INTERFACES BETWEEN ROAD INFRASTRUCTURE AND POVERTY IN AFRICA: THE CASE OF MALAWI, 1994-2013

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ABSTRACT: Critical assessment on the correlation between public investment on road infrastructure and poverty was carried out, and therefore this research paper provides an in depth analyses of the linkage between road infrastructure and poverty, as well as, other relevant macro economic variables used in the Malawi Growth and Development Strategy (MGDS) as target indicators. Using primary and secondary data from 1994-2013, dynamic time series models were applied in elaborating the various factors with thrust on road infrastructure that may influence poverty in Malawi. Noting poverty reduction as priority of Malawi Government’s development agenda since the early 1990s, MGDS provides the country’s socioeconomic growth and development platforms. According to the latest 2010 Integrated Household Survey (IHS3), the majority of Malawians (50.7 percent) are languishing in abysmal poverty; this level is remotely far from the MDGS target of 27 percent by end 2015. The country has a high inequality index (Gini 0.38) reflecting profound inequalities in access to assets, services and opportunities across the population. The distribution of the benefits of economic growth is also important for the alleviation of poverty. However, the distribution of income and wealth are highly skewed, with a majority of the population living in a state of absolute poverty. Based on NSO surveys (1998-2010), the poorest 20 percent of the population control only around 10 percent of national consumption implying inequality is not decreasing at all for long time. Hosts of factors explaining why poverty level continues to be rampant are: share of agricultural as a percent of GDP (proxy to agricultural production) and export as percent of GDP (proxy to exports). However, this paper findings show that there is significant (p=0.000<0.05) relationship between road network and poverty levels. Estimates from Granger Causality analysis indicate that for one percent increase in road network, a reduction of 7.2 percent in poverty level is perhaps achievable. Average inflation rate over the last 20 years stands at 22.41 percent, and this has an immense impact on poverty level since it dramatically reduces the purchasing power of the majority of the population. For a one percent increase in the inflation rate, there is a consequence of about 3.7 percent increase in the average poverty level. Average Gross Domestic Product (GDP) growth rate is 4.7 percent annually with a minimum of -4.9 percent and a maximum of 10.2 percent in the last 20 years. Poverty level appears to significantly respond to (GDP). There is a 4.27 percent reduction in poverty level if a one percent GDP increment takes place as shown in the dynamic time series analysis. In fact, the declining of agricultural production for export and the growing gap in balance of payment (average Malawi Kwacha -498.92 billions or approximately US$1.1 billion) would immensely influence GDP negatively and therefore poverty becomes abysmal as GDP growth plummets. In a nutshell, the findings confirm that in the long run economic growth is the key to alleviation of extreme poverty since it creates the resources to raise incomes. Given the importance of agriculture in contributing towards GDP in Malawi, the positive impact that this sector has on poverty is evident. For agriculture to meaningfully impact economic growth, road infrastructure plays a great role. Other pro-poor variables such as development...
roads and other investment on infrastructure are vital for economic growth and hence poverty alleviation.

**KEYWORDS**: Infrastructure, Vector Autoregression, Granger Causality, Public Investment, Poverty, Malawi

**INTRODUCTION**

Recently the issue of poverty reduction has been at the centre of global policymaking. Although the link between public investment in rural roads and poverty reduction has been at the centre of development discourse for some time, not much has been investigated on the impact of public investment in rural development in Africa and particularly in Malawi. In fact, the majority of poor people in the world including Malawi live in rural areas, where the level of public infrastructure especially roads is low. The inadequate roads and poor access escalates the high cost of transportation, and limits the use of local markets for sales of their products, the purchase of consumer goods and opportunities for off-farm employment, as well as reducing access to high quality inputs. In addition, poor road access has immense constraint for rural poor in terms of access to other social infrastructures such as education and health facilities.

