

Price Volatility and Demand for Oil: A Comparative Analysis of Developed and Developing Countries

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JEL Classification JEL Q37, Q41, Q43

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1. Introduction

The volatility of global oil prices in the past decade is widely reported, yet the disparate consumption pattern between developing and developed countries is scarcely analysed. Oil prices charts have looked like a rollercoaster ride throughout the last thirty years. The introduction of netback pricing temporarily saw the brakes applied and instability slowed. However, this could not be sustained in the face of rapid economic growth from emerging economies. The fundamental change in living standards experienced in these countries has seen increased oil usage. The developing countries are climbing the ‘energy ladder’, moving to industrial enterprise.

Global energy consumption is valued at around \$3.65 trillion annually, making up approximately 5 percent of global GDP (BP, 2014; World Bank, 2014). The oil market is a significant portion of the global economy thus the shocks to this market can cause significant macroeconomic turbulence. Currently, the distribution of oil consumption is changing. Demand in China is expected to surpass the US in 2029 and non-OECD Asian demand is expected to grow by six million-barrels per day through to 2030 (BP, 2013). In contrast to OECD countries, demand fell between 2006-08 with the price increases suspected to have played a role (Wurzel *et al.*, 2009).

More than one-third of energy use comes from oil, more than any other source of fuel (Rubin, 2012). Oil’s role as a transit fuel is of paramount importance in the daily lives of most citizens worldwide. The price of oil is a significant driver of economic development and is so inextricably a part of modern society that an oil shock can have a significant impact on an economy. Rubin (2012, p. 33) states an oil shock can “deep-six” an economy, as other consumption suffers as the effect of high oil prices is transmitted to the wider economy.

Oil price shocks have previously been considered to be caused primarily by supply disruption. Historical shocks are strongly correlated with major geopolitical disturbances (Hamilton, 2003). In the recent episode no one major war or instability can be attributed to the shock. In fact supply remained constant (BP, 2014). The fundamental cause suggested was a shock to demand. Hamilton (2009) postulated the cause was a classic economic shortage situation. The increased global demand would have required significant supply increases to sustain the pre-shock price level. This was not experienced. Moreover, non-OPEC countries reduced oil supply as we will discover later in this paper. The primary economic problem was an issue of scarcity occurring from a demand shock.

An emerging hypothesis is that the elasticities for income and price vary between developed and developing countries. Consideration of elasticities is taken by previous papers e.g. Gately and Huntington (2002); Hamilton (1983). Despite the simplicity of the technique the expanse of comparable studies for different time periods enables cross-study comparison. This study uses two separate cross-country panels which enables the comparison of consumption patterns between country groups of varied development (see Appendix 1). This paper will also use decomposition of the price and income series to account for asymmetric effect of these on consumption. The period of analysis is 1980-2012, as this provides a wide

coverage of oil price instability and considers a more up to date panel than has previously been published.

This paper analyses the varying oil consumption patterns of developed and developing countries during the period of an oil price shock. The use of asymmetric price and income decomposition will allow a greater understanding of the impact of not only the price shock but also the global financial crisis. In developing countries demand is expected to be fuelled by economic expansion (income effect), whereas, in developed economies, the substitution effect plays a larger role. Price elasticity of demand for oil should vary between the two panels.

The paper is structured as follows: Section 2 reviews the literature surrounding the oil market and the development in estimation techniques. Section 3 considers the driving forces of an oil price shock and analyses their significance in the 2007-08 price shock. Section 4 sets out the methodology which underpins the analysis undertaken. Section 5 presents and interprets the results and establishes policy implications. Section 6 provides a summative conclusion.

2. Previous Studies of Oil Price and Consumption

The literature surrounding the economics of the oil market has evolved considerably since the 1970's. Studies initially sought to establish how oil consumption and other macroeconomic indicators were interrelated. Since the late 1980's, the use of asymmetric decomposition has been applied however there is varied support for its relevance. Newer studies have focused more on the developing economies and their growing energy demands. A cross-country panel data analysis can aid understanding of global consumption patterns. However, due to the proximity of the recent oil shock there is a scarcity of literature investigating its economic impact.

Hamilton (1983) in his seminal work discussed the relationship with oil consumption and the wider economy. The paper highlights the evidence that all US recessions are preceded by an oil shock, where typically an increase in oil prices is followed 3-4 quarters later with a shock to domestic output. The study illustrated one of the first explicit relationships between oil and the lagged transmission to the economy. Hamilton undertakes causality testing to establish that the oil price was a key factor in the signalling of recession.

Numerous studies following Hamilton (1983) have found a relationship between oil consumption and variables such as income, inflation, and exchange rates. The effect of a price shock is estimated by Burbidge and Harrison (1984). They find that during the period 1962-1981 in Canada, Japan, West Germany, USA and UK a shock of one-deviation to oil price leads to a rise in wages and inflation. They also found a significant decline in industrial production (a proxy for growth) following the shock in the US and UK though the effect is less significant in other countries.

Despite these findings the macroeconomic relationship has been questioned. Mork (1989) noted that in the period used in Hamilton's work all price movements were positive and questioned if the relationship would hold in periods of price decline. Through developing

separate variables for price increases/decreases Mork mirrors the results of Hamilton (1983) during price increases but for prices decline the result is statistically insignificant. Consequently, only positive price changes appear to transmit to the US economy. One must consider if such results would also hold for developing economies.

The asymmetry of oil prices was investigated using a Koyck (1954) lag-demand function with an additional price and/or income decomposition term in Gately and Huntington (2002). Results for OECD countries echo the findings of Mork (1989) - that price increases have greater effect on demand for oil than price decreases. Therefore failure to account for price asymmetry may bias estimations of demand. OECD countries experience an income response, which is “dramatically different from its effect in the developing countries” (Gately and Huntington, 2002, p. 16).

Indeed, the *global* effect may be negligible but that is a net effect. Comparison of developing versus developed economies yields different results. Rubin (2012) considers in the present scenario where oil demand is outstripping supply to be a ‘zero-sum’ situation. The analysis shows that emerging economies are price insensitive to oil in comparison to the industrialised economies. In part this is attributable to China and India experiencing large increases in income. This is due to what he describes as a “transformational development in the way life is lived” (Rubin, 2012, p. 142). It is argued that as the living standards change the income elasticity will be nearing unity.

There was a considerable gap in the literature surrounding the economic impact of oil price shocks in developing countries. Tang *et al.* (2010) consider the Chinese economy, and observed that a rise in oil price leads to an increase in inflation and interest, but a negative effect on output and investment. The authors highlight the price controls and that they prevent the market from operating. Nonetheless, China reacts to price shocks, however this is significantly dampened as a result of price controls. They show that price elasticity of demand for energy is very low in the short-term, but does note that in the longer term the consumption of energy is more sensitive to higher prices.

A number of studies have calculated demand elasticities. Cooper (2003) estimates the price elasticity of demand for 23 countries in the period 1971-2000 and established the demand for oil to be price-inelastic in the short term and ranging from -0.016 (Finland) to -0.109 (Iceland). In the long run countries experience greater responsiveness to price. Cooper does not consider asymmetric decomposition. Dargay (1992) looked at the UK elasticities between 1960-1990. Price elasticities were found to react significantly to the 1979-81 oil price rise, with a short-run elasticity of -0.5 deviating from a mean elasticity of -0.1. Nonetheless, the period saw great increases in energy efficiency. The study notes there would need to be a large price rise to incite such changes in energy demand. One could consider if this has been the case in the recent price shock.

A study by the IMF demonstrated the economic transmission of an unexpected \$5 increase in oil prices. The study applies the global economic model MULTIMOD to establish the effects. It discovered that the price shock would cause a 0.3 per cent decline in global output in the two years after the event and the effect would be 0.4 per cent in the developed economies of the USA and European Union (IMF, 2001). This indicates a difference in developed economies response which could be attributed to greater reliance on energy intensive technology and a slower rate of adjustment.

According to Fisher-Vanden *et al.* (2004) China has experienced energy efficiency gains since its economic liberalisation in the 1970's. The scope for efficiency gains is present in a developing nation, as manufacturing processes have not yet reached maturity. They report a gain of 30 per cent in efficiency of oil products, and to an extent attribute this to rising energy costs. Hang and Tu (2007) validate the results through elasticities, noting pre-1995 oil price elasticity was -0.544 and post-1995 at -0.059; confirming the declining ability to adjust for prices as the ability for further energy efficiency gains approaches a minima. The makeup of the Chinese economy as a newly industrialising nation could have caused a disparity from industrialised economies and consequently may react uniquely to an oil price shock.

Hamilton (2009) considers the recent oil price shock in the United States. The paper provided a comprehensive theoretical and empirical analysis of the recent price shock. It argued that demand was the main driving factor in the recent episode, yet this was multiplied due to speculative activity. It is estimated that the increase in oil prices caused real US GDP growth to be 0.7 per cent lower than expected during 2007Q4 - 2008Q3. A main cause of price rise is attributed to increased demand from developing countries, yet Hamilton does not analyse these nations.

Aastveit *et al.* (2012) considers the effect of the price shock on a global scale. They find that different geographical regions have varying responses to oil shocks, with the developed economies suffering greater negative shocks. They report that European and US economic activity is permanently reduced by an oil shock but in emerging economies activity falls to a lesser degree and in some cases a temporary increase in activity is experienced following a supply shock. Yet, the approach is time-series and not a cross-country pooled approach, hence a shortcoming in the literature exists.

