OIL PRICES AND ECONOMIC GROWTH IN KENYA: A TRIVARIATE SIMULATION

Nicholas M. Odhiambo
Sheilla Nyasha

Working Paper 15/2018
June 2018

Nicholas M. Odhiambo
Department of Economics
University of South Africa
P. O. Box 392, UNISA
0003, Pretoria
South Africa
Email: odhianm@unisa.ac.za / nmbaya99@yahoo.com

Sheilla Nyasha
Department of Economics
University of South Africa
P. O. Box 392, UNISA
0003, Pretoria
South Africa
Email: sheillanyasha@gmail.com

UNISA Economic Research Working Papers constitute work in progress. They are papers that are under submission or are forthcoming elsewhere. They have not been peer-reviewed; neither have they been subjected to a scientific evaluation by an editorial team. The views expressed in this paper, as well as any errors, omissions or inaccurate information, are entirely those of the author(s). Comments or questions about this paper should be sent directly to the corresponding author.

©2018 by Nicholas M. Odhiambo and Sheilla Nyasha
Abstract
In this study, the dynamic causal relationship between oil price and economic growth in Kenya has been explored during the period from 1980 to 2015. A trivariate Granger-causality framework that incorporates oil consumption as an intermittent variable – in an effort to address the omission-of-variable bias – has been employed. Using the newly developed ARDL bounds testing approach to co-integration and the Error-Correction Model-based Granger-causality framework, the results of the study reveal that there is distinct unidirectional Granger-causality flowing from economic growth to oil price in the study country. These results were found to apply both in the short run and in the long run. Thus, it can be concluded that in Kenya, it is the real sector that pushes oil prices up. Further, it is possible to predict oil price changes in Kenya – given the changes in economic growth.

Keywords: Kenya, Oil Prices, Energy Consumption, Economic Growth, Granger-Causality

JEL Classification Code: O40, Q43

1. Introduction

Although the relationship between energy and economic growth has attracted a proliferation of empirical studies in recent years; not much attention has been given to the relationship between oil prices, in particular, and economic growth. On the theoretical front, an increase in oil prices is expected to have two effects – the demand-side effect and the supply-side effect. According to the demand-side effect, an increase
in oil prices leads to an increase in transportation costs, which then translates into higher prices for consumer goods. This, in turn, lowers the consumption demand, which eventually leads to a contraction in real output. However, according to supply-side effect, a rise in oil prices leads to higher production costs, which force producers to cut back their output – thereby lowering the country’s aggregate output.

On the empirical front, a handful of studies have been conducted on the causal relationship between oil prices and economic growth, but still, with conflicting results. These studies can be conveniently grouped into four categories. The first group consists of studies that found Granger-causality to flow from oil prices to economic growth while the second group found the flow to be from economic growth to oil prices. The third group is of studies that found the feedback hypothesis to be predominant, while the fourth group constitutes studies that are consistent with the neutrality hypothesis.

Although a number of studies have been conducted on the causal relationship between oil prices and economic growth in developing countries, the majority of these studies have concentrated mainly on Asia and Latin America. Very little attention has been given to sub-Saharan African countries, which, in most cases, are hard hit by the oil price shocks.

Moreover, some previous studies on this subject have been found to suffer from two major weaknesses. Firstly, some of these studies have mainly used a bivariate causality test to examine this linkage; hence, they are prone to suffer from the omission-of-variable bias (see also Odhiambo, 2008; 2009b). Secondly, some of these studies have mainly used the cross-sectional data to examine the causal relationship between oil prices and economic growth. This, unfortunately, does not address the country-specific
effects. In order to fill this lacuna, the current study attempts to examine the dynamic causal relationship between oil prices and economic growth in Kenya using the newly developed ARDL-bounds-testing approach.