Most developing countries in the world have depended on agriculture as the main source of income and poverty reduction. In this context, Malawi being a rural country, infrastructure development in the rural areas is important. In this regard, the correlation between public investment and rural infrastructure in Malawi is relevant. For instance, farmers need good rural roads to access farm inputs and commodity markets, electricity, schools, health and telecommunication facilities. Investing in rural infrastructure can enhance rural employment and income and thereby reducing poverty. Accordingly, critical assessment on the correlation between public investment and what type of rural infrastructure must be prioritized for inducing rural productivity and poverty reduction is essential. In depth analyses have to be carried out with the aim of developing the linkage between rural infrastructure and poverty. A good understanding and analysis on public investment and rural infrastructure in Malawi is crucial in reducing poverty. This information is vital in providing policy direction on government expenditure and resource allocation and value for money.

**Background and Reviews**

Of late, there is a very strong impetus given to infrastructure investments in Sub-Saharan Africa.

For the period 2008-2010, the Chinese EXIM bank committed around $20billion in infrastructure for financing railway rehabilitation in Nigeria, Angola, building dams in Ethiopia and other places. The African Development Bank will spend over $5billion in the next three years, of which over 60 percent in infrastructure (mainly roads, energy and water). The World Bank spends more than $5 billion a year in Sub-Saharan Africa (with more than $1.5billion in roads), and as of 2006 it has spent US$16.5 million on road infrastructure development in Malawi. Moreover, aid to Africa is planned to double in the near future (G8 commitment in 2008); infrastructure investments will be of crucial importance (World Bank, 2006; 2013).
Characteristics of Road in Malawi

According to National Roads Authority (2011) report, roads continue to be the country’s most dominant mode of transport and handles over 99 percent of domestic passenger traffic, more than 70% of internal freight traffic and 90 percent of international freight and passenger traffic. National Roads Authority (NRA, 2011) reported that the road network is composed of 15,451 km of which about 26 percent are paved. The rest of the road network (74 percent) is of earth/gravel surface. Studies carried out in 2005 have identified about 10,000 km of undesignated road network serving rural communities. The road transport is also important for international trade as it handles more than 90% of freight and passenger traffic.

Based on NRA data, the paved road density is about 268 meters per 1000 population in Malawi compared to 318 meters per 1000 population for Sub-Saharan African countries. This implies the dysfunctional state of roads all over Africa, and it is even worst and a huge challenge for Malawi as a land-locked country. It is a complete deficiency in coverage.

According to UN-Habitat (2011) study, Africa’s 15 land-locked countries, which accounts for 40 percent of the region’s overall population, face special challenges. Countries that are land-locked are at the disadvantage because, on average, they take four more days to land distribution of exports and nine days to imports compared with seaport countries with equivalent distances. The geographical results in higher transport costs which hamper intra and inter-regional trade, as well as reduces openness to trade emerges as the main factor, other things being equal, that contribute to slow growth than the rest of the countries.

According to data obtained from Malawi Roads Authority, the condition of the road network is such that 60 percent (2426 km), 33 percent (1361 km), and 7 percent (286 km) are in good, fair and poor conditions, respectively. On the other hand, 44% (5000 km), 23 percent (2654 km) and 33 percent (3724 km) of the unpaved roads are in good, fair and poor condition, respectively. Thus, the total road network condition is 48 percent good, 26 percent fair and 26 percent poor. With no statistically difference between good and a combination of fair and poor roads, Malawi still does not have good road conditions; where by over 60 percent of the roads are in unacceptable conditions. These conditions of the road also likely contribute to declining of economic growth and elevating poverty, as a whole.

Extent of Poverty

According to the latest 2010 Integrated Household Survey (IHS3), the majority of Malawians are poor (50.7 percent). The rural areas have proportionately more people living in abysmal poverty (55.9 percent) than urban areas (25.4 percent) (National Statistical Office, 2005-2010). The poverty is manifested in low incomes, high illiteracy rates, high infant (under-5) and maternal mortality rates, high levels of child malnutrition, low life expectancy and a high HIV and AIDS prevalence rate. The country has a high inequality index (Gini 0.38) reflecting profound inequalities in access to assets, services and opportunities across the population (World Bank, 2006). In fact, according to the 1998, 2005 and 2010 Integrated Household Surveys, the poorest 20 percent of the population control only around 10 percent of national consumption implying inequality is not decreasing at all.
Ostensibly, poverty has continued to manifest itself in different ways and more pronounced in the rural areas of Malawi where 85 percent of the population reside. As shown in Figure 3.4, poverty in Malawi has become almost endemic and has been about 52 percent in the last fifteen years. Over half of the entire population is in abysmal poverty. Eradicating extreme poverty is the first of the eight MDGs goals, which targets to halve the level of poverty to 27 percent by 2015. This is remotely impossible to achieve as poverty stands at about 52 percent at the moment; implying that Malawians have continued to remain in poverty and prone to vulnerability due to exogenous shocks including natural disasters, external economic forces, internal policy failures, mismanagement of government funds and virtually with no meaningful development.

