Quantifying the Impact of Exogenous Non-Economic Factors on UK Transport Oil Demand

David C Broadstock and Lester C Hunt

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QUANTIFYING THE IMPACT OF EXOGENOUS
NON-ECONOMIC FACTORS ON
UK TRANSPORT OIL DEMAND

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ABSTRACT

This paper attempts to quantify the impact of exogenous non-economic factors on UK transport oil demand (in addition to income, price, and fuel efficiency) by estimating the demand relationship for oil transport for 1960-2007 using the Structural Time Series Model. From this, the relative impact on UK transport oil demand from income, price, and efficiency are quantified. Moreover, the relative impact of the non-economic factors is also quantified, based on the premise that the estimated stochastic trend represents behavioural responses to changes in socio-economic factors and changes in lifestyles and attitudes. The estimated elasticities for income, price and efficiency are 0.6, -0.1, and -0.3 respectively and it is shown that for efficiency and price the overall contribution is relatively small, whereas the contribution from income and non-economic factors is relatively large. This has important implications for policy makers keen to reduce transport oil consumption and associated emissions, but not willing to reduce the trend rate of economic growth. Taxes and improved efficiency only have a limited impact; hence, a major thrust of policy should perhaps be on educating and informing consumers to persuade them to change their lifestyle and attitudes and thus reduce their consumption through the non-economic instruments route.

JEL Classification: C22, Q41.

Key Words: Transport oil demand; Structural Time Series Model, STSM; Underlying Energy Demand Trend, UEDT; Exogenous Non-Economic Factors, ExNEF.
Quantifying the Impact of Exogenous Non-Economic Factors on UK Transport Oil Demand

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1. Introduction

As a level one student of economics is taught very early on, demand for a good or service is a function of income, price, and ‘tastes’. Tastes describe the utility a consumer gets from the goods consumed and, according to Begg’s (2009) level one text, “depend on culture, history, familiarity, relationships with others, advertising, and so on” (p. 34). Moreover, explaining the influence of ‘tastes’ is not the role of economists it is the “role of other social sciences, like psychology and sociology” (Begg, 2009, p. 34). So for economists these ‘tastes’ are often regarded as ‘exogenous’ to the model; part of the ‘ceteris paribus’ assumption behind a particular demand curve and although there is usually some discussion and analysis of how demand might be affected by a change in ‘tastes’, there is no real discussion of the importance of these ‘tastes’ relative to the economic factors.

However, given the ‘urgency’ of the energy and environmental policy agenda and the arguments put forward in some quarters for non-economic solutions it appears imperative that an attempt is made to quantify the impact of these exogenous ‘taste’ factors. Given also that carbon predominately arises from the consumption of energy
then there is arguably a need to understand what contribution behaviourally focussed interventions, such as educational schemes and other government initiatives, might have on transport oil demand. In other words what contribution do exogenous non-economic factors (of ExNEF for short) play in driving transport oil demand? ¹

This paper therefore considers the modelling of UK transport oil demand and attempts to isolate the traditional ‘economic drivers’ of income and price from the ‘exogenous non-economic factors’. A summary of elasticities from previous fuel demand studies, taken from Goodwin et al. (2004), is given in Table 1. This will serve as a benchmark for the empirical results presented in this paper, however, the results summarised in Table 1 are based upon studies in which the consumer ‘tastes’ or ExNEF were not controlled – effectively ignored. This is arguably a shortcoming given the current energy and environmental policy agenda with policy makers desperate to devise a range of workable and achievable policies to help curtail future energy demand and hence CO₂ emissions. However, a useful review of the manner in which surface transport carbon emissions reductions might be achieved through behavioural and technological change is given in UKERC (2009), further reinforcing that the ExNEF should at least feature in a general transport (energy) demand modelling framework.