3. Determination of Oil Price

3.1 Recent Price Volatility

Energy price shocks are determined by a number of factors which synthesise in a complex transmissions channel. Oil prices during the period of study have seen great volatility, entering into what could be regarded as the third epoch of crude oil pricing (Dvir and Rogoff, 2010). Figure 1 illustrates the clear structural breaks in oil pricing since 1861. Epoch I is characterised by price instability as the new discovery of oil is commercialised and unsteady supply is prevalent. Epoch II is the era of 'cheap' energy, steady prices, and the formalisation of the oil market occurred which resulted in a stable pricing policy. It is Epoch III, which is of great interest. During this period the market experienced high levels of volatility. The 1980 price shock was followed by what appeared to be a return to stability, but this was soon upturned by the price shock in the 2000's.

Volatility in the price of oil has been evident in Epoch III. There is a significant peak in 1986 following a reversal of the oil industry shock at the turn of the 1980's. The introduction of formula netback pricing took place in 1986 which links the production sale price to a crude oil benchmark price (Downey, 2009). The introduction of price benchmarking could be seen to have remedied the issue of mispricing.

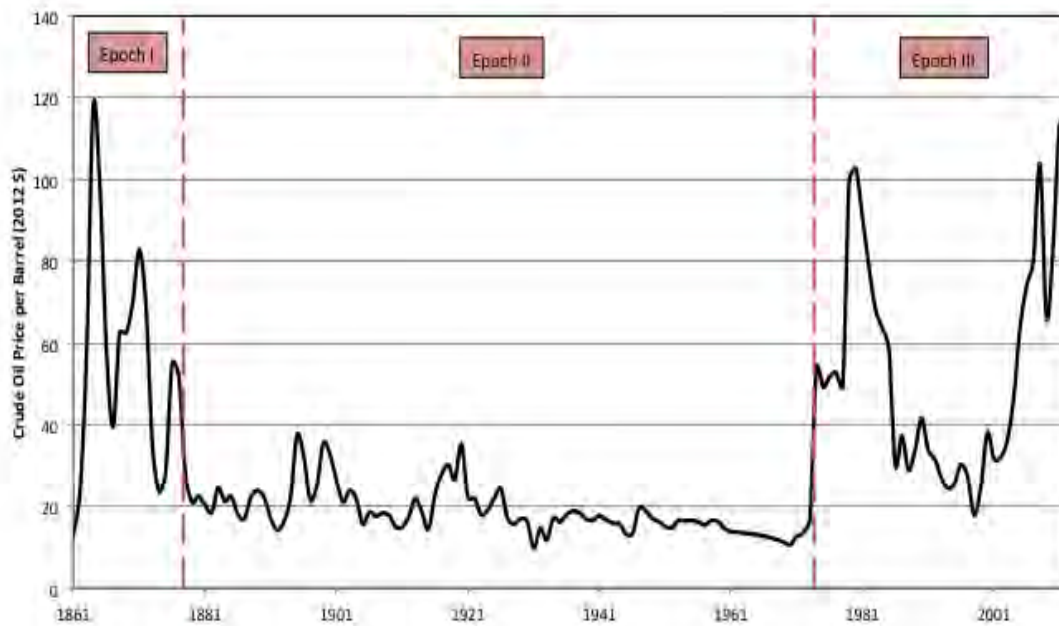


Figure 1: Historical Prices 1861-2012 separated into three epochs.

Data Source: BP (2014)

The recent price shock is of particular interest to this study. The real price of oil in 2008 was the highest since 1864. There have been a number of explanations for the cause of this rapid surge in prices; the main theories are focused around disequilibrium in supply and demand (Cantore *et al.*, 2012). Emerging economies are increasing their demand for oil; at the same time advanced industrialised economies are reducing consumption (Rubin, 2012). Nonetheless, global oil consumption is increasing year-on-year (BP, 2013). The recent shock is primarily characterised by a rapid increase in demand as opposed to a supply shock. This is in stark contrast to previous oil shocks (Hamilton, 2009). The remainder of this chapter will consider the factors using the framework of Cantore *et al.*, (2012) and apply them to the 2008 price shock to establish the key driving factors.

3.2 Supply

The global supply of oil is deemed not to provide a limiting factor in meeting projected demand till 2035 (IEA, 2012). According to the *BP Statistical Review of World Energy* in the past decade global proved reserves have increased by 26 per cent; and the R/P ratio¹ has increased to 52.9 years from 48.3 years (BP, 2014).

The increase in global proven reserves has arisen in 70 percent of cases from revisions to estimations of already discovered fields. As technology improves a high recoverable rate is

¹ R/P ratio is the “reserves-to-production” ratio and represents the length of time production could continue based on the current reserves and production rate.

achievable (IEA, 2012). New fields are no longer solely conventional oil wells. Technological advances and the higher energy prices have lead to development of tar sands, tight oil extraction, and deep-water drilling operations (Rubin, 2012). Interest in offshore oil in the Artic has been described as the “final frontier” (IEA, 2012, p. 110); the level of production is low and the technical and environmental risks are recognised. As conventional routes of supply change, the oil majors must adapt to supply the ever-present market demand.

As previously conjectured, the recent oil shock is assumed to have occurred due to disequilibrium in the market fundamentals. Researchers have suggested the fundamental cause of the price shock was not due to a specific supply shock as witnessed in previous oil price shocks (Hamilton, 2009). Consideration of the level of output by OPEC and non-OPEC countries respectively highlights a stalling of oil production during the period 2004-2008 in comparison to the forecast. Kaufmann (2011) hypothesises that the halt to production in non-OPEC countries caused a reduction in OPEC spare capacity. Production hiatus from the non-OPEC producers was not expected as Kaufmann (2011) reports the US Energy Information Administration expected that non-OPEC supply would increase by approximately 9 million barrels daily (EIA, 2005). A slowdown in oil supply would lead to a situation of shortage and a rise in prices.

During the oil price shock a number of global events could have influenced the supply of oil causing negative supply shocks. Table 1 outlines major events, which contributed to supply disruption. The potential of conflict over Iranian nuclear programme in 2007-08 could have influenced the price of oil significantly; it was estimated that the effect of intervention would have caused prices to increase by one-third (te Velde, 2007). Despite deviations in supply this has merely caused temporary shocks to the production schedule which has had limited supply side effect. The principal trend has not been one of supply reduction but rather a failure to increase production (Hamilton, 2009).

Table 1: Geopolitical disturbances and their impact on supply
Source: Hamilton (2009, 2011) and BP (2014)

Date	Event	Impact on Supply
2001-onwards	War on Terror	Iraq war cut 2.2 MBD during April-July 2003
2003	Venezuela Unrest	General strike cut supply for two months by 2.1 MBD on previous year
2005	Gulf of Mexico Hurricanes	Reduction of 0.064 MBD on previous year Damage to offshore oil platforms
2006-08	Nigerian Conflict	Cut of 0.5 MBD during turmoil

Saudi Arabia has an influential position as the world's largest crude oil reserves (BP, 2014). Saudi exports oil to a diverse and varied consumer base including Europe, US and the Far East. The Kingdom maintains a large reserve of crude oil, which has compensated global supply in times of supply interruptions (Huntington *et al.*, 2012). Yet, the Saudi production was 850,000 barrels a day lower in 2007 than in 2005 (Hamilton, 2009).

In addition to stagnation of crude oil production a shortage of refinery capacity emerged in the early 2000's. This was following a long period of surplus in refinery capacity since the 1970's, which maintained low refining margins. At the turn of the millennium, refinery shortage was an issue, which increased refinery margins during the time of already rising prices (Downey, 2009). Refinery capacity was soon increased following a number of construction projects, and once again spare capacity is realised so that supply shocks can be softened. However, as Downey (2009) points out, while capacity in refining may no longer be an issue, extracting the crude may be a different problem.

In the lead up and during the recent oil price shock, there was no significant long-term decline in production of oil, rather a slowing of growth in production. The transient deviations in oil supply as a consequence of geopolitics may have caused short-term market deviations in price. Nonetheless, production inertia was the primary trend in oil supply and this is significantly different to any previous oil price shock.

3.3 Demand

Since the 1980's global oil consumption has increased at a staggering rate. There has been a doubling in oil demand from non-OECD countries since 1980, representing an additional 24 million barrels of oil a day (BP, 2014). The compounding growth of non-OECD demand is shown in Figure 2. Projections by Petrie (2011) estimate an additional 1.4 million barrels per day on average will be demanded until 2015. However, one must note over the same time basis OECD countries are seen to be reducing oil demand. This is especially evident between 2005 and 2009 and indicated potential differences in elasticity between the two country groups.

The growth of oil demand particularly in the Chinese market has been postulated to be one of the main causes of the rapid rise in oil prices during the 2000's (Hamilton, 2009; Kilian, 2009). The growth in Chinese oil demand has looked not short of exponential in the past-two decades. China accounts for 30 percent of the increased global consumption since 1980 (BP, 2013). One of the main constituents of their increased demand requirements is that of growing domestic consumption, with citizens now having disposable income (Rubin, 2012). Consequently, there have been large increases in personal transport, sales of cars and light commercial vehicle now exceed 20 million cars per year (Petrie, 2011). The increased use of oil in transport fuels for China is a very significant change to the global oil market, and significantly adjusts the worldwide distribution of oil.