**METHODOLOGY**

According to the World Bank (2006), insufficient and inadequate economic infrastructure is among the most pressing obstacles to achieving pro-poor growth. Among infrastructure, roads are considered of first interest to reduce poverty due to the widely accepted consensus that transport infrastructure has a significant, positive and substantial impact on economic growth and poverty. Within this context, the rural access index (RAI) (proportion of rural people who live within two kilometers (typically equivalent to a 20-minute walk) of an all-season road has been set as the most important indicator for the World Bank in its current and future investments in road transport. Based on ADB (2013) and macroeconomic conceptual framework, Figure 3.1 illustrates the links from infrastructure investments (areas of intervention) through these determinants (areas of influence) to the poor’s wages and employment (direct channel), on the other hand, rural economic growth (indirect channel) that influences the supply and prices of basic goods. The final links are to real income/consumption of the poor and consequently, poverty reduction (area of concern).

The various links can be illustrated with an example from the tentative conceptual framework. For example, a road investment could result in an increase in agricultural productivity, nonfarm employment and productivity, directly raising the wages and employment of the poor and, hence, their economic welfare. This is the (direct) income

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1 MDG’s first target is to halve, between 1990 and 2015 the proportion of people whose income is less than one dollar per day
distribution effect. In addition, higher productivity and expanded employment lead to higher economic growth, affecting the supply and prices of goods and, thus, the poor’s well-being. This is the (indirect) growth effect. Similar links can arise from irrigation, electricity and other infrastructure investments.

Figure 3.1: Conceptual framework

Source: Adopted from Ali and Pernia (2003)
The framework herein was used in designing the empirical work that aimed to trace more carefully and systematically the links of interest. Various econometric studies available generally do trace some links described above. They provide useful assessments of the more important links, indicating their quantitative and statistical significance. These measures are typically represented as elasticities denoting the responsiveness of a variable to a determinant. The few available studies covered in the review offer examples from Asian and few African countries. While differences in econometric model specifications, data, and definitions call for caution in the interpretation of results across countries, they do lend helpful insights into the connection between physical infrastructure and poverty reduction. This road map will be utilized to formulate relevant econometric models to understand the influences of various variables in reducing poverty.

Data requirements and Sources

Both primary and secondary sources of data were explored. Primary data were gathered through formal interviews with different stakeholders including officials from government and World Bank representative/s working on the ISP, community leaders and ordinary members of communities in rural areas. Structured questionnaires and checklists were be used for the data collection. Different Focus Group Discussion (FGDs), government ministries, universities, research centers, UN organizations and NGOs, were consulted and information were used where deemed necessary.

Secondary sources of information involved a review of literature relevant to the subject matter including project documents, public expenditure statements and government annual report. Sources of secondary data included the following; Ministry of Transport and Public Works, Ministry of Agriculture, Ministry of Planning and Economic Development, Ministry of Finance, Reserve Bank of Malawi, National Roads Authority (NRA), National Statistical Office (NSO), Malawi Social Action Fund (MASAF) and the World Bank.

Analytical Techniques

In order to link the roadmap or conceptual framework illustrated previously, various analytical techniques were reviewed from studies in developing countries, and in which the best analytical technique or a combination of methods was adopted for Malawi’s scenarios (based on available data). Some of the methods used to analyze impact of rural road infrastructure on poverty are summarized in Table 3.1.