By incorporating oil consumption in the bivariate model between oil prices and economic growth, a simple trivariate-causality model between oil prices, oil consumption and economic growth is examined. Contrary to the results of some previous studies, our results show that there is a distinct unidirectional causal flow from oil price to economic growth in Kenya. The rest of the paper is organised as follows: Section two covers the dynamics of oil prices and economic growth in Kenya; while section 3 reviews the literature. Section 4 presents the methodology used in the study, and section 5 presents and analyses the results. Section 6 concludes the study.

2. Oil price increases and economic growth in Kenya

According to Omagwa et al. (2017), the pricing of oil products in Kenya is often controlled by the relevant government department, making it a complex process. In 2016, according to the Kenya National Bureau of Statistics (KBNS, 2018), the average crude oil price increased 20.3% compared to February prices of the same year. In the same period, the Brent oil price increased by $5.9 per barrel, reaching $39.07 per barrel. Historically, crude oil prices reached a maximum of $132.83 per barrel in July 2008, while record low prices of $1.17 per barrel were recorded in February 1946 (KBNS, 2018).

A number of oil shocks have been experienced during the last +/- 50 years. Most of these oil shocks have been somewhat linked to the disruption of oil production in the
Middle East due to conflicts (Hamilton, 2012). According to Hamilton (2012), these conflicts include:

i) The closure of the Suez Canal following the conflict between Egypt, Israel, Britain, and France in October 1956
ii) The oil embargo implemented by the Arab members of OPEC following the Arab-Israeli War in October 1973
iii) The Iranian revolution beginning in November 1978
iv) The first Persian Gulf war beginning in August 1990.

Besides the oil shock, there are other events that were linked to the disruption of oil supply; and these were:

i) the combined effects of the second Persian Gulf War and strikes in Venezuela beginning in December 2002
ii) the Libyan revolution in February 2011.

Furthermore, oil price increases were part of the world energy landscape. The notable historical factors that have led to the oil price increases include:

i) the economic recovery from the East Asian Crisis in 1997
ii) the dislocations associated with post-World War II growth in 1947
iii) the Korean conflict in 1952-53.

Table 1 presents a summary of events that significantly affected the post-independence Kenya.

