The Impact of Oil Prices on the Exchange Rate and Economic Growth in Norway

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Abstract

This study examines the impact of oil shocks on the real exchange rate and the gross domestic product in Norway using time series data from 1975 to 2008. The vector autoregressive has been implemented using the cointegration and the Granger causality test. The results of the study show that the increase in oil price is the reason behind Norway’s GDP increase and the increase of its competitiveness to trade by its real exchange rate depreciation. So it seems that oil price in this case is a blessing due to two reasons. First Norway uses the floating exchange rate regime which is a good shock absorber, increases the freedom of the monetary authority, and makes the adjustment smoother and less expensive. The second reason is that Norway has more flexible labor markets, improvements in monetary policy and smaller share of oil in production.

1. Introduction

Norway is one of the wealthiest nations in the world because its GDP per capita is the second highest in several years. Norway is also one of the largest oil exporting country in Europe and the sixth largest in the world in 2008 whose petroleum plays a role in its economic growth. In 2001 Norway’s oil production reached its peak at 3.22 million barrels per day. Since the discovery of oil in 1969, Norway did not join the OPEC and decides to evaluate its energy prices in line with the global market. Contrary to other producing countries, Norway uses its energy revenue wisely because it increases its gross savings during the increase of oil revenues.

The main goal of this study is to examine the impact of oil prices on Norway’s real exchange rate and gross domestic product to show clearly if the oil shock is a blessing or a curse. To achieve this goal, the vector autoregressive VAR model will be used using time series data from 1975-2008 covering the three famous oil shocks.

2. Studies on Oil Price and the Macroeconomy

Many writers found that oil prices has a significant negative impact on the macroeconomy in both developed and the developing countries. Jayaraman & choong (2009) found that oil
prices have a long run and short run relationship with the economic growth in the Asian pacific countries. The same results are found in Japan, (Hanabusa, 2009), in Thailand (Rafiq et. Al, 2008), in Malaysia, Japan, south Korea, Thailand, Philippines and Singapore (Cunado & Gracia, 2005), in China (Du et. Al, 2010) and in Greece, (Papapetrou, 2001). It also seems that oil prices has a long run relationship with the economic activates in the US and the European countries, (Lardic & Mignon, 2006). The same results are found by Cun˜ado & Gracia (2003) and Blanchard et. al (2007) in the OECD countries. Garratt, et. al (2003) also found that oil prices increase economic activities and inflation in the UK. Doroodian & Boyd (2003) found that the impact of oil prices on the US economy was negatively significant during the 1970’s and this impact began to fall after the 1980’s because the US economy has been transformed from manufacturing based economy to a service based economy. However, Hamilton (1983), Barsky (2004), Leduc (2002), Gisser et. al (1986), Brown et. al (1995) and Anzuini (2007) found that oil prices have a negative impact on the US economic growth even after the 1980s. However, in Norway, oil prices have an insignificant impact on the economic activates, (Robalo, 2007) while in the Republic of Trinidad and Tobago oil prices are one of the major determinants on growth, (Lorde et. Al, 2009).

3. Studies on Oil Prices and the Exchange Rate

Amano (1998) states that an oil price shock is a major source on the real exchange rate movement of the US dollar. The same results are found in Turkey, (Ozturk, 2008), and in Russia, (Rautava, 2004). However, China’s exchange rate has an insignificant response to the changes in oil prices because China is less dependent on imported oil. It also pegs its exchange rate to a basket of currencies. This helped the Chinese exchange rate to remain stable despite the oil shocks, (Huang, 2006). In the 1970’s the changes in oil prices have caused a real exchange rate appreciation in the US dollar exchange rate in 1973-74 oil shocks, while the US dollar depreciated in the 1979-80 oil shock due to the decrease in the US dependents on the OPEC oil, (Golub, 1983). However, in Nigeria, oil prices will cause its real exchange rate to appreciate, (Aliyu, 2008). The same results are found in Fiji, (Narayan et. al, 2008).

4. Methodology

In this study the vector autoregressive (VAR) model which is commonly used for forecasting a system of interrelated time series and for analyzing the dynamic impact of
random disturbances on the system of variables is used. It is useful because it is less restrictive compared to other models. This model introduced by Sims (1980) can be written as follows:

\[ Y_t = B_1 y_{t-1} + \ldots + B_p y_{t-p} + U x_t + \varepsilon_t \]  

(1)

Where \( Y_t \) is the \( k \)-vector of endogenous variables, \( x_t \) is a \( d \)-vector of exogenous variables, \( B_1, \ldots, B_p \) and \( U \) are the matrices of coefficients to be estimated, and \( \varepsilon_t \) is a vector of novelty that may be contemporaneously correlated but are uncorrelated with their own lagged values and uncorrelated with all of the right hand side variables. In this study two models the GDP growth and the real exchange rate are built.

