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## **TESTING THE PPP HYPOTHESIS IN THE SUB-SAHARAN COUNTRIES**

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**ABSTRACT**

This paper examines the PPP hypothesis in a number of Sub-Saharan countries by testing the order of integration in the log of their real exchange rate vis-à-vis the US dollar. I(d) techniques based on both asymptotic and finite sample results are used. The test results lead to the rejection of PPP in all cases: although orders of integration below 1 are found in fourteen countries, the unit root null cannot be rejected.

**Keywords:** Purchasing Power Parity (PPP), fractional integration

**JEL classification:** C22, F31

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## 1. Introduction

This paper examines the Purchasing Power Parity (PPP) hypothesis in a number of Sub-Saharan countries using a time series approach. Froot and Rogoff (1995) distinguish three stages in this literature on PPP. In stage one possible non-stationarities of the series of interest were not taken into account. In stage two unit root tests were carried out to establish whether or not the real exchange rate follows a random walk, the alternative being that PPP holds in the long run. However, it soon became apparent that such tests have very low power, and with relatively few observations cannot distinguish between a random walk process, and one which reverts very slowly toward PPP (see, e.g., Frankel, 1986, and Lothian and Taylor, 1997). This led to the so-called “embarrassing resiliency of the random walk model” (see Rogoff, 1996). Over longer time spans mean-reverting real exchange rate behaviour was instead found (see, e.g., Lothian and Taylor, 1996, and Cheung and Lai, 1994). In stage three cointegration tests (between the nominal exchange rate, domestic and foreign prices) were applied, but they also appeared to be affected by small sample bias.

The present study makes a twofold contribution. First, it adopts a more general framework than the standard stage-two unit root tests to investigate the presence of mean-reverting behaviour in the real exchange rate. Specifically, it uses fractional integration or  $I(d)$  techniques allowing the degree of integration  $d$  to be any real number, therefore introducing a higher degree of flexibility in the dynamic specification of the stochastic processes followed by the variables of interest. Second, it focuses on a long span of data for a large set of 44 Sub-Saharan countries whose exchange rates to our knowledge have not been previously analysed using advanced time series methods. The only previous empirical study is due to Olayungbo (2011), but it considers a smaller Subset of 16 countries over a relatively short sample period and carries out standard unit root tests whose low power has already been mentioned as well as panel unit root tests, the limitations of which have also

been highlighted and extensively discussed by Caporale and Cerrato (2006). Evidence on PPP in the Sub-Saharan countries is particularly interesting in view of the current discussion on creating an African Union that would eventually have its own currency and central bank, as its feasibility would also depend on the degree of conformity to PPP.

The layout of this paper is as follows. Section 2 outlines the econometric approach. Section 3 describes the data and presents the empirical results. Section 4 offers some concluding remarks.

## 2. Methodology

We consider the following model

$$y_t = \alpha + \beta t + x_t, \quad t = 1, 2, \dots, \quad (1)$$

where  $y_t$  is the observed time series,  $\alpha$  and  $\beta$  are the coefficients on the deterministic terms (an intercept and a linear trend), and  $x_t$  is assumed to be  $I(d)$  and defined as

$$(1 - L)^d x_t = u_t, \quad t = 1, 2, \dots, \quad (2)$$

with  $x_t = 0$  for  $t \leq 0$ , and where  $L$  is the lag operator ( $Lx_t = x_{t-1}$ ),  $d$  can take any real value and  $u_t$  is assumed to be  $I(0)$ . Thus,  $d$  may be equal to 0, a fraction between 0 and 1, 1, or even above 1. When it is not an integer, the process is said to be fractionally integrated. In this context, the parameter  $d$  plays a crucial role for the degree of persistence of the series. If  $d = 0$  in (2),  $x_t = u_t$ , and the series is  $I(0)$ . If  $d$  belongs to the interval  $(0, 0.5)$  the series is still covariance stationary but the autocorrelations take a longer time to disappear than in the  $I(0)$  case. If  $d$  is in the interval  $[0.5, 1)$  the series is no longer stationary; however, it is still mean-reverting in the sense that shocks affecting it disappear in the long run. Finally, if  $d \geq 1$  the series is nonstationary and non-mean-reverting. Thus, by allowing  $d$  to be any real value, we introduce more flexibility in the dynamic specification of the series than in the classical  $I(0)$  and  $I(1)$  representations. These processes (with non-integer  $d$ ) were first considered by

Granger (1980), Granger and Joyeux (1980) and Hosking (1981) and since then have been widely employed to describe the behaviour of many economic time series.

