

Daily Egyptian Pound / Nigerian Naira Exchange Rates Intervention Modelling

Igboye Simon Aboko & Ette Harrison Etuk

Department of Mathematics Rivers State University, Port Harcourt **Email:** ettetuk@yahoo.com

ABSTRACT

A look at the time plot of the exchange rates of the Egyptian pound EGP and the Nigerian Naira NGN across the year 2017 reveals an abrupt jump in the amount of the latter in one unit of the former on the 4th of August 2017 and thenceforth prompting an intervention modeling. This situation is due to the ongoing economic recession Nigeria has fallen into. With a realization starting from March 17, 2017 and ending September 8, 2017, the pre-intervention data are adjudged stationary by the Augmented Dickey Fuller test. Fitted to it is an ARMA (13, 12) on which basis post-intervention forecasts are obtained. Intervention modeling produces very close post-intervention forecasts to the real data.

Keywords: Egyptian pound, Nigerian Naira, Arima modeling, intervention modeling.

INTRODUCTION

The exchange rate between the currencies of two nations is very important in the transaction of business between the two countries. It measures the relative strength of the two currencies. Egyptian pound EGP is the legal tender in Egypt. It is divided into 100 piastres, or ersh or 1000 milliemes. (Wikipedia, 2019). Nigerian Naira NGN is made up of 100 kobo. The notes are in denominations of 50 naira, 100 naira, 200 naira, 500 naira and 1000 naira and the coins which are hardly used are 5, 10 and 50 kobo only. It has been observed that there was jump in the value of the naira per EGP on 4th August 2017 in the exchange rate series of that year. Since the series has not returned back it is seen that an intervention analysis is called for. The modeling approach adopted is the one proposed by Box and Tiao (1975), an approach widely applied successfully. For instance, Lakshman et al. (1989) applied this approach in a field experiment with test and control panels connected to a split-cable TV system. Gilmour *et al.* (2006) show that Australian Heroin shortage of 2001 improved crime rate amongst the populace. Igboanugo and Ekhuemelo (2007) have provided adequate representation of 7-year monthly road accident statistics in Edo State of Nigeria. This approach was used by Min (2008) to evaluate the impact of the September 21 Earthquake in 1999 and the Severe Acute Respiratory Syndrome of 2003 on Taiwan's inbound tourism. Changes in heart rate of cows receiving both programmed and non-programmed cues have been observed and noted by Anderson et al. (2010). Masukawa et al. (2014) have verified that a statistically significant reduction on hospitalization rates of children I year or younger after an introduction of rotavirus vaccine. An adequate representation of the US Dollar/NGN exchange rates was given by Mosugu and Anieting (2016). Etuk et al. (2019) have fitted an adequate intervention model to daily Gambian Dalasi / Nigerian Naira exchange rates.



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MATERIALS AND METHODS

Data

The data used for this work are 177 daily exchange rates of the Egyptian pounds / Nigerian naira exchange rates from 17^{th} March 2017 to 9^{th} September 2017 copied from the website www.exchangerates.org.uk/EGP-NGN-exchange-rate-history.html accessed on 10th September 2017. They are to be read as the amounts of NGN in one EGP.

