The Demand for Food in South Africa

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Abstract

Food consumption is an important issue in South Africa, not only in its relation to poverty and deprivation, but also given the importance of nutrition in allowing HIV/AIDS sufferers to lead extended, productive lives. With the pressing need to increase food security and the enormity of the epidemic, understanding the demand for food has become a vital task. It is important that the determinants of the demand for food are understood, so that responses of household food consumption to changes in the prices of foodstuffs, prices of other commodities, and total expenditure can be anticipated. There is, however, surprisingly little economic research on this topic. This paper provides an empirical analysis of the demand for food in South Africa for the years 1970 to 2002. It uses two modelling approaches, a general dynamic log-linear demand equation and a dynamic version of the almost ideal demand system to provide estimates of the short- and long-run price and expenditure demand elasticities.

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1. INTRODUCTION

Studies of food consumption and expenditure have often been the subject of research in the developed and developing world. They provide important inputs into food and related nutritional policy initiatives, by providing estimates of how food consumption is likely to change with changes in prices, incomes and taxation. In the context of South Africa food policy is inextricably linked to food security. Despite its status as a lower-middle-income-country (World Bank website), it has considerable inequality and deprivation. It is estimated that about 35 per cent of the population in South Africa are vulnerable to food insecurity and approximately a quarter of the children under the age of 6 years are estimated to have had their growth stunted by malnutrition (Human Sciences Research Council, 2004, 3). Thus, the Department of Agriculture devised ‘An Integrated Food Security Strategy for South Africa’ (IFSS) in July 2002. Food also has special significance, in the fight against HIV/AIDS. Nutrition is known to be important in determining the body’s ability to cope with HIV/AIDS and a well nourished body is essential for ensuring those living with the disease lead extended, productive lives. This means that it is important to be able to anticipate the types of changes that can take place in the food consumption of households when there are changes in incomes, prices and taxation.

Understanding the demand for food is, therefore, a vital task, but there is, however, surprisingly little economic research on this topic. The main recent studies are Balyamujura et al. (2000) which replicates some earlier studies in order to project impact of HIV/AIDS on the demand for food and Selvanathan and Selvanathan (2003), who estimate a demand system for South African consumption expenditure, incorporating food as one of the categories of the demand system. There are also a couple of studies using cross-sectional data; Agbola et al. (2003) and Human Sciences Research Council (2004). The studies usually start from a similar theoretical background, but differ markedly in the way they operationalise their models, some using single equation
estimation, others using systems of equations, some sticking with static models and others introducing dynamic models in a number of ways. Add to this the different definitions in the data and the different samples and it is no surprise that they get a range of results.

In this paper a time series analysis of the demand for food in South Africa for 1970 – 2002 is undertaken. Rather than focus on one particular method it takes both an ‘ad hoc’ approach, estimating a general dynamic log-linear demand equation, as well as a ‘demand system’ approach, estimating a dynamic form of the almost ideal demand system (Deaton and Muellbauer, 1980). These provide estimates of price and income elasticities in both the short run and the long run. In the next section a brief description of food and related data trends in South Africa is provided, followed in section 3 by a brief review of demand theory and the econometric approaches used in empirical work. Sections 4 and 5 then consider the South African data, discussing the empirical models and the result of estimating them on the South African data. The elasticity estimates are summarised and discussed in section 6. Finally, section 7 presents some conclusions and considers how the work could be developed in future.

2. Food and related data trends in South Africa

The main source of consistent time series data on food and other consumption is the Reserve Bank of South Africa web site, ‘Quarterly Bulletin Time Series”. Under this data the category for ‘food’ is classified as ‘food, beverages and tobacco’. The variables include expenditure on food (in constant prices) ($Q_t$), total consumers’ expenditure in current prices ($X_t$), the price of food ($PF_t$), and a general price index ($PG_t$).

Figure 1 shows the growth of real food expenditure, real total consumer expenditure and real gross domestic product (GDP)$^1$. This shows the period under consideration, 1970-

$^1$ Table A1 in the appendix provides a summary of the growth rates.
2002 to be one of poor growth in GDP. In the 1960s the real average annual growth was 5.8% (Akinboade et al., 2004), slowing to 3.31% in the 1970s and 1.50% per annum in the 1980s. Growth then improved in the 1990s to average 1.73% per annum and 3.5% in 2002. As expected, the growth of consumer expenditure followed a similar pattern to the growth of real GDP, averaging 3.78% per annum in the 1970s, 2.77% per annum in the 1980s and 2.34% per annum for the 1990s. Between 2001 and 2002 total consumption expenditure grew at 3.19%.

Figure 1: Growth rates of real gross domestic product, food expenditure and total consumer expenditure

While food expenditure followed a similar pattern, the deterioration of growth was more marked, with average growth in the 1970s at 3.59% per annum, declining to 2.62% per annum in the 1980s and a meagre 0.66% per annum for the 1990s. 2001-2 saw growth of 1.61%.
As Table 1 shows, the share of food in expenditure also fell over this period\(^2\). In 1970 the share of consumer expenditure devoted to food was 35.6\%, by 1990 it was 34.5\%, declining markedly in the 1990s, to reach 28.4\% in 2002. The categories that showed increases in their share were transport, housing, clothing and medical services, with expenditure on furniture showing a small decline over the period.

Table 1: Budget shares (percentages) of consumer spending, South Africa, 1970 – 2002

