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REMITTANCE INFLOWS AND POVERTY DYNAMICS IN SOUTH AFRICA: AN EMPIRICAL INVESTIGATION

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REMITTANCE INFLOWS AND POVERTY DYNAMICS IN SOUTH AFRICA: AN

EMPIRICAL INVESTIGATION

Mercy T. Musakwa 1 and Nicholas M. Odhiambo

Abstract

In this study, we investigate the impact of remittance inflows on poverty reduction in South Africa, using

time series data from 1980 to 2017. The main objective of this study is to establish if South Africa can

harness remittance inflows to alleviate poverty. Two poverty proxies, namely household consumption

expenditure and infant mortality are used in this study. To ensure robustness of the results, both income

and non-income proxies of poverty are employed. Using the autoregressive distributed lag (ARDL) bound

approach, the study found that remittance has a negative impact on poverty in the short run and in the long

run when household consumption expenditure is used as a proxy for poverty. However, when infant

mortality rate is used as a proxy, remittance is found to have no impact on poverty. It can be concluded

that the impact of remittance on poverty is sensitive to the proxy used. The study concludes that South Africa

could benefit immensely from some forms of remittances in its quest to poverty alleviation.

Key Words: Remittance; poverty reduction; autoregressive distributed lag (ARDL); household

consumption expenditure expenditure; infant mortality rate; South Africa

JEL Classification: F24, I31

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Introduction

The end of the Millennium Development Goals (MDGs) in 2015 with some of the goals partially achieved, paved the way for the Sustainable Development Goals (SDGs) 2030. Under the MDGs, the poverty reduction target was achieved in aggregate; but at country level, a number of developing countries are still struggling with high poverty levels, and South Africa is not an exception. Thus, many countries are still seeking solutions to poverty by focusing on different macroeconomic variables such as foreign direct investment; and most recently, attention has been drawn to the poverty-reducing effect of remittances. The literature on the impact of remittances on economic growth is well documented, although the findings are inconclusive (Kumar, 2013; Imai et al., 2014; Goschin, 2014; Atanda and Charles, 2014; Lim and Simmons, 2015; Adams and Klobodu, 2016; Meyer and Shera, 2017; Makun, 2018). However, little has been done to establish if remittances have a positive impact on poverty alleviation, after discovering that economic growth does not directly translate to poverty reduction. The SDGs also support favorable migration policies based on the potential that remittances have on the policy thrust of inclusive development as spelt out by SDG 10.7 and subsequent subsections (United Nations, 2017). Given a dearth of literature on the impact of remittances on poverty levels on one hand, and the thrust of the SDG spearheaded by the United Nations on the other hand, another study that explores the direct impact of remittances on poverty levels in South Africa cannot be overemphasised, as it will shed more light on the country's poverty alleviation policies.

According to Ratha *et al.* (2018), remittance inflow in low and middle income countries has significantly increased in the past years and is projected to reach \$528 million in 2018 – a growth

of 10.8% from the 2017 level. This has caused remittance inflows to be three times greater than official development assistance (ODA), and larger than foreign direct investment (FDI) when China is excluded in low and middle income countries (Ratha *et al.*, 2018:1). Remittances have grown to be an important source of income not only for low and middle income countries at macro level (national level) but also at micro level (household level). This makes another study on the impact of remittance on poverty in South Africa important, given this perceived importance of remittances on the one hand, and extremely high poverty levels on the other hand.

Studies that have been done on South Africa on this subject are largely based on panel data and cross sectional analysis. Given the limitations of cross sectional data in capturing country-specific effects; and that of panel data in pooling countries with different economic dynamics (Blonigen and Wang, 2004), this study employs a time-series methodology, autoregressive distributed lag (ARDL) in particular, to capture the South African country-specific remittance-poverty nexus.

To improve the robustness of the results, and given the debate that is still raging on what a more comprehensive measure of poverty is between income and non-income measures, this study uses two poverty proxies – household consumption expenditure, which is an income measure and infant mortality rate, a non-income measure. These poverty proxies have been selected based on the need to capture poverty from income and non-income perspectives. Moreover, unavailability of time series data for other poverty measures like poverty headcount, poverty gap, poverty gap squared and human development index, has contributed to the selection of these proxies. Other studies

have used these poverty proxies, for example, Ravallion (2001), Rehman and Shahbaz (2014) and Kaidi *et al.* (2018) used household consumption expenditure as a proxy for poverty and Van Multzahn and Durrheim (2008) and Abosedra *et al.* (2016) used infant mortality rate as a poverty proxy.

To the best knowledge of the author, no study has been done employing a time series analysis in general – and autoregressive distributed lag approach (ARDL), in particular – to analyse the remittance-poverty relationship in South Africa. Therefore, the contribution of this study to the body of knowledge and South Africa's relevant policy makers cannot be overstated.

Although South Africa is a major source of outward remittance, especially in the Southern Africa Development Community (SADC), the country has also experienced emigration, with over 900 000 South Africans in 2017, estimated to be working outside the country (Businesstech, 2018). The question that needs an investigation is whether the money that is sent in by South Africans working abroad can help to alleviate poverty in the country. This study will inform policy makers on whether remittances are useful in reducing poverty so that they can, in an informed way, devise measures that can be taken by the government to harness remittances in order to reduce poverty.

The rest of the study is set out as follows: Section 2 discusses the literature review both empirical and theoretical; Section 3 outlines estimation techniques; Section 4 presents and discusses the results; and Section 5 concludes the study.

1. Empirical Literature Review

2.1 Remittance and Poverty Dynamics in South Africa

The United Nations (2018) defines remittance as personal transactions from a migrant to their family and friends back in their home country. This makes remittance more targeted to the needs of the family or friends. Thus, remittances are more inclined to the special needs of the family, which are unique and vary from individual to individual, in contrast to other cash transfers from the government that are one size fit all.

South Africa has experienced notable emigration, with mostly young age groups looking for greener pastures in other countries. Among the top destinations for these migrants are the United Kingdom, Australia, New Zealand and United States (Businesstech, 2018). The South African government has put in place policies that focus much on regulating remittance outflows. Unlike the foreign direct investment drive that is well supported in the economic policies and different arms of government, the remittance issue is still on the back burner, though it is gradually gaining attention.