In the empirical analysis we test the null hypothesis:

$$H_0: d = d_0, \quad (3)$$

in a model given by the equations (1) and (2) where  $d_0$  in (3) can be any real value. Thus, under the null hypothesis (3) the model becomes:

$$y_t = \alpha + \beta t + x_t; \quad (1 - L)^{d_0} x_t = u_t; \quad t = 1, 2, \dots \quad (4)$$

This is a very general specification that includes many cases of interest. Thus, for example, if we cannot reject the null with  $d_0 = 0$ , we are in the classical trend-stationary representation with or without weak (ARMA) autocorrelation in  $u_t$ .<sup>1</sup> On the other hand, if we cannot reject the null with  $d_0 = 1$ , the unit root model advocated by many authors is given support. Moreover, we can also consider cases where  $d_0$  can be a real value between 0 and 1, or even above 1. As mentioned before, the estimation of  $d_0$  is crucial to determine the degree of persistence: the higher is the degree of integration, the higher is the level of dependence across the observations, and if  $d_0 < 1$  the series will be mean-reverting with shocks disappearing in the long run.

To test the null hypothesis (3), we employ a parametric approach developed by Robinson (1994). This is a general testing procedure based on the Lagrange Multiplier (LM) principle that uses the Whittle function in the frequency domain. Robinson (1994) showed that, under certain very mild regularity conditions, the LM-based statistic ( $\hat{r}$ )

$$\hat{r} \xrightarrow{dtb} N(0, 1) \quad \text{as } T \rightarrow \infty,$$

where “ $\xrightarrow{dtb}$ ” stands for convergence in distribution, and this limit behaviour holds independently of the regressors used in (1) and the specific model for the  $I(0)$  disturbances

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<sup>1</sup>Note that  $u_t$  is  $I(0)$  and therefore could incorporate stationary and invertible ARMA sequences.

$u_t$ . The functional form of this procedure can be found in any of the numerous empirical applications based on these tests.

As in other standard large-sample testing situations, Wald and LR test statistics against fractional alternatives will have the same null and limit theory as the LM test of Robinson (1994). In fact, Lobato and Velasco (2007) essentially employed such a Wald testing procedure, and though this and other recent methods such as the one developed by Demetrescu et al (2008) have been shown to be robust with respect to even unconditional heteroscedasticity (Kew and Harris 2009), they require a consistent estimate of  $d$ , and therefore the LM test of Robinson (1994) seems computationally more attractive.

### **3. Data and empirical results**

We use data on real exchange rates, in logged form, for forty-four Sub-Saharan countries, for the time period 1970 – 2012 (with 2005 as the base year), obtained from the Economic Research Service, US Department of Agriculture (<http://www.ers.usda.gov>).

We consider the model given by the equations (1) and (2), testing  $H_0$  (3) for values of  $d_0$  from 0 to 2 with 0.001 increments, i.e.,  $d_0 = 0, 0.001, 0.002, \dots, 1.999$  and 2. We report in Table 1 the estimates of  $d$  based on the Whittle function in the frequency domain (Dahlhaus, 1989) along with the 95% confidence interval of non-rejection values of  $d$  using Robinson's (1994) tests, under the assumption that the error term  $u_t$  in (4) is a white noise process. Weakly (ARMA) autocorrelated errors were also considered and led to very similar results.

**[Insert Tables 1 and 2 about here]**

Table 1 displays the results for the three standard cases usually analysed in the literature, i.e., with no regressors in the undifferenced regression model in (4) (i.e.  $\alpha = \beta = 0$  a priori); with an intercept ( $\alpha$  unknown and  $\beta = 0$  a priori); and with an intercept and a linear time trend ( $\alpha$  and  $\beta$  unknown); statistically significant deterministic terms are in bold. It

