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Panicos Demetriades, University of Leicester, UK

Gregory James, Loughborough University, UK

FINANCE AND GROWTH IN AFRICA: THE BROKEN LINK

Panicos O. Demetriades[*] and Gregory A. James[ξ]

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ABSTRACT. Utilizing the latest panel cointegration methods we provide new empirical evidence from 18 countries that suggests that the link between finance and growth in Sub-Saharan Africa is 'broken'. Specifically, our findings suggest that banking system development in this region follows economic growth. They also indicate that there is no link between bank credit and economic growth.

KEYWORDS: Panel cointegration, cross-sectional dependence, African financial under-development, African credit markets

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[*] Department of Economics, University of Leicester.

[ξ] Corresponding author. Address for correspondence: Department of Economics, School of Business and Economics, Loughborough University, Leicestershire LE11 3TU, UK. Email: G.A.James@lboro.ac.uk. Tel: + 44 (0) 1509 222706. Fax: + 44 (0) 1509 223910.

1. Introduction

Although banking systems in Sub-Saharan Africa (SSA) lack depth compared to other parts of the developing world, an influential study by the World Bank has shown that savings mobilization in the region does not represent a binding constraint on financial deepening (Honohan and Beck, 2007). Drawing on surveys and other qualitative information, the World Bank study concludes that SSA banks do not lend enough - and as a result are excessively liquid compared to banks in other regions - because of a lack of acceptable or 'bankable' loan applications. Moreover, the study suggests that bank lending is often unprofitable because of severe information problems including lack of credit bureaus and weak contract enforcement. These intriguing insights lend themselves to rigorous macro-econometric analysis, not least because of their important policy implications.[1] To this end, this paper provides new macro-econometric evidence from a panel of 18 SSA countries over the period 1975 to 2006 that is consistent with the findings of the World Bank study. Specifically, we provide evidence to suggest that the relationship between finance and growth in the region is a rather loose one: finance, at best, follows growth; at worst there is no evidence of significant long run link between the two. We show that the worst-case scenario is obtained when financial development is measured by bank credit while the best-case scenario occurs when it is measured by liquid liabilities, an indicator that measures the size of bank balance sheets.[2]

We utilize the latest panel econometric techniques for non-stationary data that make the most of limited data availability for SSA countries. Such data scarcity is

1

indeed the main factor explaining why empirical studies of the finance-growth nexus in Africa are still in their infancy. Although a handful of such studies is now available - a good example of which is Gries *et al.* (2009) – these studies rely on individual country estimates obtained with relatively small samples and, as such, inference may be unreliable and/or inefficient compared to panel data methods. In contrast, the techniques we utilize circumvent these problems by combining 32 time observations with 18 cross-sections, resulting in a sample of 576 observations. This generates marked improvements in the reliability of statistical inference while respecting the non-stationary properties of the data and allowing cross-country heterogeneity. We avoid the limitations of conventional panel cointegration methods by allowing for cross-country dependence, utilizing some of the very latest advances in the panel cointegration literature.

The rest of the paper is organized as follows. Section 2 outlines the data and econometric methodology. Section 3 presents and discusses the empirical results. Section 4 summarizes and concludes.

2. Data and Methodology

The basic empirical model we postulate between financial and economic development, denoted FD_{it} and Y_{it} respectively, is the following simple log-linear relationship

$$\ln(FD)_{it} = \mu_i + \beta_i' \ln(Y_{it}) + e_{it}, \tag{1}$$

where the index $i = 1,\dots,N$ denotes countries and t denotes time.

We use a panel of 18 SSA countries that covers the period 1975-2006.[3] Annual data for financial development is obtained from the *Financial Development and Structure Database* of Beck *et al.* (2001), updated in April 2010. Specifically, we extract *bank deposits to GDP* (BD), *liquid liabilities to GDP* (LL) and *private credit by deposit money banks to GDP* (PC), each transformed in logarithms. Economic development is measured by *real GDP per capita* in international dollars (2005 constant prices), labeled LY, also transformed in logarithms, and is obtained from the *PENN World Table* (Table 6.3), compiled by Heston *et al.* (2006), updated in November 2009.

To test whether the variables are stationary or not, we consider five panel unit root tests, which can be classified into two groups depending on whether they allow for cross-sectional dependence. The first group consists of the tests of Im *et al.* (2003), Levin *et al.* (2002) and Harris and Tzavalis (1999), denoted W_{tbar}, t^* and ρ, respectively, which assume cross-sectional independence. The second group of tests allows for cross-sectional dependence and comprises the tests of Breitung (2000) and Pesaran (2007), denoted λ and Z_{tbar}, respectively.[4] The panel unit root tests of Breitung (2000), Levin *et al.* (2002), and Harris and Tzavalis (1999) assume that all cross-sectional units have a common autoregressive parameter while the tests of Im *et al.* (2003) and Pesaran (2007) both permit the individual autoregressive roots to differ across the cross-sectional units. All tests are normally distributed under the common null hypothesis of non-stationarity.