<table>
<thead>
<tr>
<th>Reference</th>
<th>Elasticity w.r.t. price</th>
<th>Elasticity w.r.t. income</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Short-run</td>
<td>Long-run</td>
</tr>
<tr>
<td>Goodwin (1992)</td>
<td>-0.27</td>
<td>-0.73</td>
</tr>
<tr>
<td>Espey (1998)</td>
<td>-0.26</td>
<td>-0.58</td>
</tr>
<tr>
<td>Graham and Glaister (2002)</td>
<td>-0.2 to -0.3</td>
<td>-0.6 to -0.8</td>
</tr>
</tbody>
</table>

Notes
- taken from Goodwin et al. (2004)

¹ This work is part of on-going research in RESOLVE attempting to quantify the impact of ExNEF on consumer demand and expenditure; see, for example, Chitnis and Hunt (2009a and 2009b).
The order of this paper is as follows. Following from this introduction the next section outlines the estimation methodology, including a brief discussion of the data. Section 3 presents the results and Section 4 summarises and concludes with a brief discussion.

2. Methodology

The model is based on the structural time series model (STSM) developed by Harvey (1989 & 1997) and advocated for the use when modelling energy demand functions by Hunt et al. (2003a & 2003b) and Hunt and Ninomiya (2003) who argue that a stochastic trend, coined the underlying energy demand trend (or UEDT for short) should be estimated. They argue that the UEDT is driven by exogenous technical progress of the capital and appliance stock, such as vehicle efficiency, etc and other socio economic factors termed ‘consumer tastes’ such as driving children to school etc. (see Hunt and Ninomiya, 2003). However, here, an attempt is made to separate out the effect of technical progress and fuel efficiency from the UEDT by incorporating an appropriate efficiency variable. Consequently, the estimated UEDT here only picks up the effects of the ‘tastes’ referred to by the simple economic model above, i.e. it captures the effects of values, lifestyles, etc.

The following STSM is therefore specified for UK transport oil demand:

\[ e_{o_{i}} = \alpha_{y}^{o} y_{i} + \alpha_{p}^{o} p_{o_{i}} + \alpha_{f}^{o} f_{i} + \mu_{o}^{o} + \varepsilon_{o_{i}}^{o} \]  \hspace{1cm} (1)

\[ \mu_{i}^{o} = \mu_{i-1}^{o} + \beta_{i-1}^{o} + \eta_{i}^{o} \]  \hspace{1cm} (2)

\[ \beta_{i}^{o} = \beta_{i-1}^{o} + \varepsilon_{i}^{o} \]  \hspace{1cm} (3)
where $\epsilon_t^o \sim NID(0, \sigma_{\epsilon_t}^2)$, $\eta_t^o \sim NID(0, \sigma_{\eta_t}^2)$ and $\xi_t^o \sim NID(0, \sigma_{\xi_t}^2)$. $e_o_t$ is the natural logarithm of transport oil consumption, $y_t$ is the natural logarithm of income, $p_o_t$ is the natural logarithm of the real transport oil price, $f_t$ is the natural logarithm of car fuel efficiency and $\epsilon_t^o$ the error term. $\alpha_{\psi}^o$, $\alpha_{p}^o$ and $\alpha_{f}^o$ represent the (long-run) oil transport demand income, price, and fuel efficiency elasticities respectively.2

Equations (2) and (3) represent the UEDT (without fuel efficiency) for transport oil demand, $\mu_t^o$, made up of the two components, the level and slope respectively. This is a stochastic trend dependent upon the variances $\sigma_{\eta_{t}}^2$ and $\sigma_{\xi_{t}}^2$ (also known as the hyperparameters), the larger the hyperparameters the greater the stochastic movements in the trend. In the limiting case when the hyperparameters are equal to zero, the model collapses to a conventional deterministic time trend regression. This therefore gives a number of alternative forms of the stochastic trend depending on the values of the hyperparameters.3

The initial model to be estimated therefore consists of an Autoregressive Distributed Lag version of Equation (1) with lags of 4 years plus Equations (2) and (3), over the period 1964-2003, keeping 2004-2007 for predictive failure tests. All disturbance terms are assumed independent and uncorrelated with each other. The estimation is carried out by maximum likelihood and the hyperparameters are obtained from a smoothing algorithm using the Kalman filter. The preferred parsimonious

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2 When actually estimating the model, a general autoregressive distributed lag model is estimated and the preferred parsimonious model found by testing down using the ‘general to specific’ methodology. However, for the preferred models presented below the parsimonious model is always that given by Equation (1).