appears that the time trend is only required for four series, namely those for the real exchange rates of Gambia, Guinea Bissau, Malawi and Sudan. In all the remaining cases, an intercept is sufficient to describe the deterministic part. Focusing now on the estimated orders of integration of the series (for the selected models, in Table 2), we see that for fourteen countries the value of  $d$  is strictly smaller than 1 - these are Malawi, Guinea Bissau, Liberia, Swaziland, Sudan, Gambia, Madagascar, Comoros, Angola, Togo, Botswana, Senegal, Ivory Coast and Central Africa. However, the confidence intervals for the values of  $d$  imply that the unit root null (i.e.,  $d = 1$ ) cannot be rejected in any single case. For the remaining countries, the estimated  $d$  is strictly above 1, and the unit root null is rejected in favour of  $d > 1$  in the cases of Djibouti, Sierra Leone, Mauritania, Cape Verde, Eritrea, Uganda, Sao Tome, Tanzania and Ghana.

**[Insert Table 3 about here]**

Table 3 summarises the results in terms of the degree of persistence. The countries are divided in three groups according to the statistical significance of the estimated values of  $d$ : mean reversion ( $d < 1$ ); unit roots (with  $d < 1$  and with  $d > 1$ ), and explosive behaviour ( $d > 1$ ). There is no single country where mean reversion is statistically significant, implying that PPP does not hold anywhere. However, although the unit root null cannot be rejected in 35 countries, in 14 of them the estimated value of  $d$  is below 1 implying that PPP might hold in the very long run. Another group of nine countries displays explosive behaviour. Overall, the evidence does not support PPP, consistently with the findings of Olayungbo (2011), who reports that it holds only in Ghana and Uganda; in fact even for these two countries PPP is rejected according to our results, since they are found to belong to the group with the highest degree of persistence.

We also examined the finite sample behaviour of sized-corrected versions of Robinson (1994) tests by means of Monte Carlo simulations, and compared the results with

those based on the asymptotic critical values. Note that in the original paper by Robinson (1994) he stressed large sample theory and suggested approximate critical values. Thus, we calculated the empirical size of the test statistic  $\hat{r}$  for a sample size  $T = 42$  as in our case, based on 10,000 replications, for the three cases of no regressors, an intercept, and an intercept with a linear time trend. In all cases, we assume  $u_t$  to be a Gaussian white noise process with zero mean and variance 1, generated by the routines GASDEV and RAN3 of Press, Flannery, Teukolsky and Vetterling (1986).

**[Insert Tables 4 and 5 about here]**

Table 4 displays for each country both the asymptotic and the finite sample 95% confidence intervals for the non-rejection values. We notice that in all cases the intervals are shifted to the right, implying higher degrees of integration, and therefore, even less evidence of PPP for the Sub-Saharan countries. These results are presented in Table 5. There are four countries (Mozambique, Seychelles, Burundi and Zambia) where the unit root cannot be rejected in Table 3, and is rejected in favour of  $d < 1$  in Table 5.

#### **4. Conclusions**

This paper applies long-range dependence or fractional integration techniques to test for PPP in a set of 44 Sub-Saharan countries. The advantage of this approach is its generality and flexibility in comparison to standard time series methods restricting the degree of integration to integer values. Previous evidence (see Olayungbo, 2011) was only available for a smaller Subset of countries and a short sample period and was based on low-power unit root tests as well as panel tests whose drawbacks are also well known (see Caporale and Cerrato, 2006).

On the whole, our results suggest that PPP does not hold in this group of countries. This is in contrast to the available evidence for developed countries based on long-memory models. For instance, using a similar version of Robinson's (1994) tests to the one adopted



here, Gil-Alana (2000) found mean reversion in the US real exchange rates vis-à-vis five major currencies with weakly autocorrelated disturbances. Similar conclusions were reached applying fractional integration and cointegration techniques by Caporale and Gil-Alana (2004) in the case of the DM/dollar and the yen/dollar real exchange rates, and by Masih and Masih (2004) for the Australian dollar real exchange rate vis-à-vis seven major OECD trading partners. Finally, Yoon (2009) applied the Exact Local Whittle estimators of Shimotsu and Phillips (2005) to estimate the long memory parameters of the real exchange rates for more than 100 years in 16 developed countries and concluded again that PPP holds in most of these countries. Overall, it appears that the degree of conformity to PPP is much less in the Sub-Saharan countries compared to the developed ones, and, as already pointed out by Olayungbo (2011), this has important implications for the proposed African Union and the creation of a common currency, namely the absence of PPP relationships between its prospective members raises some doubts about its feasibility or at least long-run sustainability.