<table>
<thead>
<tr>
<th>Year</th>
<th>Food</th>
<th>Transport</th>
<th>Housing</th>
<th>Furniture</th>
<th>Clothing</th>
<th>Medical</th>
<th>Recreation</th>
<th>All other goods and services</th>
</tr>
</thead>
<tbody>
<tr>
<td>1970</td>
<td>35.63</td>
<td>16.81</td>
<td>6.54</td>
<td>11.46</td>
<td>5.84</td>
<td>3.48</td>
<td>7.40</td>
<td>12.84</td>
</tr>
<tr>
<td>1975</td>
<td>33.67</td>
<td>16.25</td>
<td>5.70</td>
<td>12.42</td>
<td>6.22</td>
<td>4.30</td>
<td>8.26</td>
<td>13.18</td>
</tr>
<tr>
<td>1980</td>
<td>34.98</td>
<td>16.61</td>
<td>7.28</td>
<td>12.40</td>
<td>6.08</td>
<td>4.29</td>
<td>8.11</td>
<td>10.25</td>
</tr>
<tr>
<td>1985</td>
<td>36.08</td>
<td>13.42</td>
<td>11.10</td>
<td>11.61</td>
<td>5.18</td>
<td>4.69</td>
<td>7.75</td>
<td>10.17</td>
</tr>
<tr>
<td>1990</td>
<td>34.47</td>
<td>15.56</td>
<td>11.47</td>
<td>11.04</td>
<td>5.08</td>
<td>4.86</td>
<td>6.87</td>
<td>10.65</td>
</tr>
<tr>
<td>1995</td>
<td>30.81</td>
<td>17.41</td>
<td>11.80</td>
<td>10.80</td>
<td>6.13</td>
<td>6.35</td>
<td>6.80</td>
<td>9.90</td>
</tr>
<tr>
<td>2002</td>
<td>28.38</td>
<td>18.77</td>
<td>10.67</td>
<td>10.07</td>
<td>7.21</td>
<td>6.76</td>
<td>7.70</td>
<td>10.44</td>
</tr>
<tr>
<td>Mean</td>
<td>33.43</td>
<td>16.17</td>
<td>9.51</td>
<td>11.43</td>
<td>5.90</td>
<td>4.96</td>
<td>7.42</td>
<td>11.18</td>
</tr>
</tbody>
</table>

3. Analysing the Demand for Food

To analyse the demand for food a useful starting point is the neoclassical model of consumer choice. This suggests the demand is the outcome of a consumer making a utility-maximising decision, based on the prices of all products available to the consumer, and the consumer’s total expenditure. Thus the demand function is specified as

\(^2\) Reserve Bank website data on expenditures have been aggregated to obtain broad categories. For example, transport requires the aggregation of expenditure of personal transport equipment, motor car tyres, parts and accessories, transport and communication services. ‘All other goods and services’ is obtained by summing all other shares of consumer expenditure and subtracting their total from the total final consumer expenditure. The budget share in this study is calculated on the basis of a share of total final consumer expenditure from the website.
\[ q_i = q_i(p_1, p_2, \ldots, p_n, x) \quad i = 1, 2, \ldots, n. \]  

where \( q_i \) and \( p_i \) are the quantity demanded and price of the \( i \)th product, there are \( n \) products in total, and \( x = \sum p_i q_i \) is total expenditure. Empirical studies of demand using econometrics have used single demand equations of particular commodity groups and systems of demand equations.

For single equation models the theory provides limited information. It does not specify the precise form of the equation, nor variables to be included, other than prices and income. It does, however, suggest certain restrictions which can help limit the degrees of freedom lost. Demand functions are considered homogenous of degree zero in all explanatory variables and the Slutsky conditions imply that own-price substitution effects are negative and the own-price and income derivatives of a demand equation should be 
\[ \frac{\partial q_i}{\partial p_i} + q_i \frac{\partial q_i}{\partial x} < 0. \]

As theory does not provide much in the way of guidelines for the specification of a demand equation, empirical versions of demand equations are often relatively ad hoc. Functional forms are chosen to facilitate easy estimation, and explanatory price variables usually include own-price, the price of close substitutes and complements and possibly the general price level. Specifications in logarithms have the advantage that the coefficients estimated can be interpreted as elasticities. This general specification is followed for our food demand equation.

\[ q_i = \beta_0 + \beta_1 p_i + \beta_2 p_j + \beta_3 p_g + \beta_4 x + u \]  

Where \( q_i \) is the quantity demanded of good \( i \), \( p_i \) is the price of good \( i \), \( p_j \) is the price of good \( j \), \( p_g \) is the general price level, and \( x \) is current total expenditure. To allow for dynamic analysis, lags of the dependent variable and explanatory variables are introduced.

An alternative to this approach is to estimate a system of demand equations. This takes all commodity groups and treats them as a singular system. It also allows the degrees of
freedom to be increased through single equation and cross equation restrictions\(^3\). The most popular system in the literature, due to its flexibility and ease of use is the almost ideal demand system of Deaton and Muellbauer (1980):

\[
    w_i = \alpha_i + \sum \delta_k D_k + \sum \gamma_j \ln p_j + \beta_i \ln (E/P)
\]

where \(\ln P = \alpha_0 + \sum \alpha_k \ln p_k + \frac{1}{2} \sum \sum \gamma_{kj} \ln p_j \ln p_j\), \(E\) is total expenditure, \(w_i = p_i q_i / E\) and the conditioning variables \(D_k\) can determine the intercepts of the demand functions. The system can be estimated equation by equation by using the approximation for \(\ln P\), \(\ln P^* = \sum w_j \ln p_j\). This is a static model but has been made dynamic in the literature by respecifying the theory (Dunne et al., 1984), or by specifying a general dynamic estimable form of the regression (Anderson and Blundell, 1987; Smith, 1989). The latter approach is used below.

Given the importance of the issue for South Africa, there has been surprisingly little research on the demand for food. In a recent contribution Taljaard et al. (2004) focus on the demand for meat, using an almost ideal demand system, while Selvanathan and Selvanathan (2003) estimate a version of the Rotterdam demand system for a range of groups of commodities, but they do not focus particularly on food and give it little attention. Agbola et al. (2003) examine cross-sectional data from the South African Integrated Household Survey (SAIHS) of 1993\(^4\), analysing household consumption of food groups and estimating price and expenditure elasticities of demand for the food groups. There has also been some research conducted by the Integrated Rural and Regional Development research programme of the Human Sciences Research Council. It

\(^3\) In estimating systems of demand equations, two approaches are used. Firstly, the precise form of the utility function is specified as in the case of the linear expenditure system. Using this approach implies that the system of demand equations derived will satisfy at least some of the general restrictions of consumer theory. The advantage of this approach is that degrees of freedom are saved, but the disadvantages are that the restrictions cannot be tested and there is a loss of generality in specifying the form of the utility function. The second approach begins with a set of demand equations which possibly satisfy the theoretical restrictions. The advantage of this is that the restrictions can then tested, but at the cost of limiting the degrees of freedom. This is the case for the Rotterdam and almost ideal demand systems.

\(^4\) This was conducted by the Southern Africa Labour and Development Research Unit (SALDRU). This survey was part of the World Bank Living Standard Measurement Surveys and was conducted nine months before South Africa’s first democratic election in 1994.
conducted a pilot study for the Department of Agriculture ‘… to establish a methodology for monitoring and evaluating the impact of … price increases on low-income communities in both rural and urban areas.’ (Human Sciences Research Council, 2004, 59). The results of this study were presented in 2002 (Aliber and Modiselle, 2002), but the sample size and budget were limited, so the survey only provides useful insight into the impact of price increases on low-income households, and nothing more. HSRC(2004) noted that the survey indicates that during the 6 months April – October 2001, the price of maize meal rose by between 25% and 35%, the sugar price increased by about 20% and the rice price by between 25% and 65%, while average spending on all of the items increased by 28% - the closest to an elasticity measure they get. Balyamujura et al. (2000) replicated some earlier studies in order to project impact of HIV/AIDS on the demand for food.