Table 3.1: Analytical techniques and References

| Analytical methods | References |
|--------------------|------------|
| Generalized Method of Moments as employed | Seetanah, 2012; Oraboune, 2008 |
| Three log forms model with Fixed/random effects techniques | Datt and Ravallion, 2002; Ravallion and Datt, 1996 and Ghura, Leite, and Tsangarides, 2002 |
| Simultaneous equations | Fan, Hazell and Thorat (1999) |
| Neo-classical production functions such as Cobb-Douglas or log linear production function | Fan, Rao and Zhang (2004); Munnel, 1992; Gramlich, 1994; Sturm, Kuper & De Haan, 1998; Romp & DeHaan, 2005 |
Simultaneous Equations  
(A) The Human Capital Channel, (B)  The Market Access Channel, and (C) The Labor Activities Channel  
Gachassin, Najman and Raballand, 2010; Mustajab, 2009  
Panel data and Dynamic Panel Analysis  
Seenanah, Ramessur and Rojid (2009); Woodridge, 2002  
Vector Autoregression (VAR) and Vector Error Correction Models (VECM)  
Perron (1990); Toda & Phillips (1993; 1994), Dufour & Renault, 1998; Ramirez, 2004; Lütkepohl, 2005  
Structural Vector Autoregressive (SVAR)  
Sims (1980a, 1980b); Amisano & Giannini, 1997; Arellano and Bover, 1995; Saikkonen & Lutkephol, 2002; Sarte, 1997; Ogun’s (2010)  

In this study the impact of road infrastructure on Malawi’s poverty is assessed from a macroeconomic perspective. The lack of clear theoretical guidance on the choice of regressors, for the poverty equation, leads to a wide set of possible specifications and model uncertainty which in turn often results in contradictory conclusions. A challenge therefore is to motivate which macroeconomic variables to include in the poverty equation. The fact that a certain variable is available in the data set seldom provides sufficient justification for including it in the model. Thus, after reviewing various studies done in developing countries (Table 3.1), a comprehensive list of variables were put together in conjunction to MDGs target variables and indicators. The variables that are used in the multivariate time series models are:

**povertyrate** = the headcount poverty index  
**gdprate** = gross domestic product (%)  
**inflationrate** = inflation rate or cpi  
**xportpergdp** = exports as a % of GDP  
**foreigninflow** = foreign fund in flow Malawi Kwacha (MK’00 billions)  
**totalgvtexpendi** = government capital expenditure  
**totalgovrev** = government revenue in MK(billions)  
**shareagrgdp** = share of agriculture in GDP  
**lifeexpectancy** = life expectancy rate at birth or health(years)  
**literacrate** = literacy rate(%)  
**unemloyrate** = unemployment rate(%)  
**roadlength** = length of road paved/tarred(km)  
**fmd2gdp** = financial development (Money supply or M2 as a % of GDP)
With the availability of Malawi data from 1994 to 2013 on these selected variables, Vector Autoregression (VAR) and Vector Error Correction (VEC) time series models were used as econometric specifications to link poverty with different variables identified and listed previously.

**VAR and VEC Models**

The vector autoregressive (VAR) model is a general framework used to describe the dynamic interrelationship among stationary variables. So, the first step in multivariate time-series analysis should be to determine whether the levels of the data are stationary. If not, take the first difference of the series and try again. Usually, if the levels (or log levels) of the time series are not stationary, the first difference will be (Adikins and Hill, 2011).

If the time series are not stationary then the VAR framework needs to be modified to allow consistent estimation of the relationships among the series. The vector error correction (VEC or VECM) model is just a special case of the VAR for variables that are stationary in their differences. The VEC can also take into account any cointegrating relationships among the variables. Generalizing the discussion about dynamic relationships between two interrelated variables, say $y_t$ and $x_t$, the following systems of equations are used.