Intervention Analysis

Let X_{ij} X,... X_n be an n-point time series. Suppose that it is stationary. It is said to follow an autoregressive moving average time series of order p and q denoted ARMA (p, q) if X_t - $\alpha_{I}X_{t-1} - \alpha_{2}X_{t-2} - \dots - \alpha_{p}X_{t-p} = \varepsilon_{t} + \beta_{1}\varepsilon_{t-1} + \beta_{2}\varepsilon_{t-2} + \dots + \beta_{q}\varepsilon_{t-q}(1) \text{ Or } \Phi(L|X_{t} = \Theta(L)\varepsilon_{t}(2) \text{ where } L|X_{t} = \Theta(L)\varepsilon_{t}(2)$ $\{\varepsilon_t\}$ is a white noise process, $\Phi(L) = I - \alpha_1 L - \alpha_2 L^2 - \dots - \alpha_p L^p$, $\Theta(L) = I + \beta_1 L + \beta_2 L^2 + \dots$ + $\beta_{\alpha}L^{\alpha}$ and L is the backshift operator defined by $L^{k}X_{t} = X_{t-k}$. The α 's and β 's are constants defined such that the model is stationary and invertible. If the series $\{X_t\}$ is nonstationary then according to Box and Jenkins (1976) if it is differenced the differenced series ∇X_t if stationary may be modeled as an ARMA (p, q). If not, further differencing may be done, until stationarity is attained and then modeling done. If the series needed to be differenced d times before modeling, we say that the original series is modeled as an autoregressive integrated moving average model of order p, d and q denoted by ARIMA(p, d, q). Let this model be denoted by $\Phi(L)\nabla^p X_t = \Theta(L)\varepsilon_t$ [3] where $\nabla = I - L$. Let there be an intervention on the time series at the point t = m. According to Box and Tiao (1975), the preintervention series if fitted with an ARIMA (p, d, q) of the type (3). On the basis of this model, a forecast is made of the postintervention part of this series. Let it be $F_{ty} t \ge m$. Define $Z_t = X_t - F_{ty} t$ \geq m. Then $Z_t = c(I)^*(I-c(2)^*(t-m+I))/(I-c(2)), t \geq (4)$ (The Pennsylvania State University, 2016) So that the intervention model is given by $Y_t = \Theta(L)\varepsilon_t/[\Phi(L)\nabla^p] + l_t \cdot c_1/(1-c_2)^{-1}(t-c_2)$ m+1 ||/(1-c(2))| (5) where $l_t = 1$, $t \ge m$, zero otherwise.

Computer Software

The computer package used for this work is the eviews 10. It adopts the maximum likelihood estimation procedure.

RESULTS AND DISCUSSION

The time plot of the data is given below in Figure 1 shows intervention at t = 141, that is, on 4th August 2017.

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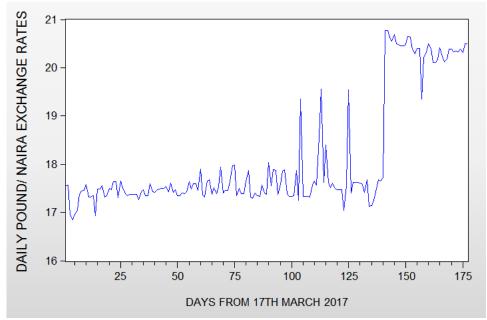


Figure 1: Time plot of EGP/NGN exchange rates.

The pre-intervention series whose time plot appears below in Figure 2 shows a stationary nature as seen in the following table 1.

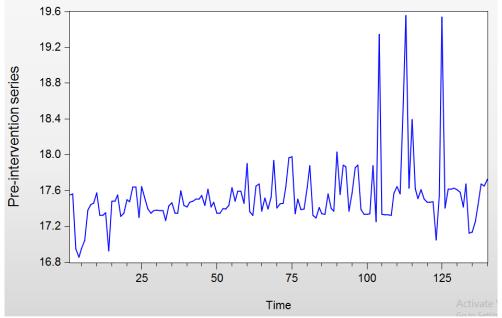


Figure 2: Time plot of the pre-intervention series



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Table 1: Unit Root Test for the Pre-intervention Series

Null Hypothesis: EGNN has a unit root Exogenous: Constant Lag Length: 0 (Automatic - based on SIC, maxlag=13)

		t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic		-9.641918	0.0000
Test critical values:	1% level	-3.477835	
	5% level	-2.882279	
	10% level	-2.577908	

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

Augmented Dickey-Fuller Test Equation Dependent Variable: D(EGNN) Method: Least Squares Date: 11/25/18 Time: 22:52 Sample (adjusted): 2 140 Included observations: 139 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
EGNN(-1) C	-0.809500 14.19214	0.083956 1.472123	-9.641918 9.640597	0.0000 0.0000
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood F-statistic Prob(F-statistic)	0.404261 0.399913 0.367904 18.54342 -57.23452 92.96658 0.000000	Mean dependent var S.D. dependent var Akaike info criterion Schwarz criterion Hannan-Quinn criter. Durbin-Watson stat		0.001245 0.474928 0.852295 0.894518 0.869453 2.039967