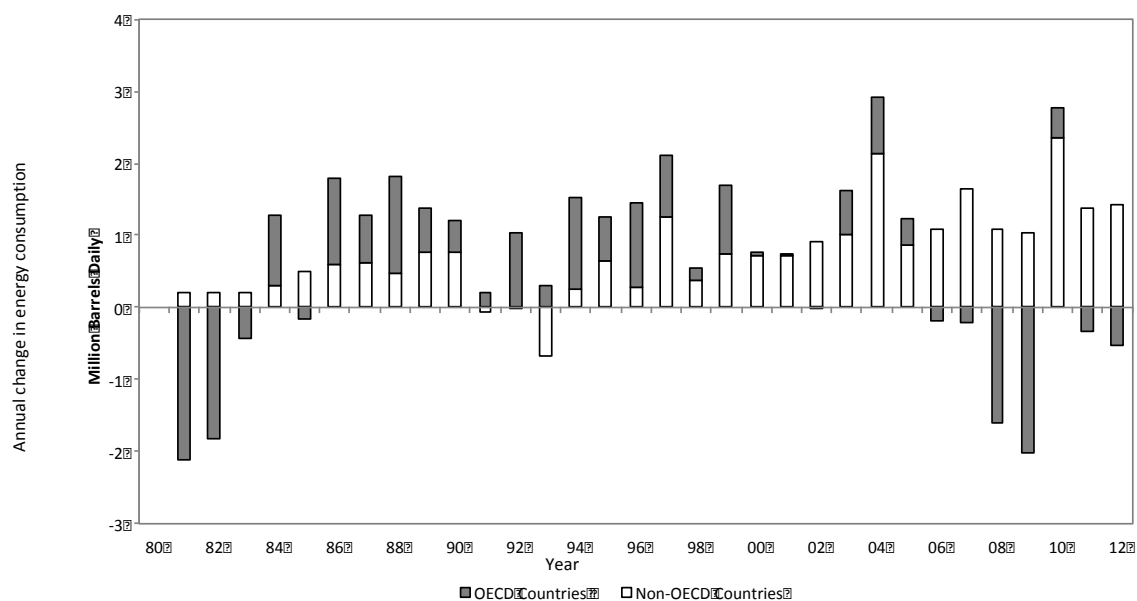


Figure 2: Annual change in oil consumption OECD vs. non-OECD countries.

Data Source: BP (2014)

Decomposition of the recent oil price shock reinforces the importance of demand from developing countries. It is estimated that the emerging markets added 55 dollars in real terms to the peak price of oil in 2008 (Aastveit *et al.*, 2012). Additionally, without reduced demand from developed countries the price would have risen even higher. Countries have heterogeneous tolerance to oil prices and are able to tolerate oil price rises to a certain point, after which they will adjust consumption patterns. Table 2 shows the estimates of this price. Despite the reliance on oil in the short term as experienced by developing countries, it is the EMEA (Europe, Middle East, and Africa) which is able to tolerate the highest prices. Therefore, the reductions in demand from developed countries over the price shock could be due to other factors such as energy efficiency or substitution to other forms of energy.

Table 2: Peak tolerable oil price.

Source: Petrie (2011)

Location	Peak Price (Barrel USD)
US	145
Eurozone	120-130
Emerging Asia	120-130
Latin America	130-140
EMEA	150+

Population is expected to grow by one billion in the next decade and Petrie (2011) argues population will be a definite driver of oil demand. Currently, 20 per cent of the world's population account for 50 per cent of global energy demand. If the remaining 80 per cent increased their energy demand per capita to the same level there would only be twenty-five years of reserves (Wagner, 2008). Economic growth is assumed to have a linear relationship with per capita oil demand *ceteris paribus* (Huntington *et al.*, 2012). The industrialisation of non-OECD economies has increased their domestic energy requirements, as industrial expansion is needed to sustain growth and based on estimates demand will increase at least as fast as GDP (Medlock and Soligo, 2001). Galli (1998) observes a correlation between energy use and economic growth in non-OECD countries, but the converse is observed in OECD countries. This non-monotonic relationship can be approximated with a squared-income term. However, recent studies have estimated that the linear relationship for developing countries still holds. This is explained due to increasing energy intensity during economic development (Bentham and Romani, 2009).

Growing oil demand may have been caused by unexpected high rates of GDP growth in non-OECD countries mainly in Asian economies, but also notably high growth in some OECD members, especially Japan. The growth rates of economies were inadequately forecast and are biased downwards. Therefore, the Chinese GDP growth was on average 0.12 per cent above *a priori* expectations each month, which is argued to have compounded over time and prompted large adjustments in global oil demand (Kilian and Hicks, 2009). As the growth rate was unexpected, there would be insufficient supply to meet the demand something which was experienced in the recent shock. Thus, in a situation of economic shortage of a commodity, theory suggests prices rises will be encountered (Barksey and Kilian, 2002).

The recent price shock is often regarded to have occurred as a result of a demand shock. Indeed, development in non-OECD countries has been extensive and this is expected to have contributed to increased oil consumption. Yet, one must consider if different countries have different experiences to the recent price shock. If the work of Kilian and Hicks (2009) is considered to hold, unexpected growth in demand would have been a major contributing factor in explaining the recent price shock.

4. Methodology

4.1 Methodological Advancements

Energy consumption models came to the forefront following the oil crisis of the 1970's. A key study was that of Pindyck (1979), the paper analysed global energy demand and estimated demand models for different oil products. Pindyck recognises the issue of data availability and reliability in developing countries. Consequently a simple log-linear model for price and income incorporating a lag term to account for the dynamic adjustment is employed. The use of a lagged term overcomes the potential model breakdown in periods of unstable energy prices (Westoby and Pearce, 1984). The model postulated by Pindyck (1979) has been used in many estimations of income and price elasticity of oil (e.g., Dargary (1992);

Gately and Huntington (2002); Cooper (2003)). Pindyck found that developing countries have high income-elasticity but low price-elasticity, thus economic growth fuels demand for energy and predicts that higher prices would impede growth prospects. This early paper recognises that energy usage is strongly correlated to the economic growth rate.

The development of econometric analysis enabled the application of pooled models and provided a better understanding of demand patterns as the use of a pooled data model can substantially increase the efficiency of an econometric model. Adelman (1984, p. 5) suggests that the global oil market is “one great pool.” Baltagi and Griffin (1997) pooled data for gasoline demand for 18 OECD countries and establish that pooling data sets increases the model efficiency especially over a longer period. However, as their data is across a set of comparable countries, OECD nations, one must consider if the pooled data model would work across a more global data set.

Fawcett and Price (2012) use a global pooled model with a more heterogeneous group of countries during the period 1984-2009 to provide a robust estimation. In order to establish the differences in groups of countries, they run cross-country panel data models for G7, remaining OECD, developing Asia and Latin America. Estimation of different panels can reduce the cross sectional heterogeneity which arise from pooling dissimilar economies following Pesaran *et al.* (1995). This study does not undertake asymmetry decomposition and therefore the effect of the shock is not established.

Despite the amass of research to date, there has been little research which considers specifically what effect the 2007-08 oil shock had on countries with varied levels of development. Research that encompasses the period of the price shock is limited as it only considers a time-series model for example Aastveit *et al.* (2012). The literature has not yet considered an asymmetric decomposition during the recent period of unstable pricing.

4.2 Conceptual Model

Considering previous research the most efficient modelling technique is to follow the conceptual ideas of Gately and Huntington (2002), van Benthem and Romani (2009), and Fawcett and Price (2012). Not only does the small number of variables increase the likelihood of data availability in developing countries, it also allows the results of study to be compared to previous research. Therefore, it is assumed oil demand can be represented as:

$$D_{i,t} = f(Y_{i,t}, P_t, \theta_i) + \varepsilon_{i,t} \quad (1)$$

Table 3 highlights the variables to be used and provides a summary of the symbols which will be employed in this paper. The dependent variable of per-capita oil demand is employed to account for the natural requirement of energy due to the size of the population. As noted previously, population growth is one of the key drivers of oil demand (Petrie, 2011). Consequently, a considerable number of studies model per-capita oil demand (Pindyck, 1979; Cooper, 2003; Fawcett and Price, 2012). Modelling demand as the dependent variable in logarithmic form provides elasticities of demand.

Table 3: Variables and symbols used

Variable	Symbol	Description	Data Source
Demand	$D_{i,t}$	Natural logarithm of per-capita oil demand, in country i in year t	BP (2014)
Income	$Y_{i,t}$	Natural logarithm of GDP per-capita (current US\$), in country i in year t	World Bank (2014)
Price	P_t	Natural logarithm of the real price of a barrel of oil	BP (2014)
Country	θ_i	Country fixed effect variable	N/A
Random Error	$\varepsilon_{i,t}$	Random stochastic error term	N/A

We consider the effect of oil demand during a period of price shock. Accordingly, a measure of oil price is included in the models. BP produced a table of crude oil prices at 2012 prices based upon the most commonly traded spot price during that period. As such, across this study 1980-1983 uses Arabian Light prices and from 1984-2012 Brent spot prices are employed. The panel assumes the price of oil to be exogenous to each country. Therefore no individual country ‘pump’ oil prices is considered which helps avoid the problem of endogeneity. Fawcett and Price (2012) utilise this assumption to combat the downward biased price elasticities as suggested by Kilian and Murphy (2010). Oil prices are expected to have a negative relationship with demand, consistent with demand theory.

