

**Seminar in Applied Econometrics**

**An Econometric Analysis of  
Japanese Consumption**

by

**Sylvain Frochaux**

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**Professor: Marius Brülhart**

**Assistant: Aleksandar Georgiev**

**University of Lausanne, Department of Econometrics**

# Contents

1. PREFACE.....	2
2. THEORETICAL BACKGROUND .....	3
2.1. Keynes .....	3
2.2. The Relative Income Hypothesis .....	4
2.3. Past Habits .....	5
3. ECONOMETRIC FRAMEWORK .....	6
3.1. Assumptions .....	6
3.2. R-squared.....	6
3.3. Dickey-Fuller Tests .....	7
3.4. Further Tests .....	8
4. TESTING THE THEORY .....	9
4.1. Data & Description of the Variables .....	9
4.1.1. <i>Consumption</i> .....	9
4.1.2. <i>Income</i> .....	9
4.2 Tests.....	10
4.2.1. <i>Deseasonalization</i> .....	10
4.2.2. <i>Cointegration</i> .....	13
4.2.3. <i>Error Correction Model</i> .....	14
4.2.4. <i>Short and Long-run</i> .....	14
4.2.5. <i>Past Habits</i> .....	16
5. CROSS-SECTION ANALYSIS .....	18
5.1. Data & Description of the Variables .....	18
5.1.1. <i>Consumption</i> .....	18
5.1.2. <i>Income</i> .....	19
5.1.3. <i>Age</i> .....	19
5.1.4. <i>Family Composition</i> .....	19
5.2. Tests.....	20
5.2.1. <i>1<sup>st</sup> Analysis</i> .....	20
5.2.2. <i>2<sup>nd</sup> Analysis</i> .....	22
5.2.3. <i>3<sup>rd</sup> analysis</i> .....	24
5.2.4. <i>4<sup>th</sup> analysis</i> .....	25
6. CONCLUSION .....	27
7. REFERENCES .....	28
8. ANNEXES .....	29

## 1. PREFACE

The improvement in consumption levels in Japan during the post-war period (especially the high-growth period, 1960-75) was truly remarkable. In 1960, Japan's consumption level was just over one-third that of the United States and lower than that of most of the other developed countries as well, but because the level of consumption in Japan showed by far the highest growth rate among developed countries, Japan was able not only to increase its consumption level to nearly three-quarters of the US level but also become one of the six most affluent countries in the developed world.

One of the key elements of such rapid growth in consumption can be seen when looking at the decline in Japan's Engel coefficient, i.e. the budget share of food, beverages, and tobacco. It was a full 44% in 1960 but had declined to 28% by 1975 and to 18% by 1991. Thus it declined by more than one-half in just 29 years, with most of the decline occurring during the high growth period. Moreover, Japan's Engel coefficient showed significant improvement during the 1960-75 period even when compared to that of the United States, declining from 2.1 times to 1.6 times to that of the US level (Horioka 1994).

The purpose of this paper is to analyse the consumption pattern of Japan after the high-growth period, that is 1976-2001. The first two chapters present a theoretical and econometrical framework to the analysis. The following section provides an empirical time series investigation of the various well-accepted consumption model using the statistical software *Stata 7.0* and *E-Views 3.1*. The last part will discuss in a cross-section context the share of various consumption categories in Japan for the year 1999.

## 2. THEORETICAL BACKGROUND

### 2.1. Keynes

The consumption function as it exists today stems from the "fundamental psychological law" stated by Keynes (1936) that "men are disposed, as a rule and on the average, to increase their consumption as their income increases, but not by as much as the increase in their income". Keynes listed a number of "principal objective factors" and modifications of this function, the most important of which, in terms of future economic investigation, can be summarized here:

1. The correct income variable should be personal disposable income, not national income (although Keynes did not use these exact terms).
2. [For]"the wealth-owning class... changes in the money-value of wealth... should be classified amongst the major factors capable of causing short-period changes in the propensity to consume"
3. The short-run marginal propensity to consume (mpc) is less than the long-term mpc because "a man's habitual standard of life usually has the first claim on his income, and he is apt to save the difference which discovers itself between his actual income and the expense of his habitual standard", while over a longer period of time, his standard of living will become more flexible.
4. Even in the long run, "as a rule, a greater proportion of income [will be] saved as real income increases". That is the proportion of income consumed to decrease as income increases, i.e.  $d(C/Y)/dY < 0$ , implying an income elasticity of consumption less than one, and from this follows the argument that redistribution of income in favour of the poor will raise aggregate demand.

In the notation which has become standard for this model I denote consumption by  $C$  and income by  $Y$  and write the consumption function as:

$$C = f(Y), \text{ with } 0 < \frac{dC}{dY} < 1 ,$$

that is consumption increases as income increases, but not by as much as the increase in income. Both time series and cross-section data are relevant to an investigation of the exact form of the consumption function. Time series data from the national income accounts permit the determination of a relation explaining variations in aggregate consumers' expenditure. Cross-section analysis, based on data collected from sample surveys of households, is often concerned with the role of other factors additional to income, such as age, household composition, and so forth.

For both types of data the simplest formulation of the consumption function, useful at least as a first approximation, is the standard linear regression model:

$$C = a + bY + e ,$$

where  $\mathbf{e}$  is a random disturbance term. Here  $C/Y$ , the share of consumption in income or average propensity to consume (apc), decreases as income rises provided that  $\mathbf{a} > 0$ . The marginal propensity to consume (mpc) is assumed to satisfy  $0 < \mathbf{b} < 1$ .

Statement 1 suggests that the appropriate income variable to use in empirical work is personal income, for it is constructed in terms of an individual determining the scale of consumption having net income in mind. Further, the absence of money illusion can be incorporated in this simple linear relation. The saving-consumption decision is assumed independent of the aggregate price level, hence the variables of the consumption function should be entered in real rather than nominal terms (statement 2). Of course with cross-section data it can be reasonably assumed that prices do not vary across the sample, but with aggregate time series data an important first step is to deflate the current values of observed variables by an appropriate price index, to ensure that the correct statistical specification is employed. Statement 3 and 4 will be part of the empirical section.

## 2.2. The Relative Income Hypothesis

Duesenberry (1949) attempted to explain these findings by a reformulation of consumption theory rather than a contribution to statistical methodology. His "relative income" hypothesis has consequences for both cross-section and time-series data. The cross-section effect derives from the hypothesis that a person's consumption behaviour is a function of his position in the income distribution or his "relative income", thus individual consumption behaviour is assumed to be interdependent, not independent. The time series effect depends on the idea that consumption behaviour is not readily reversible, and so people react differently to upward and downward income changes rather than having a simple consumption-income relationship equally applicable to income movements in either direction.

As a simple illustration of the relative income hypothesis, I shall express the  $i$ th household's consumption-income ratio as a function of its relative income:

$$\frac{C_{it}}{Y_{it}} = \mathbf{a} + \mathbf{b} \frac{\bar{Y}_t}{Y_{it}},$$

where  $\bar{Y}_t$  is the mean income of the group to which it belongs. Thus with  $\mathbf{b} > 0$  the share of consumption in income or mpc declines as individual income increases. Equally, writing

$$C_{it} = \mathbf{a}Y_{it} + \mathbf{b}\bar{Y}_t$$

shows that, at a point in time, the cross-section regression within a group has an mpc equal to  $\mathbf{a}$ , and the function supports the declining share hypothesis, having a positive intercept term  $\mathbf{b}\bar{Y}_t$ . On aggregating over both individual households and groups, however, we see that the aggregate variables  $C_t = \sum C_{it}$  and  $Y_t = \sum Y_{it}$  satisfy the relation  $C_t = (\mathbf{a} + \mathbf{b})Y_t$ , thus the aggregate time series model not only yields an mpc of  $(\mathbf{a} + \mathbf{b})$ , greater than the cross-section mpc, but also implies that the share of consumption in income remains constant as income increases over time (that is  $d(C/Y)/dY = 0$ , which contrasts sharply with Keynes's analysis).

### 2.3. Past Habits

Brown (1952) suggested that the influence of past habits is continuous in both directions rather than discrete and irreversible. He represented the dependence on past behaviour by including a lagged dependent variable of consumption:

$$C_t = a + bY_t + dC_{t-1}, \quad 0 < d < 1,$$

which modifies the Duesenberry pattern of behaviour to give continuous partial adjustment of consumption habits. Kuznets (1942) and Goldsmith (1955) empirically obtained similar results using a model without an intercept to demonstrate that the long-run mpc is equal to the apc.

## 3. ECONOMETRIC FRAMEWORK

### 3.1. Assumptions

A time series is a sequence of numerical data in which each item is associated with a particular instant in time. One can quote numerous examples: daily closing prices of stock indices, weekly measures of money supply, monthly unemployment and quarterly consumption as we will develop in this paper. An analysis of a single sequence of data is called univariate time-series analysis. An analysis of several sets of data for the same sequence of time periods is called multivariate time-series analysis or, more simply, multiple time-series analysis. The purpose of time-series analysis is to study the dynamics or temporal structure of the data.

Most aggregate time series such as income and consumption are not stationary because they exhibit time trends. But many time series with trend can be reduced to stationary processes. A process is called trend stationary if it is stationary after subtracting from it a (usually linear) function of time (which is the index  $t$ ). If a process is not stationary but its first difference,  $y_t - y_{t-1}$ , is stationary,  $\{y_t\}$  is called first-difference stationary. The usual assumptions are:

- (1) Constant and finite mean, i.e.  $\mathbf{m}(t) = E(Y_t)$
- (2) Constant and finite variance, i.e.  $\mathbf{s}^2(t) = \text{var}(Y_t)$
- (3) Autocovariances depend only on  $r$  and not on  $t$ , i.e.  $\mathbf{g}(t, t-r) = \text{cov}(Y_t, Y_{t-r})$

⇒ series is mean reverting, not trending, has limited memory.

For a strictly stationary time series the distribution of  $Y_t$  is independent of  $t$ . Thus it is not just the mean and variance that are constant. All higher order moments are independent of  $t$ , so are all higher order moments of the joint distribution of any combinations of the variables  $X_t, X_{t+1}, X_{t+2}, \dots$ . In practice this is a very strong assumption, and it is useful to define stationarity in a less restrictive way. This definition is in terms of first and second moments only.

A time series is said to be weakly stationary if its mean is constant and its autocovariance function depends only on the lag, that is,

$$E[Y_t] = \mathbf{m} \text{ and } \text{cov}[Y_t, Y_{t+r}] = \mathbf{g}(r)$$

### 3.2. R-squared

Time-series observations normally show a strong trend (upward or downward) and strong seasonal effects. Any model that is able to pick up these effects will have a high  $R^2$ . The question is: How good and reliable is this? Note that the criterion on which the usual  $R^2$  is based is the residual sum of squares from the model relative to the residual sum of squares from a naive alternative (that consists of the estimation of the mean only):

$$R^2 = 1 - \frac{RSS}{S_{yy}},$$

where  $S_{yy} = \sum (y_t - \bar{y})^2$  and  $RSS$  is the residual sum of squares. Thus the "norm"  $R^2$  judges a model compared with a naive model, where only the mean is estimated. In time-series models with strong trends and seasonals, this is not a meaningful alternative. The meaningful alternative is a random walk with drift, or with seasonal data, or with both. Harvey (1984) suggests two alternatives  $R^2$  measures. One is

$$R_D^2 = 1 - \frac{RSS}{\sum_{t=2}^T (\Delta y_t - \overline{\Delta y_t})^2}$$

The denominator  $\sum_{t=2}^T (\Delta y_t - \overline{\Delta y_t})^2$  is the residual sum of squares from a random walk with drift, that is,  $y_t = y_{t-1} + \mathbf{b} + \mathbf{e}_t$ ,  $t=2, \dots, T$ . For most time-series data this is the "naive alternative". With seasonal data, Harvey suggests

$$R_S^2 = 1 - \frac{RSS}{RSS_0}$$

where  $RSS$  is, as before, the residual sum of squares from the model and  $RSS_0$  is the residual sum of squares from the naive model - which in this case is the random walk with drift and seasonal dummies, that is,

$$\Delta y_t = \mathbf{a}_1 S_1 + \dots + \mathbf{a}_k S_k + \mathbf{e}_t$$

where  $S_1, S_2, \dots, S_k$  are the  $k$  seasonal dummies. For quarterly data  $k=4$ . A model for which  $R_S^2 < 0$  should be discarded as useless. The software *Stata 7.0* and *E-Views 3.1* do not give these alternative  $R^2$  but only the usual  $R^2$  mentioned above. As a result, the reader ought to be careful with the interpretation of the given  $R^2$ .

### 3.3. Dickey-Fuller Tests

The Dickey-Fuller (DF) test statistics are derived from the estimation of the first-order autoregressive model:

$$y_t = \mathbf{q} y_{t-1} + \mathbf{e}_t$$

If  $\mathbf{q} = 1$ , then  $\Delta y_t = (y_t - y_{t-1}) = \mathbf{e}_t$ . This is what we call a unit root. We can distinguish four typical types of unit-root series, requiring different DF critical-value tables (Maddala 2001):

- (1) random walk:  $\Delta y_t = (\mathbf{q} - 1) y_{t-1} + \mathbf{e}_t$
- (2) random walk with drift:  $\Delta y_t = \mathbf{m} + (\mathbf{q} - 1) y_{t-1} + \mathbf{e}_t$
- (3) random walk with drift and time trend:  $\Delta y_t = \mathbf{m} + \mathbf{b}t + (\mathbf{q} - 1) y_{t-1} + \mathbf{e}_t$
- (4) same as (3), but with a lag number  $> 1$ :  $\Delta y_t = \mathbf{m} + \mathbf{b}t + (\mathbf{q} - 1) y_{t-1} + \sum d_i \Delta y_{t-i} + \mathbf{e}_t$

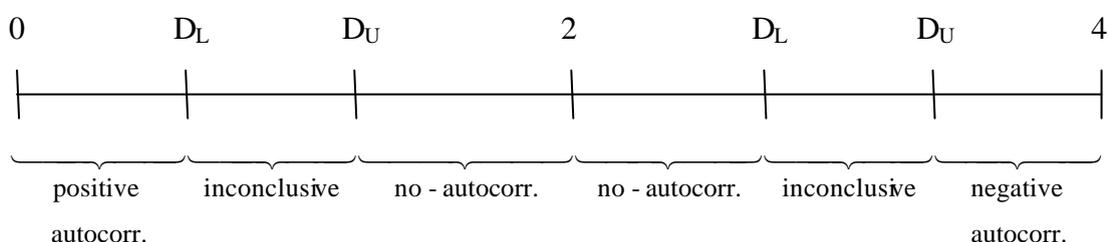
The DF and the augmented Dickey-Fuller (ADF) tests are frequently used in testing for unit roots although there are several problems (size distortions and low power) with these tests. With the ADF test there is also the problem of selection of lag length. When  $\rho < 1$ , but close to 1 (such as 0.95), the theory will predict the variable to be stationary. However, using *Stata 7.0*, the ADF test accepts the hypothesis of a unit root. In the experimental test, I shall discuss more on ADF tests.