3 A classification of the different types is given in Table 9.2 in Hunt et al. (2003b).
specification of the model is selected by testing down from the general equation by eliminating statistically insignificant variables, whilst ensuring that the equation residuals (similar to those from ordinary regression) pass a range of diagnostic tests, a set of auxiliary residuals (irregular, level and slope) do not suffer from non-normality, and the model passes predictive failure tests over the 2004-2007 period. The preferred specification is then re-estimated over the longest data period available, up to 2007 and as far back to 1960 as possible, depending on whether any lag variables remain in the model. The software package STAMP 6.3 (Koopman et al., 2000) is used for all estimation.

Using the results over the longer period for the preferred specification, the contributions of income, price, fuel efficiency and ExNEF to the annual change in oil demand are constructed from the following:

$$\Delta eo_t = \hat{\alpha}^o_t \Delta y_t + \hat{\alpha}^o_p \Delta po_t + \hat{\alpha}^o_f \Delta f_t + \Delta \hat{\mu}_t$$

(4)

where $\hat{\alpha}^o$, $\hat{\alpha}^o_p$, and $\hat{\alpha}^o_f$ are the estimated income, price and fuel elasticities respectively and $\Delta \hat{\mu}_t$ the estimated UEDT. Therefore, $\hat{\alpha}^o_y$, $\hat{\alpha}^o_p$, $\hat{\alpha}^o_f$, and $\Delta \hat{\mu}_t$ represent the estimated contributions to the change in UK transport oil demand (in logs), from income, the real price of oil, car fuel efficiency, and ExNEF respectively.4

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4 Note, that given that the estimated preferred models presented below do not have any lags the contributions are easily calculated in this way. Furthermore, given the model is in logs the change approximates the percentage change.
3. Data and Results

Data

The UK annual data set covers the period 1960-2007 with the energy consumption and price data taken from the Digest of UK Energy Statistics (DUKES) and the income data from the UK Office of National statistics (www.statistics.gov.uk). These are supported with data on total vehicle kilometres travelled in the UK taken from the UK Department for Transport (http://www.dft.gov.uk/pgr/statistics/) used to derive the efficiency term. The variables used in estimation are as follows:

\[ EO = \text{total transport oil consumption in thousand tonnes of oil} ; \]
\[ Y = \text{GDP at £million 2003 market prices} \]
\[ PO = \text{real price of transport oil consumption in 2003£/litre}^5 \]
\[ F = \text{car fuel efficiency in miles per gallon (mpg)}^6 \]

The price series for gasoline demand is made up of a number of separate grades of gasoline. Therefore, a quantity-weighted average of the series is defined including the following components: 2 Star (1960-1989), 3 Star (1967-1989), 4 Star (1960-2005), 5 Star (1961-1979), Super Premium Unleaded (1990-2005) and Premium Unleaded (1988-2005). Noting that a number of the fuel types of the period either/both entered or left the market for various reasons. Unique price series are not available for 3 Star and 5 Star fuel, therefore it is assumed that they have the same prices as 4 Star fuel, which is much closer in quality than 2 Star fuel. Using superscripts to denote the different components of gasoline, the calculation of the nominal weighted average series can be expressed as;

\[
\frac{[(p_i^{2,\text{star}} q_i^{2,\text{star}}) + (p_i^{4,\text{star}} q_i^{4,\text{star}}) + (p_i^{5,\text{star}} q_i^{5,\text{star}}) + (p_i^{\text{SPU}} q_i^{\text{PU}}) + (p_i^{\text{SPU}} q_i^{\text{SPU}})]}{[q_i^{2,\text{star}} + q_i^{3,\text{star}} + q_i^{4,\text{star}} + q_i^{5,\text{star}} + q_i^{\text{PU}} + q_i^{\text{SPU}}]}
\]

This was then deflated by the GDP deflator at 2003 prices to give the real price of gasoline (PO) at 2003£/litre.