The income variable is measured as the gross domestic product (GDP) per capita and provides a cross-country comparison of relative wealth. Data is obtained from the World Bank and the variable is calculated as current US dollar prices to account for inflationary differences. In a review of oil price elasticity studies, it is found most major studies have employed an income term using GDP or GNP per capita (Hoffman, 2012).

4.3 Model Specification

As specification can significantly adjust the parameters obtained (Dahl, 1993), a number of forms will be estimated.

4.3.1 Static Linear Model

A panel data model approach based on the work of Gately and Huntington (2002) is employed in this study. Initially, the relationship is to be estimated as:

$$D_{i,t} = \alpha_0 + \alpha_1 Y_{i,t} + \alpha_2 P_t + \theta_i + \varepsilon_{i,t} \quad (2)$$

The simple nature of this specification enables the elasticities of income and price to be estimated with ease as:

$$\epsilon_Y = \alpha_1 \text{ and } \epsilon_P = \alpha_2 \quad (3)$$

Income elasticity estimates the responsiveness of oil demand to economic growth rate. Price elasticity is expected to be negative. As the price of the commodity rises, demand theory suggests consumption will fall. Developed countries are more likely to be sensitive to prices than developing countries as they are more mature in their energy profile.

4.3.2 Dynamic Linear Model

The addition of a lagged dependent variable for one period allows for estimation of long-term price elasticity as it considers the speed of adjustment to price and income changes. The addition of a lagged term gives a dynamic model that can “provide robust forecasts” (Bhattacharyya and Timilsina, 2009, p. 44). As Gujarati (2003) notes dynamic models can account for psychological, technological and institutional factors. The oil market is subject to effects of the same factors. Psychological effects occur due to inertia in demand. Technology needs time to adapt to price changes due to the widespread use of oil. Institutional obligations in the form of futures contracts influence future demand. Gujarati (2003) states that short-run elasticities will therefore be significantly lower than the long-run counterparts. The addition of a Koyck-lag to the demand term as a lagged dependent variable specifies the elasticities to represent the short term (Koyck, 1954). The dynamic model is:

$$D_{i,t} = \alpha_0 + \alpha_1 Y_{i,t} + \alpha_2 P_t + \gamma D_{i,t-1} + \theta_i + \varepsilon_{i,t} \quad (4)$$

The addition of the lag to equation (2) enables more flexibility as it accounts for adjustments occurring overtime. The lagged term does however assume that the adjustment speed for both price and income is equal. In order to estimate the long run elasticity for equation (4), we divide the short-term elasticities by $1 - \gamma$ (van Benthem and Romani, 2009).

4.3.3 Asymmetric Decomposition

Some studies highlight the need to consider asymmetric response or imperfect price-reversibility (Mork, 1989); Hogan, 1993; Gately, 1993). A pooled model for the OECD oil demand rejected the use of a symmetric model over the period 1966-1990 (Hogan, 1993). Hogan notes that during the sharp drop in prices in 1986, one could expect a much larger demand increase than experienced. We expect a similar result following the 2008 price fall. Gately and Huntington (2002) illustrate the expected results for demand response for both symmetric and asymmetric reactions. Using real price of oil versus oil demand per capita over the period 2004-2012, diagrams in Figure 4 resemble the stylised functions in Gately and Huntington (2002) and show evidence to proceed with asymmetric decomposition.

Decomposition of price into three monotonic series follows a method used in agricultural commodity analysis in Gately (1993) noting that prior to this approach, structural breaks were used which have less explanatory power. The use of asymmetric price effects also controls for instances of technological improvement and acts as a proxy for energy saving technological change (van Benthem and Romani, 2009). Therefore, we use a similar decomposition as in Gately (1993). We expect to find greater response to prices rises than to price decreases. Gately and Huntington (2002) undertake a decomposition of income to consider asymmetry to income changes. This is of interest in this paper due to the occurrence of the global financial crisis during the period of study.

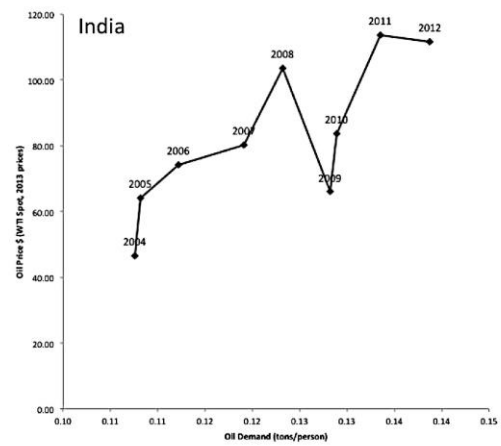
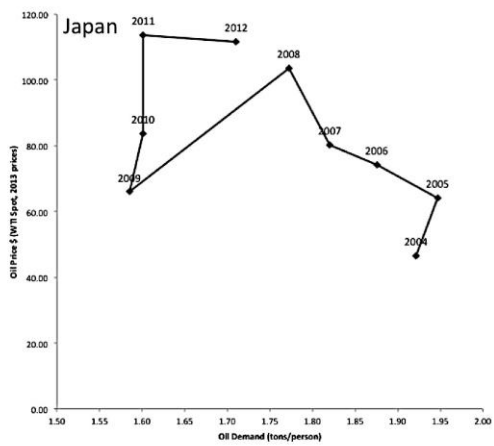
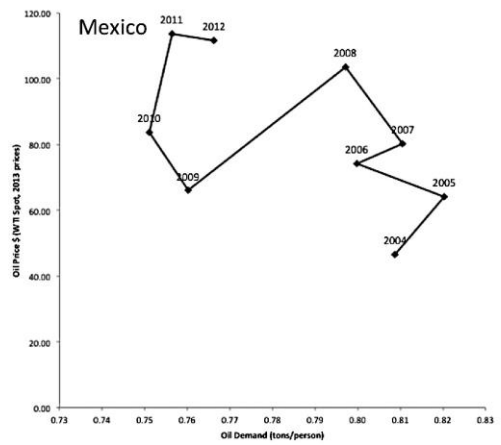
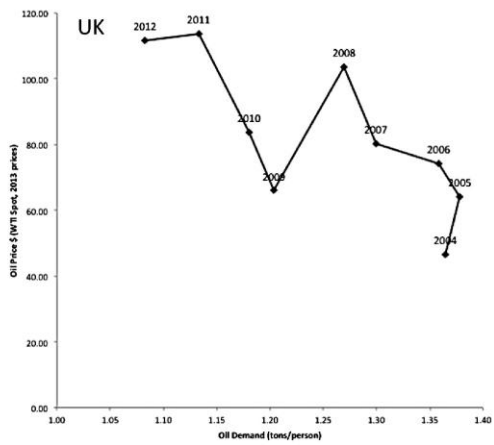
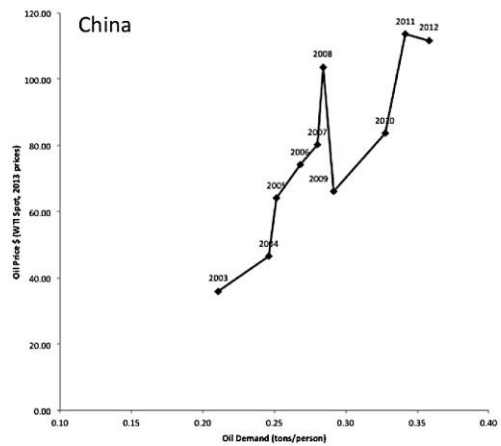
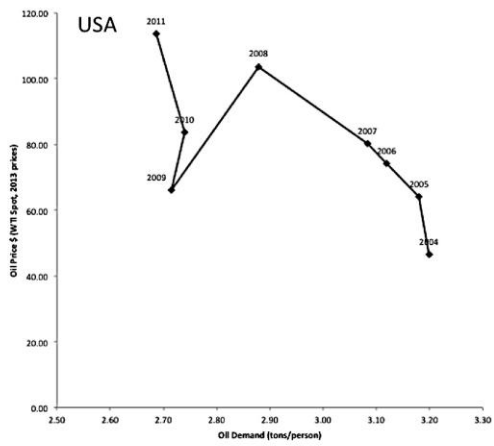


Figure 4: Price response functions (oil price versus demand).

Source: Author panel data

In order to estimate asymmetries a decomposition of the price and income is undertaken. In order to present the concept of variable decomposition the term X will be used. In our study X can represent either price (P) or income (Y). This technique can be applied to any time series independent variable.

We decompose in logarithmic form: the cumulative sum of increases in the maximum value, the cumulative sum of decreases (cuts) in the variable and the cumulative sum of sub-maximum increases (recoveries) in the variable. In order to establish the decomposition, the annual change in the variable is calculated. The base year is a given and is indicated as X_1 which in this study is 1980.