### 3.4. Further Tests

The regressions computed in this paper mainly used the method of Ordinary Least Squares (OLS). With time series analysis the first test to do is a ADF test to eliminate a non-stationary trend as discussed above. In order to test whether there is a problem of cointegration, the stationarity of the error terms should be tested.

The major OLS assumptions that can be relaxed are homoskedasticity, i.e.  $\text{var}(\mathbf{e}_i) \neq \text{var}(\mathbf{e}_j)$ , and non-autocorrelation of error terms, i.e.  $\mathbf{e}_t = \rho \mathbf{e}_{t-1} + \mathbf{u}_t$ , where  $0 < |\rho| < 1$ . We can test the first violation by using the Cook and Weisberg test for heteroskedasticity<sup>1</sup>. The null hypothesis is error terms of constant variance. Not accepting it means the error terms are heteroskedastic. If it is the case, in *Stata 7.0* the command *robust* should be used to integrate in the regression a constant variance of errors<sup>2</sup>.

For the autocorrelation of errors, one should use the Durbin-Watson test which computes a test for first-order serial correlation in the disturbance. The *d*-statistics given by Durbin-Watson is as follow (Brühlhart 2002):



where L and U stands for lower and upper-values taken out of the tables at 1% significance points.

If there is a positive or a negative autocorrelation of errors, one should use either the Prais-Winsten or Cochrane-Orcutt regression<sup>3</sup>. It estimates a linear regression that is corrected for first-order serial-correlated residuals. The command *robust* can be added to reflect heteroskedasticity<sup>4</sup>.

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<sup>1</sup> The same test was derived by Breusch and Pagan.  
<sup>2</sup> The command *robust* will not be used when the regression is homoscedastic as it might change *t*-statistic significantly.  
<sup>3</sup> In *Stata 7.0* the command for a Prais-Winsten regression is *prais y x*, and for a Cochrane-Orcutt regression is *prais y x, corc*.  
<sup>4</sup> Note that all estimates from *prais* are conditional on the estimated value of rho. This means that robust variance estimates in this case are only robust to heteroskedasticity and are not generally robust to misspecification of the functional form or omitted variables.

## 4. TESTING THE THEORY

### 4.1. Data & Description of the Variables

In this chapter all data have been computed from *DataStream*.

#### 4.1.1. Consumption

These data come from the "Family Income and Expenditure Survey" and was conducted by the Statistics Bureau of the Ministry of Public Management, Home Affairs, Posts and Telecommunications. The sampling covers the period 15.01.1976 to 15.12.2001, on a monthly base, of about 8000 workers' households<sup>5</sup> around the country excluding agricultural, forestry and fishery households and non-workers' households<sup>6</sup>.

In order to compare in real terms the different variables, nominal consumption has been deflated by a monthly consumer price index (CPI) conducted by the same bureau<sup>7</sup>. The consumption data are net of taxations and social security expenses in order to compare it with real disposable income.

#### 4.1.2. Income

In macroeconomics one often mixes Income and GDP as for the expression of  $Y$ . However, it is important to cut a clear line between those two denominations. Income relates here to what Japanese workers' households receive each month from various sources (salary, dividend, interest, and so on). On the other hand GDP is the total amount of added value of goods and services created by producers with operations in Japan, including foreign-affiliated enterprises and foreign workers. With the study of consumption, from consumers' point of view, it is meaningful to take the real disposable income<sup>8</sup>.

The difference between current and disposable income is taxation and social security expenses that a household pays. Japanese workers often receive bonuses in June or July and especially in December, which can count from a half to a three-month salary. On the other hand, as the Japanese fiscal year starts on April 1<sup>st</sup>, May is the month of payments of taxations and therefore the month of the lowest disposable income. The fourth quarter shows a seasonal tendency higher than the other three quarters<sup>9</sup>.

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<sup>5</sup> On average: 7949 households (min: 7730 ; max: 8081)

<sup>6</sup> That is households of individual business owners, independent professionals and of the unemployed.

<sup>7</sup> Unless consumer savings go directly into the purchase of capital goods, the CPI is to be preferred to a GDP or GNP deflator. In Evans (1969) it was shown that the use of the GNP deflator instead of the consumption deflator will lower the estimate of the short-run mpc but will raise the sum of the coefficients and hence the estimate of the long-run mpc.

<sup>8</sup> The data were computed by the same bureau, for the same period, and with the same number of workers' households as consumption.

<sup>9</sup> May and June are computed together in the second quarter, therefore the month of the lowest disposable income, May, is corrected by the second/third highest, June.

For consistency, since disposable income ought to be deflated with the GDP deflator<sup>10</sup> and the latest is only quarterly available, monthly consumption and disposable income have been transformed to quarterly data. Note that the quarterly data have been computed on a three-month average basis, not on a sum of the three months, and therefore reflect monthly incomes and expenditures.

## 4.2 Tests

### 4.2.1. Deseasonalization

The plot of real disposable income and real consumption shows that the two series are drifting together before 1992, suggesting cointegration, but diverging between 1992 and 2001.

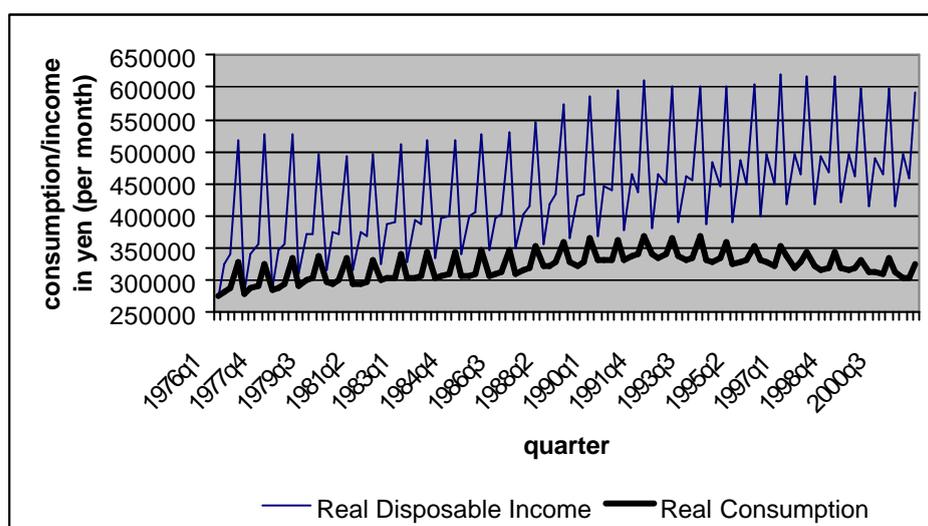


Figure 1: Trend of Consumption and Income (1976 - 2001)

The fourth-quarter clearly shows a seasonal tendency, suggesting the use of a deseasonalization adjuster. To test whether the series are stationary or not, one may use the ADF  $t$ -test without adapting it with the studied issue. The test with both series taken individually will highly reject the null hypothesis, i.e. unit root, non-stationary. The ADF  $t$ -test will in this problem gives a p-value for  $Z(t)$  of 0.0000 suggesting a stationary process of the two series<sup>11</sup>. However, by simply looking at the two graphs, one can clearly see the seasonal tendency and trend.

Several methods are appropriate for such a double problem. In Hayashi (2000) when discussing the cointegrated relationship between income and consumption he conducted an ADF with a constant and time trend to test whether the two series are individually  $I(1)$ , i.e. first-order integrated. Since the series in the United States were also diverging at some point, he suggested to break down the series in two different tests. Using the option trend with the ADF  $t$ -test Hayashi obtained significant results with the first fragmented series but none with the series taken as a whole. One has to mention that the Japanese disposable income (with its multiple bonuses in the fourth-quarter) does not follow the same trend as in the United States and therefore the test suggested here by Hayashi cannot be applied.

<sup>10</sup> Income is reflected through production, that is GDP.

<sup>11</sup> All ADF  $t$ -test are grouped in table 1-4, page 12.

Other sources suggested two different way-outs. First the *E-Views 3.1* and *Stata 7.0*, besides the option trend, offer the option lags(4) in order to adapt the ADF *t*-test to the seasonal tendency<sup>12</sup>. The fourth quarter peaks now do not slant the series and the test indicates that the series are not stationary and therefore should be differentiated.

A second approach is to use a coefficient of deseasonalization. In Bourbonnais (2001), besides the Census X-11, two main sets of seasonality analysis are suggested, multiplicative and additive, using both the ratio to moving average<sup>13</sup>. Multiplicative methods are appropriate if the series can be decomposed into a product of the trend component and a seasonal component, while additive methods are appropriate if the series can be decomposed into a sum of the trend component and a seasonal component. The "band test" allow us to chose between those two methods. This consists to connect in the non-seasonalized graph the upper-values and then the lower-values with a line. If the two lines are parallel, the additive method should be used, otherwise the multiplicative would be appropriate.

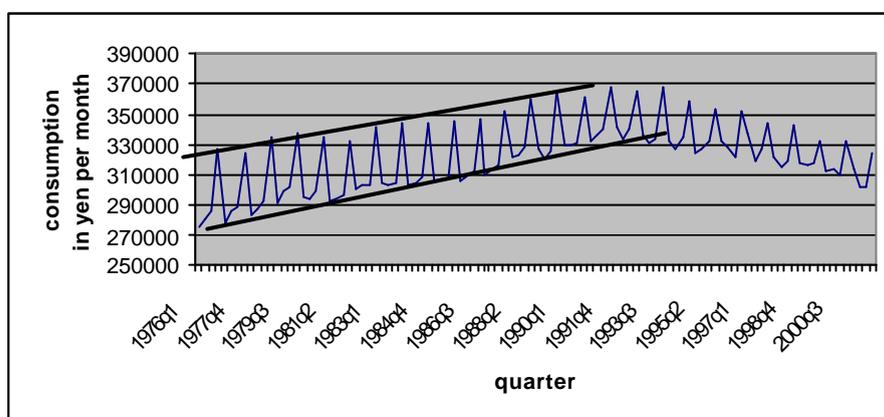


Figure 2: Trend of Consumption (1976 - 2001)

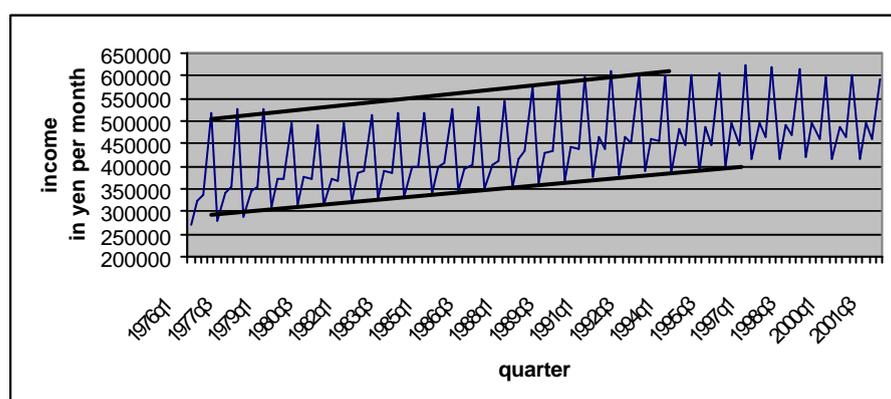


Figure 3: Trend of Income (1976 - 2001)

Visually, the additive ratio to moving average is appropriate and is described as follow:

<sup>12</sup> *E-Views 3.1* suggested in the help function to use at the same time the options trend and lags (4) when testing the relationship between income and consumption.

<sup>13</sup> The Census X-11 methods (multiplicative and additive) are the standard methods used by the U.S. Bureau of Census to seasonally adjust publicly released data. However, even though *E-Views 3.1* has the option for Census X-11, the computer was not able to compute it (not installed). The main difference between X-11 and the ratio to moving average methods is that the seasonal factors may change from year to year in X-11. The seasonal factors are assumed to be constant for the ratio to moving average method, which might distort the results.

- Denote the series to be filtered as  $y_t$ . First compute the centered moving average of  $y_t$  as:  

$$x_t = (0.5y_{t+2} + y_{t+1} + y_t + y_{t-1} + 0.5y_{t-2})/4$$
, for quarterly series
- Take the difference  $d_t = y_t - x_t$ .
- Compute the seasonal indices. For quarterly series, the seasonal index  $i_q$  for quarter  $q$  is the average of  $d_t$  using observations only for quarter  $q$ .
- Adjust the seasonal indices so that they add up to zero. This is done by setting  $s_j = i_j - \bar{i}$  where  $\bar{i}$  is the average of all seasonal indices. These  $s$  are what are reported as scaling factors in *E-Views 3.1*. The interpretation is that the series  $y$  is  $s_j$  higher in period  $j$  relative to the adjusted series.
- The seasonally adjusted series is obtained by subtracting the seasonal factors  $s_j$  from  $y_t$ .

When using the deseasonalization adjuster, the ADF  $t$ -test changes and allow us to use the normal ADF without any options. The results show that both series are non-stationary. Note that the  $t$ -test is stronger for the linear series 76-91 than the non-linear series 76-01, implying that when the series tend to diverge from the linear trend, ADF  $t$ -test has more problem to determine whether the series is stationary or not, since on average the non-linear series tend to go back to previous levels. The following table groups the different ADF  $t$ -test discussed above<sup>14</sup>.

ydisp	Without deseasonalization		overall	Without deseasonalization	
	76-01	76-91		76-01	76-91
normal	-11.155 (0.0000)	-9.463 (0.0000)	normal	-6.945 (0.0000)	-5.586 (0.0000)
trend	-17.021 (0.0000)	-12.063 (0.0000)	trend	-8.154 (0.0000)	-9.985 (0.0000)
trend, lags(4)	-0.197 (0.9928)	-1.259 (0.8969)	trend, lags(4)	0.932 (0.9989)	-1.983 (0.6111)

ydispseas	With deseasonalization		overallseas	With deseasonalization	
	76-01	76-91		76-01	76-91
normal	-2.630 (0.0869)	-2.303 (0.1709)	normal	-2.781 (0.0610)	-1.775 (0.3927)
trend	-7.417 (0.0000)	-7.740 (0.0000)	trend	-2.392 (0.3851)	-6.406 (0.0000)
trend, lags(4)	-0.098 (0.9944)	-1.704 (0.7484)	trend, lags(4)	0.909 (0.9989)	-2.145 (0.5218)

Table 1 - 4: ADF  $t$ -test on data levels

<sup>14</sup> Note that *ydisp* stands for real disposable income, *overall* for real overall consumption, net of taxations and social security expenses, and *res* for residuals. The end-of-word *seas*, if added, denotes a deseasonalized series, and the letter *d* is added in front of a word if the series have been first-order differentiated ( $t$ -statistics are in parenthesis). There is 64 observations for the period 76-91, and 104 for the period 76-01.