Recent international studies of the demand for food include a field trail experiment in applied econometrics reported in Magnus and Morgan (1999) which collected budget survey and time-series for the USA and the Netherlands. Various groups of researchers examined the data using differing techniques and methodologies, including Song, Liu and Romilly (1997). They use an error correction model and an almost ideal demand system, to obtain elasticities.

Recent food related studies using the almost ideal demand system are Eakins and Gallagher (2003) on alcohol expenditure in Ireland, Duffy (2001) on food demand in the UK, Karagiannis et al. (2000) on the demand for meat in Greece. In the context of major transformation of the US farm and food policy, LaFrance (1999a, 1999b) and LaFrance and Beatty (2001) analysed the demand for food in the US, while Edgerton et al. (1996) estimate demand systems for food for the Nordic countries.

4. Estimating the Demand Equation

To operationalise equation (2) for a time series study a dynamic model is required. Using a general first order log-linear demand model (ARDL) gives:

\[ q_t = b_0 + b_2 q_{t-1} + b_3 pf_t + b_4 pf_{t-1} + b_5 pg_t + b_6 pg_{t-1} + b_7 x_t + b_8 x_{t-1} + \varepsilon_t \]  

(4)

Reparameterising to provide a specification that allows for easy identification of the long-run effects gives:

\[ \Delta q_t = b_0 + b_3 \Delta pf_t + b_5 \Delta pg_t + b_7 \Delta x_t - \alpha_0 q_{t-1} + \alpha_1 pf_{t-1} + \alpha_2 pg_{t-1} + \alpha_3 x_{t-1} + \varepsilon_t \]  

(5)

Estimating this model for the period 1970-2002 gives the results in Table 2. There are four significant explanatory variables \( \Delta pf_t, \Delta x_t, x_{t-1} \) and \( q_{t-1} \), with the change in the general price index, \( \Delta pg_t \), the lagged food price, \( pf_{t-1} \), and lagged general price, \( pg_{t-1} \) insignificant.

Table 2: The dynamic demand model

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th>t-ratio (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Constant</td>
<td>-1.59</td>
<td>-1.53 (0.14)</td>
</tr>
<tr>
<td>( \Delta pf_t )</td>
<td>-0.44</td>
<td>-2.75 (0.01)</td>
</tr>
<tr>
<td>( \Delta pg_t )</td>
<td>-0.00</td>
<td>-0.00 (0.99)</td>
</tr>
<tr>
<td>( \Delta x_t )</td>
<td>0.54</td>
<td>3.94 (0.00)</td>
</tr>
<tr>
<td>( q_{t-1} )</td>
<td>-0.31</td>
<td>-2.24 (0.04)</td>
</tr>
<tr>
<td>( pf_{t-1} )</td>
<td>-0.28</td>
<td>-1.42 (0.17)</td>
</tr>
<tr>
<td>( pg_{t-1} )</td>
<td>-0.16</td>
<td>-1.03 (0.31)</td>
</tr>
<tr>
<td>( x_{t-1} )</td>
<td>0.41</td>
<td>3.22 (0.00)</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Diagnostic Test</th>
<th>Result (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Standard error</td>
<td>0.02</td>
</tr>
<tr>
<td>( R^2 )</td>
<td>0.65</td>
</tr>
<tr>
<td>Serial Correlation</td>
<td>F(1,23) = 0.00 (0.99)</td>
</tr>
<tr>
<td>Functional form</td>
<td>F(1,23) = 1.89 (0.18)</td>
</tr>
<tr>
<td>Heteroscedasticity</td>
<td>F(1,30) = 0.01 (0.93)</td>
</tr>
<tr>
<td>Normality</td>
<td>Chsq(2) = 0.93 (0.63)</td>
</tr>
<tr>
<td>Chow test</td>
<td>F(8,16) = 0.92 (0.53)</td>
</tr>
</tbody>
</table>

The diagnostic tests do not suggest any particular problems with the specification, though a plot of the residuals did indicate the possibility of a structural break in the data around 1990 and a Chow test was undertaken, but the result was not significant.
The restrictions of short-run and long-run homogeneity, imply that the short-run elasticities sum to zero and long-run elasticities sum to zero. This is imposed by introducing the relative price of food and real total consumer expenditure into the equation in place of the price of food and current total consumer expenditure. This gives:

\[ \Delta q_t = \rho_2 \Delta rpf_t + \rho_3 \Delta rx_t - \rho_1 (q_{t-1} - \rho_0 - \rho_4 r pf_{t-1} - \rho_5 rx_{t-1}) + v_t \]  

or

\[ \Delta q_t = \tau_0 - \rho_1 q_{t-1} + \rho_2 \Delta rpf_t + \rho_3 \Delta rx_t + \tau_1 rpf_{t-1} + \tau_2 rx_{t-1} \]  

Where \( \tau_0 = \rho_0 \rho_1, \tau_1 = \rho_1 \rho_4, \) and \( \tau_2 = \rho_1 \rho_5 \)

As the results in Table 3 show, all five explanatory variables are significant when these restrictions are imposed. The only diagnostic test to suggest a problem is that for the functional form. The \( F \)-test for the homogeneity restrictions indicate that they are not rejected.