Provided that the variables are I(1) and that the regression error is stationary, equation (1) may be viewed as representing a cointegrating relationship. The second step in our analysis is, therefore, to test whether financial and economic

development are cointegrated. We use the four panel cointegration tests of Westerlund (2007), which have good small-sample properties and high power relative to popular residual-based panel cointegration tests (e.g. Pedroni, 2004). Furthermore, asymptotic and bootstrap *p*-values are computed, the latter making inference possible under very general forms of cross-sectional dependence. The tests are designed to test the null hypothesis of no cointegration by testing whether the error correction term in a conditional error correction model is equal to zero. If the null hypothesis of no error correction is rejected, then the null hypothesis of no cointegration is also rejected. The error correction model we consider is as follows:

$$\Delta \ln(FD_{it}) = \alpha_i + \rho_i(\ln(FD_{it-1}) - \beta_i' \ln(Y_{it-1})) + \sum_{s=1}^{p_i} \delta_{is}\Delta \ln(FD_{it-s}) + \sum_{s=0}^{p_i} \lambda_{is}\Delta \ln(Y_{it-s}) + e_{it} \quad (2)$$

The panel cointegration tests of Westerlund (2007) are able to accommodate serially correlated error terms, country-specific intercept and trend terms, and country-specific slope parameters. The G_τ and G_α test statistics test the null hypothesis of no cointegration for all cross-sectional units against the alternative that there is cointegration for at least one cross-sectional unit (i.e. $H_0 : \rho_i = 0$ for all *i* versus $H_1^G : \rho_i < 0$ for at least one *i*). Rejection of the null should therefore be taken as evidence of cointegration of at least one of the cross-sectional units. The P_τ and P_α test statistics pool information over all the cross-sectional units to test the null of no cointegration for all cross-sectional units against the alternative of cointegration for all cross-sectional units (i.e. $H_0 : \rho_i = 0$ versus $H_1^G : \rho_i = \rho < 0$ for all *i*). Rejection of the null should therefore be taken as evidence of cointegration

4

for the panel as a whole. Thus, the homogenous alternative hypothesis for these two tests is more restrictive. In small datasets, as in this study with T=32, Westerlund (2007) warns that the results of the tests may be sensitive to the specific choice of lag and lead lengths. Hence, to avoid overparametrization and the resulting loss of power, we hold the short-run dynamics fixed (i.e. p_i=p=1) in equation (2).

To formally test whether the cross-sectional units are independent, we assume that e_{it} is formed by a combination of a fixed component specific to the country and a random component that captures pure noise and estimate (2) using the FE estimator. We then use the Pesaran (2004) CD test on the residuals of (2) under the FE specification. The CD test statistic is normally distributed under the null hypothesis of no cross-sectional dependence. A possible drawback of the CD test is that adding up positive and negative correlations may result in failing to reject the null hypothesis even in the presence of cross-sectional dependence in the errors. Although it is not the case here, for the sake of completeness, we also computed the average absolute value of the off-diagonal elements of the cross-sectional correlation matrix of residuals, which can help identify cases of cross-sectional dependence in case the sign of the correlations alternates.

Finally, where evidence of cointegration is obtained, the third step in our analysis consists in estimating the long-run coefficient of real income per capita (LY). For this purpose, we use the newly developed estimators of Bai *et al.* (2009) known as CupBC (continuously-updated and bias-corrected) and CupFM (continuously updated and fully-modified) estimators. These two estimators have been shown to be superior in terms of mean bias to the LSDV (least squares dummy variables) and 2sFM (2-stage fully modified) estimators. Cross-sectional

dependence is modeled by means of a small number of common factors, which are treated as parameters and estimated jointly with β using an iterated procedure. The estimators are consistent and asymptotically (mixed) normal and permit inference to be conducted using standard test statistics. Under this approach, the structural errors e_{it} are allowed to be cross-sectionally dependent, strongly persistent, non-stationary, and correlated with the explanatory variables.

3. Empirical Results

The results of the panel unit root tests reported in Table 1 suggest that the unit root null hypothesis cannot be rejected at any conventional significance level for any variables.[5] The only exception is LL, for which the null can be rejected at the 10% level when using the W_{tbar} test. Since this rejection is very marginal, we treat all four variables as non-stationary and proceed to test for cointegration.