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**Table 1: Estimates of d and 95% confidence intervals**

Country	No regressors	An intercept	A linear time trend
ANGOLA	0.936 (0.742, 1.234)	<b>0.959 (0.689, 1.367)</b>	0.957 (0.646, 1.365)
BURKINA FASO	0.867 (0.640, 1.175)	<b>1.041 (0.837, 1.358)</b>	1.041 (0.814, 1.358)
BENIN	0.896 (0.691, 1.186)	<b>1.138 (0.926, 1.458)</b>	1.134 (0.927, 1.453)
BOTSWANA	0.741 (0.401, 1.109)	<b>0.967 (0.735, 1.347)</b>	0.964 (0.639, 1.345)
BURUNDI	0.882 (0.666, 1.187)	<b>1.233 (0.989, 1.637)</b>	1.233 (0.986, 1.634)
CAPE VERDE	0.889 (0.677, 1.193)	<b>1.308(1.082, 1.663)</b>	1.304(1.078, 1.671)
CAMEROON	0.876 (0.665, 1.178)	<b>1.053 (0.775, 1.432)</b>	1.054 (0.793, 1.420)
CENTRAL AF.	0.852 (0.630, 1.158)	<b>0.997 (0.802, 1.299)</b>	0.996 (0.779, 1.297)
CHAD	0.861 (0.642, 1.163)	<b>1.035 (0.824, 1.355)</b>	1.033 (0.802, 1.355)
COMOROS	0.864(0.643, 1.165)	<b>0.957 (0.764, 1.251)</b>	0.954 (0.748, 1.250)
CONGO REP.	0.878 (0.676, 1.179)	<b>1.106 (0.629, 1.515)</b>	1.105 (0.787, 1.496)
DJIBOUTI	0.904 (0.707, 1.194)	<b>1.228 (1.033, 1.562)</b>	1.229 (1.033, 1.564)
EQ. GUINEA	0.856 (0.632, 1.177)	<b>1.085 (0.929, 1.314)</b>	1.083 (0.926, 1.312)
ERITREA	0.868 (0.636, 1.212)	<b>1.314 (1.102, 1.642)</b>	1.307 (1.099, 1.654)
ETHIOPIA	0.728 (0.495, 1.075)	<b>1.102 (0.906, 1.428)</b>	1.100 (0.890, 1.428)
GABON	0.866 (0.641, 1.179)	<b>1.115 (0.909, 1.416)</b>	1.112 (0.919, 1.398)
GAMBIA	0.841 (0.592, 1.159)	0.895 (0.729, 1.175)	<b>0.870 (0.602, 1.179)</b>
GHANA	1.385 (1.119, 1.873)	<b>1.459 (1.152, 2.003)</b>	1.457 (1.153, 2.004)
GUINEA B.	0.863 (0.637, 1.167)	0.840 (0.712, 1.079)	<b>0.831 (0.684, 1.072)</b>
GUINEA	0.748 (0.410, 1.123)	<b>1.011 (0.837, 1.292)</b>	1.006 (0.804, 1.293)
IVORY COAST	0.884 (0.672, 1.185)	<b>0.996 (0.708, 1.375)</b>	0.998 (0.729, 1.370)