The variables in the decomposition are defined as:

$X_{max,t}$ = cumulative series maximum rises in the logarithmic value of the variable; monotonically increasing: $\Delta X_{max,t} \geq 0$

$X_{cut,t}$ = cumulative falls in the logarithmic value of the variable; monotonically decreasing: $\Delta X_{cut,t} \leq 0$

$X_{rec,t}$ = cumulative sub maximum rises in the logarithmic value of the variable; monotonically increasing: $\Delta X_{rec,t} \geq 0$

Therefore, a variable e.g. P or Y , can be presented in the form:

$$X_t = X_1 + X_{max,t} + X_{cut,t} + X_{rec,t} \quad (5)$$

Illustration of the decomposition clearly indicates the effect it has on the series. Over the time period of study (1980-2012), the decomposition of price is depicted in Figure 5. Equally, income decomposition can be illustrated as in Figure 6 for the UK.

Therefore, equation (4) can be now expressed to account for asymmetric prices and income. The base asymmetric specification is presented as:

$$D_{i,t} = \alpha_0 + \alpha_1 Y_{max,i,t} + \beta_1 Y_{cut,i,t} + \beta_2 Y_{rec,i,t} + \alpha_3 P_{max,t} + \beta_3 P_{cut,t} + \beta_4 P_{rec,t} + \gamma D_{i,t-1} + \theta_i + \varepsilon_{i,t} \quad (6)$$

The model will account for period of declines in price and income, which may not produce significant changes because of the nature of technological development (Hogan, 1993). Elasticities for Model (6) will be estimated using the parameter on $Y_{max,i,t}$ and $P_{max,i,t}$, as in Gately and Huntington (2002).

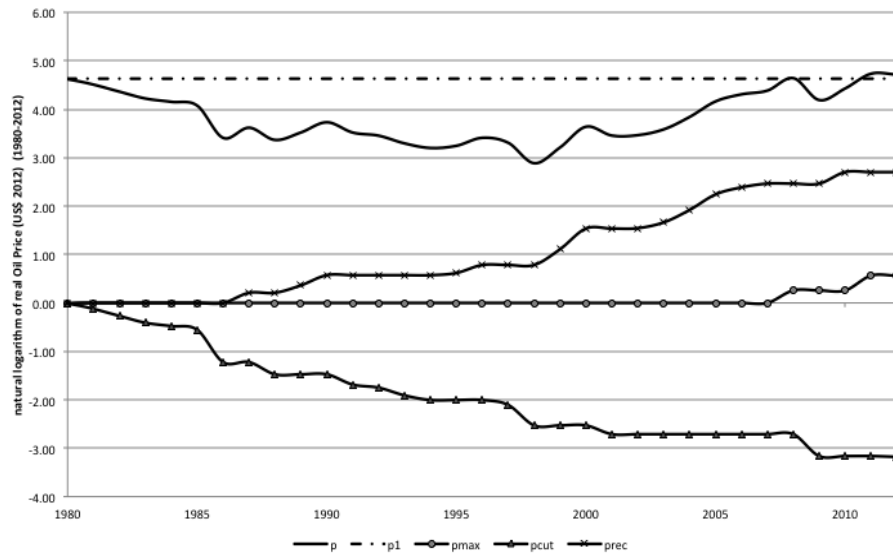


Figure 5: Decomposition of oil price in the period 1980-2012

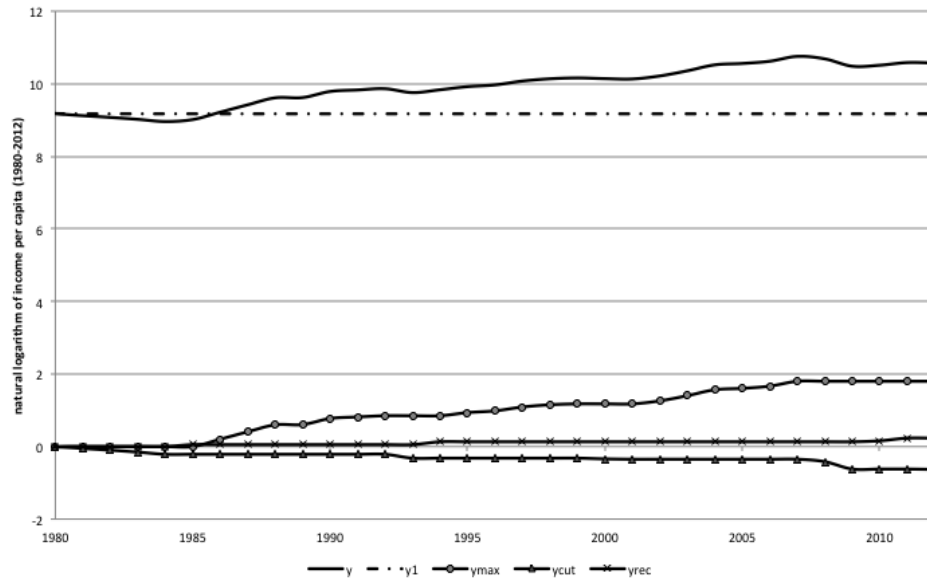


Figure 6: Decomposition of UK income in the period 1980-2012

4.4 Estimation Technique

Following Gately and Huntington (2002) we analyse multiple specifications of the model to compare the elasticities obtained. In order to establish difference between developed and developing nations a separate panel will be estimated for each group of countries. The definition of developing and developed countries for the purpose of this paper is as defined in Appendix 1, the decision for classification is based upon the average GDP per capita in that country over the period of the study. The use of two separate panels not only improves the efficiency of estimation over separate time series regressions, but also helps to correct misspecification caused by cross-sectional heterogeneity (Fawcett and Price, 2012).

We use Ordinary Least Squares (OLS) method to estimate our models. Hamilton (2003) suggests that OLS provides high quality estimation for oil demand models. In addition, cross-country fixed effects will be used. This estimates a different intercept variable (constant) for each country and is used to regulate static differences in countries (van Benthem and Romani, 2009). The intercept in a fixed effects model (FEM) is assumed to be time invariant, and consequently the slope coefficients do not change over time (Gujarati, 2003). One may consider each country to have a heterogeneous starting demand. This would theoretically fit with the FEM. The suitability of this approach is tested using the Redundant Fixed Effects – Likelihood Ratio.

Initially, symmetric static models will be estimated and then a dynamic element will be introduced through the introduction of a lagged term. If this is statistically significant it will provide further insight into the responsiveness of oil demand. The use of asymmetric decomposition for both price and income will be considered. The suitability of the approach will be tested using a Wald test. The preferred model for each panel will be established after consideration of the economic merit and statistical significance.

A dynamic panel model with fixed effects can cause bias in the results (Webb, 2006). Individual (country) FE is suggested to give unreliable estimates as the number FE estimators tend to infinity when time period is fixed (Nickell, 1981). The FE yields significant importance in this analysis as it controls for bias, which could occur from omitted variables (Wooldridge, 2010). Even through the separation of country groups there still exists heterogeneity between the country groups. Estimations without FE produce elasticities that are unrealistically large. However, when the panel is sufficiently large in time dimension, e.g. when $T=30$, average bias is reduced drastically, although the model is not the most efficient estimation (Judson and Owen, 1999). Consequently, FE is suggested to be the most suitable approach for estimation.

An alternative panel data estimation, a first-differenced generalised moment model (GMM) instead of using OLS, could increase model efficiency (Judson and Owen, 1999). Arellano and Bond (1991) proposed the GMM method for dealing with bias in dynamic panel data models. This has been applied to a study of OECD energy demand, but found similar elasticities to previous studies (Lui, 2004). Therefore, despite the potential for bias the country FE approach used in this model has been suitable and explains the prevalence of previous studies following this estimation technique.

4.5 Data

The data set employed consists of 562 observations for 16 countries over the period 1980-2012. The countries in the data set account for over 65 percent of oil demand in 2012; 32 percent of the consumption is from just the USA and China. Less developed countries (LDCs) in 1980 only accounted for 7.9 percent, but in 2012 made up 21.9 percent of global demand. Figure 7 shows a positive relationship between growth in income and growth in oil demand. A notable deviation from the linear trend is India. Based upon the growth in energy consumption in India, a higher level of GDP growth would be expected and signals energy

inefficiency has potentially occurred in these countries. Nonetheless, we expect to find a positive relationship between income and oil demand. Table 4 shows an overview of the data used in this study.

Figure 7: Average growth in GDP per capita and oil consumption (1980-2012).