Remittance inflows in South Africa improved during the period under study, although the numbers remain depressed when compared to countries like India and Brazil (United Nations Conference on Trade and Development 'UNCTAD', 2018). Remittance inflows in South Africa were depressed in 1980 when a proportion of 3% of GDP was received (UNCTAD, 2018). An average

of 5% was recorded between 1980 and 1996 (UNCTAD, 2018). The inflows significantly increased from 1997 when the inflows increased from 5% to 11% (UNCTAD, 2018). An average of 24.3% was registered between 1997 and 2017 (UNCTAD, 2018). This period received the highest remittances during the period under study (UNCTAD, 2018). South Africa, like other middle income countries, experienced a gradual increase in remittance levels from the late 90s (UNCTAD, 2018).

On the poverty front, South Africa has made great strides in implementing policies that focus on poverty alleviation. The country's national economic policy; National Development Plan 2030 has a broad objective of reducing poverty and inequality, among other policy advancements. The approach to poverty alleviation has been multifaceted and can be categorised as: (i) economic empowerment, associated with financial support for small businesses, training to improve success rates of business, marketing of products, copyright and intellectual property rights with the Competition Commission being the custodian, international exposure to business and export opportunities; (ii) improved access to social services, such as education, housing and health through free access to those who cannot afford these and increasing awareness and accessibility of the services; and (iii) direct intervention through the social safety net, where social grants are given to targeted individuals, public works programmes and social insurance. In response to these policies, there has been a gradual reduction in poverty although the figures are still high and volatile (World Bank, 2018). The poverty headcount was recorded at 29.3% in 1993 and increased to 33.8% in 1996 before falling consistently over the years to 16.5% in 2010 (World Bank, 2018). In 2014 the poverty headcount took an upward trend, recording 18.9% (World Bank, 2018). The

human development index (HDI) marginally improved from 0.618 in 1990 to 0.699 in 2017 (United Nations Development Programme "UNDP", 2018). This was a great improvement compared to 0.39 and 0.537 recorded for sub-Saharan Africa during the same period (UNDP, 2018).

2.2 A Review of Related Literature

According to Hagen-Zanker and Himmelstine (2016), remittances can be in cash or kind. Theoretical literature suggests a positive relationship between remittance and poverty reduction. Although it has the potential to mitigate poverty, a remittance option does not belong to the poorest in society because of the cost associated with migration (Hagen-Zanker and Himmelstine, 2016).

Literature further splits the positive impact of remittances on poverty reduction into direct (see De Vries, 2011; Hagen-Zanker and Himmelstine, 2016) and indirect impact (see Ratha, 2007). Remittance has a direct positive effect on households (Hagen-Zanker and Himmelstine, 2016). They reach the greater share of the overall population compared to other forms of transfers (Hagen-Zanker and Himmelstine, 2016). Another advantage of remittances is that it reaches both male and female recipients compared to targeted cash transfers (Duflo and Hendry, 2004). Apart from consumption, remittances are associated with investment – human capital, cash assets (Adam Jr and Page, 2005; Bui *et al.*, 2015), real estate and small business (Ratha, 2007).

In addition to the direct impact of remittances on poverty through positively affecting household income, remittances have an indirect effect on welfare through the multiplier effect on the economy (Ratha, 2007). Thus, remittances can mitigate poverty at a national level through the multiplier effect, which is realised through changes in consumption and investment (Ratha, 2007).

The stable and countercyclical nature of remittances make it important as a shock absorber when there are social shocks like civil wars and disasters (Kapur, 2004). The ultimate result of the multiplier effect is a growth in the economy that is more than the initial change in investment and consumption. De Vries (2011) summed the benefits of remittances as: poverty reduction, economic growth spur, savings and investment increase, and sectoral growth stimulation.

The empirical studies on the impact of remittances on poverty or welfare, have recently drawn widespread attention following the realisation of the potential that the remittances have on poverty alleviation, coupled with a steady increase of remittance inflows that developing countries have been receiving of late. However, unlike the theoretical literature, which lays bare in no uncertain terms the positive relationship between remittances and poverty reduction, the empirical literature on the subject is still subject to debate. Studies that have attempted to establish the relationship between remittances and poverty reduction largely found a positive relationship between the two (Adam Jr and Page, 2005; Acosta *et al.*, 2008; Gupta *et al.*, 2009; Fuente, 2010; Adam Jr and Cuecuecha, 2013; Akobeng, 2016; Nahar and Rashad, 2017; Vacaflores, 2018; Tsaurai, 2018). There are also some studies that found the relationship between remittances and poverty alleviation

to be a complex one, varying from one study country to the other or sensitive to the poverty proxy used (see, for example, Azam *et al.*, 2016; Wangle and Devkota, 2018).

Adam Jr and Page (2005) studied the impact of remittances on poverty using data from 71 developing countries and found remittance to reduce the depth and severity of poverty. A 10% increase in per capita international remittance was found to lead to a 3% reduction in people living in poverty. Acosta *et al.* (2008) analysed the impact of remittances on poverty in Latin America, employing panel data from 59 industrial developing countries from 1970 to 2000. The findings from the study were consistent with those by Adam Jr and Page (2005) – thus remittance was found to reduce poverty.

Gupta *et al.* (2009) also found a positive direct poverty mitigating effect in 76 countries. A 10% increase in remittances was found to lead to a 1% decrease in the poverty headcount and poverty gap. In the same vein, Fuente (2010) investigated the impact of remittance on poverty in Mexico, using data from 1998 to 2000, on rural households and found a negative relationship between poverty and remittance. Adam Jr and Cuecuecha (2013) also analysed the impact of remittance on poverty. The findings from the study were consistent with Fuente (2010) and Gupta *et al.* (2009).

Akobeng (2016) investigated the impact of remittance on poverty and inequality using micro data from sub-Saharan Africa and found remittance to reduce poverty but the extent to which poverty reduction is achieved was found to be sensitive to the poverty measure used. Nahar and Rashad

(2017) investigated the impact of remittance on poverty reduction in Indonesia using data from 1983 to 2015 and also found remittance to reduce poverty; however, only by a small margin.