For the real exchange rate model five variables are used. They are real exchange rate model as a dependent variable, consumer price index, trade balance, oil price, government final consumption expenditure as independent variables. The real exchange rate model is specified as follows:

\[ \log \text{REXCH}_t = \alpha + \beta_1 \log \text{CPI}_t + \beta_2 \log \text{OIL}_t + \beta_3 \text{TB}_t + \beta_4 \text{FDINET}_t + \varepsilon_t \]  

(2)

Where REXCH is the real effective exchange rate national currency per US dollar, CPI is the consumer price index; TB is the trade balance of goods and services measured in millions of US dollars, OIL is the oil price US dollar per barrel, FDINET is the net of foreign direct investment measured in millions of US dollars, \( \beta_1, \beta_2, \beta_3, \beta_4 \) are the coefficients of the model, \( \alpha \) is the intercept, and \( \varepsilon_t \) is the error term.

For the GDP growth model five variables are used. They are the gross domestic product as a dependent variable, inflation rate, total trade, employment, and the oil price as independent variables. The GDP growth model is specified as follows:

\[ \log \text{GDP}_t = \alpha + \beta_1 \text{INF}_t + \beta_2 \log \text{OIL}_t + \beta_3 \log \text{TDV}_t + \beta_4 \log \text{EMPLOY}_t + \varepsilon_t \]  

(3)

Where GDP is the gross domestic product measured in millions of US dollars, INF is the inflation rate, OIL is the price of oil US dollar per barrel, TDV is the total trade of goods and services measured in millions of US dollars, EMPLOY is the employment measured in thousands of workers, \( \beta_1, \beta_2, \beta_3, \beta_4 \) are the coefficients of the model, \( \alpha \) is the intercept, and \( \varepsilon_t \) is the error term.
5. Data Source

All the variables namely, gross domestic product, real effective exchange rate, government consumption expenditure, trade balance of goods and services, total trade of goods and services, consumer price index, inflation rate, and employment are taken from the World Bank data base, while the oil price is taken from the OPEC data statistics.

6. Estimation procedures

6.1 Unit Root Test

Most time series variables are not stationary because they have time trend, i.e. the time series variable mean, variance and the covariance do not change over time to avoid the possibility of incorrect regressions and wrong conclusion. The Augmented Dickey-Fuller test is used in this study to examine whether the variables are stationary or not. If the variables are stationary at levels, the variables will be integrated in order \( y_t \sim I(0) \), but if the variables are not at levels, the variables will be integrated in order \( y_t \sim I(1) \).

6.2 Cointegration Test

If the variables are stationary at the first difference, they will have a long run relationship; hence, the variables are cointegrated.

Johansen (1988) and Johansen and Juselius (1990) built a test that can help to find out whether the variables have a long relationship. In this study the Johansen-Juselius (JJ) cointegration test is used which is based on the vector autoregressive (VAR) model. The optimal lag length will be determined by the Akaike Information Criteria.

\[
y_t = \mu + \sum_{k=1}^{p} \Gamma_k y_{t-k} + \varepsilon_t \quad (4)
\]

This model can help us to use the cointegration process where \( y_t \) is a vector of \( I(1) \) variables, \( \mu \) is a vector of constants, and \( \varepsilon_t \) is a vector of white noise residuals at time \( t \) with zero mean and constant variance. Equation (5) below is the same as equation (4) but with only one difference which is \( p-1 \):

\[
\Delta y_t = \mu + \sum_{k=1}^{p-1} L_k \Delta y_{t-k} + \Gamma \Delta y_{t-1} + \varepsilon_t \quad (5)
\]
where $J_k = -(I - A_1 - ... - A_k)$, $(k = 1, ..., p-1)$ and $\Gamma = -(I - A_1 - A_2 - ... - A_k)$. $\Gamma$ is called the impact matrix that can give us information about the long run relationship between the variables. The rank $(r)$ of $JI$ is equal to the number of cointegrating vectors. If $\Gamma$ is of full-rank, that is $r = g$, then there are $g$ cointegrating vectors. If $0 < r < g$, there exists $r$ cointegrating vectors, which means that there are $r$ stationary linear combinations of $y_t$. If the rank of $\Gamma$ is 1, there exists only 1 cointegrating vector. But if the rank of $JI$ is zero, there is no cointegrating equation and the variables are not cointegrated.