For further analysis of the consumption function, I shall use the deseasonalized series. To establish cointegration, we must first check whether each series is integrated and contains a unit root. As a result, since both series are non-stationary, ADF  $t$ -tests on the first-differentiated series were conducted, presented in the following tables.

dydispseas	With deseasonalization		dooverallseas	With deseasonalization	
	76-01	76-91		76-01	76-91
normal	-13.980 (0.0000)	-12.058 (0.0000)	normal	-18.166 (0.0000)	-14.192 (0.0000)
trend, lags(4)	-5.240 (0.0001)	-4.199 (0.0045)	trend, lags(4)	-5.152 (0.0001)	-3.934 (0.0109)

Table 5 - 6: ADF  $t$ -test on first-difference levels

#### 4.2.2. Cointegration

One can assert now that real overall consumption and real disposable income are both integrated of the same order, i.e. I(1). However, in order to use the cointegration process, one has to demonstrate the stationary of the error terms.

resseas	With deseasonalization	
	76-01	76-91
normal	-2.915 (0.0436)	-9.581 (0.0000)

Table 7: ADF  $t$ -test on residuals

The results of the ADF  $t$ -test of the residuals (computed from an OLS on levels of data series) show that there is a cointegrating relationship between the two variables for the period 76-91. As expected, this result corroborates the fact that the visual drifts of the variables tend to have similar trends. However, when looking at a larger sample (76-01) this relationship is not maintain and therefore the conventional regression technique shall not be applied. The Johansen cointegration test allows us to assert that the period 76-91 is cointegrated at a 1% significant level, but the larger sample is not<sup>15</sup>.

	Period 76-01			Period 76-91	
Eigenvalue	0.085859	0.070986	Eigenvalue	0.648631	0.04625
Likelihood ratio	16.83042	7.584066	Likelihood ratio	68.87603	2.98325
5% critical value	15.41	3.76	5% critical value	15.41	3.76
1% critical value	20.04	6.65	1% critical value	20.04	6.65
Hypothesized No. of CE(s)	None *	At most 1 **	Hypothesized No. of CE(s)	None **	At most 1

Table 8 - 9: Johansen Cointegration tests

\*(\*\*) denotes rejection of the hypothesis at 5%(1%) significance level.

Likelihood ratio test indicates one cointegrating equation(s) at 5% critical value.

<sup>15</sup> At a 5% significant level it is assumed that the two variables are cointegrated. This reflects the previous result of the ADF  $t$ -test on the residuals: significant at a 5% level but not at 1%. Note that the Johansen procedure used assumes a linear deterministic trend in the data. Moreover, the interpretation of the  $t$ -statistic and the  $R^2$  should not be regarded as consistent since the two variables are cointegrated.

Having in mind that the two series are cointegrated for the period 76-91, an OLS on levels of data series is applicable. The coefficient of the regression is of **0.4301507** (24.01), that is the quarterly marginal propensity to consume (mpc) is equal to 0.43. Moreover, the regression is homoscedastic with no autocorrelation of the error terms<sup>16</sup>.

### 4.2.3. Error Correction Model

Once the quarterly mpc was computed one wonders what is the short-run adjustment parameter. Since both variables are I(1) and cointegrated, the short-run disequilibrium relationship can be described by the error correction model (ECM). This is known as the Granger representation theorem and takes the general form (Brühlhart 2002):

$$\Delta y_t = \sum_{i=1}^{P_y} d_i \Delta y_{t-i} + \sum_{i=1}^{P_x} g_i \Delta x_{t-i} - \mathbf{I} u_{t-1} + v_t$$

where  $u_{t-1}$  is the disequilibrium error computed from the OLS cointegration regression (i.e. extent of departure from the long-run relationship between  $x$  and  $y$ ), and  $0 \leq \mathbf{I} \leq 1$  is the short-run adjustment parameter. A first attempt with only one lag was performed. However the results were not satisfying as the  $\mathbf{I}$  coefficient was not significantly different from zero implying the two variables were not cointegrated (Bourbonnais 2000). Looking at the correlogram of the disequilibrium error, I noticed that more lagged should have been included. In *Stata 8.0* a lag-order selection statistics reports the final prediction error (FPE), Akaike's information criterion (AIC), the Bayesian information criterion (BIC), and the Hannan and Quinn information criterion (HQIC). All predicted a number of lag as large as 16<sup>17</sup>. However, the  $\mathbf{I}$  coefficient was still not significant with 16 lags and had the wrong sign. I have tried then with several different lags, but none were conclusive<sup>18</sup>. This can be interpreted as real disposable income is not the only variable in the model, and thus this lack of explanatory variables gives a non-significant coefficient of  $\mathbf{I}$ . The sign, however, is negative from the second lag which implies there is an error correction mechanism (that is the variables tend to a long-term relationship).

### 4.2.4. Short and Long-run

Keynes's fourth statement predicted a negative relationship between the apc and disposable income, i.e.  $d(C/Y)/dY < 0$ . For consistency only the period 76-91 has been computed since it has been demonstrated that the two variables are cointegrated. The result of the regression shows a coefficient of  $-8.84e-07$  (-18.48) which is highly significant and with the expected negative sign. This conclusion underlies that Keynes's statement was correct and Duessenberry's was not.

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<sup>16</sup> P-value of the Cook-Weisberg heteroskedasticity test: 0.5590, implying homoskedasticity. The Durbin-Watson autocorrelation test: 2.40048 which is below the DL (2.5978) described in the theory, implying non-autocorrelation of the error term. See the annex for the correlogram of the error-term.

<sup>17</sup> BIC predicted a number of lag of 17.

<sup>18</sup> 16 lags: 1.406548 (0.81); 14 lags: 0.480776 (0.53); 10 lags: -0.2838217 (-0.73); 7 lags: -0.3194889 (-1.14); 6 lags: -0.342654 (-1.24); 5 lags: -0.2743507 (-1.03); 4 lags: -0.3807905 (-1.54); 3 lags: -0.2662712 (-1.04); 2 lags: -0.2205204 (-0.86); 1 lag: 0.0806047 (0.39). Note that the sign should be negative.

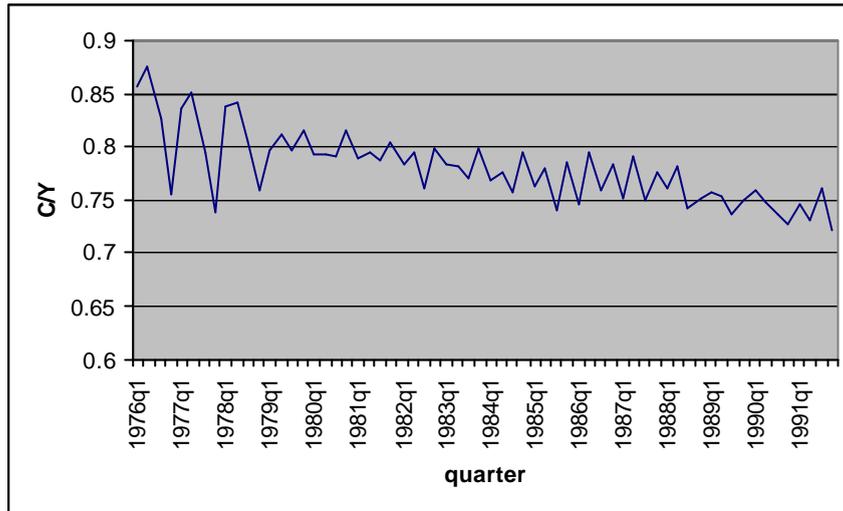


Figure 4: Trend of apc (1976 - 1991)

We can think of the short-run consumption function approaching the long-run function in at least two ways. If income is growing over time, each short-run function (with the same slope) will have a higher intercept as last period's consumption increases. If we consider the relevant levels of income at each time period, the points on the short-run functions trace out a long-run function that goes through the origin and has a slope equal to the apc. This process is illustrated in Figure 5. If income remains constant over time, we can think of the slope of the short-run consumption function as increasing continuously as time progresses until it finally merges with the long-run function. This process is shown in Figure 6. The two figures illustrate how the mpc approaches the apc over time. In the long run, after the consumer has become fully accustomed to receiving a given amount of additional income, he will spend the same proportion of this income as of all his other income.

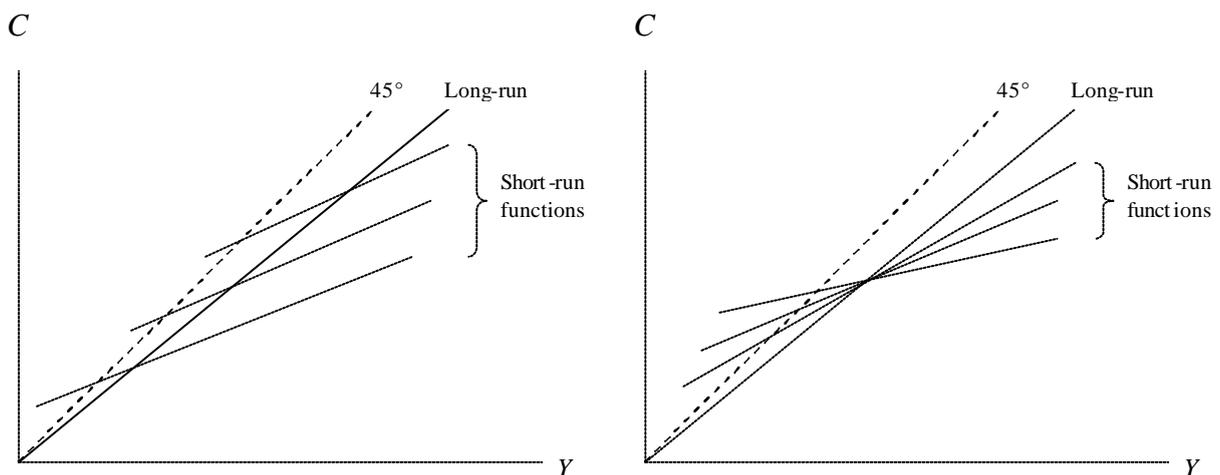


Figure 5 - 6: Dynamics of short-run functions

This view of the mpc has been advanced most strongly by Friedman (1957), but it also incorporated in the work of Duessenberry (1949), and others. Friedman showed how this concept of consumer behaviour logically leads to the conclusion that the long-run mpc must be equal to the apc. It is of interest to compare both yearly and quarterly functions to see if the

mpc of the first four quarters is approximately equal to the mpc estimated from the yearly function.

quarterly	yearly	long-run
0.4301507	0.5065242	0.7747747
(24.01)	(20.88)	(196.04)

Table 10: Various mpc's computed

This provides an additional check of internal consistency of the estimates and also may be useful in determining the approximate magnitudes of some of the biases. The coefficients of the variables in all the consumption functions chosen for comparison are highly significant in all cases and give reasonable values of the short/middle-run and long-run mpc's<sup>19</sup>. One can assert that the short-run marginal propensity to consume is less than the long-term mpc which corroborates Keynes's third statement. Note that the short-run adjustment parameter (using the ECM) ought to be smaller than the quarterly coefficient. The following graph show the trend of the computed long-term consumption and consumption<sup>20</sup>.

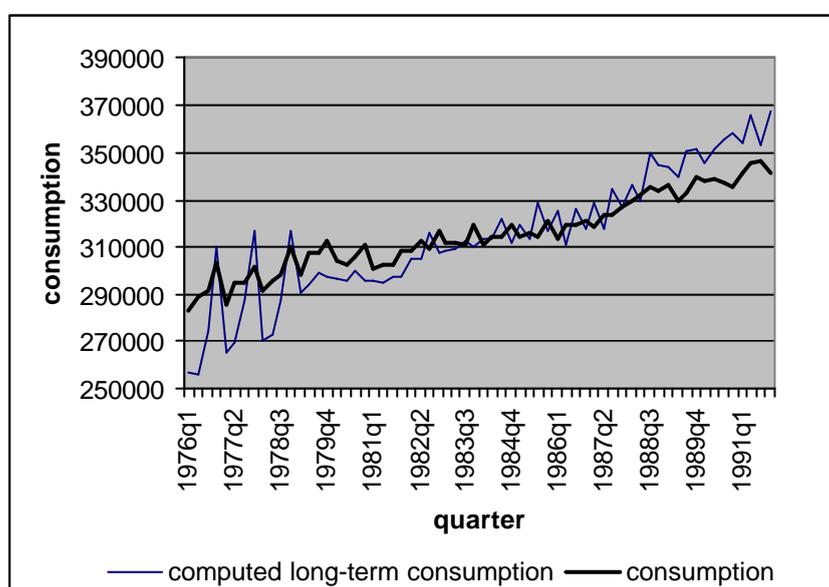


Figure 7: Trend of consumption and computed long-term consumption (1976 - 1991)

#### 4.2.5. Past Habits

Besides Keynes's fundamental consumption regression, one of the most commonly estimated forms of the consumption function is Brown's past habit model, that is:

$$C_t = a + bY_t + dC_{t-1} + u_t$$

<sup>19</sup> The long-run equation is an OLS regression without the intercept. Quarterly data have been used to compute it since it has four-times more observations. Note that the yearly-based long-term coefficient is of 0.7317906 (151.17)

<sup>20</sup> Note that the slope of the computed consumption is bigger than consumption. That is since the slope of disposable income is bigger than consumption.

The following table maintains the assumption that long-term mpc is bigger than the short-term. However, as for the lagged consumption, it seems that the influence of the previous quarter has a stronger impact on today's consumption compared with the previous year.

	quarterly	yearly
$Y_t$	0.2265756 (6.14)	0.3179475 (4.91)
$C_{t-1}$	0.5074759 (6.47)	0.340912 (2.49)

Table 11: Coefficients of Brown's regression

In time-series analysis multi-collinearity often occurs when several of the variables in a given function have very similar trends or other movements throughout the entire sample period. In this case it is not possible in a statistical sense to determine which of the independent variables is actually causing the dependent variable to change. In the case of the consumption function, both income and lagged consumption follow smooth upward patterns. Either variable could be used to explain a large part of the movements in consumption, and it is often not possible to sort out the independent contributions of each variable.

Finally, Kuznets (1942) and Goldsmith (1955) assert that a model without an intercept, that is  $C_t = bY_t + dC_{t-1} + u_t$ , will demonstrate that the long-run mpc is equal to the apc. The following table reproduces the computed coefficients.

	quarterly	yearly
$Y_t$	0.1151312 (2.87)	0.1101398 (1.84)
$C_{t-1}$	0.8538025 (16.45)	0.8581899 (10.36)

Table 12: Coefficients of Kuznets and Goldsmith's regression

Having in mind the problem of multi-collinearity, one can assert that the income coefficient is again underestimated. In this chapter, the time-series breakdown was introduced to reproduce for the period 1976-91 a Japanese long-term mpc of 0.77 and a quarterly mpc of 0.43. When taking a closer look at figure 2 and 3, it is worth noting that the period 1992-2001 is far from the Keynesian consumption pattern as real disposable income is stagnating and real consumption is decreasing. The ADF  $t$ -test of the deseasonalized variables show that income is stationary on data levels, i.e. I(0), and consumption is first-difference stationary, i.e. I(1), therefore the process of cointegration cannot be applied<sup>21</sup>. The structural basement of the Japanese economy has collapsed in the early 1990s. As a result, Japanese workers' households have received constant real disposable income for the last ten years, but have decreased their real consumption expenditures in order to save more for the uncertain future.

<sup>21</sup> See the annex for the detailed ADF  $t$ -test results.