Vacaflore (2018) investigated the relationship between remittance and poverty employing panel data from 2000-2013 from 19 Latin America countries. The results confirmed a negative relationship between remittance and poverty. Thus, remittance was found to reduce poverty in Latin America. Tsaurai (2018) also investigated the impact of remittance on poverty in selected emerging market economies using panel data from 1980 to 2012. The results were in line with the findings by Vacaflore (2018) and Wangle and Devkota (2018).

Azam *et al.* (2016) analysed the impact of remittance on poverty alleviation in 39 low middle, upper middle and high income countries using panel data. Remittance was found to have a positive impact on poverty alleviation in upper middle income countries. A 1% increase in remittance was found to lead to a 0.2% reduction in poverty. However, in the same study, remittance was found to be insignificant in high income countries. The findings by Azam *et al.* (2016) imply that results from one study country may not be generalised or used to inform policy in another country.

Wangle and Devkota (2018) analysed the impact of remittance on poverty using longitudinal panel survey data from 1996 to 2017 for Nepal. The results confirm that remittance reduces poverty in Nepal. However, the results were sensitive to time frame and poverty definition. It can be concluded that the impact of remittance on poverty reduction is largely supported in the literature.

However, studies like Azam *et al.* (2016), and Wangle and Devkota (2018) suggest that a country specific study remains important to come up with policy relevant results.

3. Methodology

An autoregressive distributed lag (ARDL) bounds testing approach is used to investigate the impact of remittances on poverty in this study. The ARDL bounds testing approach has been selected because of a number of advantages. First, the ARDL approach does not require all variables to be integrated of the same order (Pesaran *et al.*, 2001). Variables can be integrated of order [I (1)], order 0 [I (0)], or fractionally integrated (Pesaran *et al.*, 2001: 290). Second, the ARDL bounds approach uses a single reduced form equation, unlike other methods (see Duasa, 2007). Third, the ARDL approach to cointegration is robust in a small samples (Odhiambo, 2009; Solarin and Shahbaz, 2013). Fourth, the ARDL bounds testing approach to cointegration provides unbiased estimates of the long-run model, even in cases where some variables are endogenous (Odhiambo, 2009). It is against this background that the ARDL bounds approach was selected in this study.

Variables

In this study, poverty is the dependent variable measured by household consumption expenditure (Pov1) and infant mortality rate (Pov2). These two proxies have been selected based on the need to capture poverty from a multidimensional perspective so as to improve the robustness of the results. Pov1 (household consumption expenditure) is expressed as a percentage of gross domestic

product. This proxy captures income poverty. In the literature, a number of studies have employed this poverty proxy (see Ravallion, 2001; Odhimbo, 2009; Rehman and Shahbaz, 2014; Kaidi *et al.*, 2018). A positive relationship between Pov1 and remittance implies that remittances mitigate poverty. Pov2 (infant mortality rate) captures non-income poverty. The higher the infant mortality rate, the worse the poverty. A negative relationship between infant mortality rate and remittances imply that remittance leads to poverty alleviation as it reduces mortality. There are numerous studies that used infant mortality rate as a proxy for poverty (see, Laderchi, 1984; Van Multzahn and Durrheim, 2008; Abosedra *et al.*, 2016).

The main dependent variable of interest is remittance (REM), proxied by international remittance inflows. International remittance inflows are expressed as a percentage of GDP. This has the advantage of making comparisons among countries easier as it takes the size of the economy into account. A number of studies found remittance to mitigate poverty (Wangle and Devkota, 2018; Tsaurai, 2018). Remittance is expected to have a negative impact on poverty when infant mortality rate is used as a proxy, while a positive impact is expected when household consumption expenditure is used as a proxy.

To fully specify the model and to eliminate variable-omission-bias, a number of control variables have been included in the model, namely: (i) trade openness (TOP); (ii) education (EDU); (iii) real gross domestic product per capita (GDPC) and inflation rate (INF). Trade openness is measured by a sum of exports and imports divided by GDP.

Trade openness (TOP) shows how open the economy is to international transactions. Bharadwaj (2014) found trade openness to help in poverty alleviation. Trade openness is expected to reduce poverty. Another variable that is also included in the model is education (EDU). Education, which is a measure of human capital, is measured by gross primary school enrolment rates. Higher enrolment rates are expected to result in high human capital which is crucial in knowledge acquisition, training and improved access to better paying jobs; hence higher income. Education is expected to reduce poverty.

Living standards (LS) is proxied by real gross domestic product per capita (GDPC), which is an indicator of living standards in a country, has been included as one of the control variables. A high real GDP per capita indicates better living standards and a decrease in poverty levels. Akobeng (2016) and Nahar and Arshad (2017) among other studies, used gross domestic product per capita as a control variable. The LS is expected to reduce poverty. Inflation (INF) is also one of the control variables, measured as the rate of change of the consumer price index. High inflation rates erode the purchasing power of the poor, putting them in a worse off position (Mohr *et al.*, 2008, p.480). A positive relationship is expected between inflation and poverty. Thus an increase in the inflation rate is expected to put the poor on a worse off position.

Model Specification

The model specification follows Adam Jr and Page (2007) and Gupta *et al.*, (2009), but differs on the variables included in the model. The general model specification is given in Equation 1.

Where Pov_m is household consumption expenditure (Pov1) or infant mortality rates (Pov2). The two poverty proxies enter the equation one at a time, but the control variables remain the same.

The ARDL specification of the general empirical model in Equation 1 can be expressed as:

Where α_0 is a constant; $\alpha_{1i} - \alpha_{6i}$ are short-run coefficients; $\vartheta_1 - \vartheta_6$ are long-run coefficients; and μ_{1t} is the white noise error term. The rest of the variables are as defined in Equation 1.