Source: Authors' panel data

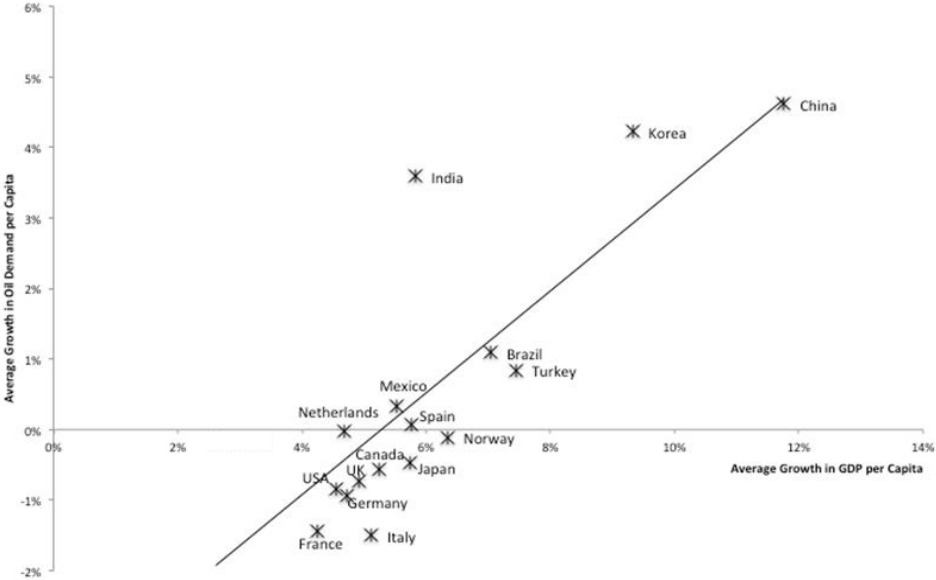


Table 4: Summary of data

Country	Oil Demand per Capita (mbd)			GDP per Capita (2012 USD)		
	1980	2012	Av. Growth	1980	2012	Av. Growth
Brazil	1,163	2,805	1.10%	1,931	11,340	7.04%
Canada	1,898	2,412	-0.56%	10,934	52,219	5.23%
China	1,690	10,221	4.61%	193	6,091	11.76%
France	3,020	2,358	-1.44%	12,500	39,772	4.25%
Germany	3,020	2,358	-0.94%	11,746	41,863	4.72%
India	644	3,652	3.59%	271	1,489	5.82%
Italy	1,930	1,345	-1.50%	8,148	33,072	5.12%
Japan	4,905	4,714	-0.46%	9,308	46,720	5.74%
Korea	476	2,458	4.22%	1,674	22,590	9.32%
Mexico	1,048	2,074	0.34%	2,763	9,749	5.53%
Netherlands	780	933	-0.02%	12,775	45,955	4.67%
Norway	199	247	-0.11%	15,595	99,558	6.34%
Spain	1,044	1,278	0.08%	6,037	28,624	5.76%
Turkey	295	685	0.84%	1,567	10,666	7.44%
UK	1,647	1,468	-0.73%	9,623	39,093	4.93%
USA	17,062	18,555	-0.85%	12,598	51,749	4.54%

5. Results and Analysis

5.1 Results

5.1.1 Symmetric Models

We first use a simple model to estimate the basic relationship in each country group is estimated (equation 2). Table 5 reports the first set of results. In the static model, developing countries have larger income and price elasticities of demand, 0.40 and -0.14 respectively. In developed countries, income elasticity is 0.01 (statistically insignificant) and price elasticity is -0.04. A clear difference in elasticities for the two panels is evident. The estimation suggests developing countries are more responsive to oil price rises. This is contrary to the literature. Yet, the model does estimate income growth to be fuelling oil demand, consistent with *a priori* expectations.

The addition of the lag term (equation 4) alters the elasticity results obtained. The long-term income elasticity falls to 0.49 in developing economies and rises to 0.11 in developed economies. Interestingly the dynamic model results do not differ greatly from the static model in developed countries, this could be due to the already mature energy profile unlike developing economies that are on a steep growth trajectory. The use of the lagged term has significant advantage for econometric analysis by accounting for long-term elasticities, and hence will be utilised in future regressions.

Table 5: Results for Models (2) and (4)

Variable	Developing		Developed	
	2	4	2	4
Equation				
Constant	-3.564 (-24.933)**	-0.2686 (-3.064)**	-0.6512 (6.217)**	0.0757 (1.7890)*
Y	0.404 (23.083)**	0.0414 (4.074)**	0.0146 (1.456)	-0.0357 (-8.824)**
P	-0.140 (-5.729)**	-0.030 (-3.617)**	-0.0384 (-3.84)**	0.8910 (39.908)**
D _(t-1)	-	0.9157 (43.216)**	-	-
FEM	Yes	Yes	Yes	Yes
Observations	198	192	330	320
Countries	6	6	10	10
R ²	0.9727	0.9976	0.9072	0.9865
R ² (adjusted)	0.9717	0.9975	0.9040	0.9859
F-statistic	967.221**	9434.766**	282.695**	1865.563**
Jarque-Bera	6.178*	8.824**	25.559**	515.418**
Durbin Watson	0.2455**	-	0.4878**	-
AR(2)	-	1.243	-	1.872
Redundant FE	166.178**	5.861**	319.300**	3.241**
Hausman RE	0.000	11.426**	0.000	23.876**
ϵ_Y	0.404	0.4911	0.0146	0.1128
ϵ_P	-0.140	-0.3559	-0.0384	-0.3275

i) t-statistics are in parenthesis

ii) Elasticities ϵ_Y and ϵ_P are long-term elasticities and are comparable

**1% significance level; *5% significance

The main consideration from the symmetric models is to establish the correct panel data estimation technique. The Hausman test is found to be inconclusive in equation (2) for both country groups, and the use of random effects is strongly rejected in the dynamic model. The use of fixed effects is strongly supported by the Redundant Fixed Effects Likelihood ratio for both equations (2) and (4). Accordingly, in all future estimations fixed effects will be applied as it fits statistically and theoretically with the approach undertaken.

5.1.2 Asymmetric Models

Theoretical and empirical support for the application of asymmetric response functions is established earlier. Econometric analysis using decomposition of the income and price series (equation 6) yields more noteworthy results that can be applied to the recent price shock. Table 6 provides the estimation and diagnostic tests.

The initial estimation combines both income and price decompositions in one model. In both panels the Wald test finds that the use of decomposition is not redundant. The P_{\max} term is the largest elasticity for both country groups. The difference in long-run price elasticities is less than 0.03, with both groups adjusting demand similarly. The elasticities for P_{cut} illustrate energy maturity in developed countries, with elasticity of -0.03 compared to -0.05 in developing countries. Hence if price fell by \$10/barrel, in the short-run, developing countries will *ceteris paribus*, demand an additional 1.04mbd in excess of developed economies.

Table 6: Results for Model (6)

Variable	Developing	Developed
Constant	-0.1829 (-5.008)**	0.0480 (2.738)**
Y_{\max}	0.0591 (3.338)**	0.0270 (2.056)*
Y_{cut}	0.1642 (5.685)**	0.0852 (2.668)**
Y_{rec}	0.0308 (1.049)	0.0799 (2.088)*
P_{\max}	-0.0473 (-1.468)*	-0.0573 (-2.945)**
P_{cut}	-0.053 (-4.474)**	-0.0281 (-3.649)**
P_{rec}	-0.017 (-1.363)	-0.0313 (-4.973)**
$D_{(t-1)}$	0.8641 (36.373)**	0.8731 (35.307)**
Country Fixed Effect (FE)	Yes	Yes
Observations	192	320
Countries	6	10
R^2	0.9980	0.9868
R^2 (adjusted)	0.9978	0.9861
F-statistic	7455.124**	1419.998**
Jarque-Bera	4.769*	677.616**
Redundant FE Likelihood	7.479**	3.722**
AR(2)	0.9111	1.892
Wald Test	24.606**	27.776**
ϵ_Y	0.4349	0.2128
ϵ_P	-0.3481	-0.3885

i) t-statistics are in parenthesis

ii) Elasticities ϵ_Y and ϵ_P are long-term elasticities and are comparable

**1% significance level; *5% significance

For developing countries, the largest income elasticity is when income falls (Y_{cut}) with an elasticity of 0.16, indicating that falls in income have the strongest effect on oil demand. Over the time period most income rises in developing countries are increases in the maximum value (i.e. Y_{max}). Long-run income elasticity based on this variable is 0.43. Contrasting to developed economies, increases in the maximum income are not as strong and shows maturity in the energy cycle. This yields a long-run elasticity of 0.21. Thus, the developing countries demand for oil is fuelled significantly by income growth.

The models do not pass autocorrelation tests indicating that the variables are correlated over time. This could be due to misspecification by the use of two decomposed time-series. In fact, van Bentham and Romani (2009) have issues with autocorrelation and re-specification helped reduce the effect.

5.1.3 Adjustments to the decomposed model

As previously highlighted the models suffer from autocorrelation. Accordingly, two variations on equation (6) will be presented, each with one decomposed series (price or income). Re-specification can be used to correct for autocorrelation (Gujarati, 2003). This is because inference may occur between the two decomposed series if included in one model.

$$D_{i,t} = \alpha_0 + \alpha_1 Y_{i,t} + \alpha_2 P_{max,t} + \beta_3 P_{cut,t} + \beta_4 P_{rec,t} + \gamma D_{i,t-1} + \theta_i + \varepsilon_{i,t} \quad (6p)$$

$$D_{i,t} = \alpha_0 \alpha_1 Y_{max,i,t} + \beta_1 Y_{cut,i,t} + \beta_2 Y_{rec,i,t} + \alpha_2 P_t + \gamma D_{i,t-1} + \theta_i + \varepsilon_{i,t} \quad (6y)$$

Estimation of the improved equations is presented in Table 7. The models do still suffer from autocorrelation in developing countries and non-normality in developed economies. However, the significance is lower and therefore Models (6p) and (6y) are better specified to highlight differences between the elasticities in the two panels. The non-normality in developed economies can be attributed to outliers for the UK in 1984 and 1985 which significantly deviates from the fitted line (see Figure 8). This might have been in part due to the miners' strike which caused greater dependency on oil (Ledger and Sallis, 1995). This could be controlled using a dummy variable, but the inference on the estimated parameters is low.