Table 1: Significant Post-Independence Events in Kenya’s Oil Sector
<table>
<thead>
<tr>
<th>Key factors</th>
<th>Business cycle peak</th>
<th>Gasoline Shortages</th>
<th>Crude oil increase</th>
<th>Crude oil or gasoline price controls</th>
</tr>
</thead>
<tbody>
<tr>
<td>Strong demands, supply constraints</td>
<td>Nov 48</td>
<td>Nov 47 – Dec 47</td>
<td>Nov 47 – Jan 48 (37%)</td>
<td>no (threatened)</td>
</tr>
<tr>
<td>Strike, control lifted</td>
<td>Jul 53</td>
<td>May 52</td>
<td>June 53 (10%)</td>
<td>yes</td>
</tr>
<tr>
<td>Suez Crisis</td>
<td>Aug 57</td>
<td>Nov 56 – Dec 56 (Europe)</td>
<td>Jan 57-Feb 57 (9%)</td>
<td>yes (Europe)</td>
</tr>
<tr>
<td>---</td>
<td>Apr 60</td>
<td>none</td>
<td>none</td>
<td>no</td>
</tr>
<tr>
<td>Strike, strong demand, supply constraints</td>
<td>Dec 69</td>
<td>None</td>
<td>Feb 69 (7%)</td>
<td>no</td>
</tr>
<tr>
<td>Strong demand, supply constraints</td>
<td>Nov 73</td>
<td>June 73 Dec 73 – Mar 74</td>
<td>Apr 73-Sep 73 (16%</td>
<td>yes</td>
</tr>
<tr>
<td>Strong demand, supply constraints, OAPEC embargo</td>
<td></td>
<td></td>
<td>Nov 73 – Feb 74 (51%)</td>
<td></td>
</tr>
<tr>
<td>Iranian revolution</td>
<td>Jan 80</td>
<td>May 79 – Jul 79</td>
<td>May 79 – Jan 80 (57%)</td>
<td>yes</td>
</tr>
<tr>
<td>Iran-Iraq War, controls lifted</td>
<td>Jul 81</td>
<td>none</td>
<td>Nov 80 – Feb 81 (45%)</td>
<td>yes</td>
</tr>
<tr>
<td>Gulf War I</td>
<td>Jul 90</td>
<td>none</td>
<td>Aug 90 – Oct 90 (93%)</td>
<td>no</td>
</tr>
<tr>
<td>Strong demand</td>
<td>Mar 01</td>
<td>none</td>
<td>Dec 99 – Nov 00 (38%)</td>
<td>no</td>
</tr>
<tr>
<td>Venezuela unrest, Gulf War II</td>
<td>none</td>
<td>none</td>
<td>Nov 02 – Mar 03 (28%)</td>
<td>no</td>
</tr>
</tbody>
</table>
On the economic growth front, Kenya’s economic growth has been significantly fluctuating since the 1970s. During the early years of independence, Kenya achieved commendable economic growth compared to other SSA countries. Between 1975 and 1985, the average annual percentage growth in GDP was 4.1% (World Bank, 2018a). During the period 1985 to 1989, the average growth in GDP increased dramatically to 5.7% (World Bank, 2018a). However, in 1991 the percentage change in GDP declined to 1.4%. In 1992, Kenya recorded a historic low GDP growth rate of about -0.8% – the lowest since independence. However, between 1993 and 1995, the GDP growth increased considerably. The GDP growth rate increased from about -0.8% in 1992 to 0.4% in 1993, before further increasing to 2.6% in 1994 (World Bank, 2018a). By 1995 the GDP growth rate had reached 4.4%. But this high growth rate did not last for long. The GDP growth rate declined again systematically from 4.1% in 1996 to 0.5% in 1997 but bounced back to 3.3% in 1998. Just before the 2007 global financial crisis (GFC), Kenya’s growth rate was above 6%. Although the country was negatively affected by the GFC, leading to faltering of economic activity – recording a growth rate of 0.2% in 2008 – it quickly recovered. By 2010, growth rate in Kenya was 8.4% (World Bank, 2018a).

From 2014 to 2016, economic growth averaged 5.6%, while in 2016 alone it was at 5.8%, placing Kenya as one of the fastest growing economies in SSA. According to the World Bank (2018b), a stable macroeconomic environment, low oil prices, rebound in tourism, strong remittance inflows and a government-led infrastructure development initiative were the key drivers of the high growth rate. However, GDP growth is
expected to decelerate to 5.5% in 2017 as a result of the on-going drought and weak credit growth. The World Bank (2018b) projects Kenya’s GDP growth rate to rebound to 5.8% and 6.1% in 2018 and 2019, respectively, on the hopes of the completion of on-going infrastructure projects, a boom in tourism, resolution of slow credit growth and the strengthening of the global economy.

3. Literature review

While a number of studies have been conducted on the relationship between energy consumption and economic growth, the same cannot be said regarding the studies on the relationship between oil prices and economic growth – the latter are scanty.

The theoretical literature has four categories in which the energy-growth causality outcomes grouped – the growth hypothesis, the conservation hypothesis, the feedback hypothesis, and the neutrality hypothesis (see Saidi et al., 2017; Tiba and Omri, 2017; Shahbaz et al., 2018). A number of studies have examined the relationship between oil prices, energy consumption and economic growth, in both developed and developing countries. In particular, the relationship between energy consumption and economic growth has been examined extensively. Although the results have been mixed, they have been consistent with theoretical literature – that has established four outcomes.