Table 3: The dynamic demand model with homogeneity imposed

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th>( t )-ratio (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Constant</td>
<td>-0.14</td>
<td>-0.41 (0.69)</td>
</tr>
<tr>
<td>( \Delta rpf_t )</td>
<td>-0.55</td>
<td>-3.39 (0.00)</td>
</tr>
<tr>
<td>( \Delta rx_t )</td>
<td>0.49</td>
<td>3.48 (0.00)</td>
</tr>
<tr>
<td>( q_{t-1} )</td>
<td>-0.31</td>
<td>-2.66 (0.01)</td>
</tr>
<tr>
<td>( rpf_{t-1} )</td>
<td>-0.39</td>
<td>-2.02 (0.05)</td>
</tr>
<tr>
<td>( rx_{t-1} )</td>
<td>0.29</td>
<td>2.40 (0.02)</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Diagnostic test</th>
<th>Result (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Standard error</td>
<td>0.02</td>
</tr>
<tr>
<td>( R^2 )</td>
<td>0.57</td>
</tr>
<tr>
<td>Serial Correlation</td>
<td>F(1,25) = 0.41 (0.53)</td>
</tr>
<tr>
<td>Functional form</td>
<td>F(1,25) = 5.14 (0.03)</td>
</tr>
<tr>
<td>Heteroscedasticity</td>
<td>F(1,30) = 3.24 (0.082)</td>
</tr>
<tr>
<td>Normality</td>
<td>Chsq(2) = 0.71 (0.69)</td>
</tr>
<tr>
<td>Homogeneity restrictions</td>
<td>( F(2,24) = 2.65 ) (( F_{0.05}(2,24) = 3.40 ))</td>
</tr>
</tbody>
</table>

A plot of the residuals suggested that 1972 might be an outlier and that this might be influencing the functional form (Reset) test result. There are three possible reasons why 1972 might have extreme values. Firstly, an exceptional rise in food prices accounted for 2 percentage points of the 7.4 per cent rise in consumer prices in 1972. This was caused
by severe cold spells in winter combined with drought conditions in summer (South African Reserve Bank Quarterly Bulletin, March 1973). Secondly, Rutter (2003) studying the determinants of the South African producer price of maize from 1970 to 2001 also observed an outlier for 1972, which he suspected was due to a deal between the US and Russia on the sale of wheat, which had a knock on effect on feed grain prices (Lutterell, 1973) and so the South African grain industry. Thirdly, the Foodstuffs, Cosmetics and Disinfectants (FCD) Act came into being in 1972 (Act 54 of 1972). It falls within the ambit of the government’s Health Department, but covers all foodstuffs sold in South Africa (The South African Market for Food and Beverages Report, 1997).

Introducing a dummy variable for 1972 to the error correction model with homogeneity imposed gives the results in Table 4. The dummy is significant and its inclusion deals with the functional form issue and the homogeneity restrictions are still not rejected\(^6\).

**Table 4: The dynamic demand model with homogeneity imposed and a 1972 dummy**

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th>t-ratio (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Constant</td>
<td>-0.05</td>
<td>-0.17 (0.87)</td>
</tr>
<tr>
<td>(\Delta rpf_t)</td>
<td>-0.60</td>
<td>-4.15 (0.002)</td>
</tr>
<tr>
<td>(\Delta rx_t)</td>
<td>0.50</td>
<td>3.98 (0.00)</td>
</tr>
<tr>
<td>(q_{t-1})</td>
<td>-0.37</td>
<td>-3.56 (0.00)</td>
</tr>
<tr>
<td>(rpf_{t-1})</td>
<td>-0.46</td>
<td>-2.63 (0.01)</td>
</tr>
<tr>
<td>(rx_{t-1})</td>
<td>0.35</td>
<td>3.15 (0.00)</td>
</tr>
<tr>
<td>(D72)</td>
<td>-0.05</td>
<td>-2.92 (0.01)</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Diagnostic Test</th>
<th>Result (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Standard error</td>
<td>0.02</td>
</tr>
<tr>
<td>(R^2)</td>
<td>0.68</td>
</tr>
<tr>
<td>Serial Correlation</td>
<td>F(1,24) = 0.20 (0.66)</td>
</tr>
<tr>
<td>Functional form</td>
<td>F(1,24) = 3.8 (0.06)</td>
</tr>
<tr>
<td>Heteroscedasticity</td>
<td>F(1,30) = 0.11 (0.74)</td>
</tr>
<tr>
<td>Normality</td>
<td>Chsq(2) = 0.61 (0.74)</td>
</tr>
<tr>
<td>Homogeneity restrictions</td>
<td>F(2,23) = 1.27 (F_{0.05}(2,23) = 3.44)</td>
</tr>
</tbody>
</table>

\(^6\) The \(F\)-test was conducted using the restricted and unrestricted models with the dummy variable included.
This means that the short-run elasticities of demand with respect to the relative price of food and real total consumer expenditure are -0.6 and 0.5 respectively, while the values of the coefficients on the lagged levels suggest a long-run relation:

\[ q = -1.24rpf + 0.95rx \]

Thus the long-run elasticities are -1.2 with respect to the relative price of food and 0.95 with respect to real total consumer expenditure. Testing for unitary elasticities by imposing them on the equation, saw the price elasticity restriction rejected, but the income elasticity is not rejected\(^7\).

These results suggest that the demand for food is more responsive to price and income changes in the long run than in the short run. This would be expected for food as short-run consumption behaviour is likely to be affected by habit and other factors, whereas in the long run consumers have time to adjust. The fact that the long-run income elasticity result is nearly unity could be the result of the high proportion of poor households in the South African population. As incomes rise it is largely matched by rising expenditure on food in the long run.

The long-run price elasticity of -1.2 is a concern, as it is much larger than one might expect for a necessity. This might result from the definition of food in the data. In fact the category is ‘food, beverages and tobacco’. While this is in common use, it could be argued that the constituent components act differently. Thus, when there is an own-price increase, the consumption of tobacco and alcoholic beverages may be reduced fairly substantially to ensure basic food needs are met. On the other hand the addictive nature of these goods might limit such adjustments in practise.

\(^7\)The joint restriction that the coefficients on \(rpf\) and \(rx\) were 1 and -1 respectively, in equation 6 with D72 included, was rejected: \(F(2,25) = 5.73\) and \(F_{0.05}(2,23) = 3.39\). Likewise, testing the coefficient on \(rpf\) was 1 was rejected: \(F(1,25) = 9.67\) and \(F_{0.05}(1,25) = 4.24\). However, the restriction that the coefficient on \(rx\) was -1 was not rejected: \(F(1,25) = 2.08\) and \(F_{0.05}(1,25) = 4.24\).
It is possible to get a variable with a less broad definition of food from the South African Reserve Bank, but only for the shorter period 1976-2003. Using this ‘only food’ variable in equation (5) gives:

\[
\Delta q_{fo_t} = \theta_0 + \theta_1 q_{fo_{t-1}} + \theta_2 \Delta p_{fo_t} + \theta_3 \Delta p_g + \theta_4 \Delta x_t + \theta_5 p_{fo_{t-1}}
\]

\[\quad + \theta_6 p_{g_{t-1}} + \theta_7 x_{t-1} + \nu_t \tag{8}\]

where \(q_{fo_t}\) is consumer expenditure on ‘only food’ and \(p_{fo_t}\) is the price of ‘only food’, and other variables are as previously defined. The results suggest a relatively well specified model, for which the homogeneity restriction is not rejected\(^8\).