## 5. CROSS-SECTION ANALYSIS

### 5.1. Data & Description of the Variables

The following analysis is based on a survey conducted by the Ministry of Public Management, Home Affairs, Posts and Telecommunications. The data were obtained in three kinds of forms, namely, Household Schedule, Family Account Book and Yearly Income Schedule. Enumerators fill in the Household Schedule with the number of household members, occupation and industry of earners, type of the dwelling, etc. Households are requested to fill in the Family Account Books with daily incomes and expenditures for six months. Furthermore, by using the Yearly Income Schedule yearly income is surveyed by self-entry for all households. The survey covers about 8000 two-or-more-person workers' households, however only the averages for the year 1999 were available. The analysis will stress on the following variables.

#### 5.1.1. Consumption

Classified in 11 different groups<sup>22</sup>:

- Food
- Housing
- Fuel (& Light & Water Charges)
- Furniture (& Household Utensils)
- Clothing (& Footwear)
- Medical Care
- Transportation (& Communication)
- Education
- Reading (& Recreation)
- Other (Expenses)
- Non-consumption (including taxations and social security expenses)

A first remark has to be made for *housing*. Considered as part of the category *housing*, expenditures on house in which one currently dwells and in which one does not currently dwell as well as expenditure on housing lot and related services. However, expenditures on house for rent, a business expense, is excluded of any category. The purchase, construction, expansion and renovation of house or land is deemed "purchase of property" while the repayment of housing loan is deemed "repayment of loans on land and house". Such expenses are not included in housing expenses but in *non-consumption expenditures*. As a result the category *housing* shares only 7.1% of workers' income.

A second remark is worthwhile. The category *other expenses* includes various types of expenditures such as beauty expenses, personal effects, cigarettes, social expenses and money sent. This group is the largest with 25.6% of all expenditures in 1999. Therefore, one should

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<sup>22</sup> The words which are not in parenthesis are the names given to the categories for the econometrical tests.

not jump to any conclusion when a coefficient of this explanatory variable is larger than average.

### 5.1.2. Income

Income here is to be considered as prior to any payments of taxations or social security, and has been coded to include open-groups. However, in the first analysis, in order to test Duessenberry's assumption, the median disposable income of each income category was used.

### 5.1.3. Age

In order to include the shape of the age curve<sup>23</sup>, the variable age-squared has also been introduced. Depending on the analysis, age and age-squared were coded as it is expressed below.

### 5.1.4. Family Composition

The 4<sup>th</sup> analysis will test whether the number of children and the level of education of the eldest is significant or not<sup>24</sup> and is coded as follow:

Income per month (thousands of yen)		Age					Education		Children	
1 <sup>st</sup> analysis	Code	3 <sup>rd</sup> & 4 <sup>th</sup> analysis	2 <sup>nd</sup> analysis	Code	Code <sup>2</sup>	3 <sup>rd</sup> analysis	4 <sup>th</sup> analysis	Code	4 <sup>th</sup> analysis	Code
<200	10	<200	<25	1	1	<30	Preschool	100	1	1
200-250	20	200-300	25-29	2	4	30-39	Elem+JHS	200	2	2
250-300	30	300-400	30-34	3	9	40-49	SHS	300	>3	3
300-350	40	400-500	35-39	4	16	50-59	Univ+Coll	400		
350-400	50	500-600	40-44	5	25	60-69	Others	500		
400-450	60	600-800	45-49	6	36	>70				
450-500	70	800-1000	50-54	7	49					
500-550	80	1000-1250	55-59	8	64					
550-600	90	1250-1500	60-64	9	81					
600-650	100	>1500	65-69	10	100					
650-700	110		>70	11	121					
700-750	120									
750-800	130									
800-900	140									
900-1000	150									
1000-1250	160									
1250-1500	170									
1500-2000	180									
>2000	190									

Table 13: Variable coding

<sup>23</sup> The theory predicts that the age for maximum consumption is at around 50 years old, and the average consumption expenditures decrease passing this age.

<sup>24</sup> The workers' households with no children are thus not considered in the 4<sup>th</sup> analysis.

## 5.2. Tests

### 5.2.1. 1<sup>st</sup> Analysis

$$C_i = f(\text{Income})$$

Duessenberry (1949) maintained that the aggregate time series model yields an mpc greater than the cross-section mpc. Several problems arise when one is willing to test this assumption. First of all, the studied Japanese time series model showed problems of consistency and had to be broken down in two distinct periods. Secondly, the series showed unusually seasonal tendencies and had to be deseasonalized. Finally, one cannot compare two models when the number of observations is different, and should not compare two different types of model, such as time series and cross-section. With all this in mind, the following table reproduced the results of the time series model and the cross-section model:

	time series n = 64	cross-section n = 17
with a constant (short-term)	0.5065242 (20.88)	0.5958333 (34.40)
without a constant (long term)	0.7747747 (196.04)	0.7371857 (40.31)

Table 14: Duessenberry's assumption

In theory, Duessenberry showed the inequality between the two mpc's. However, it has been demonstrated above that empirically the time series model does not always yield an mpc greater than the cross-section mpc. This short analysis was put in place for the sake of testing Duessenberry's assumption, but for the reason mentioned above it should not be taken seriously. The following four cross-section analysis are of greater interest. Note that all explanatory variables have been coded, therefore the coefficient are interesting only for the magnitude of each group, and not for the calculation of any mpc's. In this analysis and in the third one, income has to be considered as overall income (not disposable income).

The following graph shows the share of each category in 1999 consumption expenditures. One can easily notice that the categories *food*, *other* and *non-consumption* account for 60% of one worker's household.

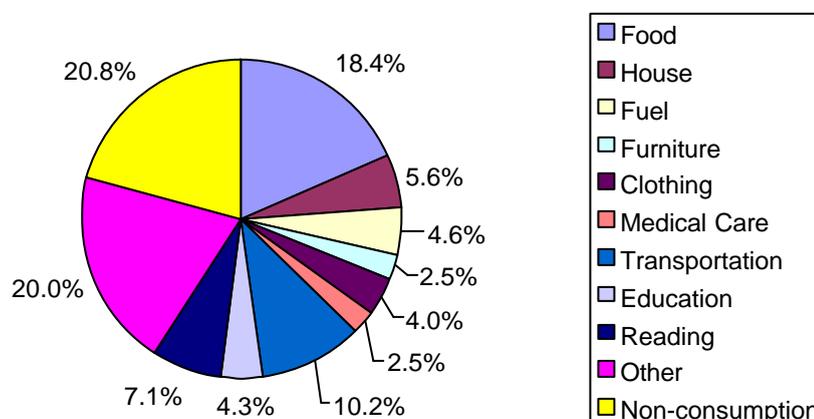


Figure 8: Share of each category (1999)

Since the size of each category is not equivalent, the coefficient of income, if significant, will not tell much. In order to measure the income-elasticity for each category, one has to divide the obtained coefficient by its overall share, that is to compute the relative increase in expenditure for an increase in income.

N = 19

Dependent Variable	Regression	Coeff. Income	Share	Relative Coeff.	R <sup>2</sup>
Overall	OLS, robust	2228.535 (9.34)	100%	1.00	0.8928
Food	OLS	353.0495 (28.68)	18.4%	0.86	0.9798
Housing	OLS	-20.12298 (-1.49)	5.6%	-0.16	0.1152
Fuel	OLS, robust	63.14719 (11.66)	4.6%	0.62	0.9325
Furniture	OLS, robust	71.0193 (8.74)	2.5%	1.26	0.8981
Clothing	OLS, robust	147.5233 (6.26)	4.0%	1.65	0.8114
Medical Care	OLS, robust	40.53298 (5.21)	2.5%	0.73	0.7141
Transportation	OLS, robust	277.3825 (7.17)	10.2%	1.22	0.8299
Education	OLS	169.8686 (18.49)	4.3%	1.77	0.9526
Reading	OLS, robust	231.2439 (13.23)	7.1%	1.47	0.9464
Other	OLS	894.9832 (10.02)	20.0%	2.00	0.8552
Non-Consumption	OLS, robust	1077.901 (5.36)	20.8%	2.32	0.7690

Table 15: Regression on income

Except for the category *housing*, income is significant for all categories. Looking at the relative coefficient, one can notice that the expenditures in *non-consumption* will increase relatively more than the other categories when income raises. This indicates that the marginal taxation rate increases as income increases. Besides the broad category *other*, the groups *education* and *clothing* show the higher income-elasticity, that is when income increases, those two categories will raise their expenditures more than the average, which implies that they are luxury goods. On the other hand, food for instance with a relative coefficient at 0.86 is categorised as a normal good. The Engel coefficient, i.e. the budget share of food and beverages drops from a share of 28% for the poorest to a share of 18% for people with an income over 2'000'000 yen per month.

### 5.2.2. 2<sup>nd</sup> Analysis

$$C_i = f(\text{Age}, \text{Age}^2), \quad N = 11$$

Dependent Variable	Coeff. Age	Coeff. Age <sup>2</sup>	R <sup>2</sup>
Overall	72985.23 (6.43)	-5204.497 (-5.65)	0.8589
Food	17719.91 (8.63)	-1220.221 (-7.32)	0.9240
Housing	-8785.706 (-6.01)	569.0816 (4.80)	0.8787
Fuel	3211.644 (7.75)	-218.4044 (-6.49)	0.9108
Furniture	1953.021 (7.75)	-113.979 (-5.58)	0.9445
Clothing	2701.001 (4.50)	-173.1259 (-3.55)	0.8078
Medical Care	-941.1573 (-2.76)	93.81235 (3.39)	0.6822
Transportation	7402.753 (3.82)	-625.3438 (-3.97)	0.6642
Education	13677.27 (3.59)	-1158.501 (-3.75)	0.6381
Reading	6095.26 (5.49)	-396.352 (-4.40)	0.8563
Other	29951.13 (3.54)	-1961.455 (-2.86)	0.7079
Non-Consumption	36140.64 (5.88)	-2919.317 (-5.85)	0.8132

Table 16: Regression on age

First of all, it is of importance to have in mind that there are only eleven observations for age and therefore the results might not reflect the real trend. However, even with such a small number of observations, all variables tend to have a high correlation with age and age-squared. Since all the regressions are homoscedastic, a simple OLS regression was used. The following graph shows the relationship between age and general consumption.

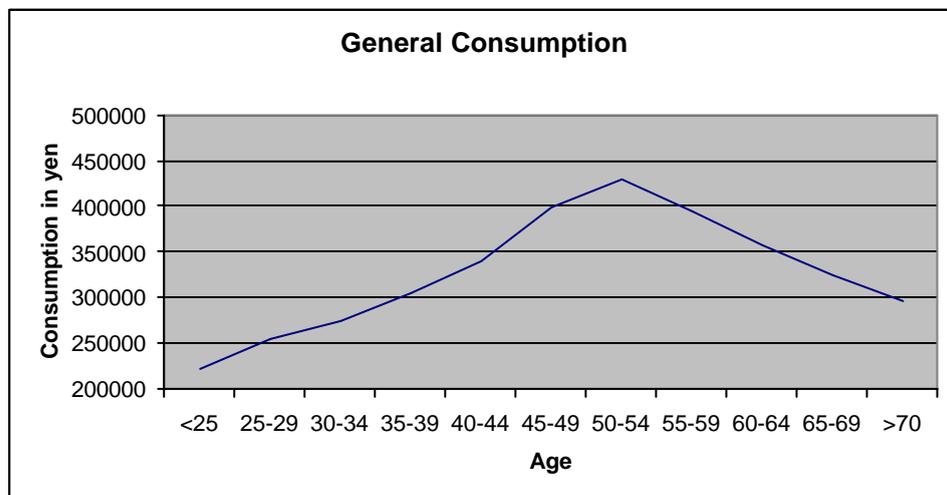


Figure 9: Breakdown of overall consumption by age

The categories *housing* and *medical care* have a sign opposite to what we would have expected, that is a positive relationship between age and consumption. This can be explained by the fact that, first, young people cannot usually afford to purchase a property and second housing as included in "business expenses" is more often part of experienced workers' contracts<sup>25</sup>.

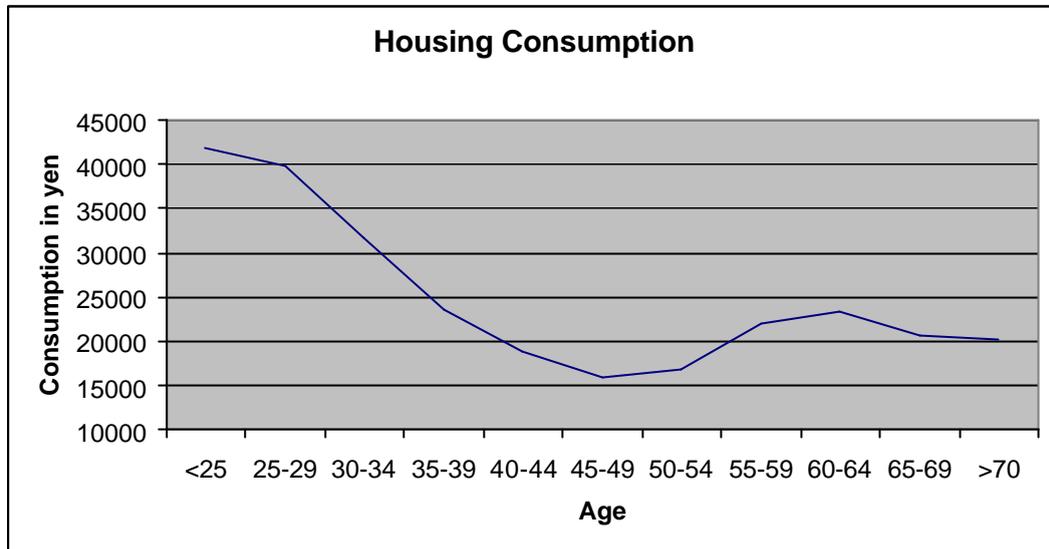


Figure 10: Breakdown of housing consumption by age

When overall consumption tends to decrease after passing 50 years old, the medical care expenditures increase to match the fact that older people are inclined to spend more on medical care than average.

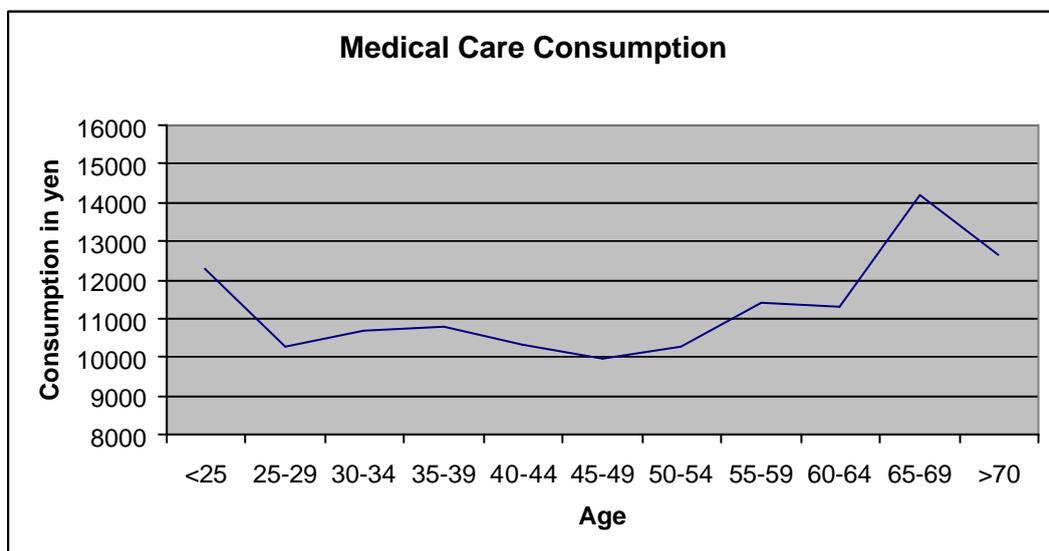


Figure 11: Breakdown of medical care consumption by age

<sup>25</sup> See the description of the consumption variable for a discussion of "business expenses".