The ARDL-based error correction model of the general empirical model is specified as follows:

$$\begin{split} \Delta Pov_{mt} &= \alpha_0 + \sum_{i=1}^n \alpha_{1i} \, \Delta Pov_{mt-i} + \sum_{i=0}^n \alpha_{2i} \, \Delta REM_{t-i} + \sum_{i=0}^n \alpha_{3i} \, \Delta TOP_{t-i} + \sum_{i=0}^n \alpha_{4i} \, \Delta EDU_{t-i} \\ &+ \sum_{i=0}^n \alpha_{5i} \, \Delta LS_{i-1} + \sum_{i=0}^n \alpha_{6i} \, \Delta INF_{t-i} + \gamma_{1i} ECM_{t-1} + \mu_t \, ... \, ..$$

Where α_0 is a constant; $\alpha_{1i} - \alpha_{6i}$ are short-run coefficients; ECM is the error correction model and γ_{1i} is the white noise error term. The rest of the variables are as defined in Equation 1.

Data Sources

The study employs time series data from 1980 to 2017 to investigate the direct impact of remittance on poverty reduction. The data on gross domestic product, education, trade openness and inflation was obtained from the World Bank Development Indicators Database, while remittance data was extracted from UNCTAD data base. The analysis of the data was done using Microfit 5.0.

4. Results

4.1 Unit Root Test

The ARDL approach employed in this study does not require a unit root test, however the test was done on all the variables included in the model to ascertain that they are integrated of order 0 [I (0)] or 1 [I(1)]. The ARDL approach falls away if variables are integrated of an order higher than one. Table 1 presents the results of the unit root tests conducted using the Dickey-Fuller Generalised Least Square (DF-GLS) and Perron (1997) (PPURoot) tests.

Table 1: Unit Root Test Results

DF-GLS Test	PPURoot Test

Variable	Stationarity	y of Variable	Stationarity	of Variable in	Stationarity	of all	Stationarity	of all
	in Levels		First Difference		Variables in Levels		Variables Difference	in First
	Without Trend	With Trend	Without Trend	With Trend	Without Trend	With Trend	Without Trend	With Trend
Pov1	-1.6113*	-2.8900***	_		-5.8455**	-6.1583***	_	_
Pov2	-0.7869	-12255**	_		-3.3657	-3.2152	-5.3584**	-5.6282**
REM	-0.4459	-2.8724	-3.9378***	-3.9736***	-6.4715***	-6.1904***	_	_
ТОР	-1.6938*	-2.3758	_	-6.0333***	-3.8475	-3.6592	-7.1419***	-7.0464***
EDU	-1.0469	-1.2385	-5.2803***	-5.3785***	-7.1622***	-7.1136***	_	_
LS	-0.1560	-1.7040	3.5096***	-4.0021***	-4.0192	-3.4146	-5.8828**	-6.3588***
INF	-1.4454	-2.8164	-5.6970***	-6.0829***	-4.1289	-4.0819	-5.2864**	-6.4695***

Note: *, ** and *** denote statistical significance at 10%, 5% and 1% levels, respectively

Unit root test results presented in Table 1 show that all variables are stationary in levels or in first difference. This confirms the appropriateness of the use of the ARDL bounds testing approach. The study, therefore, proceeds to testing the cointegration in the two models –Model 1 where Pov1 is proxied by household consumption expenditure and Model 2 where Pov2 is proxied by infant mortality rate.

Cointegration Test

Cointegration results are presented in Table 2.

Table 2: ARDL Approach to Cointegration Results

Model	Independent	Functio	n		F –statistic		Cointeg	Cointegration Status	
	Variables								
1	Pov1	F(Pov1 EDU, G		TOP,	3.8716**	**	Cointegr	rated	
2	Pov2	F(Pov2 EDU, G		TOP, NF)	3.5278**	:	Cointegr	rated	
	Asymptotic Critical Values (unrestricted intercept and no trend)								
Pesaran et d	10%		5%		1%	1%			
values (1 aur	e CI(iii) Case III)	I(0)	I(1)	1	I(0)	I(1)	I(0)	I(1)	
		2.26	3.35	5	2.62	3.79	3.41	4.68	

Note: *, ** and *** denote statistical significance at 10%, 5% and 1% levels respectively

The calculated F-statistics for Model 1 and Model 2 are 3.8716 and 3.5278 respectively and are greater than the upper-bound critical values by Pesaran *et al.* (2001). This confirms a long-run relationship among the variables in both models – Model 1 and Model 2.

1.2 Coefficient Estimation

To proceed with the analysis, after confirming cointegration in Model 1 and Model 2, ARDL is used to estimate the coefficients for the models. The optimal lag length for the two models is selected using Akaike Information Criteria (AIC) depending on the most parsimonious model. The long-run and short-run results for Model 1 and Model 2 are presented in Table 3.