The Johansen process is based on two kinds of likelihood ratio tests, the trace test and the maximum eigenvalue test. The test statistic for the trace test is given in the following equation:

$$\lambda_{\text{trace}}(r) = -T \sum_{i=r+1}^{g} \ln(1-\lambda_i) \quad (6)$$

where $\lambda_i$ is the largest eigenvalue of the $\Pi$ matrix, $r$ is the number of cointegration vectors, $g$ is the number of variables and $T$ is the number of observations. The null hypothesis under this test is that there are less than or equal to $r$ cointegrating vectors and the alternative hypothesis is a general one. For example, to test if there is at most only 1 cointegrating vector, the null and alternative hypotheses will be as follows:

$H_0$: $r \leq 1$ (there is at most 1 cointegrating vector) against

$H_1$: $r \geq 2$ (there are at least 2 cointegrating vectors)

If the test statistic is greater than the critical value, $H_0$ will be rejected.

The test statistic for the second test, the maximum eigenvalue test is written as follows:

$$\lambda_{\text{max}}(r, r+1) = -T \ln(1-\lambda_{r+1}) \quad (7)$$
The null hypothesis in this test is that there are exactly \( r \) cointegration vectors against the alternative hypothesis of \((r + 1)\) cointegrated vectors where \( r = 1, 2, \ldots, g - 1, g \). For example, to test the existence of 1 cointegrating vector, the null and alternative hypotheses are as follows:

\[
H_0: \ r = 1 \text{ (there is exactly 1 cointegrating vector) against}
\]

\[
H_1: \ r = 2 \text{ (there are exactly 2 cointegrating vectors)}
\]

If the value of the test statistic is greater than the critical value, then \( H_0 \) will be rejected.

### 6.3 Granger Causality test

The Granger approach (1969) helps us to find out the variable \( a \) granger causes variable \( b \), the granger causality exists if the past value of \( a \) can predicate the present value of \( b \). There is unidirectional causality running from \( a \) to \( b \) if the estimated coefficients on the lagged values of \( a \) are statistically significantly different from zero as a group in equation (8). The set of estimated coefficients on the lagged values of \( y \) in equation (9) below is not significantly different from zero.

\[
y_t = \sum_{i=1}^{k} \alpha_i y_{t-i} + \sum_{i=1}^{k} \beta_i x_{t-i} + u_{1t} \quad (8)
\]

\[
x_t = \sum_{i=1}^{n} \lambda_i x_{t-i} + \sum_{i=1}^{n} \theta_i y_{t-i} + u_{2t} \quad (9)
\]

Conversely, unidirectional causality from \( y \) to \( x \) exists if the set of lagged coefficients of \( y \) in equation (9) is statistically significantly different from zero but, the set of lagged coefficients of \( x \) in equation (8) is not. Bilateral causality between \( x \) and \( y \) exists when the set of lagged coefficients of \( x \) in equation (8) and the set of lagged coefficients of \( y \) in equation (9) are both statistically significantly different from zero. Finally, there is an
independence between x and y when the lagged coefficients of x in (8) and the lagged coefficients of y in (9) are both insignificantly different from zero. If we find out that the models of this study are cointegrated, we will use the vector error-correction model (VECM) to find the short run causality between the variables in our models. The VECM can help us to find the long run and the short run causality between the variables.

7. Empirical Results and Discussion of Results

Table 2: ADF Unit Root Test Results (exchange rate model)

<table>
<thead>
<tr>
<th>Variable</th>
<th>Level</th>
<th>First Difference</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Intercept</td>
<td>Intercept and Trend</td>
</tr>
<tr>
<td>LREXCH</td>
<td>-2.150352</td>
<td>-2.385633</td>
</tr>
<tr>
<td>LCPI</td>
<td>-2.112206</td>
<td>-0.761737</td>
</tr>
<tr>
<td>LOIL</td>
<td>-1.030786</td>
<td>-0.390732</td>
</tr>
<tr>
<td>TB</td>
<td>4.479656</td>
<td>3.151230</td>
</tr>
<tr>
<td>LGOVEX</td>
<td>-0.398263</td>
<td>-2.957423</td>
</tr>
</tbody>
</table>

Note: *** denotes significance at the 1% level and ** at the 5% level.