**Table 5: The dynamic demand model: ‘only food’**

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th>(t)-ratio (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Constant</td>
<td>-0.85</td>
<td>-0.84 (0.41)</td>
</tr>
<tr>
<td>(q_{fo_{t-1}})</td>
<td>-0.36</td>
<td>-2.47 (0.02)</td>
</tr>
<tr>
<td>(\Delta p_{fo_t})</td>
<td>-0.35</td>
<td>-2.69 (0.02)</td>
</tr>
<tr>
<td>(\Delta p_g)</td>
<td>-0.05</td>
<td>-0.24 (0.81)</td>
</tr>
<tr>
<td>(\Delta x_t)</td>
<td>0.43</td>
<td>2.83 (0.01)</td>
</tr>
<tr>
<td>(p_{fo_{t-1}})</td>
<td>-0.14</td>
<td>-0.76 (0.46)</td>
</tr>
<tr>
<td>(p_{g_{t-1}})</td>
<td>-0.28</td>
<td>-1.45 (0.17)</td>
</tr>
<tr>
<td>(x_{t-1})</td>
<td>0.38</td>
<td>3.01 (0.01)</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Diagnostic Test</th>
<th>Result (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Standard error</td>
<td>0.02</td>
</tr>
<tr>
<td>(R^2)</td>
<td>0.71</td>
</tr>
<tr>
<td>Serial Correlation</td>
<td>(F(1,17) = 0.47 (0.50))</td>
</tr>
<tr>
<td>Functional form</td>
<td>(F(1,17) = 3.55 (0.08))</td>
</tr>
<tr>
<td>Heteroscedasticity</td>
<td>(F(1,24) = 2.03 (0.17))</td>
</tr>
<tr>
<td>Normality</td>
<td>(\text{Chsq}(2) = 0.22 (0.90))</td>
</tr>
</tbody>
</table>

Imposing the homogeneity restriction gives the results in Table 6, with all the explanatory variables significant, except for \(r_{pfo_{t-1}}\), the lagged relative/real price of food, and no obvious specification problems.

---

\(^8\) The relevant equation to test homogeneity is:

\[
\Delta q_{fo_t} = \gamma_0 + \gamma_1 q_{fo_{t-1}} + \gamma_2 \Delta r_{pfo_t} + \gamma_3 \Delta r_x + \gamma_4 r_{pfo_{t-1}} + \gamma_5 r_{x_{t-1}} + \nu_t
\]

Where \(r_{pfo_t} = p_{fo_t} - p_g\) and \(r_{x_t} = x_t - p_g\).
With the more restricted definition of food, the elasticities are indeed lower, with the short-run price elasticity -0.5 and the income/expenditure elasticity is 0.5. The long-run relationship is:

\[ q_{fo} = -0.77r_{pfo} + 0.63r_x \]

This implies a long-run price elasticity of -0.8, though this is based on an insignificant coefficient for the relative price and an expenditure elasticity of 0.6. That the long-run relative price term should be insignificant is not unexpected. As Thomas (1997) argues in a study of food demand in the USA, food is a very basic necessity and it would be reasonable to assume that the demand for it may well be eventually unaffected by its price and only be determined by real income/expenditure.\(^9\)

**Table 6: The dynamic demand model with homogeneity imposed: only food**

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th>t-ratio (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Constant</td>
<td>1.03</td>
<td>1.93 (0.07)</td>
</tr>
<tr>
<td>(q_{fo,-1})</td>
<td>-0.35</td>
<td>-3.16 (0.01)</td>
</tr>
<tr>
<td>(\Delta rpfo_t)</td>
<td>-0.48</td>
<td>-3.63 (0.00)</td>
</tr>
<tr>
<td>(\Delta rx_t)</td>
<td>0.45</td>
<td>2.73 (0.01)</td>
</tr>
<tr>
<td>(rpfo_{t,-1})</td>
<td>-0.27</td>
<td>-1.45 (0.16)</td>
</tr>
<tr>
<td>(rx_{t,-1})</td>
<td>0.22</td>
<td>2.52 (0.02)</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Diagnostic Test</th>
<th>Result (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Standard error</td>
<td>0.02</td>
</tr>
<tr>
<td>(R^2)</td>
<td>0.63</td>
</tr>
<tr>
<td>Serial Correlation</td>
<td>(F(1,19) = 0.04 (0.84))</td>
</tr>
<tr>
<td>Functional form</td>
<td>(F(1,19) = 2.13 (0.16))</td>
</tr>
<tr>
<td>Heteroscedasticity</td>
<td>(F(1,24) = 3.62 (0.07))</td>
</tr>
<tr>
<td>Normality</td>
<td>(Chsq(2) = 0.03 (0.98))</td>
</tr>
<tr>
<td>Restrictions</td>
<td>(F(2,18) = 2.56 \ F_{0.05}(2,18) = 3.55)</td>
</tr>
</tbody>
</table>

These results do seem to support our concern with the composition of the food category, excluding beverages and tobacco from the data reduces the elasticities and suggests that own price is not an important variable in the long-run, but moving to the ‘food only’ category is at the cost of degrees of freedom. The income elasticity estimates show food

\(^9\) Thomas, 1997, p394
to be a normal good, its demand increases as income increases, and it is inelastic in both
the short-run and long-run. Again this meets expectations for a necessity.

5. Estimating a Demand System

In using the ‘demand system’ approach, the almost ideal demand system of equation (3)
is employed in a dynamic form, following Smith (1989). The system has two equations;
one for food and a residual category non-food:

\[
\Delta sq_t = \lambda_{10} + \lambda_{11} \Delta pf_t + \lambda_{12} \Delta pnf_t + \lambda_{13} \Delta (x - P*)_t - \lambda_{14} sq_{t-1} + \lambda_{15} pf_{t-1} \\
+ \lambda_{16} pnf_{t-1} + \lambda_{17} (x - P*)_{t-1} + \epsilon_t \tag{8}
\]

\[
\Delta snfq_t = \lambda_{20} + \lambda_{21} \Delta pf_t + \lambda_{22} \Delta pnf_t + \lambda_{23} \Delta (x - P*)_t - \lambda_{24} sq_{t-1} + \lambda_{25} pf_{t-1} \\
+ \lambda_{26} pnf_{t-1} + \lambda_{27} (x - P*)_{t-1} + \epsilon_t \tag{9}
\]

Where, \(sq_t\) is the share of expenditure on food and \(snfq_t\), the share of non-food
expenditure, \(pf_t\) the price of food, \(pnft\) the non-food price and \((x - P*)\) is the log of
consumption expenditure in current prices minus the log of an index of prices, \(P^*\). Note
that the dependent variable no longer sums to one, the dynamic form it is not a singular
system. The estimation results for the food equation are given in Table 8.