The partial use of asymmetry in equations (6y) and (6p) yield elasticities that indicate the substantial disparity between the two country groups for income elasticity. In equation (6y) income elasticity is 0.67 in developing and 0.18 in developed economies. The decomposition indicates short-run income elasticity to be nearly four times larger in developing countries when there is an increase in the series-maxima. The effect of a price shock that increases the series-maxima (P_{max}) is insignificant in developing countries, thus indicating the price shock had little statistical importance on consumption.

Table 7: Results for Models (6y) and (6p)

Variable	Developing		Developed	
	6y	6p	6y	6p
Constant	-0.011 (-0.364)	-0.669(-5.295)**	0.195 (8.803)**	-0.262 (-2.427)*
Y		0.081 (5.212)**		0.033 (2.748)**
Y _{max}	0.081 (6.217)**		0.021 (3.406)**	
Y _{cut}	0.141 (5.242)**		0.093 (3.029)**	
Y _{rec}	0.052 (1.873)*		0.079 (2.185)*	
P	-0.037 (-4.680)**		-0.034 (-8.217)**	
P _{max}		-0.045 (-1.354)		-0.065 (-3.456)**
P _{cut}		-0.016 (-1.595)		-0.023 (-3.485)**
P _{rec}		-0.047 (-4.203)**		-0.033 (-5.889)**
D _(t-1)	0.878 (38.513)**	0.912 (43.694)**	0.886 (39.689)**	0.867 (35.382)**
Country FE	Yes	Yes	Yes	Yes
Observations	192	192	320	320
Countries	6	6	10	10
R ²	0.9980	0.9977	0.9868	0.9867
R ² (adjusted)	0.9978	0.9976	0.9862	0.9861
F-statistic	8811.827**	7928.852**	1625.781**	1616.211**
Jarque-Bera	3.7503	5.703*	674.921**	595.079**
Redundant FE	6.5278**	7.551**	3.5999**	3.9013**
AR(2)	1.301	1.432	1.835	1.972*
Wald Test	4.960**	2.234	8.974**	48.066**
ϵ_Y	0.6680	0.9205	0.1842	0.2481
ϵ_P	-0.3032	-0.5114	-0.2982	-0.4887

i) t-statistics are in parenthesis

ii) Elasticities ϵ_Y and ϵ_P are long-term elasticities and are comparable

**1% significance level; *5% significance

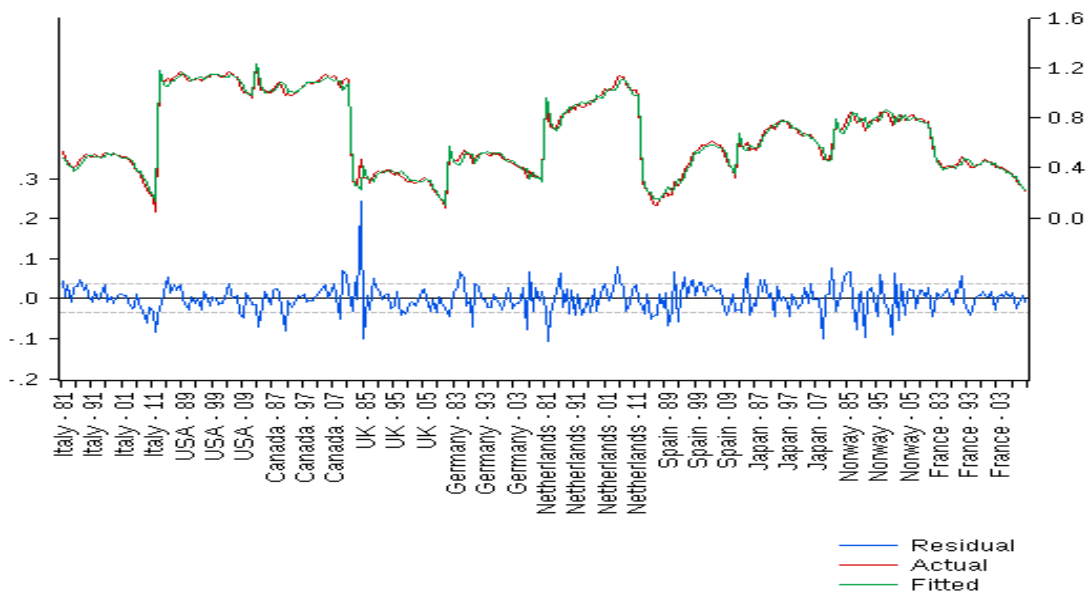


Figure 8: Residual plot for Model (6y), developing countries.

Across all regressions the income elasticity is discovered to be approximately 0.2 in developed economies and between 0.5-0.7 in developing economies. Demand has tended to adjust more to increases, rather than decreases in price, except in one instance for developing economies. The price elasticity is generally stronger in developed countries, which indicates more responsiveness to price changes. There are clear differences between the country groups and the driving factors behind their demand for oil across the time period. The heterogeneity between the two panels has implications on policy and in understanding the recent price shock.

The preferred models to be used for further analysis are Model (6y) for developing countries and Model (6p) for developed countries. In developing countries, consistent with *a priori* expectations the key driver for oil consumption has been income. Model (6y) suitably captures this and allows consideration of the income effect. In developed countries, the more stable growth has had less effect on oil consumption, thus price plays a greater role on consumption patterns. Consequently, Model (6p) will be used. The crucial elasticities are indicated in Table 8.

Table 8: Crucial elasticities

Panel	Income Elasticity	Price Elasticity
Developing	0.6680	-0.3032
Developed	0.2481	-0.4887

5.2 Income and Substitution Effect

Income elasticity is significantly closer to unity in comparison to developed economies. This indicates a dependency on oil for economic development. Yet, one must consider if oil is a key driver in development, why does developing countries consumption suffer significantly less in the event of an oil price shock? One might consider this in the context of energy intensity and the structure of industry in the countries.

Analysis of the decomposed price Model (6p) shows an insignificant parameter for P_{max} . This indicates that developing countries may have been unaffected by the price shock; as the P_{max} term is only activated in this model during the period of price shock. Additionally, considering the model with both series decomposed (6) in the lead up to the price shock (P_{rec}) short-term demand adjustment is only 1.7 percent. This is nearly half of the developed economies response.

The strength of the income effect in developing countries indicates that they are less affected by the price of oil. Calculation of the Slutsky equation for each panel provides indication of the compensated demand for oil, and thus indicates the relative income and

substitution effects.² The results in Table 9 indicate the income effect is 6.3 times stronger in developing countries showing income growth is fuelling oil demand. Additionally, the substitution effect is 2.08 larger in developed countries and indicated a larger responsiveness to the price shock.

Table 9: Income and substitution effects (derived from crucial elasticities)

Country Group	Oil Spending (% Income)	Income Effect	Sub. Effect
Developing	11.09	0.074	-0.2291
Developed	4.77	0.011	-0.4778

5.3 Explanations for Variation in Consumption Patterns

Income in developing countries has increased on average by seven per cent in the time period studied. The compound effect of this is a doubling of national income, this major change in economic activity is sure to cause changes in consumption patterns and provides an explanation to the strength of the income effect in developing countries. Nonetheless, the current per capita oil consumption is still very low and any growth in disposable income will increase oil demand further.

Behr (2009) regards the increases in oil demand to have had a disproportionate effect on global oil prices. One of the main components fuelling the surge was the increased demand for transport fuel and in particular diesel for the Asian market (Yergin, 2008). As noted in Section 3, annual car sales in China alone now exceed 20 million vehicles. To sustain this oil demand will have to increase by 128 million barrels a year (Petrie, 2011). Projecting oil demand to 2035, the key driver of oil demand will be transport fuels and this is expected to account for 40 percent of the overall demand increase (IEA, 2012). They expect road freight in non-OECD countries to grow nine-times in part due to increased construction and industrial activity.

Growing economies also require large amounts of infrastructure creation, which is very energy intensive. Such projects are inextricably linked with income growth and this is a good indication as to why the income effect is so large in developing countries. Barclays (2012) report that most developing economies are in a stage of rapid development and creation of infrastructure, with demand outstripping supply on these projects. The construction of motorways in these countries will only make car ownership more appealing so increasing fuel demand.

One could consider that during a period of rapid economic growth in developing countries little regard was paid to energy efficient technologies. The dataset used in this paper

² The Slutsky equation allows derivation of the income and substitution effects based upon the elasticities obtained and the equation in elasticity form can be expressed as:

$$\varepsilon_{ii} = \sigma_{ii} - s_i \eta_i$$

where ε_{ii} denotes price elasticity of demand for oil (i.e. α_2), the share of oil expenditure as percentage GDP (calculated from panel), η_i is the income elasticity of demand (α_1), σ_{ii} denotes substitution effect, and income effect is $s_i \eta_i$.

shows that developing countries use 3.6 times more oil per unit of GDP than developed economies. Inefficient use of energy in developing countries caused levels of consumption that could otherwise have kept global prices down (Merrill Lynch, 2012). However, energy intensity has been declining in all regions globally. In part this is attributed to China becoming less energy-intensive recently (BP, 2013).

Asian developing countries often have price controls which prevent the transmission of rises in crude prices in order for the country to continue on an uninterrupted growth trajectory. Price controls reduce the substitution effect as there is less incentive to be more energy efficient in the face of price increases (Economist, 2008). The monetary size of fuel subsidies is large. In 2013, India spent \$23bn in reducing fuel costs (Mallett, 2014). If pump costs were reflective of the global price of oil consumers may have reduced oil demand and the price shock could have been dampened. However, as this paper uses a single price we will be unable to further analyse these effects.