Table 2: ADF Unit Root Test Results (GDP model)

<table>
<thead>
<tr>
<th>Variable</th>
<th>Level</th>
<th>First Difference</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Intercept</td>
<td>Intercept and Trend</td>
</tr>
<tr>
<td>LGDP</td>
<td>0.337309</td>
<td>-2.344588</td>
</tr>
<tr>
<td>LOIL</td>
<td>-1.030786</td>
<td>-0.390732</td>
</tr>
<tr>
<td>INF</td>
<td>-2.237289</td>
<td>-2.693264</td>
</tr>
<tr>
<td>LEMPLOY</td>
<td>-0.340870</td>
<td>-2.837357</td>
</tr>
<tr>
<td>LTDV</td>
<td>0.834194</td>
<td>-1.967240</td>
</tr>
</tbody>
</table>

Note: *** denotes significance at the 1% level and ** at the 5% level.

The ADF results show that all the variables in both models are stationary at the first difference at 5% level of significance, thus we will use the Johansen and Juselius
cointegration test to find out the negative or the positive long run relationship between the independent variables and the dependent variables in both models.

7.1 Cointegration Test results

The cointegration will be tested after we have found that all the variables in both models are stationary at the first difference to examine the long run relationship between the independent variables and the dependent variable in both the GDP model and the exchange rate model. Before we conduct the test we will use the lag length criteria to find the optimal lag length for our models, due to the sensitiveness of the cointegration to the lag length.

Table 3: Lag Length Selection from VAR Estimates (The Exchange Rate Model)

<table>
<thead>
<tr>
<th>Lag</th>
<th>LogL</th>
<th>LR</th>
<th>FPE</th>
<th>AIC</th>
<th>SC</th>
<th>HQ</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>-1424.295</td>
<td>NA</td>
<td>7.68e+33</td>
<td>92.21255</td>
<td>92.44384</td>
<td>92.28795</td>
</tr>
<tr>
<td>1</td>
<td>-1269.788</td>
<td>249.2041</td>
<td>1.85e+30</td>
<td>83.85729</td>
<td>85.24502</td>
<td>84.30966</td>
</tr>
<tr>
<td>2</td>
<td>-1242.310</td>
<td>35.45531</td>
<td>1.81e+30</td>
<td>83.69743</td>
<td>86.24160</td>
<td>84.52677</td>
</tr>
<tr>
<td>3</td>
<td>-1187.984</td>
<td>52.57400*</td>
<td>4.02e+29*</td>
<td>81.80540*</td>
<td>85.50601</td>
<td>83.01171*</td>
</tr>
</tbody>
</table>

* indicates lag order selected by the criterion
LR: sequential modified LR test statistic (each test at 5% level)
FPE: Final prediction error
AIC: Akaike information criterion
SC: Schwarz information criterion
HQ: Hannan-Quinn information criterion
Table 4: Lag Length Selection from VAR Estimates (The GDP Model)

<table>
<thead>
<tr>
<th>Lag</th>
<th>LogL</th>
<th>LR</th>
<th>FPE</th>
<th>AIC</th>
<th>SC</th>
<th>HQ</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>18.49346</td>
<td>NA</td>
<td>2.96e-07</td>
<td>-0.843341</td>
<td>-0.614320</td>
<td>-0.767427</td>
</tr>
<tr>
<td>1</td>
<td>163.6585</td>
<td>235.8931</td>
<td>1.66e-10</td>
<td>-8.353654</td>
<td>-6.979526</td>
<td>-7.898169</td>
</tr>
<tr>
<td>2</td>
<td>218.9874</td>
<td>72.61930*</td>
<td>2.82e-11*</td>
<td>-10.24922*</td>
<td>-7.729982*</td>
<td>-9.414161*</td>
</tr>
</tbody>
</table>

* indicates lag order selected by the criterion
LR: sequential modified LR test statistic (each test at 5% level)
FPE: Final prediction error
AIC: Akaike information criterion
SC: Schwarz information criterion
HQ: Hannan-Quinn information criterion

Table three shows that the optimal lag length for the exchange rate model is lag three based on the minimum AIC while table four shows that the optimal lag length for the GDP model is the lag two based on the minimum AIC.