The computed values of the panel cointegration statistics are presented in Table 2 along with the asymptotic and bootstrapped p-values based on 500 replications.[6]

We begin by examining the results from equation (2) with bank deposits as the dependent variable. When using the asymptotic p-values, except for G_τ at the 10% level (i.e. when ρ_i is not restricted to be homogenous), the no cointegration null is never rejected. Based on the bootstrapped p-values (i.e. when allowance is made for cross-sectional dependence), the no cointegration null is only rejected for G_τ at the 5% level. Hence, there is little evidence of cointegration in this case. Similarly,

the results with private credit as the dependent variable show that the no cointegration null is never rejected, providing even less support for cointegration.

In contrast, the results with liquid liabilities as the dependent variable provide evidence of cointegration. With the asymptotic p-values, the no cointegration null is not only rejected for G_τ at the 1% level but also for P_τ at the 10% level (i.e. when ρ_i is restricted to be homogenous), suggesting that the whole panel is cointegrated. The results with the bootstrapped p-values provide even stronger evidence of cointegration. The no cointegration null is always rejected at least at the 10% level regardless of whether ρ_i is restricted to be homogenous or not. Since the homogenous alternative is particularly restrictive, these results provide strong evidence that the whole panel is cointegrated.

The Westerlund (2007) tests rely on the assumption of weakly exogenous regressors. In order to shed some light on the appropriateness of this assumption, we perform a series of reverse regression tests, the results of which are reported in Table 3. If LY is indeed weakly exogenous, then it should not be error-correcting, and this is exactly what we find. The results with LY as the dependent variable show that, consistent with the notion of weak exogeneity, the null of no error-correction cannot be rejected with either of the finance indicators as the regressor. Hence, there seems to be no violation of the weak exogeneity assumption.

The results of the cross-sectional independence tests are reported in Table 4.[7] The CD test always strongly rejects the null hypothesis of no cross-sectional dependence. Furthermore, the average absolute correlation of the off-diagonal elements of the cross-sectional correlation matrix of residuals ranges between

0.163-0.166. Hence, there is strong evidence of cross-sectional dependence in equation (2) under the FE specification.

Since we find evidence of cointegration between liquid liabilities and real GDP per capita, we proceed to estimate the long-run coefficient of LY with LL as the dependent variable. We use the Bai *et al.* (2009) estimators in order to account for the cross-sectional dependence in the data. The CupBC and CupFM estimation results are reported in Table 5.[8] Both estimators produce very similar results. The estimated coefficient is positive and significant in both cases. This suggests the existence of a stable and positive long-run relationship between the ratio of liquid liabilities to GDP and economic development. Furthermore, the magnitude of the estimated coefficient suggests that a 1% increase in real per capita income translates on average into a 2% increase in banking system development.

4. Summary and Policy Implications

The empirical results lead to the following three key findings relating to the relationship between finance and growth in Sub-Saharan Africa:

(i)   Banking system development as measured by liquid liabilities is positively associated to real GDP per capita. In other words, richer countries in Sub-Saharan Africa will tend to have more developed banking systems.

(ii)   Bank credit does not exhibit a long run relationship with real GDP per capita. In light of finding (i), this suggests that while banking systems may grow in tandem with economic growth, their ability to extend credit to the private sector does not follow suit.

(iii)   Real income per capita is weakly exogenous with respect to financial development, however measured.   Loosely speaking, finance does not lead economic growth in SSA.

These findings, which are consistent with the insights of Honohan and Beck (2007), Andrianova *et al.* (2010) and Demetriades and Fielding (2010), highlight the dysfunctional nature of African credit markets. Banking systems could be growing reflecting increased demand for financial services, while vital firm and household credit remains scarce. The broken link between the real economy and bank credit can go some way in explaining why financial development does not result in additional economic growth.   Fixing that link seems essential to kick start the finance and growth cycle in Sub-Saharan Africa. To this end, the strengthening of creditor protection laws and related informational infrastructure, including credit information bureaus, seems critical.

Acknowledgments

Footnotes

[1] These ideas are developed further in a recent paper by Andrianova *et al.* (2010), which shows that, when contract enforcement is weak, a credit market equilibrium with a high degree of loan defaults and low bank lending – the 'African credit trap' - can arise. Andrianova *et al.* (2010) also provide micro-econometric evidence utilizing panel data on hundreds of African banks over a ten-year period that is consistent with these predictions.

[2] This is what would be expected if banks do not lend 'enough'. In such circumstances, financial development, if measured by indicators that proxy the demand for financial services, could still exhibit a stable relationship with economic growth, although it would follow growth generated elsewhere. However, financial deepening, regardless of how it is measured, is unlikely to result in additional growth if it does not lead to an expansion in bank credit.

[3] The countries are: Burundi, Burkina Faso, Côte d'Ivoire, Cameroon, Ethiopia, Gabon, Ghana, Gambia, Kenya, Madagascar, Mauritius, Niger, Nigeria, Senegal, Sierra Leone, Swaziland, Seychelles and Togo. These countries were chosen taking into account data availability and the fact that the techniques employed in this paper require a balanced panel. We have tried to include as many countries as possible over as long a sample period as possible.