Most studies support the growth hypothesis and argue that it is energy consumption that Granger-causes economic growth (see Destek, 2016; Odhiambo, 2009a; Narayan and Smyth, 2008; Narayan and Prasad, 2008; Narayan and Singh, 2007; Altimay and Karagol, 2005; Wolde-Rufael, 2004; Shiu and Lam, 2004; Chang et al., 2001; Yang, 2000; Cheng, 1997; Masih and Masih, 1996, among others). There is, however, another strand that supports the conservation hypothesis and argues that it is the growth of the
real sector that drives the demand for energy consumption (see, among others, Shahbaz et al., 2017; Odhiambo, 2016; Mozumder and Marathe, 2007; Hatemi-J and Irandoust, 2005; Narayan and Smyth, 2005; Gosh, 2002; Cheng, 1999; Cheng and Lai, 1997; Abosedra and Baghestani, 1989; Kraft and Kraft, 1978).

Between these two extremes, there are studies that support bidirectional causality; hence they maintain that both energy consumption and economic growth Granger-cause each other. Studies that support this middle-ground view include Saidi et al. (2017), Odhiambo (2009b), Paul and Bhattacharya (2004), Yang (2002), Glasure (2002) and Masih and Masih (1997). Though uncommon, there are also studies that support the fourth view the neutrality hypothesis – that contends that there is no Granger-causality between oil consumption and economic growth (see Rahman and Mamun, 2016; Cheng, 1997; Cheng, 1995; Yu and Hwang, 1984; Akarca and Long, 1980).

Unlike the causal relationship between energy consumption and economic growth, the causal relationship between oil prices and economic growth has not been fully explored. Very few studies have fully examined the nexus between oil prices and economic growth. Some of the studies that have examined the relationship between oil prices and economic growth include Hanabusa (2009), Jayaraman and Chooing (2009), Prasad et al. (2007), Rautava (2004), Glasure and Lee (2002), Kim and Willet (2000) and Darrat and Gilley (1996), among others.

Darrat and Gilley (1996), for example, find that oil price shocks are not a major cause of US business cycles. In addition, the study finds that both oil prices and real output cause significant changes in oil consumption without feedback causal effects. While examining the relationship between oil price and economic growth in the Organisation

Asafu-Adjaye (2000) estimated the causal relationships between energy consumption and income in Asian developing countries – India, Indonesia, the Philippines and Thailand – cointegration and error-correction modelling techniques. The results indicated the presence of bidirectional Granger-causality between oil prices and economic growth in the case of Thailand and the Philippines.

In an attempt to investigate the causal relationship between the price of oil and economic growth in Japan, Hanabusa (2009) finds that there is a feedback relationship between the price of oil and economic growth in Japan. While examining the causal relationship between growth and oil price in small Pacific Island countries, Jayaraman and Choong (2009) find that there is a unidirectional causal flow from oil price and international reserves to economic growth.

Although the bulk of the empirical studies support a negative relationship between oil price and economic growth, some recent studies have shown that this relationship may not be strictly negative for all countries. Prasad et al. (2007), for example, while examining the relationship between oil prices and real GDP nexus in the Fiji Islands, find that an increase in the oil price has a positive, albeit inelastic, impact on real GDP. The authors conclude that although their finding is inconsistent with the bulk of the
previous literature, it is not a surprising result for the Fiji Islands. Specifically, the authors argue that since the actual output in Fiji has been around 50% lower than its potential output, it has not reached a threshold level at which oil prices can negatively impact on output. Moreover, this finding, according to the authors, is consistent with the results from some emerging countries studied by the International Monetary Fund (IMF) (2000).

4. Estimation techniques and empirical analysis

In order to empirically examine the causality between oil prices and economic growth in Kenya, the study utilises a trivariate Granger-causality model that incorporates oil consumption as an intermittent variable – so as to address the omission-of-variable bias associated with a bivariate model (see Loizides and Vamvoukas, 2005; Nyasha and Odhiambo, 2015).