In contrast, developed countries already have a mature energy profile. Oil is primarily used as a transportation fuel, with electricity production now undertaken using other fossil fuels or alternative forms of energy (Merrill Lynch, 2012). The consumer faces the brunt of oil price shocks in economies where transportation fuel is the primary use of oil, as crude prices are easily transferred to the pumps. This is the case in most OECD countries, with minimal fuel subsidies in place (Economist, 2008). Therefore, the substitution effect is likely to be greater. Indeed, the USA witnessed a slowdown in automobile purchases in 2008 in light of higher oil prices (Hamilton, 2009).

The difference in consumption patterns between the two panels is noteworthy due to three factors. *Firstly*, growth has caused a structural change to the developing economies. The move to greater personal transportation shows no sign of slowing and consumers are unlikely to reverse their lifestyle because of fuel price increase. Therefore the income effect is greater in developing countries as the middle classes seek higher standard of living. In contrast, developed countries have mature energy profiles and have little rationale to rapidly increase long-term transport fuel demand. Once a consumer has a car, there is a finite amount of driving one can do, whereas developing countries are starting to use personal cars and consumption changes will be substantial. *Secondly*, the rapid expansion may have neglected energy saving. If alternative energy forms were in place there would be a greater substitution effect from a shock in developing countries. *Finally*, oil subsidies prevent the transmission of global prices hence fuelling consumption and biasing the substitution effect downwards in the developing economies.

5.4 Policy Implications

Preventing spikes in oil prices is difficult to achieve due to the complex nature of the market as highlighted in this paper. Nonetheless, economic policy could be used in order to smooth prices based upon the work of Behr (2009) and Wurzel *et al.* (2009). This paper has a number of policy recommendations.

5.4.1 Reducing Oil Intensity

Oil intensity has already reduced significantly in China and other developing countries (see Figure 9). The indicator provides a measure of consumption per unit of output in the economy. Reducing oil intensity acts as a safeguard if oil prices do increase, as spare capacity exists within the economy (Wurzel *et al.*, 2009). They note in the event of a demand shock as witnessed in the last oil price shock lower energy intensity would have reduced the impact of rising incomes. The need to lower intensity is also a recommendation reported by Gately and Huntington (2002). This paper has found a strong link between economic development and increased oil consumption. Unless intensity reduces, the potential for future demand shocks will still exist.

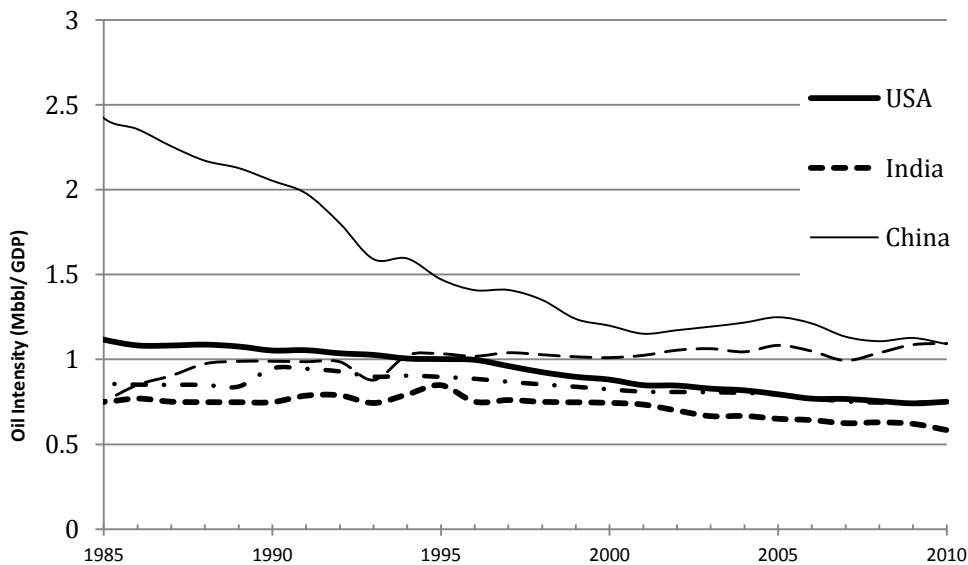


Figure 9: Global oil intensity

Source: BP (2013)

5.4.2 Recognising Regional Differences

This paper has clearly highlighted a difference in consumption patterns between the two panels. Clearly demand functions differ based on the level of development in an economy, therefore policy cannot be globally homogenous. Additionally, each panel has different responses when asymmetric decomposition is applied. If this is not taken into account the elasticities could be biased and thus incorrect policy decisions would be made. Each country has an individual energy mix, with varied uses of oil. OECD countries use 80 percent of their oil consumption for transportation. This value is around 15 percent lower in developing economies (Merrill Lynch, 2012). Consequently, blanket reduction policy cannot be applied. Therefore international economic organisations such as the IEA should target policy to enable policy address regional specific issues. The use of meaningful heterogeneous policy which reduces oil intensity will help to buffer future oil price shocks.

6. Conclusions

This paper investigated price and income elasticities in a period of global oil price volatility. Following discussion which captured the development of the literature, it was established that a scarcity of research has been undertaken in light of the recent price shock. Previous studies had introduced the use of asymmetric decomposition and cross country panel analysis, but until this study such techniques had not been applied to a dataset that covered the recent price volatility. Analysis of late has concentrated on a time-series approach and this has failed to capture the efficiency gains obtained from panel data.

Currently in Epoch III oil pricing, the previously established global trade flows appear to be changing. Fundamental economic analysis established the price rises following the millennium to have occurred as a result of disequilibrium. Supply hiatus from non-OPEC producers coupled with a number of geopolitical events resulted in supply growth stalling. The recent price shock was characterised not by a significant supply shock as in previous episodes, but rather a failure to escalate production to satisfy market demands. It was discovered that unexpected levels of demand was the primary contributing factor in causing the disequilibrium.

A dynamic cross country fixed effects specification was used in order to capture the heterogeneous intercept level of demand in each country. As the period of study was greater than thirty-years the potential for parameter bias resulting from FE was substantially reduced. Estimation of the models including asymmetric decomposition for both income and price yielded two noteworthy consumption patterns. *Firstly*, the effect of income falls is larger than income rises, with income elasticities being stronger for developing countries in all specifications. *Secondly*, when oil price rises the elasticity is larger in contrast to price falls, but this effect is only significant in developed countries.

Subsequent analysis of these observations through the use of Slutsky decomposition emphasised the results obtained in the regression. Developing countries have an income effect which is 6.3 times larger than in developed countries. Thus, in a period of significant economic growth for developing countries it is evident as to why oil consumption increased so dramatically. Moreover, the substitution effect in developed economies was 2.1 stronger than in developing countries. This explains the more responsive adjustment of consumption of developed countries in reaction to the price shock of 2007/08.

The results obtained are explainable in relation to the difference in economic structure between the two panels. Developing countries are assumed to have not yet reached relative energy maturity in relation to fossil fuels. Consumers are increasingly using oil as a transportation fuel as new found disposable income is being expended on car ownership. This explains the relative strength of the income effect in developing countries. The paper also recognises economic development has been at the expense of lower energy efficiency, again reinforcing the size of the income effect in developing countries.

Reflection of the analysis indicated a number of policy implications for the oil market. This paper recommended the implementation of oil-intensity reduction policy in order to maximise efficiency. Such policy would enable a buffering if oil prices rise as a result of a demand shock. A crucial policy consideration postulated is the recognition of heterogeneous

consumption patterns between country groups and that a one size fits all policy should not be undertaken.

There is potential for future research. This paper considered two independent panels separated by the level of economic development. However, by increasing the dataset and incorporation of more countries, alternative panels could be estimated. These could include oil importers, oil exporters and geographical regions. Granulation of the current approach would allow more tailored policy development and a more comprehensive understanding of income and price responses which would aid economic forecasting. Furthermore, research could be considered into the effect of subsidies on consumption by utilising country price data.

This paper has contributed to the existing literature and has studied oil consumption patterns during Epoch III pricing. The dominance of the income effect in developing countries has been staggering. In conclusion, oil consumption patterns during the period 1980-2012 varied significantly between developing and developed countries and failure to account for this will bias the effectiveness of energy policies.

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Appendix 1: Country Groups

Tables 10 and 11 outline the classification of developed and developing country applied in this study. The study has separated the countries based on their GDP per capita into these country groups. The difficulty in obtaining data for low income (developing country) reduces the number of available countries of analysis.

Table 10: Developing country group

Country	GDP Growth Rate	Demand Growth
Brazil	7.04%	1.10%
China	11.76%	4.61%
India	5.82%	3.59%
Korea	9.32%	4.22%
Mexico	5.53%	0.34%
Turkey	7.44%	0.84%
Average	7.82%	2.45%

Table 11: Developed country group

Country	GDP Growth Rate	Demand Growth
Canada	5.23%	-0.56%
France	4.25%	-1.44%
Germany	4.72%	-0.94%
Italy	5.12%	-1.50%
Japan	5.74%	-0.46%
Netherlands	4.67%	-0.02%
Norway	6.34%	-0.11%
Spain	5.76%	0.08%
UK	4.93%	-0.73%
USA	4.54%	-0.85%
Average	5.13%	-0.65%