Autocorrelation	Partial Correlation		AC	PAC	Q-Stat	Prob
. 🗖		1	0.190	0.190	5.1699	0.023
	1 1	2	0.141	0.109	8.0359	0.018
1 🗐 1	1 1 1	3	0.089	0.046	9.1743	0.027
10	1 10	4	-0.038	-0.080	9.3904	0.052
1 1	1 1	5	-0.002	0.002	9.3910	0.094
1 1 1	1 1 1 1	6	0.019	0.031	9.4456	0.150
10	1 11	7	-0.015	-0.015	9.4796	0.220
i þi	1 1 10	8	0.076	0.075	10.341	0.242
· 🗖 ·		9	0.167	0.151	14.558	0.104
10	1 10	10	-0.019	-0.091	14.615	0.147
1 🗐 1	1 1	11	0.110	0.084	16.496	0.124
· 🗖		12	0.245	0.235	25.834	0.011
ים י	1 101	13	0.104	0.037	27.542	0.010
1 🗓 1	1 10	14	0.052	-0.058	27.970	0.014
i 🗊 i	1 11	15	0.076	0.053	28.893	0.017
i þi	1 1	16	0.081	0.111	29.935	0.018
i þi	1 11	17	0.052	-0.010	30.380	0.024
111	1 10	18	0.018	-0.041	30.434	0.033
i 🖬 i	1 10	19	-0.084	-0.069	31.585	0.035
1 👔 1	1 11	20	0.035	0.022	31.783	0.046
· 🗖		21	0.199	0.176	38.402	0.012
10	1 10	22	-0.034	-0.096	38.598	0.016
· 🖻	1 10	23	0.132	0.071	41.563	0.010
10	□ ·	24	-0.019	-0.156	41.627	0.014
1 1	1 1	25	0.007	-0.003	41.635	0.020
111	1 11	26	-0.014	-0.012	41.668	0.027
1 1	1 101	27	0.006	0.027	41.674	0.035
ւիս	1 1	28	0.032	0.000	41.853	0.045
1 🗐 1	1 1	29	0.085	0.002	43.152	0.044
10	1 10	30	-0.027	-0.090	43.284	0.055
i 🖬 i	1 111	31	-0.057	0.015	43.883	0.062
ւիւ	1 11	32	0.045	0.016	44.248	0.073
ւիւ	1 1	33	0.029	-0.002	44.401	0.089
111	101	34	0.012	-0.026	44.427	0.109
ı 🗖 i		35	0.120	0.146	47.171	0.082

Figure 3: Correlogram of the pre-intervention series

п



Table 2: An ARIMA(1,0,1)(1,0,1) Model for the Pre-intervention Series

Dependent Variable: EGNN Method: ARMA Maximum Likelihood (OPG - BHHH) Date: 11/25/18 Time: 23:11 Sample: 1 140 Included observations: 140 Failure to improve objective (non-zero gradients) after 34 iterations Coefficient covariance computed using outer product of gradients

Variable	Coefficient	Std. Error t-Statistic		Prob.	
AR(1) AR(12) AR(13) MA(1) MA(12) MA(13) SIGMASQ	0.491968 0.607357 -0.099329 -0.240162 -0.359555 -0.044685 0.129789	0.107622 4.571274 0.130395 4.657822 0.002150 -46.20451 0.119384 -2.011676 0.119644 -3.005204 0.075929 -0.588506 0.007244 17.91703		0.0000 0.0000 0.0463 0.0032 0.5572	
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.055625 0.013021 0.369622 18.17049 -61.58125 2.226881	Mean depen S.D. depend Akaike info o Schwarz crit Hannan-Qui	17.53184 0.372052 0.979732 1.126814 1.039502		
Inverted AR Roots	1.00 .51+.82i 94 .95 .49+.79i 4379i 89	.87+.47i .16 46+.83i .83+.46i .0392i 43+.79i	.8747i .0295i 8148i .8346i .03+.92i 77+.46i	.5182i .02+.95i 81+.48i .4979i 12 7746i	

So that the pre-Intervention Series is modeled by

 $X_{t} = 0.4920 X_{t-1} + 0.6074 X_{t-12} - 0.0993 X_{t-13} - 0.2402 \epsilon_{t-1} - 0.3596 \epsilon_{t-12} + \epsilon_{t}$

On the basis of that model forecasts are made for the post-intervention series and z_t are obtained and modeled on the basis of (4).