KENYA	0.899 (0.685, 1.204)	<b>1.068 (0.907, 1.305)</b>	1.070 (0.908, 1.310)
LESOTHO	0.733 (0.176, 1.134)	<b>1.009 (0.717, 1.465)</b>	1.009 (0.681, 1.464)
LIBERIA	0.808 (0.648, 1.081)	<b>0.845 (0.695, 1.102)</b>	0.825 (0.633, 1.105)
MADAGASCAR	0.902 (0.702, 1.193)	<b>0.937 (0.775, 1.196)</b>	0.936 (0.768, 1.199)
MALAWI	0.856 (0.629, 1.166)	0.828 (0.683, 1.174)	<b>0.744 (0.375, 1.177)</b>
MAURITANIA	0.846 (0.609, 1.164)	<b>1.293 (1.109, 1.565)</b>	1.291 (1.109, 1.543)
MAURITIUS	0.879 (0.655, 1.195)	<b>1.051 (0.810, 1.427)</b>	1.050 (0.786, 1.427)
MOZAMBIQUE	0.916 (0.699, 1.271)	<b>1.189 (0.972, 1.561)</b>	1.191 (0.961, 1.557)
NAMIBIA	0.789 (0.466, 1.153)	<b>1.131 (0.841, 1.580)</b>	1.132 (0.842, 1.583)
NIGER	0.866 (0.633, 1.184)	<b>1.078 (0.903, 1.345)</b>	1.078 (0.895, 1.343)
NIGERIA	0.848 (0.580, 1.192)	<b>1.122 (0.822, 1.543)</b>	1.122 (0.816, 1.540)
REUNION	0.857 (0.633, 1.165)	<b>1.002 (0.823, 1.270)</b>	1.001 (0.817, 1.265)
SIERRA LEONE	0.841 (0.606, 1.159)	<b>1.265 (1.042, 1.635)</b>	1.261 (1.042, 1.607)
SOUTH AFRICA	0.144 (0.087, 0.686)	<b>1.239 (0.527, 2.217)</b>	1.227 (0.306, 2.194)
SAO TOME	0.856 (0.629, 1.171)	<b>1.371 (1.149, 1.758)</b>	1.366 (1.149, 1.747)
SENEGAL	0.867 (0.646, 1.174)	<b>0.986 (0.782, 1.285)</b>	0.985 (0.763, 1.282)
SEYCHELLES	0.843 (0.645, 1.144)	<b>1.223 (0.952, 1.623)</b>	1.205 (0.963, 1.544)
SUDAN	0.917 (0.741, 1.193)	0.875 (0.719, 1.139)	<b>0.861 (0.678, 1.144)</b>
SWAZILAND	0.853 (0.576, 1.222)	<b>0.859 (0.580, 1.327)</b>	0.856 (0.541, 1.327)
TANZANIA	0.893 (0.679, 1.194)	<b>1.428 (1.191, 1.797)</b>	1.428 (1.190, 1.798)
TOGO	0.883 (0.664, 1.194)	<b>0.959 (0.749, 1.276)</b>	0.958 (0.730, 1.274)
UGANDA	0.781 (0.472, 1.140)	<b>1.358 (1.117, 1.707)</b>	1.349 (1.167, 1.683)
ZAMBIA	0.880 (0.669, 1.178)	<b>1.304 (0.995, 1.811)</b>	1.306 (0.998, 1.810)