$$y_t = \beta_{20} + \beta_{11}y_{t-1} + \beta_{12}x_{t-1} + \nu_{yt}$$

$$x_t = \beta_{30} + \beta_{21}y_{t-1} + \beta_{22}x_{t-1} + \nu_{xt}$$

These equations describe a system in which each variable is a function of its own lag, and the lag of the other variable in the system. In this case the system contains two variables $y$ and $x$. Together the equations constitute a system known as a vector autoregression (VAR). However, if $y$ and $x$ are stationary, the system can be estimated using least squares applied to each equation. If $y$ and $x$ are not stationary in their levels, but stationary in their differences (i.e., I(1)), the differences estimate:

$$\Delta y_t = \beta_{20} + \beta_{11}\Delta y_{t-1} + \beta_{12}\Delta x_{t-1} + \nu_{\Delta yt}$$

$$\Delta x_t = \beta_{30} + \beta_{21}\Delta y_{t-1} + \beta_{22}\Delta x_{t-1} + \nu_{\Delta xt}$$

Using least squares, if $y$ and $x$ are I(1) and cointegrated, then the system of equations can be modified to allow for the cointegrated relationship between I(1) variables. Introducing the cointegrating relationship leads to a model known as the vector error correction (VEC or VECM) model (Adikins and Hill, 2011). These are the dynamic multivariate models applied in this analysis with all assorted and necessary unit root tests including Dickey-Fuller tests and remedial measures are taken to make sure the estimates are consistent as expected and required.

**RESULTS AND DISCUSSIONS**

The level of poverty index shows that the percentage of the population with consumption of food and non-food essentials lowers than the poverty line of US$1/day (currently MK420/day). As shown in Table 4.1, the average for the last two decades stands at 52.8 percent with no significant reduction at all comparing to current level of 51 percent. With the
current population growth rate of 2.8 percent, Malawi would need GDP growth of 8.4 percent to have meaningful economic development and prevent an increase in poverty level.

Table 4.1 displays statistics for some selected variables, which were used in the multivariate time series modeling, for the last two decades.

1. In fact, eradicating extreme poverty is the first of the eight MDGs goals, which targets to halve the level of poverty to 27 percent by 2015. This is remotely impossible with the current pace of economic growth and development when already in the year 2015.

Table 4.1: Descriptive statistics for selected variables (1994-2013)

| Variable          | Obs  | Mean   | Std. Dev. | Min  | Max  |
|-------------------|------|--------|-----------|------|------|
| povertyrate       | 20   | 52.865 | 1.34801   | 50.6 | 55   |
| gdpkmkblili-s     | 20   | 407.715| 358.145   | 103.25| 1055 |
| gdprate           | 20   | 4.665  | 3.4087    | -4.9 | 10.2 |
| xportpergd          | 20  | 125.5  | 0.960523  | 0.01 | 29   |
| inflationrate-f     | 20  | 22.405 | 18.33089  | 7.4  | 83.3 |
| foreigninf-w      | 20   | 1.6985 | 1.156193  | .22  | 4.1  |
| totalgovrev       | 17   | 6.04e+10| 6.50e+10 | 2.02e+09| 1.78e+11|
| totalgvtx-k       | 12   | 13350.78 | 15909.8 | 162.549| 63038.5|
| fiscaldefi-k      | 20   | -3432.07 | 5499.323 | -22700.5 | 1787.7 |
| bopmillion-k     | 20   | -498.92 | 7241.064 | -12483 | 21710.2 |
| shareagrdp        | 20   | 37.046 | 8.915974 | 25.07651 | 60.2567 |
| lifeexpect-y      | 20   | 44.66  | 5.386181  | 38   | 55.1 |
| literate          | 20   | 61.4   | 7.676553  | 50   | 79.6 |
| money supply-n    | 18   | 117.775 | 115.8445 | 24.89 | 397.3 |
| fdom2gdp          | 18   | 26.64167 | 11.14851 | 14.41 | 62.53 |
| unemployment-e    | 20   | 21.82  | 1.255556  | 19.5 | 23.9 |

It should be noted that poverty is essentially a dynamic phenomenon such that those who were in the poverty trap last year are more likely to be still in it this year. Thus a dynamic time series approach is applied so that it minimizes such endogeneity problems, as well as control for lagged and feedback effects. Results from VAR and VEC time series models are reported in Table 4.2, and determinants of poverty in Malawi as of 2013. Note that not surprisingly, the positive and significant coefficient of povertyrate-l, which is 0.5789, from Table 4.2 implies that poverty is a vicious cycle, since the responsiveness of current period poverty measures with respect to their respective in the previous year is significant. This confirms the existence of dynamism and endogeneity in the modeling framework.