Table 3: Empirical results for Model 1 and Model 2

C 88.9503*** 5.5447 22.6921*** 9.9782 REM 0.1832* 1.8045 0.0361 1.0719 TOP 0.0257 0.1357 0.0361 1.3703 EDU 0.0150 0.1610 -0.0399** -3.3337 LS -0.1917 -1.2033 -0.2408*** -8.1183 INF -0.3841* -1.8712 -0.1627*** -4.2013 Panel B (Short run results) APovI 0.1597 0.4549 - - AREM 0.3200° 1.8855 0.0093 1.0064 AREM (I) -0.1671 -1.2380 -0.0056 -0.7026 AREM(2) 0.0042 0.0339 - - ATOP (I) 0.1671* 1.8319 - - ATOP (I) 0.1671* 1.8319 - - AEDU -0.0138 -0.2030 -0.0011 -0.3871 AEDU(3) 0.1815* 1.8723 0.0043 1.2861 AEDU(3) 0.15	Model	Model 1 (Dependent Variable is Pov1)		Model 2 (Dependent Variable is Pov2)		
Regressor Coefficient T-ratio Coefficient T-ratio C 88.9503*** 5.5447 22.6921**** 9.9782 REM 0.1832* 1.8045 0.0361 1.0719 TOP 0.0257 0.1357 0.0361 1.3703 EDU 0.0150 0.1610 -0.0399** -3.3337 LS -0.1917 -1.2033 -0.2408*** -8.1183 INF -0.3841* -1.8712 -0.1627*** -4.2013 Panel B (Short run results) ACREM APov1 0.1597 0.4549 - - AREM (1) -0.1671 -1.2380 -0.0056 -0.7026 AREM (2) 0.0042 0.0339 - - - ΔTOP (1) 0.1671* 1.8319 - - - - ΔTOP (2) 0.1268* 1.8764 - - - - - ΔEDU (2) 0.138 -0.2030 -0.0011 -0.3871 - -<		ARDL (2, 3, 3, 4, 3, 0)		ARDL (1, 2, 1, 3, 0, 2)		
C 88.9503*** 5.5447 22.6921*** 9.9782 REM 0.1832* 1.8045 0.0361 1.0719 TOP 0.0257 0.1357 0.0361 1.3703 EDU 0.0150 0.1610 -0.0399** -3.3337 LS -0.1917 -1.2033 -0.2408*** -4.2013 Panel B (Short run results) APov1 0.1597 0.4549 AREM 0.3200* 1.8855 0.0093 1.0064 AREM (1) -0.1671 -1.2380 -0.0056 -0.7026 AREM (2) 0.0042 0.0339 ATOP 0.0589 0.5158 0.0127** 2.9016 ATOP (1) 0.1671* 1.8319 ATOP (1) 0.1671* 1.8319 ΔTOP 0.0188 -0.2030 -0.0011 -0.3871 ΔEDU -0.0138 -0.2030 -0.0011 -0.3871 ΔEDU 0.1741* 2.4465 0.0051 1.6083 ΔEDU 1 0.1815* 1.8723 0.0043 1.2861 ΔEDU(2) 0.1815* 1.8723 0.0043 1.2861 ΔEDU(3) 0.1560* 2.2216 ΔLS -0.0393 -0.1526 -0.0437*** -5.8825 ΔLS(1) -0.2980 -1.2242 ΔLS(2) -0.3385 -1.3690 ΔLNF(1)	Panel A: Long-Run Resu	ılts		1		
REM 0.1832° 1.8045 0.0361 1.0719 TOP 0.0257 0.1357 0.0361 1.3703 EDU 0.0150 0.1610 -0.0399°* -3.3337 LS -0.1917 -1.2033 -0.2408°** -8.1183 INF -0.3841° -1.8712 -0.1627*** -4.2013 Panel B (Short run results)	Regressor	Coefficient	T-ratio	Coefficient	T-ratio	
TOP 0.0257 0.1357 0.0361 1.3703 EDU 0.0150 0.1610 -0.0399** -3.3337 LS -0.1917 -1.2033 -0.2408**** -8.1183 INF -0.3841* -1.8712 -0.1627*** -4.2013 Panel B (Short run results) ΔPov1 0.1597 0.4549 - - ΔREM 0.3200* 1.8855 0.0093 1.0064 ΔREM (1) -0.1671 -1.2380 -0.0056 -0.7026 ΔREM(2) 0.0042 0.0339 - - ΔTOP 0.0589 0.5158 0.0127*** 2.9016 ΔTOP (1) 0.1671* 1.8319 - - ΔTOP (2) 0.1268* 1.8764 - - ΔEDU -0.0138 -0.2030 -0.0011 -0.3871 ΔEDU(1) 0.1741* 2.4465 0.0051 1.6083 ΔEDU(2) 0.1815* 1.8723 0.0043 1.2861	С	88.9503***	5.5447	22.6921***	9.9782	
EDU 0.0150 0.1610 -0.0399** -3.3337 LS -0.1917 -1.2033 -0.2408*** -8.1183 INF -0.3841* -1.8712 -0.1627*** -4.2013 Panel B (Short run results) ΔPov1 0.1597 0.4549	REM	0.1832*	1.8045	0.0361	1.0719	
LS	TOP	0.0257	0.1357	0.0361	1.3703	
NF	EDU	0.0150	0.1610	-0.0399**	-3.3337	
Panel B (Short run results) ΔPov1 0.1597 0.4549 - - ΔREM 0.3200° 1.8855 0.0093 1.0064 ΔREM (1) -0.1671 -1.2380 -0.0056 -0.7026 ΔREM(2) 0.0042 0.0339 - - ΔTOP 0.0589 0.5158 0.0127** 2.9016 ΔTOP (1) 0.1671* 1.8319 - - ΔTOP (2) 0.1268* 1.8764 - - ΔEDU -0.0138 -0.2030 -0.0011 -0.3871 ΔEDU(3) 0.1741* 2.4465 0.0051 1.6083 ΔEDU(2) 0.1815* 1.8723 0.0043 1.2861 ΔEDU(3) 0.1560* 2.2216 - - ΔLS -0.0393 -0.1526 -0.0437**** -5.8825 ΔLS(1) -0.2980 -1.2242 - - ΔINF -0.2933* -1.7843 0.0084 1.2116 ΔINF(1) -	LS	-0.1917	-1.2033	-0.2408***	-8.1183	