### 5.2.3. 3<sup>rd</sup> analysis

$$C_i = f(\text{Age}, \text{Age}^2, \text{Income}), \quad N = 60$$

In this section, age and income variables have been grouped to smaller entities but overall the number of observations has increased to 60, which gives a stronger base for analysis.

Dependent Variable	Regression	Coeff. Age	Coeff. Age <sup>2</sup>	Coeff. Income	R <sup>2</sup>
Overall	OLS, robust	20659.72 (0.49)	-3810.87 (-0.77)	4487.358 (8.96)	0.7561
Food	OLS, robust	16462.8 (7.75)	-2019.918 (-6.52)	675.5609 (26.38)	0.9366
Housing	OLS, robust	-16376.85 (-2.81)	1719.332 (2.43)	98.31545 (1.45)	0.4113
Fuel	OLS	3364.708 (4.97)	-380.1607 (-4.01)	124.6362 (15.14)	0.8289
Furniture	OLS, robust	1501.346 (2.06)	-148.6911 (-1.27)	123.1354 (12.90)	0.7952
Clothing	OLS, robust	-3303.349 (-0.91)	429.7107 (0.92)	333.0736 (8.17)	0.6692
Medical Care	OLS, robust	-3738.783 (-3.49)	512.8196 (3.59)	115.2737 (10.09)	0.7134
Transportation	OLS, robust	-12204.93 (-0.73)	987.8232 (0.51)	698.6341 (3.53)	0.3795
Education	OLS, robust	20514.46 (7.59)	-3144.17 (-7.89)	243.7387 (6.05)	0.6578
Reading	OLS, robust	-7342.816 (-1.12)	882.5125 (1.14)	609.0838 (8.04)	0.7109
Other	OLS, robust	21682.29 (1.78)	-2717.584 (-1.74)	1500.205 (9.19)	0.7458
Non-Consumption	OLS, robust	38152.54 (3.84)	-5891.116 (-4.19)	1708.611 (11.54)	0.8270

Table 17: Regression on age and income

With more observations the regressions become heteroskedastic and therefore the OLS was in ten out of the eleven cases adjusted with the *robust* command. First of all, one can note that except for the category *housing* income is always a significant factor for the determination of consumption. However, even though the signs stay the same, as for the age and age-squared variables the impact is lower than expected and is significant in only six out of the eleven regressions. This might be the result of multi-collinearity between income and age which can be explained by the fact that in Japan large companies follow the triple principle: seniority-based wage system, lifetime employment and unionism. The seniority-based wage system means that the longer one works for a specific company, the higher the income one receives. Therefore there is a strong relationship between age and income<sup>26</sup>. Moreover, as one can see on the following figures, there is a larger dispersion in age compared to income, that is the coefficient for age will be less significant<sup>27</sup>.

<sup>26</sup> In *Stata 7.0* the command *vif*, which calculates the variance inflation factor for the independent variables specified in the fitted model, computed a mean *vif* of 16.31 implying high multi-collinearity, especially since age-squared is perfectly correlated with the variable age.

<sup>27</sup> *Stata 7.0* takes the codes indicated above for the scales of the explanatory variables.

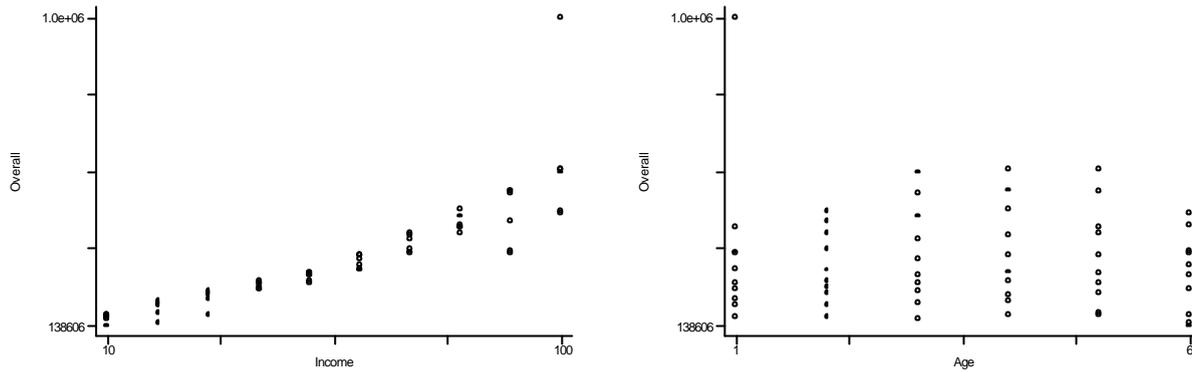


Figure 12 - 13: Income and age distribution

#### 5.2.4. 4<sup>th</sup> analysis

$$C_i = f(\text{Children}, \text{Education}, \text{Income}), \quad N = 150$$

Dependent Variable	Regression	Coeff. Children	Coeff. Education	Coeff. Income	R <sup>2</sup>
Overall	OLS, robust	7962.813 (1.26)	171.8814 (5.21)	4221.751 (23.03)	0.8163
Food	OLS, robust	9187.397 (9.04)	38.58507 (6.66)	620.6672 (20.57)	0.8113
Housing	OLS, robust	-101.042 (-0.09)	-35.1802 (-5.45)	23.14982 (0.69)	0.1820
Fuel	OLS	1905.858 (8.06)	12.51027 (9.28)	65.98612 (9.40)	0.6577
Furniture	OLS, robust	-237.004 (-0.75)	2.233849 (1.40)	120.4733 (8.43)	0.4619
Clothing	OLS, robust	-831.8619 (-2.29)	-5.19541 (-2.55)	304.7768 (22.73)	0.8575
Medical Care	OLS, robust	572.4554 (1.51)	-9.596638 (-4.44)	65.79839 (5.84)	0.2807
Transportation	OLS, robust	1316.764 (0.75)	20.17403 (1.95)	449.7743 (7.46)	0.4334
Education	OLS, robust	12940.61 (3.31)	83.4779 (3.84)	544.5716 (4.46)	0.2703
Reading	OLS, robust	819.8321 (0.99)	-29.4431 (-2.90)	596.8813 (10.12)	0.6504
Other	OLS, robust	-18128.44 (-6.47)	100.7372 (7.94)	1392.624 (15.22)	0.7707
Non-Consumption	OLS, robust	-4596.395 (-1.85)	10.29091 (0.79)	2153.589 (21.61)	0.8647

Table 18: Regression on children, education and income

Except the category *housing* all eleven other variables have a highly significant correlation with income<sup>28</sup>. The explanatory variable *education* was mainly introduced to test the assumption that as the level of education in one workers' household increases the expenditures for education will also increase since private institutions in Japan can get extremely costly.

<sup>28</sup> Note that the category *fuel* is as in the 3<sup>rd</sup> analysis homoscedastic, and therefore the *robust* command has not been used.

Except the categories *other* and *non-consumption*, which are always the category with the bigger share, the category *education* encounters an important increase of its expenditures when the level of education of workers' children increases. However, education is not the only category to be influenced by the level of education of worker's children. Nine out of the eleven variables are highly correlated with it.

As for the impact of having more children, the expected categories *food* and *education* are the ones which have the absolute and relative stronger correlation with the explanatory variable *children*. Nevertheless, only four categories are significantly related to the number of children in one's workers' household. Note that there should not be a problem of multicollinearity since the variable *education* reflects the level of education of the eldest independent of the number of children.

## 6. CONCLUSION

The investigation of Japanese consumption pattern since 1976 showed that Japan, which pursued the well-accepted Keynesian consumption model until the early nineties, has encountered an unprecedented divergence of consumption to its disposable income. The crisis striking for more than a decade now has forced the Japanese to save more than before which illustrates the fear of a country that has not seen its real national income increase more than a percent and a half per year since 1991<sup>29</sup>.

Summing up, this paper set a specific econometric discussion on unit root tests, the deseasonalization process, and cointegration. The various time series tests confirmed that the neo-classical consumption functions were not always empirically valid, but offered a short and long-run perspective of the Japanese marginal propensity to consume. The cross-section analysis allowed the reader to take a closer look at the income-elasticity of various consumption categories, the relative importance of the family composition and the curve of the age factor.

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<sup>29</sup>Except 1996: 3.3%. Note that the average for the period 76-91 was of 4.1%, and the high-growth period (60-75) of 7.7%

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## 8. ANNEXES

**TABLE 1-2:**

**76-01**

dfuller ydisp

Dickey-Fuller test for unit root                      Number of obs = 103

----- Interpolated Dickey-Fuller -----			
Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-11.155	-3.509	-2.890

\* MacKinnon approximate p-value for Z(t) = 0.0000

. dfuller ydisp, trend

Dickey-Fuller test for unit root                      Number of obs = 103

----- Interpolated Dickey-Fuller -----			
Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-17.021	-4.039	-3.450

\* MacKinnon approximate p-value for Z(t) = 0.0000

. dfuller ydisp, trend lags(4)

Augmented Dickey-Fuller test for unit root      Number of obs = 99

----- Interpolated Dickey-Fuller -----			
Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-0.197	-4.042	-3.451

\* MacKinnon approximate p-value for Z(t) = 0.9928

. dfuller overall

Dickey-Fuller test for unit root                      Number of obs = 103

----- Interpolated Dickey-Fuller -----			
Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-6.945	-3.509	-2.890

\* MacKinnon approximate p-value for Z(t) = 0.0000

. dfuller overall, trend

Dickey-Fuller test for unit root                      Number of obs = 103

----- Interpolated Dickey-Fuller -----			
Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-8.154	-4.039	-3.450

\* MacKinnon approximate p-value for Z(t) = 0.0000

. dfuller overall, trend lags(4)

Augmented Dickey-Fuller test for unit root      Number of obs = 99

----- Interpolated Dickey-Fuller -----

Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	0.932	-4.042	-3.151

\* MacKinnon approximate p-value for Z(t) = 0.9989

76-91

. dfuller ydisp91

Dickey-Fuller test for unit root                      Number of obs =    63

----- Interpolated Dickey-Fuller -----			
Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-9.463	-3.562	-2.595

\* MacKinnon approximate p-value for Z(t) = 0.0000

. dfuller ydisp91, trend

Dickey-Fuller test for unit root                      Number of obs =    63

----- Interpolated Dickey-Fuller -----			
Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-12.063	-4.121	-3.172

\* MacKinnon approximate p-value for Z(t) = 0.0000

. dfuller ydisp91, trend lags(4)

Augmented Dickey-Fuller test for unit root        Number of obs =    59

----- Interpolated Dickey-Fuller -----			
Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-1.259	-4.130	-3.175

\* MacKinnon approximate p-value for Z(t) = 0.8969

. dfuller overall91

Dickey-Fuller test for unit root                      Number of obs =    63

----- Interpolated Dickey-Fuller -----			
Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-5.586	-3.562	-2.595

\* MacKinnon approximate p-value for Z(t) = 0.0000

. dfuller overall91, trend

Dickey-Fuller test for unit root                      Number of obs =    63

----- Interpolated Dickey-Fuller -----			
Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-9.985	-4.121	-3.172

\* MacKinnon approximate p-value for Z(t) = 0.0000

. dfuller overall91, trend lags(4)

Augmented Dickey-Fuller test for unit root        Number of obs =    59

----- Interpolated Dickey-Fuller -----			
Test	1% Critical	5% Critical	10% Critical

Statistic	Value	Value	Value
Z(t)	-1.983	-4.130	-3.175

\* MacKinnon approximate p-value for Z(t) = 0.6111

**TABLE 3 - 4:**

**76-91**

. dfuller ydispseas

Dickey-Fuller test for unit root                      Number of obs =    63

----- Interpolated Dickey-Fuller -----

Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-2.303	-3.562	-2.920

\* MacKinnon approximate p-value for Z(t) = 0.1709

. dfuller ydispseas, trend

Dickey-Fuller test for unit root                      Number of obs =    63

----- Interpolated Dickey-Fuller -----

Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-7.740	-4.121	-3.487

\* MacKinnon approximate p-value for Z(t) = 0.0000

. dfuller ydispseas, trend lags(4)

Augmented Dickey-Fuller test for unit root        Number of obs =    59

----- Interpolated Dickey-Fuller -----

Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-1.704	-4.130	-3.491

\* MacKinnon approximate p-value for Z(t) = 0.7484

. dfuller overallseas

Dickey-Fuller test for unit root                      Number of obs =    63

----- Interpolated Dickey-Fuller -----

Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-1.775	-3.562	-2.920

\* MacKinnon approximate p-value for Z(t) = 0.3927

. dfuller overallseas, trend

Dickey-Fuller test for unit root                      Number of obs =    63

----- Interpolated Dickey-Fuller -----

Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-6.406	-4.121	-3.487

\* MacKinnon approximate p-value for Z(t) = 0.0000

. dfuller overallseas, trend lags(4)

Augmented Dickey-Fuller test for unit root        Number of obs =    59

```

----- Interpolated Dickey-Fuller -----
Test      1% Critical  5% Critical  10% Critical
Statistic   Value      Value      Value
Z(t)      -2.145     -4.130     -3.491     -3.175

```

\* MacKinnon approximate p-value for Z(t) = 0.5218

## 76-01

```
. dfuller ydispseas
```

Dickey-Fuller test for unit root          Number of obs =    103

```

----- Interpolated Dickey-Fuller -----
Test      1% Critical  5% Critical  10% Critical
Statistic   Value      Value      Value
Z(t)      -2.630     -3.509     -2.890     -2.580

```

\* MacKinnon approximate p-value for Z(t) = 0.0869

```
. dfuller ydispseas, trend
```

Dickey-Fuller test for unit root          Number of obs =    103

```

----- Interpolated Dickey-Fuller -----
Test      1% Critical  5% Critical  10% Critical
Statistic   Value      Value      Value
Z(t)      -7.417     -4.039     -3.450     -3.150

```

\* MacKinnon approximate p-value for Z(t) = 0.0000

```
. dfuller ydispseas, trend lags(4)
```

Augmented Dickey-Fuller test for unit root          Number of obs =    99

```

----- Interpolated Dickey-Fuller -----
Test      1% Critical  5% Critical  10% Critical
Statistic   Value      Value      Value
Z(t)      -0.098     -4.042     -3.451     -3.151

```

\* MacKinnon approximate p-value for Z(t) = 0.9944

```
. dfuller overallseas
```

Dickey-Fuller test for unit root          Number of obs =    103

```

----- Interpolated Dickey-Fuller -----
Test      1% Critical  5% Critical  10% Critical
Statistic   Value      Value      Value
Z(t)      -2.781     -3.509     -2.890     -2.580

```

\* MacKinnon approximate p-value for Z(t) = 0.0610

```
. dfuller overallseas, trend
```