To further distinguish itself from other previous studies, the study used an autoregressive distributed lag (ARDL) bounds-testing technique to examine this dynamic linkage between oil prices and economic growth in Kenya. The ARDL is a contemporary estimation technique that has been widely used of late because of numerous advantages it offers as compared to the its conventional counterparts – residual-based technique and the Full-Maximum Likelihood test (Duasa, 2007). With the ARDL approach, estimation can be carried out with variable integrated of order 0 or one or a mixture of both. Thus it does not restrict the variables to be integrated of the same order. In addition, even with endogenous regressors, the technique provides unbiased long-run estimates and valid t-statistics (Nyasha and Odhiambo, 2017). Unlike the conventional co-integration methods that estimate the long-run relationship using a system of equations, the ARDL technique uses only a single reduced form
equation, making the estimation process simpler and easier without compromising the quality of results flowing from the analysis (Pesaran and Shin, 1999; Duasa, 2007). Furthermore, with the ARDL estimation procedure, a sufficient number of lags are generated in order to obtain optimal lag length per variable via the data-generating process within a general-to-specific modelling framework. A list of the numerous advantages offered by the ARDL estimation procedure would not be complete without mention of its superior small-sample properties. This property enables the estimation of a model based on a limited dataset (Odhiambo, 2016). The ARDL is, thus, considered the most suitable analysis method for this study.

In order to overcome the traditional weaknesses associated with many conventional cointegration techniques, the study uses the recently introduced ARDL-bounds testing approach to examine the long-run relationship between oil prices and economic growth – within a trivariate setting. The cointegration equations associated with the trivariate Granger-causality models in this study are expressed as:

### 4.1 ECM-based co-integration model

\[
\Delta \ln y_t = \alpha_0 + \sum_{i=1}^{n} \alpha_{1i} \Delta \ln y_{t-i} + \sum_{i=0}^{n} \alpha_{2i} \Delta \ln \text{OP}_{t-i} + \sum_{i=0}^{n} \alpha_{3i} \Delta \ln \text{OC}_{t-i} + \alpha_4 \ln y_{t-1} + \alpha_5 \ln \text{OP}_{t-1} + \alpha_6 \ln \text{OC}_{t-1} + \mu_{1t} \ldots \ldots \ldots (1)
\]

\[
\Delta \ln \text{OP}_t = \alpha_0 + \sum_{i=0}^{n} \alpha_{1i} \Delta \ln y_{t-i} + \sum_{i=1}^{n} \alpha_{2i} \Delta \ln \text{OP}_{t-i} + \sum_{i=1}^{n} \alpha_{3i} \Delta \ln \text{OC}_{t-i} + \alpha_4 \ln y_{t-1} + \alpha_5 \ln \text{OP}_{t-1} + \alpha_6 \ln \text{OC}_{t-1} + \mu_{2t} \ldots \ldots \ldots (2)
\]
\[ \Delta l n OC_t = \alpha_0 + \sum_{i=0}^{n} \alpha_1 \Delta l n y_{t-i} + \sum_{i=0}^{n} \alpha_2 \Delta l n OP_{t-i} + \sum_{i=1}^{n} \alpha_3 \Delta l n OC_{t-i} + \alpha_4 l n y_{t-1} \\
+ \alpha_5 l n OP_{t-1} + \alpha_6 l n OC_{t-1} + \mu_{3t} \ldots \ldots \ldots (3) \]

where:
\( y \) = growth rate of real gross domestic product (a proxy for economic growth)
\( OP \) = oil prices
\( OC \) = oil consumption
\( \alpha_0 \) = respective constant; \( \alpha_1 \) – \( \alpha_3 \) = respective short-run coefficients; \( \alpha_4 \) – \( \alpha_6 \) = respective long-run coefficients; \( l n \) = log operator; \( \Delta \) = difference operator; \( n \) = lag length; \( t \) = time period; and \( \mu_{it} \) = white-noise error terms.