Table 8: Dynamic almost ideal demand system

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th>(t)-ratio (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Constant</td>
<td>-0.24</td>
<td>-0.62 (0.54)</td>
</tr>
<tr>
<td>(\Delta pf_t)</td>
<td>-0.1</td>
<td>-2.23 (0.04)</td>
</tr>
<tr>
<td>(\Delta pnf_t)</td>
<td>0.14</td>
<td>2.6 (0.02)</td>
</tr>
<tr>
<td>(\Delta (x - P*)_t)</td>
<td>-0.16</td>
<td>-3.39 (0.00)</td>
</tr>
<tr>
<td>(sq_{t-1})</td>
<td>-0.35</td>
<td>-2.56 (0.02)</td>
</tr>
<tr>
<td>(pf_{t-1})</td>
<td>-0.06</td>
<td>-1.41 (0.17)</td>
</tr>
<tr>
<td>(pnf_{t-1})</td>
<td>0.06</td>
<td>1.20 (0.24)</td>
</tr>
<tr>
<td>((x - P*)_{t-1})</td>
<td>0.03</td>
<td>0.94 (0.36)</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Measure/ Diagnostic Test</th>
<th>Result (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Standard error</td>
<td>0.01</td>
</tr>
<tr>
<td>(R^2)</td>
<td>0.68</td>
</tr>
<tr>
<td>Serial Correlation</td>
<td>F(1,23) = 0.01 (0.93)</td>
</tr>
<tr>
<td>Functional form</td>
<td>F(1,23) = 0.05 (0.83)</td>
</tr>
</tbody>
</table>

\(^{10}\) In this case \(P^* = (sq_t, pf_t) + (snfq_t, pnft)\)
All the coefficients on the changes variables are significant, as is the lagged dependent variable $sq_{t-1}$. However, the three long-run coefficients are all insignificant. The results for the diagnostic tests suggest no particular problems with the specification of the model.

Imposing homogeneity on the system implies using the relative price of food, so taking $\Delta pff_t = \Delta pf_t - \Delta pnf_t$ and $pff_{t-1} = pf_{t-1} - pnf_{t-1}$ gives for food:

$$\Delta sq_t = \delta_{10} + \delta_{11} \Delta pff_t + \delta_{12} \Delta (x - P^*_t) + \delta_{13} sq_{t-1} + \delta_{14} pff_{t-1}$$

$$+ \delta_{15} (x - P^*)_{t-1} + \nu_t$$  \hspace{1cm} (10)

Estimating equation (10) gives the results in Table 9.

### Table 9: Dynamic almost ideal demand system with homogeneity imposed:

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th>$t$-ratio (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Constant</td>
<td>0.17</td>
<td>1.74 (0.09)</td>
</tr>
<tr>
<td>$\Delta pff_t$</td>
<td>-0.14</td>
<td>-3.73 (0.00)</td>
</tr>
<tr>
<td>$\Delta (x - P^*_t)$</td>
<td>-0.17</td>
<td>-3.52 (0.00)</td>
</tr>
<tr>
<td>sq$_{t-1}$</td>
<td>-0.34</td>
<td>-2.78 (0.01)</td>
</tr>
<tr>
<td>pff$_{t-1}$</td>
<td>-0.09</td>
<td>-2.04 (0.05)</td>
</tr>
<tr>
<td>$(x - P^*)_{t-1}$</td>
<td>-0.01</td>
<td>-0.67 (0.51)</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Measure/ Diagnostic Test</th>
<th>Result (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Standard error</td>
<td>0.01</td>
</tr>
<tr>
<td>$R^2$</td>
<td>0.61</td>
</tr>
<tr>
<td>Serial Correlation</td>
<td>F(1,25) = 0.33 (0.57)</td>
</tr>
<tr>
<td>Functional form</td>
<td>F(1,25) = 0.01 (0.94)</td>
</tr>
<tr>
<td>Heteroscedasticity</td>
<td>F(1,30) = 2.32 (0.14)</td>
</tr>
<tr>
<td>Normality</td>
<td>Chsq(2) = 0.50 (0.78)</td>
</tr>
<tr>
<td>Homogeneity restrictions</td>
<td>F(2,24) = 2.62 (F$_{0.05}(2,24)$ = 3.40)</td>
</tr>
</tbody>
</table>

Homogeneity is not rejected and the only insignificant explanatory variable is $(x - P^*)_{t-1}$, though the significance of the relative price, $pff_{t-1}$, coefficient is borderline. The four tests diagnostic tests do not show any evidence of misspecification.
Within the demand system the coefficients are not elasticities, they are measuring the response of shares to changes in price and income, though a negative sign on the income/expenditure coefficient does indicate that food is an inferior good.

For the long-run (static) model:

\[ sq_t = \kappa_{10} + \kappa_{11}pff_t + \kappa_{12} (x - P^*)_t \]  \hspace{1cm} (16)

the expenditure elasticity is:

\[ e_f = 1 + \frac{\kappa_{12}}{sq_t} \]

meaning that if \( \kappa_{12} \) is not significantly different to zero, there is a unit income elasticity.

The compensated own price elasticity of for food is:

\[ pe_f = \frac{1}{sq} \left[ \frac{1}{\kappa_{11} + \kappa_{12}(x - P^*)} \right] - 1 + sq \]

and the uncompensated is \( pe_f - e_f \frac{sq}{sq} \). For the short-run elasticities we replace the long-run coefficients with the short-run ones. Computing the short-run elasticities gives a value of 0.5 for income elasticity and 0.0 for the compensated price elasticity, with the latter being rather different to that calculated using the log-linear model. The long-run elasticities are 1 for income (as expected given the insignificance of the total real expenditure term) and -0.9 for price.

Estimating the model for 'only food' gives the results in Table 10, which shows insignificant coefficients on the change of real total expenditure change and its lagged level, but with no obvious specification problems.