Dickey-Fuller test for unit root          Number of obs =    103

```

----- Interpolated Dickey-Fuller -----
Test      1% Critical  5% Critical  10% Critical
Statistic   Value      Value      Value
Z(t)      -2.392     -4.039     -3.450     -3.150

```

\* MacKinnon approximate p-value for Z(t) = 0.3851

```
. dfuller overallseas, trend lags(4)
```

Augmented Dickey-Fuller test for unit root          Number of obs =    99

```

----- Interpolated Dickey-Fuller -----
Test      1% Critical  5% Critical  10% Critical
Statistic   Value      Value      Value
Z(t)      0.909      -4.042     -3.451     -3.151

```

\* MacKinnon approximate p-value for Z(t) = 0.9989

**TABLE 5 - 6:**

**76-01**

. dfuller doverallseas

Dickey-Fuller test for unit root      Number of obs =    102

```

----- Interpolated Dickey-Fuller -----
Test      1% Critical  5% Critical  10% Critical
Statistic   Value      Value      Value
Z(t)      -18.166     -3.509     -2.890     -2.580

```

\* MacKinnon approximate p-value for Z(t) = 0.0000

. dfuller doverallseas, trend lags(4)

Augmented Dickey-Fuller test for unit root      Number of obs =    98

```

----- Interpolated Dickey-Fuller -----
Test      1% Critical  5% Critical  10% Critical
Statistic   Value      Value      Value
Z(t)      -5.152      -4.044     -3.452     -3.151

```

\* MacKinnon approximate p-value for Z(t) = 0.0001

. dfuller dydispseas

Dickey-Fuller test for unit root      Number of obs =    102

```

----- Interpolated Dickey-Fuller -----
Test      1% Critical  5% Critical  10% Critical
Statistic   Value      Value      Value
Z(t)      -13.980     -3.509     -2.890     -2.580

```

\* MacKinnon approximate p-value for Z(t) = 0.0000

. dfuller dydispseas, trend lags(4)

Augmented Dickey-Fuller test for unit root      Number of obs =    98

```

----- Interpolated Dickey-Fuller -----
Test      1% Critical  5% Critical  10% Critical
Statistic   Value      Value      Value
Z(t)      -5.240      -4.044     -3.452     -3.151

```

\* MacKinnon approximate p-value for Z(t) = 0.0001

**76-91**

. dfuller doverallseas

Dickey-Fuller test for unit root      Number of obs =    62

```

----- Interpolated Dickey-Fuller -----
Test      1% Critical  5% Critical  10% Critical
Statistic   Value      Value      Value
Z(t)      -14.192     -3.563     -2.920     -2.595

```

\* MacKinnon approximate p-value for  $Z(t) = 0.0000$

. dfuller doverallseas, trend lags(4)

Augmented Dickey-Fuller test for unit root      Number of obs =    58

----- Interpolated Dickey-Fuller -----

Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
-------------------	----------------------	----------------------	-----------------------

Z(t)	-3.934	-4.132	-3.492	-3.175
------	--------	--------	--------	--------

\* MacKinnon approximate p-value for  $Z(t) = 0.0109$

. dfuller dydispseas

Dickey-Fuller test for unit root                      Number of obs =    62

----- Interpolated Dickey-Fuller -----

Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
-------------------	----------------------	----------------------	-----------------------

Z(t)	-12.058	-3.563	-2.920	-2.595
------	---------	--------	--------	--------

\* MacKinnon approximate p-value for  $Z(t) = 0.0000$

. dfuller dydispseas, trend lags(4)

Augmented Dickey-Fuller test for unit root      Number of obs =    58

----- Interpolated Dickey-Fuller -----

Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
-------------------	----------------------	----------------------	-----------------------

Z(t)	-4.199	-4.132	-3.492	-3.175
------	--------	--------	--------	--------

\* MacKinnon approximate p-value for  $Z(t) = 0.0045$

Residuals 76-91 seas

### **TABLE 7:**

#### **76-91**

. dfuller res

Dickey-Fuller test for unit root                      Number of obs =    63

----- Interpolated Dickey-Fuller -----

Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
-------------------	----------------------	----------------------	-----------------------

Z(t)	-9.581	-3.562	-2.920	-2.595
------	--------	--------	--------	--------

\* MacKinnon approximate p-value for  $Z(t) = 0.0000$

76-01

. dfuller res

Dickey-Fuller test for unit root                      Number of obs =    103

----- Interpolated Dickey-Fuller -----

Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
-------------------	----------------------	----------------------	-----------------------

Z(t)	-2.915	-3.509	-2.890	-2.580
------	--------	--------	--------	--------

\* MacKinnon approximate p-value for  $Z(t) = 0.0436$

### **TABLE 8 - 9**

#### **76-01**

Date: 05/30/03 Time: 14:38

Sample: 1976:1 2001:4

Included observations: 103

Test assumption:

Linear deterministic

trend in the data

Series: OVERALLSEAS YDISPSEAS

Lags interval: No lags

Eigenvalue	Likelihood Ratio	5 Percent Critical Value	1 Percent Critical Value	Hypothesized No. of CE(s)
0.085859	16.83042	15.41	20.04	None *
0.070986	7.584066	3.76	6.65	At most 1 **

\*(\*\*) denotes rejection of the hypothesis at 5% (1%) significance level  
 L.R. test indicates 2 cointegrating equation(s) at 5% significance level

Unnormalized Cointegrating Coefficients:

OVERALLSEAS	YDISPSEAS
-8.04E-06	3.22E-06
5.36E-06	3.70E-07

Normalized Cointegrating Coefficients: 1 Cointegrating Equation(s)

OVERALLSEAS	YDISPSEAS	C
1.000000	-0.400858 (0.10073)	-146830.7
Log likelihood	-2198.649	

### 76-91

Date: 05/30/03 Time: 14:37

Sample: 1976:1 1991:4

Included observations: 63

Test assumption:

Linear deterministic

trend in the data

Series: OVERALLSEAS YDISPSEAS

Lags interval: No lags

Eigenvalue	Likelihood Ratio	5 Percent Critical Value	1 Percent Critical Value	Hypothesized No. of CE(s)
0.648631	68.87603	15.41	20.04	None **
0.046250	2.983250	3.76	6.65	At most 1

\*(\*\*) denotes rejection of the hypothesis at 5% (1%) significance level  
 L.R. test indicates 1 cointegrating equation(s) at 5% significance level

Unnormalized Cointegrating Coefficients:

OVERALLSEAS	YDISPSEAS
-2.52E-05	1.18E-05
6.45E-06	8.58E-07

Normalized Cointegrating

Coefficients: 1  
Cointegrating  
Equation(s)

OVERALLSEAS	YDISPSEAS	C
1.000000	-0.470423 (0.01435)	-124717.4
Log likelihood	-1302.056	

**TABLE 10:**

**Quarterly**

.reg overallseas ydispseas

Source	SS	df	MS	Number of obs = 64
Model	1.3879e+10	1	1.3879e+10	F( 1, 62) = 576.29
Residual	1.4932e+09	62	24084070.6	Prob > F = 0.0000
Total	1.5373e+10	63	244009007	R-squared = 0.9029
				Adj R-squared = 0.9013
				Root MSE = 4907.6

overallseas	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
ydispseas	.4301507	.0179185	24.01	0.000	.3943322 .4659692
_cons	140983.9	7304.457	19.30	0.000	126382.5 155585.3

.hettest

Cook-Weisberg test for heteroskedasticity using fitted values of overallseas  
 Ho: Constant variance  
 chi2(1) = 0.34  
 Prob > chi2 = 0.5590

.dwstat

Durbin-Watson d-statistic( 2, 64) = 2.40048

**Yearly**

.prais overall ydisp, robust

Iteration 0: rho = 0.0000  
 Iteration 1: rho = 0.6914  
 Iteration 2: rho = 0.7548  
 Iteration 3: rho = 0.7707  
 Iteration 4: rho = 0.7752  
 Iteration 5: rho = 0.7766  
 Iteration 6: rho = 0.7770  
 Iteration 7: rho = 0.7771  
 Iteration 8: rho = 0.7771  
 Iteration 9: rho = 0.7771  
 Iteration 10: rho = 0.7771  
 Iteration 11: rho = 0.7771

Prais-Winsten AR(1) regression -- iterated estimates

Regression with robust standard errors      Number of obs = 17  
 F( 2, 15) = 23728.06  
 Prob > F = 0.0000  
 R-squared = 0.9963  
 Root MSE = 1700.7

overall	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
ydisp	.5065242	.0242617	20.88	0.000	.4548116 .5582367
_cons	97634.73	10984.22	8.89	0.000	74222.42 121047
rho	.7771288				

Durbin-Watson statistic (original) 0.465445

Durbin-Watson statistic (transformed) 1.098292

### Long-run

quarterly, nocon

```
. reg overallseas ydispseas, nocon
```

Source	SS	df	MS	Number of obs =	64
				F( 1, 63) =	38432.10
Model	6.3842e+12	1	6.3842e+12	Prob > F	= 0.0000
Residual	1.0465e+10	63	166115665	R-squared	= 0.9984
				Adj R-squared =	0.9983
Total	6.3946e+12	64	9.9916e+10	Root MSE	= 12889

overallseas	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
ydispseas	.7747747	.0039521	196.04	0.000	.7668771 .7826724

### APC

```
. reg C_Y ydispseas
```

Source	SS	df	MS	Number of obs =	64
				F( 1, 62) =	341.45
Model	.058616888	1	.058616888	Prob > F	= 0.0000
Residual	.010643692	62	.000171672	R-squared	= 0.8463
				Adj R-squared =	0.8438
Total	.06926058	63	.001099374	Root MSE	= .0131

C_Y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
ydispseas	-8.84e-07	4.78e-08	-18.48	0.000	-9.80e-07 -7.88e-07
_cons	1.138858	.0195017	58.40	0.000	1.099875 1.177842

### TABLE 11

#### Quarterly

```
. prais overallseas ydispseas looverallseas
```

Iteration 0: rho = 0.0000

Iteration 1: rho = -0.4311

Iteration 2: rho = -0.5067

Iteration 3: rho = -0.5121

Iteration 4: rho = -0.5125

Iteration 5: rho = -0.5125

Iteration 6: rho = -0.5125

Iteration 7: rho = -0.5125

Prais-Winsten AR(1) regression -- iterated estimates

Source	SS	df	MS	Number of obs =	63
				F( 2, 60) =	2822.27
Model	8.1343e+10	2	4.0671e+10	Prob > F	= 0.0000
Residual	864652103	60	14410868.4	R-squared	= 0.9895
				Adj R-squared =	0.9891
Total	8.2207e+10	62	1.3259e+09	Root MSE	= 3796.2

overallseas	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
ydispseas	.2265756	.0368836	6.14	0.000	.1527973 .3003538
looverallseas	.5074759	.0783783	6.47	0.000	.350696 .6642558
_cons	63910.2	11071.29	5.77	0.000	41764.31 86056.08

rho | -.5125137

Durbin-Watson statistic (original) 2.800995

Durbin-Watson statistic (transformed) 1.948829

## Yearly

. reg overall ydisp loverall

Source	SS	df	MS	Number of obs =	16
				F( 2, 13) =	410.86
Model	3.6006e+09	2	1.8003e+09	Prob > F	= 0.0000
Residual	56963301.3	13	4381792.41	R-squared	= 0.9844
				Adj R-squared =	0.9820
Total	3.6576e+09	15	243837560	Root MSE	= 2093.3

overall	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
ydisp	.3179475	.0647003	4.91	0.000	.178171 .457724
loverall	.340912	.1370782	2.49	0.027	.0447724 .6370515
_cons	72936.07	17613.86	4.14	0.001	34883.64 110988.5

## TABLE 12

### Quarterly

. reg overallseas ydispseas loverallseas, nocon

Source	SS	df	MS	Number of obs =	63
				F( 2, 61) =	.
Model	6.3127e+12	2	3.1563e+12	Prob > F	= 0.0000
Residual	1.7934e+09	61	29399189.9	R-squared	= 0.9997
				Adj R-squared =	0.9997
Total	6.3145e+12	63	1.0023e+11	Root MSE	= 5422.1

overallseas	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
ydispseas	.1151312	.0400867	2.87	0.006	.034973 .1952895
loverallseas	.8538025	.0519066	16.45	0.000	.7500088 .9575961

### Yearly

. reg overall ydisp loverall, nocon

Source	SS	df	MS	Number of obs =	16
				F( 2, 14) =	86690.12
Model	1.6359e+12	2	8.1796e+11	Prob > F	= 0.0000
Residual	132095740	14	9435410.02	R-squared	= 0.9999
				Adj R-squared =	0.9999
Total	1.6360e+12	16	1.0225e+11	Root MSE	= 3071.7

overall	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
ydisp	.1101398	.0599239	1.84	0.087	-.0183842 .2386638
loverall	.8581899	.0828169	10.36	0.000	.6805654 1.035814

## 91-01

. dfuller overallseas

Dickey-Fuller test for unit root                      Number of obs = 39

Test Statistic	----- Interpolated Dickey-Fuller -----		
	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-2.024	-3.655	-2.961

\* MacKinnon approximate p-value for Z(t) = 0.2763

. dfuller ydispseas

Dickey-Fuller test for unit root                      Number of obs = 39

```

----- Interpolated Dickey-Fuller -----
      Test      1% Critical   5% Critical   10% Critical
      Statistic   Value         Value         Value
-----
Z(t)      -4.621         -3.655         -2.961         -2.613
-----

```

\* MacKinnon approximate p-value for Z(t) = 0.0001

. dfuller doverallseas

Dickey-Fuller test for unit root                      Number of obs = 38

```

----- Interpolated Dickey-Fuller -----
      Test      1% Critical   5% Critical   10% Critical
      Statistic   Value         Value         Value
-----
Z(t)      -11.484          -3.662         -2.964         -2.614
-----

```

\* MacKinnon approximate p-value for Z(t) = 0.0000

. reg overallseas ydispsseas

```

Source |   SS    df    MS      Number of obs = 40
-----+-----
Model | 91282891.2   1 91282891.2      F( 1, 38) = 0.61
Residual | 5.6597e+09  38 148938219      Prob > F   = 0.4386
Total | 5.7509e+09  39 147459877      R-squared  = 0.0159
                                           Adj R-squared = -0.0100
                                           Root MSE   = 12204
-----

```

```

-----
overallseas |   Coef.   Std. Err.   t   P>|t|   [95% Conf. Interval]
-----+-----
ydispsseas | -0.964208   .1231627   -0.78  0.439   -1.3457505   -.5826585
_cons | 376347.4   59458.38   6.33  0.000   255980.2   496714.6
-----

```

**TABLE 14:**

. reg overall ydisp

```

Source |   SS    df    MS      Number of obs = 19
-----+-----
Model | 3.1260e+11   1 3.1260e+11      F( 1, 17) = 1183.49
Residual | 4.4902e+09  17 264131168      Prob > F   = 0.0000
Total | 3.1709e+11  18 1.7616e+10      R-squared  = 0.9858
                                           Adj R-squared = 0.9850
                                           Root MSE   = 16252
-----

```

```

-----
overall |   Coef.   Std. Err.   t   P>|t|   [95% Conf. Interval]
-----+-----
ydisp | .5958333   .0173198   34.40  0.000   .5592917   .6323748
_cons | 76052.99   8334.149   9.13  0.000   58469.47   93636.5
-----