ΔPoV1 0.1597 0.4549 - - ΔREM 0.3200° 1.8855 0.0093 1.0064 ΔREM (1) -0.1671 -1.2380 -0.0056 -0.7026 ΔREM(2) 0.0042 0.0339 - - ΔTOP 0.0589 0.5158 0.0127** 2.9016 ΔTOP (1) 0.1671° 1.8319 - - ΔTOP (2) 0.1268° 1.8764 - - ΔEDU -0.0138 -0.2030 -0.0011 -0.3871 ΔEDU (1) 0.1741° 2.4465 0.0051 1.6083 ΔEDU(2) 0.1815° 1.8723 0.0043 1.2861 ΔEDU(3) 0.1560° 2.2216 - - ΔLS -0.0393 -0.1526 -0.0437**** -5.8825 ΔLS(1) -0.2980 -1.2242 - - ΔINF -0.2933* -1.7843 0.0084 1.2116 ΔINF(1) - - - ECM(-1)	INF	-0.3841*	-1.8712	-0.1627***	-4.2013	
ΔREM 0.3200° 1.8855 0.0093 1.0064 ΔREM (1) -0.1671 -1.2380 -0.0056 -0.7026 ΔREM(2) 0.0042 0.0339 - - ΔTOP 0.0589 0.5158 0.0127*** 2.9016 ΔTOP (1) 0.1671* 1.8319 - - ΔTOP(2) 0.1268* 1.8764 - - ΔEDU -0.0138 -0.2030 -0.0011 -0.3871 ΔEDU(1) 0.1741* 2.4465 0.0051 1.6083 ΔEDU(2) 0.1815* 1.8723 0.0043 1.2861 ΔEDU(3) 0.1560° 2.2216 - - ΔLS -0.0393 -0.1526 -0.0437*** -5.8825 ΔLS(1) -0.2980 -1.2242 - - ΔLS(2) -0.3385 -1.3690 - - ΔINF -0.2933* -1.7843 0.0084 1.2116 ΔINF(1) - - - ECM(-1) -0.7637* -2.583 -0.1816*** -5.0250 Model 1 </th <th>Panel B (Short run resul</th> <th>ts)</th> <th>•</th> <th>-</th> <th>-</th>	Panel B (Short run resul	ts)	•	-	-	
AREM (1)	ΔPov1	0.1597	0.4549	-	-	
AREM(2) 0.0042 0.0339	ΔREM	0.3200*	1.8855	0.0093	1.0064	
ΔTOP 0.0589 0.5158 0.0127** 2.9016 ΔTOP (1) 0.1671* 1.8319 - - ΔTOP(2) 0.1268* 1.8764 - - ΔEDU -0.0138 -0.2030 -0.0011 -0.3871 ΔEDU(1) 0.1741* 2.4465 0.0051 1.6083 ΔEDU(2) 0.1815* 1.8723 0.0043 1.2861 ΔEDU(3) 0.1560* 2.2216 - - ΔLS -0.0393 -0.1526 -0.0437**** -5.8825 ΔLS(1) -0.2980 -1.2242 - - ΔINF -0.2933* -1.7843 0.0084 1.2116 ΔINF(1) - - - ECM(-1) -0.7637* -2.583 -0.1816**** -5.0250 Model 1 Model 2 R-squared 0.4451 0.8305 - - F-statistic 2.9045 17.5648 - - Prob (F-statistic) 0.018 0.000 - - DW statistic 1.7928 1.6245 <th< th=""><th>ΔREM (1)</th><th>-0.1671</th><th>-1.2380</th><th>-0.0056</th><th>-0.7026</th></th<>	ΔREM (1)	-0.1671	-1.2380	-0.0056	-0.7026	
ΔΤΟΡ (1) 0.1671* 1.8319 - - ΔΤΟΡ(2) 0.1268* 1.8764 - - ΔΕDU -0.0138 -0.2030 -0.0011 -0.3871 ΔΕDU(1) 0.1741* 2.4465 0.0051 1.6083 ΔΕDU(2) 0.1815* 1.8723 0.0043 1.2861 ΔΕDU(3) 0.1560* 2.2216 - - ΔLS -0.0393 -0.1526 -0.0437*** -5.8825 ΔLS(1) -0.2980 -1.2242 - - ΔINF -0.2933* -1.3690 - - ΔINF -0.2933* -1.7843 0.0084 1.2116 ΔINF(1) - - - ECM(-1) -0.7637* -2.583 -0.1816*** -5.0250 Model 1 Model 2 R-squared 0.4451 0.8305 F-statistic 2.9045 17.5648 Prob (F-statistic) 0.018 0.000 DW statistic 1.7928 1.6245 SE of Regression 0.8330 0.0604 <t< th=""><th>ΔREM(2)</th><th>0.0042</th><th>0.0339</th><th>-</th><th>-</th></t<>	ΔREM(2)	0.0042	0.0339	-	-	
ATOP(2) 0.1268* 1.8764 - - ΔEDU -0.0138 -0.2030 -0.0011 -0.3871 ΔEDU(1) 0.1741* 2.4465 0.0051 1.6083 ΔEDU(2) 0.1815* 1.8723 0.0043 1.2861 ΔEDU(3) 0.1560* 2.2216 - - ΔLS -0.0393 -0.1526 -0.0437**** -5.8825 ΔLS(1) -0.2980 -1.2242 - - ΔLS(2) -0.3385 -1.3690 - - ΔINF -0.2933* -1.7843 0.0084 1.2116 ΔINF(1) - - - ECM(-1) -0.7637* -2.583 -0.1816**** -5.0250 Model 1 Model 2 R-squared 0.7814 0.8305 F-statistic 2.9045 17.5648 Prob (F-statistic) 0.018 0.000 DW statistic 1.7928 1.6245 SE of Regression 0.8330 0.0564 Residual Sum of 8.9237 0.0604	ΔΤΟΡ	0.0589	0.5158	0.0127**	2.9016	
ΔEDU -0.0138 -0.2030 -0.0011 -0.3871 ΔEDU(1) 0.1741* 2.4465 0.0051 1.6083 ΔEDU(2) 0.1815* 1.8723 0.0043 1.2861 ΔEDU(3) 0.1560* 2.2216 ΔLS -0.0393 -0.1526 -0.0437*** -5.8825 ΔLS(1) -0.2980 -1.2242 ΔLS(2) -0.3385 -1.3690 ΔINF -0.2933* -1.7843 0.0084 1.2116 ΔINF(1) ECM(-1) -0.7637* -2.583 -0.1816*** -5.0250 Model 1 Model 2 R-squared 0.4451 0.8305 F-statistic 2.9045 17.5648 Prob (F-statistic) 0.018 0.000 DW statistic 1.7928 1.6245 SE of Regression 0.8330 0.0564 Residual Sum of 8.9237 0.0604	ΔΤΟΡ (1)	0.1671*	1.8319	-	-	