### Table 10: Dynamic almost ideal demand system for ‘only food’

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th>t-ratio (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Constant</td>
<td>0.13</td>
<td>0.42 (0.68)</td>
</tr>
<tr>
<td>( \Delta pfo_t )</td>
<td>-0.11</td>
<td>-3.33 (0.00)</td>
</tr>
<tr>
<td>( \Delta pnf.o_t )</td>
<td>0.16</td>
<td>3.11 (0.01)</td>
</tr>
<tr>
<td>( \Delta (x - PO^*)_t )</td>
<td>-0.06</td>
<td>-0.79 (0.44)</td>
</tr>
<tr>
<td>( sq.o_{t-1} )</td>
<td>-0.60</td>
<td>-3.44 (0.00)</td>
</tr>
<tr>
<td>( pfo.o_{t-1} )</td>
<td>-0.10</td>
<td>-2.71 (0.01)</td>
</tr>
</tbody>
</table>
Imposing homogeneity gives the results in Table 11 and the restriction is not rejected. It shows the coefficients on the real expenditure terms to be insignificant, though the lagged level is significant at 6%, suggesting that the calculated short-run income elasticity of demand is one. All the other regressors are significant and there are no obvious problems with the specification.

Table 11: Dynamic almost ideal demand system for ‘only food’ with homogeneity imposed:

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th>t-ratio (probability)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Constant</td>
<td>0.23</td>
<td>2.78 (0.01)</td>
</tr>
<tr>
<td>Δ pffo_t</td>
<td>-0.12</td>
<td>-4.11 (0.00)</td>
</tr>
<tr>
<td>Δ (x – PO*)_t</td>
<td>-0.03</td>
<td>-0.50 (0.62)</td>
</tr>
<tr>
<td>sqo_t-1</td>
<td>-0.52</td>
<td>-3.95 (0.00)</td>
</tr>
<tr>
<td>pffo_t-1</td>
<td>-0.12</td>
<td>-3.29 (0.00)</td>
</tr>
<tr>
<td>(x – PO*)_t-1</td>
<td>-0.01</td>
<td>-1.98 (0.06)</td>
</tr>
</tbody>
</table>

Contrary to expectations, the move to the more restrictive category of food does not reduce the price coefficients and elasticities as it did in the previous section. The insignificance of both real expenditure terms suggests unitary income elasticities in the short and long run, while the price elasticities estimates are -1.2 for the short run and -1.3 for the long run. The next section brings together and discusses these results.
6. The Demand for Food

Table 12 presents the elasticity estimates found in the empirical analysis above. Looking at food, tobacco and alcohol results the two methods give similar estimates for the income elasticities 0.5 for the short run and 1 for the long run. They differ in their price elasticity estimates, though both show lower short-run than long-run values. The demand system gives smaller price elasticities, in the short-run it is equal to zero, the long-run elasticities are closer, with the demand equation just over one and the demand system just under one. Using food only, at the cost of degrees of freedom, gives smaller long-run elasticities for the demand equation as we were expecting, but not for the system.

Table 12: Elasticity Estimates

Demand equation: Food, tobacco, alcohol

<table>
<thead>
<tr>
<th>Time period</th>
<th>Price elasticity of demand</th>
<th>Uncompensated</th>
<th>Income elasticity of demand</th>
</tr>
</thead>
<tbody>
<tr>
<td>Short-run</td>
<td>-0.6</td>
<td>-0.5</td>
<td>0.5</td>
</tr>
<tr>
<td>Long-run</td>
<td>-1.2</td>
<td>-1.6</td>
<td>0.9</td>
</tr>
</tbody>
</table>

Demand equation: Food only

<table>
<thead>
<tr>
<th>Time period</th>
<th>Price elasticity of demand</th>
<th>Uncompensated</th>
<th>Income elasticity of demand</th>
</tr>
</thead>
<tbody>
<tr>
<td>Short-run</td>
<td>-0.5</td>
<td>-0.4</td>
<td>0.5</td>
</tr>
<tr>
<td>Long-run</td>
<td>-0.8</td>
<td>-0.9</td>
<td>0.6</td>
</tr>
</tbody>
</table>

Demand system: Food, tobacco, alcohol

<table>
<thead>
<tr>
<th>Time period</th>
<th>Price elasticity of demand</th>
<th>Uncompensated</th>
<th>Income elasticity of demand</th>
</tr>
</thead>
<tbody>
<tr>
<td>Short-run</td>
<td>-0.0</td>
<td>-0.2</td>
<td>0.5</td>
</tr>
<tr>
<td>Long-run</td>
<td>-0.9</td>
<td>-1.3</td>
<td>1.0</td>
</tr>
</tbody>
</table>

Demand system: Food only

<table>
<thead>
<tr>
<th>Time period</th>
<th>Price elasticity of demand</th>
<th>Uncompensated</th>
<th>Income elasticity of demand</th>
</tr>
</thead>
<tbody>
<tr>
<td>Short-run</td>
<td>-1.2</td>
<td>-1.4</td>
<td>0.9</td>
</tr>
<tr>
<td>Long-run</td>
<td>-1.3</td>
<td>-1.5</td>
<td>1.0</td>
</tr>
</tbody>
</table>

Overall, while there are some caveats, the results do allow some conclusions to be drawn. Looking at the long-run elasticities, food, alcohol and tobacco would appear to be price
elastic (although the estimate is -0.9 for the system) suggesting that an increase in price will lead to a more than proportionate decline in demand. Food, the more restrictive category, is price inelastic, though close to one for the demand equation, but price elastic for the system. Food, alcohol and tobacco seem to have unit income elasticity, implying that it will increase proportionately with income. Food alone, again produces different results for the equation, it is income inelastic, and for the system, the income elasticity is unity.

The short-run elasticity estimates are also not completely consistent across the two estimation approaches, but they are generally smaller than the long-run elasticities. Food, alcohol and tobacco would seem to be price and income inelastic, while food alone, is the same for the demand equation, but is price elastic for the system, with almost unit elasticity.

Considering other recent estimates, the Selvanathan and Selvanathan (2003) consumer demand system study, 1960-2001, gives an estimate for the price elasticity of food of -0.32 and for income elasticity of food is 0.81. Balyamujura et al. (2000), using a single equation model of demand described earlier in the paper, estimate that the price elasticity of demand, in terms of the consumer price index, is -0.90 and the income elasticity of the demand for food is 0.58. Other earlier studies found somewhat lower estimates.