```

. reg overall ydisp, nocon

```

Source |   SS    df    MS      Number of obs = 19
-----+-----
Model | 2.3908e+12   1 2.3908e+12      F( 1, 18) = 1624.83
Residual | 2.6486e+10  18 1.4714e+09      Prob > F   = 0.0000
Total | 2.4173e+12  19 1.2723e+11      R-squared  = 0.9890
                                           Adj R-squared = 0.9884
                                           Root MSE   = 38359
-----

```

```

-----
overall |   Coef.   Std. Err.   t   P>|t|   [95% Conf. Interval]
-----+-----
ydisp | .7371857   .0182883   40.31  0.000   .6987635   .7756079
-----

```

-----  
**CROSS-SECTION: TABLE 15**

OVERALL

. reg overall income, robust

Regression with robust standard errors      Number of obs = 19  
 F( 1, 17) = 87.18  
 Prob > F = 0.0000  
 R-squared = 0.8928  
 Root MSE = 44724

-----  
 |                    Robust  
 overall |    Coef. Std. Err.    t   P>|t|   [95% Conf. Interval]  
 -----+-----  
 income | 2228.535 238.6706    9.34 0.000   1724.984 2732.086  
 \_cons | 109618 18799.82    5.83 0.000   69953.89 149282.2  
 -----

FOOD

. reg food income

Source |    SS    df    MS                    Number of obs = 19  
 -----+-----                    F( 1, 17) = 822.53  
 Model | 7.1047e+09    1 7.1047e+09      Prob > F = 0.0000  
 Residual | 146840091    17 8637652.42      R-squared = 0.9798  
 -----+-----                    Adj R-squared = 0.9786  
 Total | 7.2515e+09    18 402863564      Root MSE = 2939.0

-----  
 food |    Coef. Std. Err.    t   P>|t|   [95% Conf. Interval]  
 -----+-----  
 income | 353.0495 12.31007    28.68 0.000   327.0775 379.0214  
 \_cons | 41879.58 1403.564    29.84 0.000   38918.32 44840.84  
 -----

HOUSING

. reg housing income

Source |    SS    df    MS                    Number of obs = 19  
 -----+-----                    F( 1, 17) = 2.21  
 Model | 23081262.1    1 23081262.1      Prob > F = 0.1551  
 Residual | 177214295    17 10424370.3      R-squared = 0.1152  
 -----+-----                    Adj R-squared = 0.0632  
 Total | 200295557    18 11127530.9      Root MSE = 3228.7

-----  
 housing |    Coef. Std. Err.    t   P>|t|   [95% Conf. Interval]  
 -----+-----  
 income | -20.12298 13.52345    -1.49 0.155   -48.65497 8.409002  
 \_cons | 25508.77 1541.91    16.54 0.000   22255.63 28761.92  
 -----

FUEL

. reg fuel income, robust

Regression with robust standard errors      Number of obs = 19  
 F( 1, 17) = 135.91  
 Prob > F = 0.0000  
 R-squared = 0.9325  
 Root MSE = 983.66

-----  
 |                    Robust  
 fuel |    Coef. Std. Err.    t   P>|t|   [95% Conf. Interval]  
 -----+-----  
 income | 63.14719 5.416571    11.66 0.000   51.71923 74.57516  
 \_cons | 12952.75 434.4355    29.82 0.000   12036.18 13869.33  
 -----

-----  
FURNITURE

. reg furnit income, robust

Regression with robust standard errors      Number of obs = 19  
F( 1, 17) = 76.44  
Prob > F = 0.0000  
R-squared = 0.8981  
Root MSE = 1385.4

-----

	Robust						
furnit	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]		
income	71.0193	8.123116	8.74	0.000	53.88102	88.15758	
_cons	3554.123	607.9691	5.85	0.000	2271.42	4836.826	

-----

CLOTHING

. reg clothing income, robust

Regression with robust standard errors      Number of obs = 19  
F( 1, 17) = 39.23  
Prob > F = 0.0000  
R-squared = 0.8114  
Root MSE = 4118.3

-----

	Robust						
clothing	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]		
income	147.5233	23.55393	6.26	0.000	97.82888	197.2178	
_cons	2044.93	1854.495	1.10	0.286	-1867.712	5957.572	

-----

MEDICAL CARE

. reg medic income, robust

Regression with robust standard errors      Number of obs = 19  
F( 1, 17) = 27.13  
Prob > F = 0.0001  
R-squared = 0.7141  
Root MSE = 1484.9

-----

	Robust						
medic	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]		
income	40.53298	7.782327	5.21	0.000	24.11371	56.95226	
_cons	6437.649	569.7267	11.30	0.000	5235.631	7639.667	

-----

TRANSPORTATION

. reg transp income, robust

Regression with robust standard errors      Number of obs = 19  
F( 1, 17) = 51.36  
Prob > F = 0.0000  
R-squared = 0.8299  
Root MSE = 7272.7

-----

	Robust						
transp	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]		
income	277.3825	38.70666	7.17	0.000	195.7185	359.0464	
_cons	15008.28	2948.711	5.09	0.000	8787.045	21229.52	

-----

EDUCATION

. reg educ income

```

Source | SS      df    MS      Number of obs =   19
-----+-----
Model | 1.6448e+09  1 1.6448e+09    Prob > F   = 0.0000
Residual | 81791677.5 17 4811275.15    R-squared   = 0.9526
-----+-----
Total | 1.7265e+09 18 95919225.6    Adj R-squared = 0.9498
                                           Root MSE   = 2193.5

```

```

-----+-----
educ |  Coef.  Std. Err.   t   P>|t|   [95% Conf. Interval]
-----+-----
income | 169.8686  9.187401  18.49  0.000   150.4849  189.2523
_cons | 1053.877 1047.525   1.01  0.328  -1156.207  3263.961
-----+-----

```

READING

. reg read income, robust

```

Regression with robust standard errors      Number of obs =   19
                                           F( 1, 17) = 175.16
                                           Prob > F   = 0.0000
                                           R-squared   = 0.9464
                                           Root MSE   = 3186.6

```

```

-----+-----
|               Robust
read |  Coef.  Std. Err.   t   P>|t|   [95% Conf. Interval]
-----+-----
income | 231.2439 17.47248  13.23  0.000   194.3802  268.1076
_cons | 6553.456 1306.098   5.02  0.000  3797.831  9309.081
-----+-----

```

OTHER

. reg other income

```

Source | SS      df    MS      Number of obs =   19
-----+-----
Model | 4.5657e+10  1 4.5657e+10    Prob > F   = 0.0000
Residual | 7.7325e+09 17 454851707    R-squared   = 0.8552
-----+-----
Total | 5.3389e+10 18 2.9661e+09    Adj R-squared = 0.8466
                                           Root MSE   = 21327

```

```

-----+-----
other |  Coef.  Std. Err.   t   P>|t|   [95% Conf. Interval]
-----+-----
income | 894.9832 89.33003  10.02  0.000   706.5133 1083.453
_cons | -5388.211 10185.19  -0.53  0.604  -26877.08 16100.66
-----+-----

```

NON-CONSUMPTION

. reg non\_cons income, robust

```

Regression with robust standard errors      Number of obs =   19
                                           F( 1, 17) = 28.74
                                           Prob > F   = 0.0001
                                           R-squared   = 0.7690
                                           Root MSE   = 34207

```

```

-----+-----
|               Robust
non_cons |  Coef.  Std. Err.   t   P>|t|   [95% Conf. Interval]
-----+-----
income | 1077.901 201.0546   5.36  0.000   653.7124 1502.089
_cons | -20337.21 15026.3  -1.35  0.194  -52039.92 11365.5
-----+-----

```

**TABLE 16**

OVERALL

. reg overall age age2

```
Source | SS      df    MS      Number of obs = 11
-----+-----
Model | 3.5440e+10  2  1.7720e+10    F( 2, 8) = 24.35
Residual | 5.8222e+09  8  7.27775891    Prob > F = 0.0004
-----+-----
Total | 4.1263e+10 10  4.1263e+09    R-squared = 0.8589
                          Adj R-squared = 0.8236
                          Root MSE = 26977
```

```
-----+-----
overall | Coef. Std. Err.  t  P>|t|  [95% Conf. Interval]
-----+-----
age | 72985.23  11347.26  6.43  0.000  46818.4  99152.06
age2 | -5204.497  920.9906  -5.65  0.000  -7328.305 -3080.688
_cons | 128241.1  29626.71  4.33  0.003  59921.78  196560.4
-----+-----
```

#### FOOD

. reg food age age2

```
Source | SS      df    MS      Number of obs = 11
-----+-----
Model | 2.3192e+09  2  1.1596e+09    F( 2, 8) = 48.65
Residual | 190669873  8  23833734.1    Prob > F = 0.0000
-----+-----
Total | 2.5098e+09 10  250982522    R-squared = 0.9240
                          Adj R-squared = 0.9050
                          Root MSE = 4882.0
```

```
-----+-----
food | Coef. Std. Err.  t  P>|t|  [95% Conf. Interval]
-----+-----
age | 17719.91  2053.47  8.63  0.000  12984.6  22455.22
age2 | -1220.221  166.6681  -7.32  0.000  -1604.559 -835.8842
_cons | 25565.44  5361.43  4.77  0.001  13201.96  37928.92
-----+-----
```

#### HOUSING

. reg housing age age2

```
Source | SS      df    MS      Number of obs = 11
-----+-----
Model | 699032582  2  349516291    F( 2, 8) = 28.97
Residual | 96532239.0  8  12066529.9    Prob > F = 0.0002
-----+-----
Total | 795564821 10  79556482.1    R-squared = 0.8787
                          Adj R-squared = 0.8483
                          Root MSE = 3473.7
```

```
-----+-----
housing | Coef. Std. Err.  t  P>|t|  [95% Conf. Interval]
-----+-----
age | -8785.706  1461.112  -6.01  0.000  -12155.04 -5416.377
age2 | 569.0816  118.5899  4.80  0.001  295.6129  842.5503
_cons | 51508.58  3814.835  13.50  0.000  42711.55  60305.6
-----+-----
```

#### FUEL

. reg fuel age age2

```
Source | SS      df    MS      Number of obs = 11
-----+-----
Model | 79320753.1  2  39660376.6    F( 2, 8) = 40.84
Residual | 7769561.58  8  971195.198    Prob > F = 0.0001
-----+-----
Total | 87090314.7 10  8709031.47    R-squared = 0.9108
                          Adj R-squared = 0.8885
                          Root MSE = 985.49
```

```
-----+-----
fuel | Coef. Std. Err.  t  P>|t|  [95% Conf. Interval]
-----+-----
age | 3211.644  414.52  7.75  0.000  2255.759  4167.529
age2 | -218.4044  33.64415  -6.49  0.000  -295.988 -140.8209
_cons | 9617.194  1082.276  8.89  0.000  7121.462  12112.93
-----+-----
```

#### FURNITURE

. reg furnit age age2

Source	SS	df	MS	Number of obs =	11
-----				F( 2, 8) =	68.09
Model	48826322.6	2	24413161.3	Prob > F =	0.0000
Residual	2868500.17	8	358562.521	R-squared =	0.9445
-----				Adj R-squared =	0.9306
Total	51694822.7	10	5169482.27	Root MSE =	598.80

furnit	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	1953.021	251.869	7.75	0.000	1372.21	2533.832
age2	-113.979	20.44273	-5.58	0.001	-161.12	-66.83801
_cons	4358.364	657.6079	6.63	0.000	2841.917	5874.81

#### CLOTHING

. reg clothing age age2

Source	SS	df	MS	Number of obs =	11
-----				F( 2, 8) =	16.81
Model	68477964.1	2	34238982.1	Prob > F =	0.0014
Residual	16290797.5	8	2036349.69	R-squared =	0.8078
-----				Adj R-squared =	0.7598
Total	84768761.6	10	8476876.16	Root MSE =	1427.0

clothing	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	2701.001	600.231	4.50	0.002	1316.866	4085.137
age2	-173.1259	48.71722	-3.55	0.007	-285.468	-60.78375
_cons	8439.964	1567.151	5.39	0.001	4826.108	12053.82

#### MEDICAL CARE

. reg medic age age2

Source	SS	df	MS	Number of obs =	11
-----				F( 2, 8) =	8.59
Model	11299168.6	2	5649584.31	Prob > F =	0.0102
Residual	5263727.02	8	657965.877	R-squared =	0.6822
-----				Adj R-squared =	0.6027
Total	16562895.6	10	1656289.56	Root MSE =	811.15

medic	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	-941.1573	341.1881	-2.76	0.025	-1727.939	-154.3761
age2	93.81235	27.69224	3.39	0.010	29.95394	157.6708
_cons	12629.39	890.8124	14.18	0.000	10575.18	14683.61

#### TRANSPORTATION

. reg transp age age2

Source	SS	df	MS	Number of obs =	11
-----				F( 2, 8) =	7.91
Model	336655509	2	168327754	Prob > F =	0.0127
Residual	170182900	8	21272862.5	R-squared =	0.6642
-----				Adj R-squared =	0.5803
Total	506838409	10	50683840.9	Root MSE =	4612.3

transp	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	7402.753	1940.016	3.82	0.005	2929.069	11876.44
age2	-625.3438	157.4597	-3.97	0.004	-988.4465	-262.2412
_cons	25920.66	5065.211	5.12	0.001	14240.26	37601.06

#### EDUCATION

. reg educ age age2

Source	SS	df	MS	Number of obs =	11
				F( 2, 8) =	7.05
Model	1.1571e+09	2	578549682	Prob > F	= 0.0171
Residual	656152509	8	82019063.6	R-squared	= 0.6381
				Adj R-squared =	0.5477
Total	1.8133e+09	10	181325187	Root MSE	= 9056.4

educ	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
age	13677.27	3809.339	3.59	0.007	4892.917 22461.62
age2	-1158.501	309.1817	-3.75	0.006	-1871.475 -445.5269
_cons	-14756.74	9945.851	-1.48	0.176	-37691.91 8178.435

READING

. reg read age age2

Source	SS	df	MS	Number of obs =	11
				F( 2, 8) =	23.84
Model	332019440	2	166009720	Prob > F	= 0.0004
Residual	55703732.1	8	6962966.51	R-squared	= 0.8563
				Adj R-squared =	0.8204
Total	387723173	10	38772317.3	Root MSE	= 2638.7

read	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
age	6095.26	1109.914	5.49	0.001	3535.793 8654.727
age2	-396.352	90.08521	-4.40	0.002	-604.0889 -188.6151
_cons	12157.99	2897.889	4.20	0.003	5475.45 18840.54

OTHER

. reg other age age2

Source	SS	df	MS	Number of obs =	11
				F( 2, 8) =	9.69
Model	7.8259e+09	2	3.9129e+09	Prob > F	= 0.0073
Residual	3.2299e+09	8	403735318	R-squared	= 0.7079
				Adj R-squared =	0.6348
Total	1.1056e+10	10	1.1056e+09	Root MSE	= 20093

other	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
age	29951.13	8451.633	3.54	0.008	10461.63 49440.63
age2	-1961.455	685.9694	-2.86	0.021	-3543.303 -379.6063
_cons	-7199.491	22066.47	-0.33	0.753	-58084.87 43685.89