The fuel efficiency is expressed as mpg; however, it should be noted that some authors measure this in litres per 100km; although many previous studies do use as mpg as the efficiency metric, as shown in Table 2 in Bonilla and Foxon (2009, p. 67),
Figure 1: UK Transport oil consumption
a) $EO_t$ (million tonnes)

Figure 2: GDP
a) $Y_t$ (£million 2003 Market Prices)

Figure 3: Real price of UK transport oil
a) $PO_t$ (2003pence/gallon)

Figure 4: Average vehicle fuel efficiency in mpg($F_t$)
 a) $F_t$ (mpg)
The data for oil consumption and the economic drivers (income and price) are shown in Figures 1 to 3. Figure 1a illustrates the general strong growth in UK transport oil consumption through the 1960s, 1970s and 1980s, but a general slowing in growth since the early 1990s. This is also shown in Figure 1b, which illustrates the annual changes in UK transport oil demand (in natural logarithms) that the ‘contributions methodology’ attempts to explain. Figures 2a and 2b illustrate the general upward trend in UK GDP other than during the recessions in the 1970s, 1980s and 1990s, and Figures 2a and 2b illustrate the volatility in real transport oil prices including the large increases in the early and late 1970s, the fall in the mid 1980s and the general increase since then. Figures 4a and 4b illustrate that car fuel efficiency has generally increased since the early 1980s.

Results.

The results for the estimation of the model above are given in Table 2 with the estimated trends given in Figure 5. Specification I is the preferred model over the period 1964-2003 after the testing procedure outlined above, which has no dynamics despite starting with lags of 4 years on all the data. Nevertheless, it still fits the data well; passing all diagnostic tests for serial correlation, heteroscedasticity, non-normality⁷ and predictive failure. Furthermore, the restriction of a deterministic trend is rejected via the Likelihood Ratio (LR) test.

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⁷ Note, that following Harvey and Koopman (1992), impulse dummies are included for 1979 and 1980 given some evidence of non-normality of the auxiliary residuals.
Table 2: Estimated UK Transport oil demand equation

<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>$y_t$</td>
<td>0.53**</td>
<td>0.57**</td>
</tr>
<tr>
<td>$p_{ot}$</td>
<td>-0.12**</td>
<td>-0.12**</td>
</tr>
<tr>
<td>$f_t$</td>
<td>-0.32*</td>
<td>-0.27*</td>
</tr>
<tr>
<td>Interventions</td>
<td>Irr79 &amp; Irr80</td>
<td>Irr79 &amp; Irr80</td>
</tr>
<tr>
<td>Trend</td>
<td>-0.87% p.a.</td>
<td>-0.61% p.a.</td>
</tr>
<tr>
<td>Diagnostics</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Equation residuals</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Standard error</td>
<td>1.13%</td>
<td>1.13%</td>
</tr>
<tr>
<td>Normality</td>
<td>0.18</td>
<td>0.38</td>
</tr>
<tr>
<td>Heteroscedasticity</td>
<td>$H_{(5)} = 0.68$</td>
<td>$H_{(5)} = 0.58$</td>
</tr>
<tr>
<td>$r(1)$</td>
<td>-0.05</td>
<td>-0.00</td>
</tr>
<tr>
<td>$r(2)$</td>
<td>-0.02</td>
<td>-0.09</td>
</tr>
<tr>
<td>$r(3)$</td>
<td>-0.05</td>
<td>-0.12</td>
</tr>
<tr>
<td>$r(4)$</td>
<td>0.08</td>
<td>0.02</td>
</tr>
<tr>
<td>DW</td>
<td>1.92</td>
<td>1.90</td>
</tr>
<tr>
<td>Box-Ljung statistic</td>
<td>2.76</td>
<td>4.87</td>
</tr>
<tr>
<td>$R^2$</td>
<td>0.99</td>
<td>0.99</td>
</tr>
<tr>
<td>Auxiliary residuals</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Irregular: Normality</td>
<td>1.37</td>
<td>2.86</td>
</tr>
<tr>
<td>Level: Normality</td>
<td>n/a</td>
<td>n/a</td>
</tr>
<tr>
<td>Slope: Normality</td>
<td>2.48</td>
<td>2.33</td>
</tr>
<tr>
<td>Post-sample predictive tests</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Failure</td>
<td>0.96</td>
<td>n/a</td>
</tr>
<tr>
<td>Cusum</td>
<td>0.28</td>
<td>n/a</td>
</tr>
<tr>
<td>Hyperparameter Zero Restriction</td>
<td>$LR_{(1)} = 60.60^{**}$</td>
<td>$LR_{(1)} = 114.02^{**}$</td>
</tr>
</tbody>
</table>