### 4.2 ECM-based Granger-causality model

Following Odhiambo (2010) and based on the work of Pesaran and Shin (1999) and Pesaran et al. (2001), the ARDL-bounds testing approach adopted in this study can be expressed as:

\[ \Delta l n y_t = \alpha_0 + \sum_{i=1}^{n} \alpha_1 \Delta l n y_{t-i} + \sum_{i=1}^{n} \alpha_2 \Delta l n OP_{t-i} + \sum_{i=1}^{n} \alpha_3 \Delta l n OC_{t-i} + \delta ECM_{t-1} \\
+ \mu_{1t} \ldots \ldots \ldots (4) \]

\[ \Delta l n y_t = \alpha_0 + \sum_{i=1}^{n} \alpha_1 \Delta l n y_{t-i} + \sum_{i=1}^{n} \alpha_2 \Delta l n OP_{t-i} + \sum_{i=1}^{n} \alpha_3 \Delta l n OC_{t-i} + \delta ECM_{t-1} \\
+ \mu_{2t} \ldots \ldots \ldots (5) \]

\[ \Delta l n y_t = \alpha_0 + \sum_{i=1}^{n} \alpha_1 \Delta l n y_{t-i} + \sum_{i=1}^{n} \alpha_2 \Delta l n OP_{t-i} + \sum_{i=1}^{n} \alpha_3 \Delta l n OC_{t-i} + \delta ECM_{t-1} \\
+ \mu_{3t} \ldots \ldots \ldots (6) \]
where ECM is the error-correction term and $\delta$ is its coefficient.

5. Results and discussion

5.1 Stationarity test

Although the ARDL-bounds testing approach does not require that the variables be tested for stationarity prior to analysis, the approach is not applicable if the variables are integrated of order two [I(2)] or higher. For this reason, stationarity tests were carried out using Dickey-Fuller generalised least squares (DF-GLS) and Phillips-Perron (PP) tests. The stationarity results confirmed that the variables where a mixture of those integrated of order zero and those integrated of order one – thereby fulfilling the ARDL stationarity condition.

5.2 Cointegration results

Having confirmed that all the variables included in the causality test are integrated of order one, the next step is to test for the existence of a cointegration relationship between economic growth, oil prices and oil consumption – in a two-step process. The null hypothesis of no co-integration is tested against the alternative hypothesis of co-integration. First, the order of lags on the first differenced variables in the set of cointegration equations (1-3) is determined. The second step is the application of the bounds F-test to the same equations to determine the presence or absence of a long-run relationship between the variables under study.

If the calculated F-statistic is above the upper-bound level of the critical values provided by Pesaran et al. (2001), the null hypothesis of no co-integration is rejected – and a conclusion that a long-run relationship exists, is reached. Should the calculated F-statistic be below the lower-bound level, the null hypothesis of no co-integration cannot
be rejected. However, in the event that the calculated F-statistic falls within the upper- and the lower-bound levels, the results are deemed inconclusive. The results of the bounds F-test for co-integration are given in Table 1.

Table 1: Bounds F-test for Cointegration

<table>
<thead>
<tr>
<th>Dependent Variable</th>
<th>Function</th>
<th>F-statistic</th>
<th>Cointegration Status</th>
</tr>
</thead>
<tbody>
<tr>
<td>y</td>
<td>F(y</td>
<td>OP, OC)</td>
<td>2.76</td>
</tr>
<tr>
<td>OP</td>
<td>F(OP</td>
<td>y, OC)</td>
<td>5.03**</td>
</tr>
<tr>
<td>OC</td>
<td>F(OC</td>
<td>y, OP)</td>
<td>0.46</td>
</tr>
</tbody>
</table>

Asymptotic Critical Values

<table>
<thead>
<tr>
<th>Pesaran et al. (2001), p.300 Table CI(iii) Case III</th>
<th>1%</th>
<th>5%</th>
<th>10%</th>
</tr>
</thead>
<tbody>
<tr>
<td>I(0)</td>
<td>5.15</td>
<td>3.79</td>
<td>3.17</td>
</tr>
<tr>
<td>I(1)</td>
<td>6.36</td>
<td>4.85</td>
<td>4.14</td>
</tr>
</tbody>
</table>

Note: ** statistical significance at 5% level

The co-integration results in Table 1 confirm the existence of one cointegrating vector; hence, Granger-causality can be tested.