2. These dynamic model estimates suggest that the most important factors affecting the level of household poverty are GDP level, inflation, export, foreign inflow (investment & grants), agricultural production (proxy share of agriculture of GDP), life expectancy (proxy health), infrastructure development (proxy roads), literacy rate and money supply.

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2 MDG’s first target is to halve, between 1990 and 2015 the proportion of people whose income is less than one dollar per day

2053-2199 (Print), 2053-2202 (Online)
Table 4.2: Coefficient estimates of VECM time series model

| Variables               | Coef.  | Std. err. | P>|t| |
|-------------------------|--------|-----------|-----|
| povertyrate_{t-1}       | 0.5789 | 0.1984    | 0.0100 |
| gdprate                 | -0.04268 | 0.02001 | 0.0048 |
| inflationrate           | 0.03657 | 0.00345 | 0.0030 |
| xportpergdp             | -0.0180 | 0.00314 | 0.0000 |
| foreigninflow           | -0.01810 | 0.00220 | 0.0042 |
| totalgvtxpendi          | -0.0351 | 0.00230 | 0.0000 |
| totalgovrev             | -0.01364 | 0.00264 | 0.0010 |
| shareagrgdp             | -0.04865 | 0.01994 | 0.0033 |
| lifeexpectancy          | -0.02421 | 0.00804 | 0.0030 |
| unemploymentrate        | 0.03235 | 0.00278 | 0.0025 |
| roadlength              | -0.07203 | 0.00120 | 0.0000 |
| fmd2gdp                 | -0.02871 | 0.00273 | 0.0013 |
| literacrate             | -0.01758 | 0.00451 | 0.0000 |

Note that the estimated equations has passed diagnostic tests such the Dickey-Fuller test and other related tests to have consistent estimates with associated p-values (last column) as shown in table. All the variables are in ln form.

These dynamic model analyses quantify the impact of individual factors contributing to poverty level, controlling for the impact of all the other factors, and provides indication of relative importance of each factor; and thus the analysis provides an estimate of the impact of these selected variables on the probability of being poor in Malawi.

3. The total road network (paved and unpaved) is composed of 15, 451 kilometers of which about 26% (4017 kilometers) are paved, which is a density of 0.27kms per 1000 residents, which is by far less than the even dismal Africa’s density of 0.32kms per 1000 persons. On average, it is about 20 kilometers for the rural household to reach paved road and reaching even 40 kilometers in some regions of the country.

![Figure 4.1: Determinants of Poverty in Malawi, 1994-2013](image-url)
As shown in Table 4.2 and Figure 4.1, results from VECM model show that there is significant (p=0.000<0.05) relationship between road network and poverty levels. Estimates indicate that for a one percent increase in road network, this leads to a reduction of 7.2 percent in poverty level.

Furthermore, access to roads and markets is a critical determinant of poverty. Poverty increases with distance from paved roads as it hinders accessibility to market centers and therefore difficult to sale their agricultural products and lowers their income. Also economic opportunities outside agriculture are limited since accessibility to trading or marketing center or district administrative center is inaccessible. In support of this conclusion, World Bank and NSO (2005) reported that limited access to financial services and transport infrastructure reveals a dearth of opportunities for the poor; substantial portions of the population remain isolated from the rest of the country both physically and in terms of economic activity.

3. Hosts of factors explaining why poverty level continues to be rampant are: share of agricultural as a percent of GDP (proxy to agricultural production) and export as percent of GDP (proxy to exports). Analysis of the interface between poverty and export as a percent of GDP indicate that a one percent increment in export reduces poverty level by 1.8 percent; in fact, the gap between import and export is exponential growing for the last 20 years. Malawi has become a consuming country rather than producing country. With balance of payment gaps (MK498.92 billion) on average, the current export level, which is only 12.5% percent of GDP, offsets the imports very minimally and has also minimal (1.8%) impact on the poverty level over the years.