Table 3: Intervention transfer function modelling

Dependent Variable: Z Method: Least Squares (Gauss-Newton / Marquardt steps) Date: 05/28/19 Time: 13:52 Sample: 141 177 Included observations: 37 Convergence achieved after 17 iterations Coefficient covariance computed using outer product of gradients Z = C(1)*(1-C(2)^{(T-140))/(1-C(2))}

	Coefficient	Std. Error	t-Statistic	Prob.
C(1) C(2)	2.990606 -0.102804	0.263632 0.098246	11.34388 -1.046391	0.0000 0.3026
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.037086 0.009574 0.269102 2.534554 -2.904069 1.126859	Mean dependent var S.D. dependent var Akaike info criterion Schwarz criterion Hannan-Quinn criter.		2.718549 0.270399 0.265085 0.352161 0.295783

Hence the intervention model is given by

 $Y_t = (I - 0.2402L - 0.3596L^n) \epsilon_t/(I - 0.4920L - 0.6074L^n + 0.0993L^n) + l_t.(2.9906)(I - (-0.1028)^(t - 140))/I.1028$

Where $l_t = I_1 t \ge I_4 I_1$ zero elsewhere.

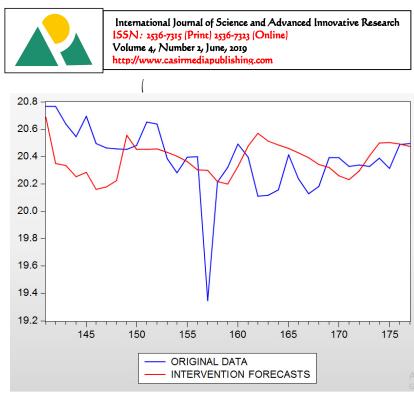


Figure 4: Comparison between the post-intervention data and their intervention forecasts

CONCLUSION

It may be noted the closeness of the post-intervention data and their intervention forecasts. Therefore the intervention model (6) is adequate for the exchange rates. Therefore decision should based on it to help manage the relationship between the two currencies.

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APPENDIX

DATA

March 2017 (Starting from the 17th) 17.5531 17.5628 16.9466 16.8534 16.9546 17.0412 17.3807 17.4481 17.4578 17.5761 17.3255 17.3255 17.3506 16.9272 17.4815. April 2017 17.4815 17.5522 17.3132 17.3463 17.5004 17.4744 17.6399 17.6398 17.2988 17.6443 17.5178 17.3921 17.3493 17.3784 17.3804 17.3761 17.3734 17.2647 17.4313 17.4653 17.3485 17.3485 17.5960 17.4338 17. 4161 17.4703 17.4803 17.5027 17.5028 17.5437. May 2017 17.4359 17.6127 17.4191 17.4690 17.3452 17.3452 17.3993 17.3909 17.4334 17.6308 17.4822 17.5900 17.5900 17.4559 17.9029 17.3623 17.3224 17.6495 17.6727 17.3688 17.5185 17.3950 17.5305 17.9360 17.4033 17.4535 17.4565 17.6391 17.9656 17.9767 17.3423. June 2017. 17.5046 17.3877 17.3916 17.6305 17.8773 17.3258 17.2932 17.4085 17.3410 17.3372 17.5636 17.4012 17.3677 18.0333 17.5563 17.8842 17.8679 17.3692 17.5767 17.8537 17.8862 17.3898 17.3352 17.3352 17.3404 17.8773 17.2542 19.3465 17.3270. July 2017. 17.3313 17.3246 17.5699 17.6468 17.5639 18.5690 19.5558 17.6270 18.3953 17.6189 17.5109 17.6080 17.5048 17.4714 17.4714 17.4760 17.0467 17.5481 19.5385 17.4054 17.6164 17.6164 17.6165 17.6124 17.5785 17.4168 17.6738 17.1257 17.1339 17.2533 17.4316. August 2017. 17.6742 17.6504 17.7262 20.7661 20.7661 20.6385 20.5442 20.6936 20.4963 20.4640 20.4565 20.4531 20.4804 20.6505 20.6388 20.3844 20.2818 20.3961 20.3979 19.3465 20.2128 20.3194 20.4916 20.3960 20.1103 20.1159 20.1581 20.4122 20.2378 20.1270 20.1807. September 2017 20.3912 20.3912 20.3290 20.3373 20.3296 20.3890 20.3147 20.4898 20.7943