23.669 23.5908 23.8318  
 24.0345 23.9172 23.9172 24.1594 24.3869 23.9853 23.7188 23.6305 23.629 23.6237 23.7681

May, 2017

23.5492 23.8868 23.4604 23.2021 23.3659 23.3659 23.4115 23.1335 23.1521 23.546 23.5951 23.7439  
 23.7439 23.5709 24.5963 24.0614 23.7668 23.8092 24.1028 23.6935 23.9378 23.7486 24.1943 25.1348  
 24.3583 24.5584 24.5621 24.9419 25.0445 24.6481 23.9868

June, 2017

24.4448 24.4273 24.4318 24.8244 25.4841 24.5018 24.4439 24.4284 24.2915 24.2898 24.6364 24.5645  
 24.6745 25.709 24.7103 25.3182 25.3104 24.5916 24.3556 24.7777 24.8158 24.321 24.3159 24.3233  
 24.3309 25.2022 24.0523 27.101 24.1359 24.0143

July, 2017

24.0146 24.0382 23.8657 23.8568 23.4679 24.7274 26.1048 23.5249 24.5589 23.4084 23.1287 23.8288  
 23.7874 24.0755 24.0755 24.0801 23.6156 24.3837 27.1174 23.867 24.3793 24.3701 24.3485 24.2944  
 24.0763 24.1721 24.2736 23.5482 23.5471 23.6938 23.6923

August, 2017

23.8031 23.8368 23.5397 27.475 27.475 27.3212 27.6022 27.5007 27.1538 27.035 26.9673 26.9677  
 27.0885 27.5085 27.498 27.494 27.1319 27.5555 27.5604 26.1284 27.2352 27.2911 27.6126 27.3427  
 27.3211 27.3211 27.3424 27.5523 27.4856 27.2707 27.3757



September, 2017

27.7879 27.7879 27.6565 27.6696 27.7573 28.0961 27.9646 27.9558 27.9533 27.9753 27.8304 27.4451  
 27.0086 27.4882 26.8915 26.8915 27.3531 27.1941 27.1584 26.9013 26.6634 26.6821 26.9874 26.9874  
 27.0868 26.8636 26.7126 26.3026 26.4343 26.4765 26.4765

October, 2017

26.4603 26.3501 26.2365 26.2256 26.1884 26.1818 26.0848 26.0761 26.2458 26.3173 26.7492 27.1473  
 27.1493 26.9207 27.0052 26.6054 26.4282 26.6015 26.4566 26.4446 26.3065 26.258 26.1655 25.4991  
 25.1871 25.5526 25.5526 25.4486 25.5791 25.4497

November, 2017

25.3041 25.7172 25.3881 25.3881 25.2533 25.2306 25.2872 25.4208 25.1191 24.9753 24.9753 24.7446  
 24.8523 24.8668 25.1427 25.4155 25.7576 25.7576 25.7136 25.6997 25.7271 26.0196 25.9093 25.4562  
 25.5674 25.5065 26.2483 26.4214 26.332 26.1048

December, 2017

26.201 26.201 26.1487 26.5993 26.8359 26.5933 26.2383 26.2673 26.283 26.2746 26.5158 26.395  
 26.7416 26.7392 27.5546 27.5546 27.5428 28.2873 28.1655 28.1476 28.1808 28.4985 28.5145 28.5092  
 28.4271 28.765