While Table five shows the cointegration results for the trace statistics, table six shows the cointegration results for the maximum eigenvalue for the exchange rate model.

Table 5: Johansen-Juselius Cointegration Test Results Based on the Trace Statistic

<table>
<thead>
<tr>
<th>Hypothesized No. of CE(s)</th>
<th>Trace Eigenvalue</th>
<th>Trace Statistic</th>
<th>0.05 Critical Value</th>
<th>Prob.**</th>
</tr>
</thead>
<tbody>
<tr>
<td>None *</td>
<td>0.824542</td>
<td>142.2757</td>
<td>76.97277</td>
<td>0.0000</td>
</tr>
<tr>
<td>At most 1 *</td>
<td>0.768647</td>
<td>90.06492</td>
<td>54.07904</td>
<td>0.0000</td>
</tr>
<tr>
<td>At most 2 *</td>
<td>0.567264</td>
<td>46.15062</td>
<td>35.19275</td>
<td>0.0023</td>
</tr>
<tr>
<td>At most 3 *</td>
<td>0.413619</td>
<td>21.02183</td>
<td>20.26184</td>
<td>0.0392</td>
</tr>
<tr>
<td>At most 4</td>
<td>0.153752</td>
<td>5.008279</td>
<td>9.164546</td>
<td>0.2825</td>
</tr>
</tbody>
</table>

Trace test indicates 4 cointegrating eqn(s) at the 0.05 level
* denotes rejection of the hypothesis at the 0.05 level
**MacKinnon-Haug-Michelis (1999) p-values
Table 6: Johansen-Juselius Cointegration Test Results Based on the Maximum Eigenvalue Statistic

<table>
<thead>
<tr>
<th>Hypothesized No. of CE(s)</th>
<th>Eigenvalue</th>
<th>Max-Eigen Statistic</th>
<th>0.05 Critical Value</th>
<th>Prob.**</th>
</tr>
</thead>
<tbody>
<tr>
<td>None *</td>
<td>0.824542</td>
<td>52.21073</td>
<td>34.80587</td>
<td>0.0002</td>
</tr>
<tr>
<td>At most 1 *</td>
<td>0.768647</td>
<td>43.91430</td>
<td>28.58808</td>
<td>0.0003</td>
</tr>
<tr>
<td>At most 2 *</td>
<td>0.567264</td>
<td>25.12879</td>
<td>22.29962</td>
<td>0.0196</td>
</tr>
<tr>
<td>At most 3 *</td>
<td>0.413619</td>
<td>16.01355</td>
<td>15.89210</td>
<td>0.0479</td>
</tr>
<tr>
<td>At most 4</td>
<td>0.153752</td>
<td>5.008279</td>
<td>9.164546</td>
<td>0.2825</td>
</tr>
</tbody>
</table>

Max-eigenvalue test indicates 4 cointegrating eqn(s) at the 0.05 level
* denotes rejection of the hypothesis at the 0.05 level
**MacKinnon-Haug-Michelis (1999) p-values

Table 5 and table 6 show that the trace and the maximum eigenvalue indicates four cointegration equations at 5% level of significance, indicating a long run relationship between the independent variables namely, consumer price index, trade balance, oil price, and net foreign direct investment and the dependent variable the real exchange rate.

Table seven below shows the normalized cointegration vector.

Table 7: Cointegration Equation Normalized With Respect To LREXCH

<table>
<thead>
<tr>
<th>LREXCH</th>
<th>LCPI</th>
<th>TB</th>
<th>LOIL</th>
<th>FDINET</th>
<th>C</th>
</tr>
</thead>
<tbody>
<tr>
<td>1.000000</td>
<td>-0.514806</td>
<td>4.10E-12</td>
<td>-0.221097</td>
<td>4.90E-11</td>
<td>-2.004132</td>
</tr>
<tr>
<td>(0.08945)</td>
<td>(2.8E-12)</td>
<td>(0.09114)</td>
<td>(1.0E-11)</td>
<td>(0.42122)</td>
<td></td>
</tr>
</tbody>
</table>

From table seven, the long run equation for the real exchange rate model can be written as:

\[ \text{Log REXCH} = 2.004132 + 0.514806 \text{ log CPI} - 4.10E-12 \text{ TB} + 0.221097 \text{ log OIL} - 4.90E-11 \text{FDI} \quad (10) \]

The equation above shows that the consumer price index and oil price have a long run positive relationship with the real exchange rate while the trade balance and the net foreign direct investment have a long run negative relationship with the real exchange rate.
1% increase in the consumer price will depreciate the real exchange rate by 0.5%. This relationship is obvious because when the price level or inflation increases, it will reduce the value of the local currency.