Results in Table 4.3 indicate that for one percent change in road network, there is perhaps a significant 1.56 percent change in export as percent of GDP, which also implies that an increase in efficient transport of produces to international market.

Table 4.3: VECM estimate export per GDP and road network

| Source   | SS         | df   | MS         | Number of obs = 20 |
|----------|------------|------|------------|------------------|
| Model    | 0.298598591| 1    | 0.298598591| F( 1, 19) = 29.59 |
| Residual | 0.191701402| 19   | 0.010089547| Prob > F = 0.0000|
| Total    | 0.490299993| 20   | 0.024515   | R-squared = 0.6090 |
|          |            |      |            | Adj R-squared = 0.5884 |
|          |            |      |            | Root MSE = 0.10045 |

| xportpergdp | Coef. | Std. Err. | t     | P>|t| | [95% Conf. Interval] |
|-------------|-------|-----------|-------|------|----------------------|
| loglengthp-d| 0.0156072 | 0.0028689 | 5.44  | 0.000 | 0.0096025 - 0.0216118 |

4. Average inflation rate over the last 20 years is 22.41 percent. This has an immense impact on poverty level since it dramatically reduces the purchasing power of the majority of the population. For a one percent increase in the inflation rate, there is a consequence of about 3.7% increase in the average poverty level.
Table 4.4: Estimates for road network and inflation rate

|                      | Coef.   | Std. Err. | t      | P>|t|    | [95% Conf. Interval] |
|----------------------|---------|-----------|--------|--------|----------------------|
| D. inflation ~ e     | -.5217325 | .2109174 | -2.47  | 0.024  | -.9667292 to -.0767358 |
| L1.                  | -.1287842 | 3.967054 | -0.03  | 0.974  | -8.498536 to 8.240968 |
| _cons                |         |           |        |        |                      |

Table 4.4 also displays that there is significant negative relationship between road network and inflation rate. For one percent change in road network, there is a possibility of reducing inflation by 52.1 percent; this is possible that if there are good road network, the transport cost could be significantly reduced from 45 percent transport costs of commodities currently charged in Malawi; thus good road conditions and effectiveness of road would have significantly impact on inflation, in general.

5. Over the last 20 years, on average the share of agricultural to GDP is 37 percent with a minimum of 25 percent and a maximum of 60.1 percent. The annual agriculture growth has fallen to 6.7 percent in 2011 from 12.3 percent in 2007 (MEPD, 2013). These results suggest a high fluctuation of agricultural production, especially tobacco, maize, rice, tea and sugarcane. As seen previously, the declining of agricultural production would have immediate and higher negative impact on export, and consequently the share of agriculture to GDP dropped from 60.2 percent to an average of 37 percent in recent years. This domino effect has dire consequences on government revenue and hence on expenditures for development including road infrastructure.

Using Granger causality analysis (Granger 1969), Table 4.5 provides results for some selected variables associated with road network.

Table 4.5: Granger Causality analysis of some selected variables

| Hypothesis (H₀)                  | Coefficient Estimates | p-value < 0.10 |
|----------------------------------|-----------------------|----------------|
| Road network → Poverty            | -0.07203              | 0.0013         |
| Road network → Export as percent GDP | 0.0156              | 0.0000         |
| Road network → unemployment rate  | -0.3292              | 0.0054         |
| Road network → Inflation rate     | -0.5217              | 0.024          |
| Road network → Agri-Share as percent GDP | 0.04865           | 0.033          |
| Road network → Gross Domestic Product | 0.1900               | 0.0000         |
| Gross Domestic Product → Road network | 0.0519             | 0.0000         |

6. Again there is significant and positive relationship between road infrastructure and agricultural GDP (proxy to agricultural production); implying that for a unit change in road infrastructure, agricultural production responds by 4.87 percent, on average (Tables 4.2 & 4.5). This high response corresponds to various activities surrounding agriculture production such as employment, input and output marketing, demand and supply and so forth.
7. Average Gross Domestic Product growth rate is 4.7 percent annually with a minimum of 4.9 percent and a maximum of 10.2 percent in the last 20 years. Poverty level appears to significantly respond to GDP. This average level of GDP rate itself shows lack of economic growth and development as it is far below the threshold of GDP level (8.4 percent) required to have meaningful investment in various sectors in Malawi.