**Table 2: Estimates of the coefficients of the selected models**

Country	d (95% conf. intv.)	Intercept (t-value)	Time trend (t-value)
ANGOLA	0.959 (0.689, 1.367)	6.20967 (12.707)	xxx
BURKINA FASO	1.041 (0.837, 1.358)	5.66871 (43.262)	xxx
BENIN	1.138 (0.926, 1.458)	5.39981 (42.137)	xxx
BOTSWANA	0.967 (0.735, 1.347)	1.35546 (14.791)	xxx
BURUNDI	1.233 (0.989, 1.637)	6.46216 (75.155)	xxx
CAPE VERDE	1.308 (1.082, 1.663)	3.81408 (42.779)	xxx
CAMEROON	1.053 (0.775, 1.432)	6.40020 (50.700)	xxx
CENTRAL AF.	0.997 (0.802, 1.299)	5.81394 (41.872)	xxx
CHAD	1.035 (0.824, 1.355)	5.70189 (48.731)	xxx
COMOROS	0.957(0.764, 1.251)	5.47811 (34.271)	xxx
CONGO REP.	1.106 (0.629, 1.515)	6.91253 (55.646)	xxx
DJIBOUTI	1.228 (1.033, 1.562)	5.08840 (128.831)	xxx
EQ. GUINEA	1.085 (0.929, 1.314)	5.57779 (45.230)	xxx
ERITREA	1.314 (1.102, 1.642)	3.59273 (34.883)	xxx
ETHIOPIA	1.102 (0.906, 1.428)	1.04824 (8.315)	xxx
GABON	1.115 (0.909, 1.416)	6.65016 (48.584)	xxx
GAMBIA	0.870 (0.602, 1.179)	2.53210 (24.089)	0.01906 (1.783)
GHANA	1.459 (1.152, 2.003)	-0.63878 (-2.670)	xxx
GUINEA B.	0.831 (0.684, 1.072)	5.09053 (35.901)	0.02781 (2.168)
GUINEA	1.011 (0.837, 1.292)	4.85004 (12.119)	xxx
IVORY COAST	0.996 (0.708, 1.375)	6.39430 (46.111)	xxx
KENYA	1.068 (0.907, 1.305)	4.48362 (49.601)	xxx
LESOTHO	1.009 (0.717, 1.465)	1.70969 (12.518)	xxx
LIBERIA	0.845 (0.695, 1.102)	1.14239 (2.047)	xxx
MADAGASCAR	0.937 (0.775, 1.196)	8.54757 (31.018)	xxx
MALAWI	0.744 (0.375, 1.177)	3.77871 (35.692)	0.02360 (3.137)
MAURITANIA	1.293 (1.109, 1.565)	5.05752 (65.745)	xxx
MAURITIUS	1.051 (0.810, 1.427)	3.08119 (43.292)	xxx
MOZAMBIQUE	1.189 (0.972, 1.561)	1.26190 (5.498)	xxx
NAMIBIA	1.131 (0.841, 1.580)	1.93323 (16.374)	xxx
NIGER	1.078 (0.903, 1.345)	5.79951 (48.370)	xxx
NIGERIA	1.122 (0.822, 1.543)	4.49616 (14.982)	xxx
REUNION	1.002 (0.823, 1.270)	5.95668 (47.192)	xxx
SIERRA LEONE	1.265(1.042, 1.635)	7.64194 (41.282)	xxx
SOUTH AFRICA	1.239 (0.527, 2.217)	1.49571 (7.966)	xxx
SAO TOME	1.371 (1.149, 1.758)	8.59936 (64.862)	xxx
SENEGAL	0.986 (0.782, 1.285)	6.00882 (45.434)	xxx
SEYCHELLES	1.223 (0.952, 1.623)	2.31383 (33.982)	xxx
SUDAN	0.861 (0.678, 1.144)	5.62238 (7.303)	-0.12472 (-1.637)
SWAZILAND	0.859 (0.580, 1.327)	1.71516 (14.499)	xxx
TANZANIA	1.428 (1.191, 1.797)	6.35066 (57.359)	xxx
TOGO	0.959 (0.749, 1.276)	6.10023 (50.497)	xxx
UGANDA	1.358 (1.117, 1.707)	7.12414 (27.905)	xxx
ZAMBIA	1.304 (0.995, 1.811)	8.46670 (56.915)	xxx

**Table 3: Summary based on the asymptotic results**

Mean Reversion ( $d < 1$ )	Unit Root ( $d = 1$ )		Explosive Behaviour ( $d > 1$ )
	$d < 1$	$d > 1$	
xxx	Malawi (0.744) Guinea Bis. (0.831) Liberia (0.845) Swaziland (0.859) Sudan (0.861) Gambia (0.870) Madagascar (0.937) Comoros (0.957) Angola (0.959) Togo (0.959) Botswana (0.967) Senegal (0.986) Ivory Coast (0.996) Centr. Africa (0.997)	Reunion (1.002) Lesotho (1.009) Guinea (1.011) Chad (1.035) Burkina Faso (1.041) Mauritius (1.051) Cameroon (1.053) Kenya (1.068) Niger (1.078) Eq. Guinea (1.085) Ethiopia (1.102) Congo Rep. (1.106) Gabon (1.115) Nigeria (1.122) Namibia (1.131) Benin (1.138) Mozambique (1.189) Seychelles (1.223) Burundi (1.233) South Africa (1.239) Zambia (1.304)	Djibouti (1.228) Sierra Leone (1.265) Mauritania (1.293) Cape Verde (1.308) Eritrea (1.314) Uganda (1.358) Sao Tome (1.371) Tanzania (1.428) Ghana (1.459)