5.3 ECM-based granger-causality results

The short-run causality is established by the F-statistics on the explanatory variables derived from the Wald Test, while the long-run causality is determined by the negative sign and significance of the coefficient of the error-correction term. The results obtained from the estimation of Granger-causality model (equations 4-6) are presented in Table 2.

Table 2: Results of Granger-Causality Tests
<table>
<thead>
<tr>
<th>Dependent Variable</th>
<th>F-statistics [probability]</th>
<th>ECT_{t-1} [t-statistics]</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \Delta y_t )</td>
<td>-</td>
<td>0.363</td>
</tr>
<tr>
<td></td>
<td></td>
<td>[0.551]</td>
</tr>
<tr>
<td>( \Delta OP_t )</td>
<td>3.294*</td>
<td></td>
</tr>
<tr>
<td></td>
<td>[0.080]</td>
<td>0.744</td>
</tr>
<tr>
<td>( \Delta OC_t )</td>
<td>6.497**</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td>[0.016]</td>
<td>[0.968]</td>
</tr>
</tbody>
</table>

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

The results of the Granger-causality model show that there is a distinct unidirectional causal flow from economic growth to oil prices in Kenya. These results apply irrespective of whether the estimation is in the long run or in the short run.

The short-run results are confirmed by the F-statistics of economic growth (\( \Delta y \)) in the oil price function (\( \Delta OP \)) that is statistically significant – and the long-run results are supported by the error-correction term (ECM_{t-1}) in the same function, that is both negative and statistically significant at 10% level. These results are consistent with the conservation hypothesis – one of the four hypotheses postulated in the energy-growth theoretical literature – that states that it is the increase in economic development that causes the demand for energy to increase. Thus, in this case, it is the growth of the real sector that pushes oil prices, implying that Kenyan consumers have the ability to thrive even when prices are high.

The results further show that there is bidirectional causality between economic growth and oil consumption – but only in the short run – as firmed by the coefficients of oil consumption (\( \Delta OC \)) and economic growth (\( \Delta y \)) in the economic growth and oil
consumption functions, respectively, that are statistically significant at 10% and 5% levels, respectively.

6. Conclusion

In this study, we have explored the dynamic causal relationship between oil prices and economic growth in Kenya during the period from 1980 to 2015. A trivariate Granger-causality model, which incorporated oil consumption as an intermittent variable, was used. Although the energy consumption and economic growth nexus is gaining attention of researchers of late, little has been done on the specific relationship between oil prices and economic growth, in general, and in Kenya in particular. In addition, a few of the studies available on the subject mostly suffer from a number of methodology-related weaknesses – such as the omission-of-variable bias emanating from the use of bivariate causality models, and use of cross-sectional methodologies that fail to incorporate country-specific issues. Based on the ARDL bounds testing approach to co-integration and the ECM-based Granger-causality tests, results of this study reveal that there is distinct unidirectional Granger-causality flowing from economic growth to oil prices in the study country. These results were found to apply both in the short run and in the long run. Thus, it can be concluded that in Kenya, it is the real sector that pushes oil prices up. Further, it is possible to predict oil price changes in Kenya – given the changes in economic growth. However, the reverse – predicting the changes in economic growth given the changes in oil prices – is not possible. Hence, manipulation of the oil prices can be achieved without affecting the performance of the real sector – both in the short and long run. Nonetheless, it is oil consumption that was found to have feedback effect on economic growth, but only in the short run.

References


Energy Information Administration (EIA), 2008. Country Analysis Brief, South Africa (October)