ΔEDU(1) 0.1741* 2.4465 0.0051 1.6083 ΔEDU(2) 0.1815* 1.8723 0.0043 1.2861 ΔEDU(3) 0.1560* 2.2216 - - ΔLS -0.0393 -0.1526 -0.0437**** -5.8825 ΔLS(1) -0.2980 -1.2242 - - ΔLS(2) -0.3385 -1.3690 - - ΔINF -0.2933* -1.7843 0.0084 1.2116 ΔINF(1) - - - ECM(-1) -0.7637* -2.583 -0.1816**** -5.0250 Model 1 Model 2 R-squared 0.4451 0.8305 F-statistic 2.9045 17.5648 Prob (F-statistic) 0.018 0.000 DW statistic 1.7928 1.6245 SE of Regression 0.8330 0.0564 Squares 0.0604	ΔΤΟΡ(2)	0.1268*	1.8764	-	-	
ΔEDU(2) 0.1815* 1.8723 0.0043 1.2861 ΔEDU(3) 0.1560* 2.2216	ΔΕDU	-0.0138	-0.2030	-0.0011	-0.3871	
ΔEDU(3) 0.1560° 2.2216	ΔEDU(1)	0.1741*	2.4465	0.0051	1.6083	
ΔLS -0.0393 -0.1526 -0.0437*** -5.8825 ΔLS(1) -0.2980 -1.2242	ΔEDU(2)	0.1815*	1.8723	0.0043	1.2861	
ΔLS(1) -0.2980 -1.2242 -	ΔEDU(3)	0.1560*	2.2216	-	-	
ΔLS(2) -0.3385 -1.3690	ΔLS	-0.0393	-0.1526	-0.0437***	-5.8825	
∆INF -0.2933* -1.7843 0.0084 1.2116 ∆INF(1) - - - ECM(-1) -0.7637* -2.583 -0.1816**** -5.0250 Model 1 Model 2 R-squared 0.7814 0.9024 R-bar-squared 0.4451 0.8305 F-statistic 2.9045 17.5648 Prob (F-statistic) 0.018 0.000 DW statistic 1.7928 1.6245 SE of Regression 0.8330 0.0564 Residual Sum of Squares 8.9237 0.0604	ΔLS(1)	-0.2980	-1.2242	-	-	
ΔINF(1) - - ECM(-1) -0.7637* -2.583 -0.1816**** -5.0250 Model 1 Model 2 R-squared 0.7814 0.9024 R-bar-squared 0.4451 0.8305 F-statistic 2.9045 17.5648 Prob (F-statistic) 0.018 0.000 DW statistic 1.7928 1.6245 SE of Regression 0.8330 0.0564 Residual Sum of Squares 8.9237 0.0604	ΔLS(2)	-0.3385	-1.3690	-	-	
ECM(-1) -0.7637* -2.583 -0.1816*** -5.0250 Model 1 Model 2 R-squared 0.7814 0.9024 R-bar-squared 0.4451 0.8305 F-statistic 2.9045 17.5648 Prob (F-statistic) 0.018 0.000 DW statistic 1.7928 1.6245 SE of Regression 0.8330 0.0564 Residual Sum of 8.9237 0.0604	ΔΙΝΓ	-0.2933*	-1.7843	0.0084	1.2116	
Model 1 Model 2 R-squared 0.7814 0.9024 R-bar-squared 0.4451 0.8305 F-statistic 2.9045 17.5648 Prob (F-statistic) 0.018 0.000 DW statistic 1.7928 1.6245 SE of Regression 0.8330 0.0564 Residual Sum of Squares 8.9237 0.0604	ΔINF(1)	-	-			
R-squared 0.7814 0.9024 R-bar-squared 0.4451 0.8305 F-statistic 2.9045 17.5648 Prob (F-statistic) 0.018 0.000 DW statistic 1.7928 1.6245 SE of Regression 0.8330 0.0564 Residual Sum of Squares 8.9237 0.0604	ECM(-1)	-0.7637*	-2.583	-0.1816***	-5.0250	
R-bar-squared 0.4451 0.8305 F-statistic 2.9045 17.5648 Prob (F-statistic) 0.018 0.000 DW statistic 1.7928 1.6245 SE of Regression 0.8330 0.0564 Residual Sum of Squares 8.9237 0.0604		Model 1	•	Model 2		
R-bar-squared 0.4451 0.8305 F-statistic 2.9045 17.5648 Prob (F-statistic) 0.018 0.000 DW statistic 1.7928 1.6245 SE of Regression 0.8330 0.0564 Residual Sum of Squares 8.9237 0.0604	R-squared	0.7814		0.9024		
F-statistic 2.9045 17.5648 Prob (F-statistic) 0.018 0.000 DW statistic 1.7928 1.6245 SE of Regression 0.8330 0.0564 Residual Sum of Squares 8.9237 0.0604	R-bar-squared					
DW statistic 1.7928 1.6245 SE of Regression 0.8330 0.0564 Residual Sum of Squares 8.9237 0.0604	F-statistic					
SE of Regression Residual Sum of 8.9237 Squares 0.8330 0.0564 0.0604	Prob (F-statistic)	0.018		0.000		
Residual Sum of 8.9237 0.0604 Squares	DW statistic	1.7928		1.6245		
Squares	SE of Regression	0.8330		0.0564		
	Residual Sum of	8.9237		0.0604		
Akaike Info. Criterion -46.5040 44.4189	Squares					
	Akaike Info. Criterion	-46.5040		44.4189		