**Table 4: Asymptotic and finite sample confidence intervals for the values of d**

	Asymptotic	Finite samples
ANGOLA	(0.689, 1.367)	(0.756, 1.501)
BURKINA FASO	(0.837, 1.358)	(0.889, 1.473)
BENIN	(0.926, 1.458)	(0.982, 1.576)
BOTSWANA	(0.735, 1.347)	(0.793, 1.485)
BURUNDI	(0.989, 1.637)	(1.051, 1.796)
CAPE VERDE	(1.082, 1.663)	(1.140, 1.797)
CAMEROON	(0.775, 1.432)	(0.853, 1.568)
CENTRAL AF.	(0.802, 1.299)	(0.852, 1.408)
CHAD	(0.824, 1.3 55)	(0.878, 1.468)
COMOROS	(0.764, 1.251)	(0.813, 1.354)
CONGO REP.	(0.629, 1.515)	(0.837, 1.644)
DJIBOUTI	(1.033, 1.562)	(1.083, 1.703)
EQ. GUINEA	(0.929, 1.314)	(0.969, 1.397)
ERITREA	(1.102, 1.642)	(1.158, 1.781)
ETHIOPIA	(0.906, 1.428)	(0.956, 1.564)
GABON	(0.909, 1.416)	(0.966, 1.528)
GAMBIA	(0.602, 1.179)	(0.682, 1.279)
GHANA	(1.152, 2.003)	(1.222, 2.226)
GUINEA B.	(0.684, 1.072)	(0.743, 1.181)
GUINEA	(0.837, 1.292)	(0.878, 1.394)
IVORY COAST	(0.708, 1.375)	(0.791, 1.507)
KENYA	(0.907, 1.305)	(0.949, 1.388)
LESOTHO	(0.717, 1.465)	(0.789, 1.636)
LIBERIA	(0.695, 1.102)	(0.737, 1.203)
MADAGASCAR	(0.775, 1.196)	(0.813, 1.294)
MALAWI	(0.375, 1.177)	(0.484, 1.337)
MAURITANIA	(1.109, 1.565)	(1.154, 1.662)
MAURITIUS	(0.810, 1.427)	(0.869, 1.563)
MOZAMBIQUE	(0.972, 1.561)	(1.026, 1.707)
NAMIBIA	(0.841, 1.580)	(0.924, 1.755)
NIGER	(0.903, 1.345)	(0.943, 1.442)
NIGERIA	(0.822, 1.543)	(0.902, 1.684)
REUNION	(0.823, 1.270)	(0.872, 1.366)
SIERRA LEONE	(1.042, 1.635)	(1.094, 1.773)
SOUTH AFRICA	(0.527, 2.217)	(0.636, 2.447)
SAO TOME	(1.149, 1.758)	(1.206, 1.915)
SENEGAL	(0.782, 1.285)	(0.835, 1.389)
SEYCHELLES	(0.952, 1.623)	(1.026, 1.757)
SUDAN	(0.678, 1.144)	(0.759, 1.243)
SWAZILAND	(0.580, 1.327)	(0.655, 1.511)
TANZANIA	(1.191, 1. 797)	(1.253, 1.935)
TOGO	(0.749, 1.276)	(0.807, 1.392)
UGANDA	(1.117, 1.707)	(1.180, 1.825)
ZAMBIA	(0.995, 1.811)	(1.073, 1.998)

**Table 5: Summary based on the finite sample results**

Mean Reversion ( $d < 1$ )	Unitroot ( $d = 1$ )		Explosivebehavior ( $d > 1$ )
	$d < 1$	$d > 1$	
xxx	Malawi (0.744) Guinea Bis. (0.831) Liberia (0.845) SWAZILAND (0.859) Sudan (0.861) Gambia (0.870) Madagascar (0.937) Comoros (0.957) Angola (0.959) Togo (0.959) Botswana (0.967) Senegal (0.986) Ivory Coast (0.996) Centr. Africa (0.997)	Reunion (1.002) Lesotho (1.009) Guinea (1.011) Chad (1.035) Burkina Faso (1.041) Mauritius (1.051) Cameroon (1.053) Kenya (1.068) Niger (1.078) Eq. Guinea (1.085) Ethiopia (1.102) Congo Rep. (1.106) Gabon (1.115) Nigeria (1.122) Namibia (1.131) Benin (1.138) South Africa (1.239)	Mozambique (1.189) Seychelles (1.223) Djibouti (1.228) Burundi (1.233) Sierra Leone (1.265) Mauritania (1.293) Zambia (1.304) Cape Verde (1.308) Eritrea (1.314) Uganda (1.358) Sao Tome (1.371) Tanzania (1.428) Ghana (1.459)