One million increase in trade balance will appreciate the real exchange rate by 4.10E-12, because the increase trade balance will increase the foreign capital inflows which comes into the country causing local currency to appreciate.

1% increase in oil price will cause the real exchange rate to depreciate by 0.22%. It seems that the oil price will make the value of the local currency lower which will increase the comparative advantage for Norway to export more since exports play more than 47% of Norway total GDP. This has a positive effect on its growth.

One million increase in net foreign direct investment will appreciate the exchange rate by 4.90E-11, because foreign investors always bring foreign currency to invest in the country, so the increase in foreign capital inflows will increase the value of Norway’s local currency.

Table 8 below shows the cointegration results for the trace statistics whereas table 9 shows the cointegration results for the maximum eigenvalue for the GDP growth model.

**Table 8: Johansen-Juselius Cointegration Test Results Based on the Trace Statistic**

<table>
<thead>
<tr>
<th>Hypothesized No. of CE(s)</th>
<th>Eigenvalue</th>
<th>Trace Statistic</th>
<th>0.05 Critical Value</th>
<th>Prob.**</th>
</tr>
</thead>
<tbody>
<tr>
<td>None *</td>
<td>0.800583</td>
<td>117.7600</td>
<td>76.97277</td>
<td>0.0000</td>
</tr>
<tr>
<td>At most 1 *</td>
<td>0.701815</td>
<td>67.77696</td>
<td>54.07904</td>
<td>0.0019</td>
</tr>
<tr>
<td>At most 2</td>
<td>0.447111</td>
<td>30.26571</td>
<td>35.19275</td>
<td>0.1544</td>
</tr>
<tr>
<td>At most 3</td>
<td>0.230216</td>
<td>11.89519</td>
<td>20.26184</td>
<td>0.4582</td>
</tr>
<tr>
<td>At most 4</td>
<td>0.114914</td>
<td>3.784182</td>
<td>9.164546</td>
<td>0.4452</td>
</tr>
</tbody>
</table>

Trace test indicates 2 cointegrating eqn(s) at the 0.05 level
* denotes rejection of the hypothesis at the 0.05 level
**MacKinnon-Haug-Michelis (1999) p-values
For both trace and maximum eigenvalue statistics we find at least two cointegration equations at 5% level of significance indicating that the long run relationship exists between the independent variables, namely inflation rate, oil price, employment, and total trade value and the GDP as a dependent variable. Table 10 shows the normalized cointegration vector.

**Table 10: Cointegration Equation Normalized With Respect To LGDP**

<table>
<thead>
<tr>
<th>LGDP</th>
<th>LOIL</th>
<th>INF</th>
<th>LTDV</th>
<th>LEMPLOY</th>
<th>C</th>
</tr>
</thead>
<tbody>
<tr>
<td>1.000000</td>
<td>-0.096420</td>
<td>0.033770</td>
<td>-0.921073</td>
<td>-0.044341</td>
<td>-1.957496</td>
</tr>
<tr>
<td>(0.02423)</td>
<td>(0.00550)</td>
<td>(0.03907)</td>
<td>(0.18005)</td>
<td>(0.72142)</td>
<td></td>
</tr>
</tbody>
</table>

Hence, the long run equation for the real exchange rate model can be written as

\[
\log GDP = 1.957496 + 0.096420 \log OIL - 0.033770 \log INF + 0.921073 \log LTDV - 0.044341 \log EMPLOY
\]

(11)

The equation above shows that oil price, total trade value, and employment have a long run positive relationship with the gross domestic product, while inflation rate has a long run negative relationship with the gross domestic product.

1% increase in oil price will increase Norway’s gross domestic product by 0.0957%, indicating the higher oil price the better economic growth in Norway.
One unit increase in inflation rate will slow Norway’s GDP growth by 0.0337%. This relationship is obvious because the increase in price level will decrease the demand for goods and services. This has its negative effect on the GDP. This result is supported by Sarel (1996), Javier (1999), and Barro (1998).

1% increase in total trade value will increase Norway’s gross domestic product by 0.921%. This positive relationship is due to the fact that Norway is an open economy and its trade plays more than 73% of total GDP. It is clear that the increase in total trade will lead to a higher growth.