Notes
- ** indicates significant at 1% level and * indicates significant at the 5% level.
- Normality is tested via the Bowman-Shenton statistic, approximately distributed as $\chi^2(2)$.
- $H_{(n)}$ is the test for heteroscedasticity, distributed approximately as $F(n, 1/T)$.
- $r_t$, the residual autocorrelation at lag $\tau$ distributed approximately as $N(0, 1/T)$.
- DW-Durbin-Watson statistic.
- The Box-Ljung statistic is based on the first 10 residuals autocorrelations and distributed approximately as $\chi^2(8)$.
- $R^2$ is the coefficient of determination.
- Failure is the post-sample predictive failure test for 2004-2007, approximately distributed as $\chi^2(n_\alpha)$ and Cusum is the test of parameter consistency, approximately distributed as $t(n_\alpha)$.
- $LR(h)$ is the likelihood ratio statistic for the zero-restriction of the hyperparameter(s) to zero, approximately distributed as $\chi^2(h)$.

The estimated income, price, and efficiency elasticities are about 0.5, -0.1, and -0.3 respectively. The estimated price elasticity is similar to the previous estimates over shorter periods of -0.12 using quarterly data in Hunt and Ninomiya (2003) and -0.11.
using annual data in Dimitropoulos et al. (2003). However, the estimated income elasticity is a little lower compared to the 0.80 in Hunt and Ninomiya (2003) and the 0.81 in Dimitropoulos et al. (2003). These, however, did not include an efficiency term (instead it was assumed that this was captured in the UEDT); but despite this, the estimated UEDT, shown in Figure 5a, has a very similar shape to the estimated UEDTs in Hunt and Ninomiya (2003) and Dimitropoulos et al. (2003).

Moreover, comparing the estimated price and income elasticities with those shown in Table 1, it can be seen that they are smaller (in absolute terms) than those observed in the previous studies surveyed. The results imply that the majority of unobserved taste information and/or efficiency changes are probably reflected through bias in the price elasticity. Although it should also be borne in mind that the values given in Table 1 are from a range of studies with different data sources, time periods, estimation approaches etc. which probably also partially explains the variation in the results.

Figure 5: Estimated UEDTs for UK transport oil demand, in logs (μt)


Given Specification I passes the predictive failure test, it is re-estimated over the whole period 1960-2007 and given as Specification II in Table 2 (with the associated estimated UEDT given in Figure 5b). This specification also passes all diagnostic
tests and the estimated parameters are very similar: about 0.6, -0.1, and -0.3 for the income, price and efficiency elasticities respectively. Moreover, the estimated UEDT has a similar shape to that for Specification I. Therefore, Specification II is used for the remainder of the analysis and discussion; it is used to estimate the contributions of the different drivers of the change in UK transportation oil demand, as specified in Equation (4) above. This is illustrated in Figure 6 and summarised in Table 3.\(^8\)

**Figure 6: Contributions to change in UK transport oil demand, in logs\((Δeo)\)**

![Graph showing contributions to change in UK transport oil demand](image)

It is clear from Figure 6 that the contribution of GDP in driving the change in transport oil demand is important and relatively stable other than in times of recessions; remaining a relatively important positive driver in all periods identified in Table 3. In contrast, the contribution of prices is relatively small other than during the