8. Analysis of dynamic models shows that there is a possibility of 4.27 percent poverty reduction if a one percent GDP increment takes place. In fact, the declining of agricultural production for export and the growing gap in balance of payment (average –MK498.92 billions) would immensely influence GDP negatively and therefore poverty becomes abysmal as GDP growth plummets. As reported in Table 4.5, road network has also significant impact on GDP, for one percent change in the road network; there is a possibility of changing GDP by 19% from its average level. Thus investing in road infrastructure enhances GDP through various channels, transport cost, agricultural production, export, etc.

9. From both dynamic modeling and descriptive statistics, it is obvious that there is correlation between poverty and unemployment rate. Being aware of seasonality of employment rate and the majority living in rural areas, unemployment is still high using proxy variables of labor participation rate in farming and manufacturing industries. It is estimated that on average the unemployment rate is about 22 percent in the last two decades, and this has tremendous incremental impact on poverty. This high rate of unemployment is not surprising when the share of women in wage employment in non-agricultural sector was less than 30 percent by 2011 (NSO, 2012).

10. The dynamic analysis shows that for a one percent increase in unemployment rate there leads to an increase of 3.25 percent on the average poverty level (Table 4.3). According to World Bank (2010), those most vulnerable to poverty usually have no investment income and receive little or no income in the form of interpersonal transfers from family to friends. Thus, unemployment fuels poverty.

As shown in Tables 4.5 and 4.6, if there is one percent in road infrastructure network, the unemployment rate declines by 32.9 percent, on average. As such employment is the chief income source for these people. Therefore through road infrastructural development, it is possible to increase employment levels, thereby increasing income and the welfare of households to significantly combat poverty, in general.

Table 4.6: Estimates road network and unemployment rate

| D. unemployme-e | Coef. | Std. Err. | t     | P>|t|  | [95% Conf. Interval] |
|-----------------|-------|-----------|-------|------|---------------------|
| ehatroadun-p    | -.3292803 | .1588375 | -2.07 | 0.054 | -.6643981 ,0058376  |
| L1. __cons      | .1857367 | .1855777 | 1.00  | 0.331 | -.205798 ,5772714  |

11. The elasticity of national poverty with respect to government revenue is significant and negative (-0.01364); implying that when government revenue increases by 1% the national poverty is reduced by about 1.4%. This gives an indication that government revenue perhaps could be used as a tool to redistribute income to the poor either in the
form of direct or indirect targeting such as infrastructure development, agricultural programs and so forth.

In a nutshell, the findings confirm that in the long run economic growth is the key to alleviation of extreme poverty since it creates the resources to raise incomes. Given the importance of agriculture in contributing towards GDP in Malawi, the impact that this sector has on poverty is evident. Other pro-poor variables such as development roads and other investment on infrastructure are vital for economic growth and hence poverty alleviation.

CONCLUSIONS

The identification and analysis of infrastructure investment and its effect on economic growth, development and poverty reduction is of considerable interest from a policy perspective. To address this issue, this research focused on Malawi as a developing country, and assessed the impact of road infrastructure on poverty level, investigated the link between poverty and road infrastructure, as well as other variables related to MGDs in Malawi.

Rigorous time series data analysis was done using dynamic models, VAR and VECM, on several selected variables on data from 1994 to 2013. The findings from both descriptive and model analyses have confirmed the theoretical link between road infrastructure and poverty reduction in Malawi. As expected, the dynamic analysis has also validated that other variables such as literacy rate, export, government expenditure, government revenue, unemployment rate, inflation and gross domestic product do have immense impact on poverty.

These research findings provide important insights into the determinants and positive effect of public infrastructure policies such that they can help to allocate scarce resources in the fight against abysmal poverty. These macro analyses provide insights in creation of better infrastructure development in rural areas as equivalent to the whole country. Noting that road sector is very important for the land-locked Malawi, not only creates access to economic activities to the majority of the population, but also cuts the high cost of transport in reducing inflation and increase of purchasing power the entire population.

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