1% increase in employment will increase Norway’s gross domestic product by 0.0443%. The increase in employment means that there will be more people who have the ability to buy goods and services and that in general will increase the demand for goods and services. This will have a positive impact on the growth.

### 7.2 Granger Causality test

After finding the long run relationship between the dependent and the independent variables in both the exchange rate model and the GDP growth model, we find it obligatory to use the Granger causality test (VECM). Table 11 shows the F-statistics results for the exchange rate model that only the oil price granger causes the real exchange rate in the short run. The significance of the ect (-1) indicates that all the variables, namely oil price, trade balance, consumer price index, net foreign investment granger cause the real exchange rate in the long run.

Table 12 shows that only the total trade value granger causes the real gross domestic product in the short run. The significance of the ect (-1) indicates that all the variables, namely oil price, inflation rate, total trade value, and employment granger causes the gross domestic product in the long run.
Table 11: Granger Causality Results with LOG REXCH as the Dependent Variable

<table>
<thead>
<tr>
<th></th>
<th>∑DLOG REXCH</th>
<th>∑DLOG OIL</th>
<th>∑D TB</th>
<th>∑D cpi</th>
<th>∑D FDINET</th>
<th>ect(-1)</th>
</tr>
</thead>
<tbody>
<tr>
<td>F-stats.</td>
<td>1.045567</td>
<td>2.406444 **(4)</td>
<td>0.951094 (2)</td>
<td>1.795367 (1)</td>
<td>0.858794 (2)</td>
<td>0.0576**</td>
</tr>
</tbody>
</table>

Notes: ect(-1) represents the error correction term lagged one period. The numbers in the brackets show the optimal lag based on the AIC. D represents the first difference. Only F-statistics for the explanatory lagged variables in first differences are reported here. For the ect(-1) the t-statistic is reported instead. ** denotes significance at the 5% level and * indicates significance at the 10% level.

Table 12: Granger Causality Results with LOG GDP as the Dependent Variable

<table>
<thead>
<tr>
<th></th>
<th>∑DLOG GDP</th>
<th>∑DLOG OIL</th>
<th>∑D INF</th>
<th>∑DLOG TDV</th>
<th>∑D LOGEMPLOY</th>
<th>ect(-1)</th>
</tr>
</thead>
<tbody>
<tr>
<td>F-stats.</td>
<td>2.125915 (4)</td>
<td>0.976487 (3)</td>
<td>2.016342(1)</td>
<td>3.918640 **(4)</td>
<td>1.986354 (1)</td>
<td>0.1252*</td>
</tr>
</tbody>
</table>

Notes: ect(-1) represents the error correction term lagged one period. The numbers in the brackets show the optimal lag based on the AIC. D represents the first difference. Only F-statistics for the explanatory lagged variables in first differences are reported here. For the ect(-1) the t-statistic is reported instead. ** denotes significance at the 5% level and * indicates significance at the 10% level.

The results emanated from the cointegration and the Granger causality show that oil windfall is not a curse because the increase in oil prices will depreciate the real exchange rate of the country. This helps Norway to increase its comparative advantage to trade. Oil price will also increase Norway gross domestic product. Therefore, it can be stated that oil price is blessing for Norway due to two reasons. The first reason is that Norway uses the floating exchange rate regime which always acts as a shock absorber. The floating exchange rate regime can achieve policy autonomy as well. The monetary authority will be independent which can help to choose the inflation rate independently. The floating exchange rate regime can help to achieve the adjustments smoother and less expensive during the external shocks.

The other reason is that despite Norway is oil producing country; it has a flexible labor market, smaller share of oil in production, and Indirect Tax Analogy. All these policies help Norway’s economy to be less vulnerable to the oil shocks.
8. Conclusion

This study aims at finding out the impact of oil shocks on the real exchange rate and the gross domestic product in Norway using time series data from 1975 to 2008. The vector autoregressive was implemented using the cointegration and the Granger causality test. The results show that the increase in oil price will cause Norway’s GDP to increase. It also increases its competitiveness to trade by its real exchange rate depreciation. It seems that the increase in oil price in Norway is a blessing due to two reasons. First, Norway uses the floating exchange rate regime, which is a good shock absorber. It increases the freedom of the monetary authority, and makes the adjustment smoother and less expensive. The second reason is that Norway has more flexible labor markets and smaller share of oil in production.

References