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\(^8\) Admittedly, the chosen dates for the different periods in Table 3 is arbitrary, but roughly coincides with the oil price hikes (and subsequent recessions) in the 1970s, the oil price collapse of the mid 1980s, and the recession of the early 1990s.
very high price changes, in particular the period 1985-1991. Similarly, the
contribution from fuel efficiency changes is relatively small over all periods, whereas,
ExNEF appears to have had an important positive impact until the early 1990s. From
1960 to 1973 ExNEF appears to have dominated the contribution to the change in
transport oil demand - being the main reason for the strong growth during this period.
Interestingly the strong positive contribution of ExNEF up to the 1990s is generally
reversed from the early 1990s. ExNEF makes a negative average contribution since
1991, so that despite the relative strong positive contribution from income, the actual
growth in oil transport demand slowed down considerably during the period 1991-
2007, driven primarily by the negative contribution from ExNEF (and the smaller
ccontributions from price and efficiency). Thus there appears to be a marked
‘behaviour change’ over this latest period, perhaps reflecting greater consumer
environmental awareness and a resultant change in ‘lifestyle’ and attitudes’.

Table 3: Summary of the contribution to the average percentage per annum
change in UK transport oil demand, in logs ($\Delta eo_t$)

<table>
<thead>
<tr>
<th>Period</th>
<th>Contribution from:</th>
<th>Total change in $eo_t$</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$y$</td>
<td>$po$</td>
</tr>
<tr>
<td>1960-1973</td>
<td>1.76</td>
<td>0.20</td>
</tr>
<tr>
<td>1973-1985</td>
<td>0.81</td>
<td>-0.28</td>
</tr>
<tr>
<td>1985-1991</td>
<td>1.41</td>
<td>0.59</td>
</tr>
<tr>
<td>1991-2007</td>
<td>1.56</td>
<td>-0.23</td>
</tr>
<tr>
<td>1960-2007</td>
<td>1.41</td>
<td>-0.02</td>
</tr>
</tbody>
</table>

Notes
- Following from Equation (4) the annual changes per annum contributions are approximated as

\[
\frac{\Delta y}{n}, \frac{\Delta po}{n}, \frac{\Delta f}{n}, \text{ and } \frac{\Delta eo_t}{n} \times 100
\]

for the contributions of GPD ($y$), price ($po$), fuel
efficiency ($f$), and exogenous non-economic factors (ExNEF) respectively. (The total change
being approximated by $\frac{\Delta eo_t}{n} \times 100$.) Where $n$ is the span of years that the change is calculated.
5. Concluding remarks

This paper is, as far as is known, the first attempt to use time series econometrics to quantify both the economic and the non-economic drivers of UK transport oil demand. This quantification shows that income generally makes an important positive contribution to driving the change in UK road transport oil demand given the estimated income elasticity of about 0.6 and the size of the changes in income. Whereas the estimated price and fuel efficiency elasticities of about -0.1 and -0.3 are smaller (in absolute terms); hence the estimated contributions from price and fuel efficiency are relatively small. Moreover, the quantification using the estimated non-linear UEDT shows that the non-economic behavioural factor ExNEF makes a non-trivial and positive contribution up until the early 1990s. However, since then, there appears to be an important behaviour change with the non-economic factors making a small but still important negative contribution to the change in the UK road transport oil demand. Moreover, the estimated ExNEF in the later periods suggests that consumer tastes are no longer driving demand up, or put another way; rises in transport oil demand are being driven predominantly by income more than ever before; particularly given fuel efficiency continues to improve.

This finding has important implications for policy makers wishing to curtail the growth in transport oil demand in the UK. The analysis suggests that the effect of taxes and increased fuel efficiency would be very limited given the estimated elasticities and subsequent contributions; although in a rounded policy package, there is no reason why these should not be used – provided their limitations are recognised. This therefore leaves policy makers with the main choice between attempting to reduce the trend rate of growth of the economy and/or acting on the non-economic
aspects of behaviour change. However, in a political world where, over the longer term, reduced economic growth is unlikely to be an option for elected policy makers, it leaves action on behaviour as crucial. In other words, assuming policy makers wish to reduce the growth in transport oil demand and hence emissions, but do not (or cannot) curtail economic growth and the effect of taxes and increased fuel efficiency are limited, then the policy action needs to be on changing non-economic behaviour – or those ‘simple’ tastes alluded to in level one economics texts. A main thrust of policy would therefore need to concentrate on educating and informing consumers in an attempt to encourage them to reduce their consumption of transport oil and hence emissions through the non-economic instruments route.

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