[4] We consider a variant of the Breitung (2000) test made robust to cross-sectional dependence.

[5] The Pesaran (2007) test is performed using the Stata "pescadf" command written by Piotr Lewandoski.

[6] The tests are performed using the Stata "xtwest" command (see Persyn and Westerlund, 2008).

[7] The test is performed using the Stata "xtcsd" command (see De Hoyos and Sarafidis, 2006).

[8] The estimations are conducted using GAUSS.

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TABLE 1: *Unit Root Tests*

Variable	Test values					*p*-values				
	W_{tbar}	t^*	ρ	λ	Z_{tbar}	W_{tbar}	t^*	ρ	λ	Z_{tbar}
BD	-0.134	0.742	0.820	0.312	1.273	0.446	0.771	0.893	0.622	0.899
LL	-1.574	-1.201	0.792	0.170	-0.933	0.057	0.114	0.652	0.567	0.175
PC	-0.646	-1.238	0.863	1.378	2.093	0.259	0.107	0.994	0.916	0.982
LY	0.352	1.082	0.819	0.955	0.927	0.637	0.860	0.883	0.830	0.823

Notes: All unit-root tests are implemented with a constant and trend in the test regression and take a unit-root as the null hypothesis. For semi-parametric corrections, the Bartlett kernel is employed with the Newey and West (1994) bandwidth selection algorithm. All bandwidth and lag orders are set according to the rule $4(T/100)^{2/9}$. The lags are chosen according to the Akaike criterion. The *p*-values are for a one-sided test based on the normal distribution.

TABLE 2: *Cointegration Tests*

Test	BD			LL			PC		
	Value	*p*-val[a]	*p*-val[b]	Value	*p*-val[a]	*p*-val[b]	Value	*p*-val[a]	*p*-val[b]
G_r	-2.100	0.064	0.042	-2.289	0.008	0.004	-1.682	0.675	0.400
G_a	-6.704	0.634	0.146	-7.741	0.320	0.026	-5.226	0.928	0.460
P_r	-6.744	0.272	0.262	-7.633	0.067	0.080	-5.306	0.800	0.540
P_a	-4.316	0.467	0.282	-5.244	0.166	0.096	-3.617	0.721	0.364

Notes: The Westerlund (2007) tests take no cointegration as the null. The test regression is fitted with a constant and one lag and lead. The kernel bandwidth is set according to the rule $4(T/100)^{2/9}$. [a]The *p*-values are for a one-sided test based on the normal distribution. [b]The *p*-values are for a one-sided test based on 500 bootstrap replications.

TABLE 3: *Weak Exogeneity Tests*

LY (dependent)	BD (regressor)			LL (regressor)			PC (regressor)		
Test	Value	p-val^a	p-val^b	Value	p-val^a	p-val^b	Value	p-val^a	p-val^b
G_τ	-1.821	0.418	0.278	-1.794	0.470	0.312	-1.876	0.332	0.208
G_a	-6.458	0.703	0.264	-5.837	0.845	0.418	-6.069	0.798	0.380
P_τ	-7.228	0.124	0.210	-7.362	0.110	0.146	-7.159	0.153	0.206
P_a	-4.956	0.244	0.212	-4.915	0.256	0.176	-4.591	0.365	0.244

Notes: The Westerlund (2007) tests are implemented with LY as the dependent variable. The test regression is fitted with a constant and one lag and lead. The kernel bandwidth is set according to the rule $4(T/100)^{2/9}$. [a]The p-values are for a one-sided test based on the normal distribution. [b]The p-values are for a one-sided test based on 500 bootstrap replications.

TABLE 4: *Cross-Sectional Independence Tests*

Test	BD			LL			PC		
	Value	*p*-value	Correlation	Value	*p*-value	Correlation	Value	*p*-value	Correlation
CD	4.880	0.000	0.166	4.901	0.000	0.163	4.858	0.000	0.163

Notes: The Pesaran (2004) *CD* test takes cross-sectional independence as the null. The *p*-values are for a one-sided test based on the normal distribution. Correlation is the average absolute value of the off-diagonal elements of the cross-sectional correlation matrix of residuals obtained from estimating (2) under an FE specification.

TABLE 5: *Cointegration Estimation Results*

CupBC		CupFM	
β	SE	β	SE
2.0173	0.007	2.0170	0.008

Notes: The value β refers to the estimated long-run coefficient of LY in the model with LL as the dependent variable. SE refers to the standard error. CupBC refers to the continuously updated and bias corrected estimator. CupFM refers to the continuously updated and fully modified estimator. The estimated model includes a constant.