Schwartz	Bayesian		
Criterion		-62.5307	32.9712

Note: *, ** and *** denote statistical significance at 10%, 5% and 1% levels respectively

The results presented in Table 3 Panel A and Panel B show that remittance is significant in both the long run and the short run when household consumption expenditure is used as a proxy for poverty. This finding is consistent with other previous studies (see Vacaflore, 2018; Wangle and Devkota, 2018). The results confirm that remittance has a poverty reducing effect, which is also supported in the literature through a number of advantages highlighted by Hagen-Zanker and Himmelstine (2016). This finding implies that South Africa can benefit from policies that support emigration. However, when infant mortality rate was used as a proxy for poverty, remittance was found to be insignificant. Thus, remittances do not lead to a reduction in the infant mortality rate, neither does it increases it, hence it does not have a significant impact on poverty in South Africa. These findings are not unique to South Africa alone, as Azam *et al.* (2016) also found the same results in upper income countries in a study on global evidence.

Other results presented in Table 3 Panel A and Panel B reveal that when household consumption expenditure (Pov1) is used as a proxy, trade openness is positive and significant in the short run, confirming a poverty mitigating effect. This can be explained by the fact that trade increases domestic consumption of a wide range of goods; and competition that results in improvement in the quality of goods produced. Education was found to have a positive effect on household consumption expenditure, leading to poverty reduction in the short run. Education increases the chances of the poor in securing better paying jobs that increase household income and access to

goods and services. In South Africa, given the economy has a challenge of high unemployment, education gives a platform for the poor to be employable (Mohr and Associates, 2015).

Gross domestic product per capita has an insignificant impact on poverty reduction. The possible explanation in South Africa is the high income inequality with a 63% gini index recorded in 2014 (World Bank, 2018). This makes real gross domestic product a less reliable measure of welfare as the real income could be in the hands of a few. Inflation leads to high poverty levels when Pov1 (household consumption expenditure) is used as a proxy in the long run and the short run. This finding is supported in the literature as an increase in prices erodes the purchasing power of income earned by the poor and makes them worse off. The error correction model [ECM (-1)] is 76%. Whenever there is a disequilibrium in the economy, it takes one year and three months to return to equilibrium.

Other results presented in Panel A and Panel B where infant mortality rate (Pov2) is used as a proxy, confirm that trade openness has a positive and significant impact on poverty in the short run. Thus, trade openness worsens poverty thereby making the poor worse off. This can be explained by increased competition for local producers who may be forced out of business due to stiff competition from imports or who may be forced to lay off some workers. Education has a negative effect on poverty in the long run when the infant mortality rate is used as a proxy. Education helps in reducing poverty and the possible explanation is increased awareness by mothers on the importance of health care, such as immunisation and primary health care.

Real gross domestic product per capita has a negative effect on infant mortality rates, hence it leads to poverty reduction. The increase in the general standard of living also implies the ability of the government to provide better social services such as health. Inflation has a negative and significant impact on poverty in both the long run and the short run when Pov2 is used as a proxy. The findings from this study reveal that there is a certain level of inflation that gives firms an incentive to continue producing (Mohr and Associates, 2015). The error correction model [ECM(-1)] for Model 2 is 18% implying that it takes more than 5 years to achieve full adjustment in the event of an economic disequilibrium in South Africa. The model is a perfect fit with an adjusted R² of 83%.

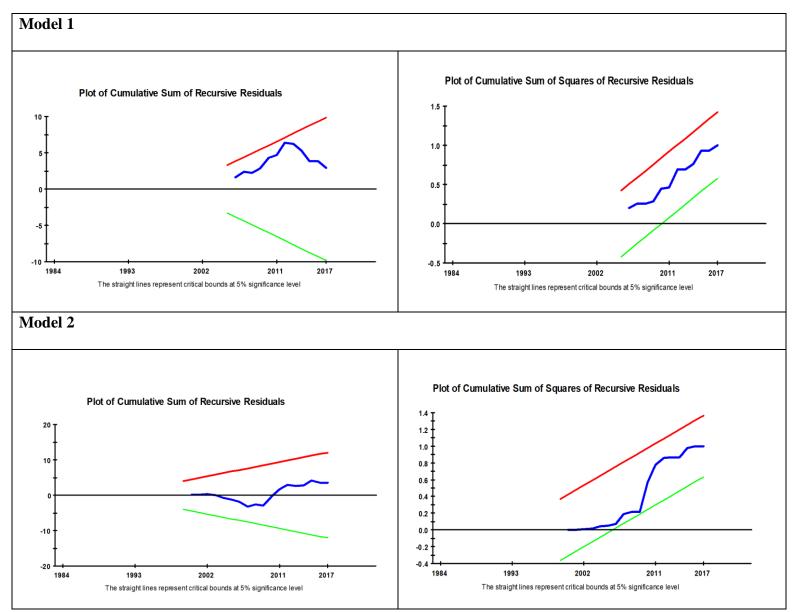
Overall, the results presented in Pane A and Panel B reveal that the impact of remittance on poverty reduction is sensitive to the proxy employed. When household consumption expenditure is used as a proxy for poverty, a significantly positive impact was found, while no impact was found when infant mortality rate is used as a proxy.

Diagnostic tests were performed, for Models 1 and 2, on serial correlation, functional form, normality and heteroscedasticity. Both models passed serial correlation, normality and heteroscedasticity tests but failed the functional form test. Upon inspection of the cumulative sum of recursive residuals (CUSUM) and cumulative sum of squares of recursive residuals (CUSUMSQ), both models were found to be stable at 5% level of significance. The results of the tests are presented in Table 4.

Table 4: Diagnostic Results for Models 1 and 2

LM Test Statistic	Results (Probability)		
	Model 1	Model 2	
Serial Correlation CHSQ(1)	0.6232 (0.430)	0.5812 (0.446)	
Functional Form CHSQ(1)	17.1715 (0.000)	5.1100 (0.024)	
Normality CHSQ(1)	1.5619 (0.458)	0.5263 (0.769)	
Heteroscedasticity CHSQ(1)	0.1053 (0.746)	0.1609 (0.688)	

Figure 1: CUSUM and CUSUMQ Results for Model 1 and 2



Model 1 and Model 2 are stable at 5% level of significance as presented in Figure 1.

5. Conclusion

This study investigated the impact of remittance inflows on poverty reduction in South Africa using data from 1980 to 2017. The study employed the ECM-based ARDL bound testing approach to examine this linkage. To improve robustness of the results, two poverty proxies were employed; namely household consumption expenditure and infant mortality rate. The empirical findings of the study revealed that remittances have a poverty-mitigating effect, both in the long run and in the short run when household consumption expenditure is used as a proxy. However, when the infant mortality rate was used as a proxy, no impact was found between poverty and remittance, irrespective of whether the estimation is in the long or short run. The impact of remittance on poverty is, therefore, sensitive to poverty proxy used. Despite the varying results based on poverty proxy under consideration, on the whole, South Africa can benefit from remittance inflows in reducing poverty. Based on the findings of this study, the study concludes that South Africa could benefit immensely from some forms of remittances in its quest to poverty alleviation.

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