Table 13: A summary of earlier studies on elasticity estimates

<table>
<thead>
<tr>
<th></th>
<th>Period</th>
<th>Price Elasticity</th>
<th>Income Elasticity</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dockel and Groenwald (1970)</td>
<td>1947-68</td>
<td>-0.3</td>
<td>0.6</td>
</tr>
<tr>
<td>Barr (1983)</td>
<td>1946-73</td>
<td>SR -0.4</td>
<td>SR 0.95</td>
</tr>
<tr>
<td></td>
<td></td>
<td>LR -0.2</td>
<td>LR 0.62</td>
</tr>
<tr>
<td>Contigiannis (1982)</td>
<td>1960-79</td>
<td>SR -0.2</td>
<td>SR 0.2</td>
</tr>
<tr>
<td></td>
<td></td>
<td>LR -0.1</td>
<td>LR 1.4</td>
</tr>
<tr>
<td>Selvanathan and Selvanathan (2003)</td>
<td>1960-2001</td>
<td>-0.3</td>
<td>0.8</td>
</tr>
<tr>
<td>Balyamujura et al. (2000)</td>
<td>1970-1998</td>
<td>-0.9</td>
<td>0.6</td>
</tr>
</tbody>
</table>

While care must be taken in comparing estimates across different time periods and different studies, the later studies do seem to be getting estimates that suggest food has
higher elasticities (larger negative in the case of prices), but our results are suggesting values beyond the other studies. This illustrates the importance of recognising the need to compare the results of different empirical methods and to be careful of taking the results of any particular empirical analysis as the basis for policy.

7. CONCLUSION

This paper has provided an empirical analysis of the demand for food in South Africa for the period 1970 – 2003. Rather than focussing on one particular method it has taken an ‘ad hoc’ approach, estimating a general dynamic log-linear demand equation, as well as a ‘demand system’ approach, estimating a dynamic form of the almost ideal demand system (Deaton and Muellbauer, 1980). These have both provided estimates of long-run and short-run price and income elasticities, but they were not wholly consistent. Further concerns with the definition of food led to estimates being made for food, alcohol and tobacco and for a more restrictive definition of food, though this was at the expense of degrees of freedom.

The empirical analysis has allowed us to make some tentative conclusions, but possibly more importantly provides an important warning of the need to consider a range of approaches and studies if feeding the results into policy recommendations. Previous studies have often started from similar theoretical background, but differ markedly in the way they operationalise their models, some using single equation estimation, others using systems of equations, some sticking with static models and others introducing dynamic models in a number of ways. Together with possible different definitions of the data and lengths of time series, it is not surprising that there can be a range of results.

With caveats, the results do allow some conclusions to be drawn. Food would seem to be increasingly elastic both in price and income from previous periods. Food, alcohol and tobacco would appear to be price elastic in the long run, suggesting that an increase in price will lead to the same or a more than proportionate decline in demand. It also seems
to have unit income elasticity, implying that it will increase proportionately with income. Food, the more restrictive category, is less elastic, but the results suggest unit elasticity or higher. For income it is either inelastic or unit elasticity.

The short-run elasticity estimates were not completely consistent across the two estimation approaches, but they were generally smaller than the long run. Food, alcohol and tobacco would seem to be price and income inelastic, while for food the result veers between price inelastic and price elastic, with almost unit income elasticity. Comparing these results with earlier studies the elasticities are considerably higher.

Information on elasticity estimates can provide policy makers with an indication of how South African consumers react to changes in the price of food, taxes and changes in income. Such changes may arise directly through food and nutrition polices, or from adjustments within the economy, whether it is the growth of the economy or, for example, some adverse shock to the agriculture sector. With a knowledge of price and expenditure elasticities, policies which aim to alleviate food insecurity and have a positive impact on nutrition, and reduce the impact of HIV/AIDS impact, can be more readily assessed. The size of the elasticities we have estimated should be a clear cause for concern to policymakers, suggesting an increase in the response of food demand to changes in prices.

This paper has made a contribution towards understanding the demand for food, it also acts as a warning of the care needed to undertake and interpret empirical work on the subject. It is also important to recognise the limitations of the study. We are working at an aggregate level and with aggregated categories of food and cannot capture consumer adjustment of their choice of food or spending on other essential goods and services, due

\footnote{Policies aimed directly and indirectly at this, will improve food consumption and help to provide conditions for bodies better able to fight AIDS. Obviously an array of policies can be utilized, as the IFSS (2002, 6) recognizes, from growth of employment opportunities to income supplements and grants. Juxtaposed with such expenditure enhancing policies are also policies which aim to directly increase food consumption by enhancing household food production at it simplest level in the form of vegetable gardens.}
to changes in prices and incomes\textsuperscript{12}. Future work will need to operate at a more detailed level, consider changes in household composition and attempt to understand the source of the difference in the estimates for the different approaches to the modelling used here. In addition, it is important to develop household survey studies, such as Agbola (2003), further to enhance our understanding of the important nutritional issues surrounding the demand for food.

\textbf{References}


\textsuperscript{12} Consumers turning to less nutritious substitutes, or choosing to spend less on other essential consumer goods and services, could impact adversely on health and welfare of households. Similarly, increases in household income could lead to substitution away from basic food products to meat, which may not necessarily improve health. Such aspects suggest that food consumption data need to be more disaggregated.


http://web.utk.edu/_leon/stats571/Project/Rutter_Project.pdf


World Bank website, 2004:
Appendix 1: Average annual growth rates

(a) Gross domestic product (constant prices)

<table>
<thead>
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<th>Period</th>
<th>Mean</th>
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<tbody>
<tr>
<td>1971 – 1980</td>
<td>3.31</td>
</tr>
<tr>
<td>1981 - 1990</td>
<td>1.50</td>
</tr>
<tr>
<td>1991 - 2000</td>
<td>1.73</td>
</tr>
<tr>
<td>2001- 2002</td>
<td>3.50</td>
</tr>
<tr>
<td>1971 - 2002</td>
<td>2.24</td>
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(b) Total consumer spending (constant prices)

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</tr>
<tr>
<td>1981 – 1990</td>
<td>2.77</td>
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<tr>
<td>1991 – 2000</td>
<td>2.34</td>
</tr>
<tr>
<td>2001 – 2002</td>
<td>3.19</td>
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<tr>
<td>1971 - 2002</td>
<td>2.97</td>
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</table>

(c) Food expenditure (constant prices)

<table>
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<td>1971 – 1980</td>
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<tr>
<td>1981 – 1990</td>
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<tr>
<td>1991 – 2000</td>
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<tr>
<td>2001 - 2002</td>
<td>1.61</td>
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<tr>
<td>1971 - 2002</td>
<td>2.26</td>
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</tbody>
</table>

Appendix 2

[Graph showing RGDP, RCONEX, and REXP over years from 1970 to 2002]