NON-CONSUMPTION

. reg non\_cons age age2

Source	SS	df	MS	Number of obs =	11
				F( 2, 8) =	17.42
Model	7.4475e+09	2	3.7237e+09	Prob > F	= 0.0012
Residual	1.7103e+09	8	213785980	R-squared	= 0.8132
				Adj R-squared =	0.7666
Total	9.1578e+09	10	915776419	Root MSE	= 14621

non_cons	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
age	36140.64	6150.095	5.88	0.000	21958.5 50322.79
age2	-2919.317	499.1671	-5.85	0.000	-4070.398 -1768.236
_cons	-6776.079	16057.36	-0.42	0.684	-43804.42 30252.26

**TABLE 17:**

OVERALL

. reg overall age age2 income, robust

Regression with robust standard errors      Number of obs = 60  
 F( 3, 56) = 159.25  
 Prob > F = 0.0000  
 R-squared = 0.7561  
 Root MSE = 76224

	Robust					
overall	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	20659.72	41833.84	0.49	0.623	-63143.52	104463
age2	-3810.87	4977.184	-0.77	0.447	-13781.37	6159.628
income	4487.358	500.8916	8.96	0.000	3483.952	5490.764
_cons	68011.59	56362.81	1.21	0.233	-44896.7	180919.9

FOOD

. reg food age age2 income, robust

Regression with robust standard errors      Number of obs = 60  
 F( 3, 56) = 251.08  
 Prob > F = 0.0000  
 R-squared = 0.9366  
 Root MSE = 5501.7

	Robust					
food	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	16462.8	2123.38	7.75	0.000	12209.16	20716.44
age2	-2019.918	309.5891	-6.52	0.000	-2640.099	-1399.736
income	675.5609	25.61335	26.38	0.000	624.2512	726.8706
_cons	11924.59	3409.178	3.50	0.001	5095.185	18753.99

HOUSING

. reg housing age age2 income, robust

Regression with robust standard errors      Number of obs = 60  
 F( 3, 56) = 5.98  
 Prob > F = 0.0013  
 R-squared = 0.4113  
 Root MSE = 11168

	Robust					
housing	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	-16376.85	5819.822	-2.81	0.007	-28035.36	-4718.348
age2	1719.332	706.5324	2.43	0.018	303.9776	3134.687
income	98.31545	67.5886	1.45	0.151	-37.08079	233.7117
_cons	50689.55	8302.987	6.10	0.000	34056.67	67322.43

FUEL

. reg fuel age age2 income

Source	SS	df	MS	Number of obs = 60
Model	909526542	3	303175514	F( 3, 56) = 90.43
Residual	187741316	56	3352523.49	Prob > F = 0.0000
Total	1.0973e+09	59	18597760.3	R-squared = 0.8289
				Adj R-squared = 0.8197
				Root MSE = 1831.0

	Robust					
fuel	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	3364.708	677.6253	4.97	0.000	2007.261	4722.154

```

age2 | -380.1607  94.76272  -4.01  0.000  -569.9933  -190.3282
income | 124.6362  8.229687  15.14  0.000  108.1501  141.1222
_cons | 6156.371  1130.347  5.45  0.000  3892.014  8420.728

```

FURNITURE

```
. reg furnit age age2 income, robust
```

```

Regression with robust standard errors      Number of obs =   60
                                           F( 3, 56) = 65.45
                                           Prob > F   = 0.0000
                                           R-squared  = 0.7952
                                           Root MSE   = 1913.0

```

```

-----+-----
|               Robust
|               Coef. Std. Err.  t  P>|t|  [95% Conf. Interval]
-----+-----
age | 1501.346  730.0694   2.06  0.044  38.84134  2963.851
age2 | -148.6911  117.2261  -1.27  0.210  -383.5231  86.14097
income | 123.1354  9.547059  12.90  0.000  104.0103  142.2604
_cons | 730.7256  1043.794   0.70  0.487  -1360.245  2821.697

```

CLOTHING

```
. reg clothing age age2 income, robust
```

```

Regression with robust standard errors      Number of obs =   60
                                           F( 3, 56) = 98.50
                                           Prob > F   = 0.0000
                                           R-squared  = 0.6692
                                           Root MSE   = 7015.6

```

```

-----+-----
|               Robust
|               Coef. Std. Err.  t  P>|t|  [95% Conf. Interval]
-----+-----
age | -3303.349  3636.02  -0.91  0.368  -10587.17  3980.473
age2 | 429.7107  467.5691   0.92  0.362  -506.9428  1366.364
income | 333.0736  40.75094  8.17  0.000  251.4397  414.7076
_cons | 4302.01  4759.186   0.90  0.370  -5231.785  13835.81

```

MEDICAL CARE

```
. reg medic age age2 income, robust
```

```

Regression with robust standard errors      Number of obs =   60
                                           F( 3, 56) = 40.95
                                           Prob > F   = 0.0000
                                           R-squared  = 0.7134
                                           Root MSE   = 2334.7

```

```

-----+-----
|               Robust
|               Coef. Std. Err.  t  P>|t|  [95% Conf. Interval]
-----+-----
age | -3738.783  1070.674  -3.49  0.001  -5883.601  -1593.965
age2 | 512.8196  142.7443   3.59  0.001  226.8685  798.7707
income | 115.2737  11.42903  10.09  0.000  92.37864  138.1688
_cons | 10514.35  1582.034   6.65  0.000  7345.159  13683.55

```

TRANSPORTATION

```
. reg transp age age2 income, robust
```

```

Regression with robust standard errors      Number of obs =   60
                                           F( 3, 56) = 25.52
                                           Prob > F   = 0.0000
                                           R-squared  = 0.3795
                                           Root MSE   = 29307

```

	Robust					
transp	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	-12204.93	16617.47	-0.73	0.466	-45493.71	21083.86
age2	987.8232	1953.536	0.51	0.615	-2925.579	4901.226
income	698.6341	197.6897	3.53	0.001	302.6141	1094.654
_cons	33015.45	22060.04	1.50	0.140	-11176.13	77207.03

#### EDUCATION

. reg educ age age2 income, robust

Regression with robust standard errors      Number of obs = 57  
 F( 3, 53) = 23.29  
 Prob > F = 0.0000  
 R-squared = 0.6578  
 Root MSE = 7557.1

	Robust					
educ	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	20514.46	2702.882	7.59	0.000	15093.17	25935.76
age2	-3144.17	398.322	-7.89	0.000	-3943.103	-2345.237
income	243.7387	40.30335	6.05	0.000	162.9004	324.577
_cons	-25467.89	4637.859	-5.49	0.000	-34770.26	-16165.52

#### READING

. reg read age age2 income, robust

Regression with robust standard errors      Number of obs = 60  
 F( 3, 56) = 76.90  
 Prob > F = 0.0000  
 R-squared = 0.7109  
 Root MSE = 11714

	Robust					
read	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	-7342.816	6575.969	-1.12	0.269	-20516.06	5830.433
age2	882.5125	775.3725	1.14	0.260	-670.7452	2435.77
income	609.0838	75.72576	8.04	0.000	457.3869	760.7808
_cons	11302.24	9130.514	1.24	0.221	-6988.379	29592.86

#### OTHER

. reg other age age2 income, robust

Regression with robust standard errors      Number of obs = 60  
 F( 3, 56) = 52.05  
 Prob > F = 0.0000  
 R-squared = 0.7458  
 Root MSE = 26505

	Robust					
other	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	21682.29	12207.56	1.78	0.081	-2772.401	46136.98
age2	-2717.584	1557.6	-1.74	0.087	-5837.831	402.6632
income	1500.205	163.1849	9.19	0.000	1173.306	1827.103
_cons	-35791.96	17548.36	-2.04	0.046	-70945.55	-638.3707

#### NON-CONSUMPTION

. reg non\_cons age age2 income, robust

Regression with robust standard errors      Number of obs = 60  
 F( 3, 56) = 54.99  
 Prob > F = 0.0000

R-squared = 0.8270  
 Root MSE = 24383

```
-----+-----
      |           Robust
non_cons |   Coef. Std. Err.   t  P>|t|   [95% Conf. Interval]
-----+-----
      age | 38152.54  9936.282   3.84  0.000   18247.77   58057.3
      age2 | -5891.116  1407.522  -4.19  0.000  -8710.721  -3071.511
      income | 1708.611  148.1043  11.54  0.000   1411.922  2005.299
      _cons | -63485.27  16293.88   -3.90  0.000  -96125.84  -30844.7
-----+-----
```

**.vif**

```
-----+-----
Variable |   VIF   1/VIF
-----+-----
      age | 23.97  0.041721
      age2 | 23.97  0.041721
      income | 1.00  1.000000
-----+-----
Mean VIF | 16.31
-----+-----
```

**TABLE 18**

OVERALL

. reg overall children level income, robust

Regression with robust standard errors      Number of obs = 138  
 F( 3, 134) = 191.60  
 Prob > F = 0.0000  
 R-squared = 0.8163  
 Root MSE = 57561

```
-----+-----
      |           Robust
overall |   Coef. Std. Err.   t  P>|t|   [95% Conf. Interval]
-----+-----
children | 7962.813  6312.441   1.26  0.209  -4522.095  20447.72
      level | 171.8814  32.99854   5.21  0.000   106.616  237.1467
      income | 4221.751  183.3525  23.03  0.000   3859.112  4584.39
      _cons | 61555.61  18314.29   3.36  0.001  25333.13  97778.09
-----+-----
```

FOOD

. reg food children level income

Source | SS df MS Number of obs = 138  
 -----+----- F( 3, 134) = 192.09  
 Model | 5.3476e+10 3 1.7825e+10 Prob > F = 0.0000  
 Residual | 1.2435e+10 134 92794781.9 R-squared = 0.8113  
 -----+----- Adj R-squared = 0.8071  
 Total | 6.5910e+10 137 481095946 Root MSE = 9633.0

```
-----+-----
      food |   Coef. Std. Err.   t  P>|t|   [95% Conf. Interval]
-----+-----
children | 9187.397  1016.38   9.04  0.000   7177.175  11197.62
      level | 38.58507  5.795446   6.66  0.000   27.12269  50.04746
      income | 620.6672  30.1769   20.57  0.000   560.9825  680.3518
      _cons | 19928.42  3170.954   6.28  0.000  13656.83  26200.01
-----+-----
```

HOUSING

. reg housing children level income

Source | SS df MS Number of obs = 138  
 -----+----- F( 3, 134) = 9.94  
 Model | 3.4346e+09 3 1.1449e+09 Prob > F = 0.0000  
 Residual | 1.5437e+10 134 115204534 R-squared = 0.1820

```
-----+-----
Adj R-squared = 0.1637
Total | 1.8872e+10 137 137751746      Root MSE = 10733
```

```
-----+-----
housing | Coef. Std. Err. t P>|t| [95% Conf. Interval]
-----+-----
children | -101.042 1132.476 -0.09 0.929 -2340.882 2138.798
level | -35.1802 6.457432 -5.45 0.000 -47.95188 -22.40853
income | 23.14982 33.62387 0.69 0.492 -43.35233 89.65197
_cons | 31132.69 3533.157 8.81 0.000 24144.72 38120.66
-----+-----
```

FUEL

```
. reg fuel children level income
```

```
Source | SS df MS Number of obs = 138
-----+----- F( 3, 134) = 85.83
Model | 1.2940e+09 3 431347908 Prob > F = 0.0000
Residual | 673429535 134 5025593.54 R-squared = 0.6577
-----+----- Adj R-squared = 0.6501
Total | 1.9675e+09 137 14361118.7 Root MSE = 2241.8
```

```
-----+-----
fuel | Coef. Std. Err. t P>|t| [95% Conf. Interval]
-----+-----
children | 1905.858 236.5309 8.06 0.000 1438.041 2373.675
level | 12.51027 1.348711 9.28 0.000 9.84276 15.17779
income | 65.98612 7.02274 9.40 0.000 52.09636 79.87587
_cons | 9546.792 737.9413 12.94 0.000 8087.273 11006.31
-----+-----
```

FURNITURE

```
. reg furnit children level income, robust
```

```
Regression with robust standard errors Number of obs = 138
F( 3, 134) = 31.38
Prob > F = 0.0000
R-squared = 0.4619
Root MSE = 3642.8
```

```
-----+-----
| Robust
furnit | Coef. Std. Err. t P>|t| [95% Conf. Interval]
-----+-----
children | -237.004 317.0967 -0.75 0.456 -864.1661 390.158
level | 2.233849 1.598964 1.40 0.165 -9286227 5.39632
income | 120.4733 14.28669 8.43 0.000 92.21676 148.7299
_cons | 3699.113 1035.608 3.57 0.000 1650.862 5747.365
-----+-----
```

CLOTHING

```
. reg clothing children level income, robust
```

```
Regression with robust standard errors Number of obs = 138
F( 3, 134) = 175.65
Prob > F = 0.0000
R-squared = 0.8575
Root MSE = 3436.9
```

```
-----+-----
| Robust
clothing | Coef. Std. Err. t P>|t| [95% Conf. Interval]
-----+-----
children | -831.8619 363.0933 -2.29 0.024 -1549.997 -113.7266
level | -5.19541 2.037406 -2.55 0.012 -9.225044 -1.165775
income | 304.7768 13.40835 22.73 0.000 278.2575 331.2962
_cons | 2944.686 1130.203 2.61 0.010 709.3421 5180.031
-----+-----
```

MEDICAL CARE

```
. reg medic children level income
```

```

Source | SS      df    MS          Number of obs = 138
-----+-----
Model | 675583908  3  225194636      F( 3, 134) = 17.43
Residual | 1.7315e+09 134 12921894.2    Prob > F   = 0.0000
-----+-----
Total | 2.4071e+09 137 17570202.4    R-squared  = 0.2807
                                           Adj R-squared = 0.2646
                                           Root MSE   = 3594.7

```

```

-----+-----
      medic |   Coef.  Std. Err.   t  P>|t|  [95% Conf. Interval]
-----+-----
children |  572.4554  379.2776   1.51  0.134  -177.6896  1322.6
level | -9.596638  2.16266  -4.44  0.000  -13.874  -5.319274
income |  65.79839  11.26097   5.84  0.000  43.52615  88.07064
_cons |  8906.002  1183.29   7.53  0.000  6565.661  11246.34
-----+-----

```

#### TRANSPORTATION

```
. reg transp children level income, robust
```

```

Regression with robust standard errors      Number of obs = 138
                                           F( 3, 134) = 40.10
                                           Prob > F   = 0.0000
                                           R-squared  = 0.4334
                                           Root MSE   = 14902

```

```

-----+-----
      |           Robust
transp |   Coef.  Std. Err.   t  P>|t|  [95% Conf. Interval]
-----+-----
children | 1316.764 1757.538   0.75  0.455  -2159.341  4792.868
level | 20.17403 10.33672   1.95  0.053  -2701937  40.61826
income | 449.7743 60.27957   7.46  0.000  330.5519  568.9968
_cons | 11779.26 4505.085   2.61  0.010  2868.984  20689.53
-----+-----

```

#### EDUCATION

```
. reg educ children level income, robust
```

```

Regression with robust standard errors      Number of obs = 135
                                           F( 3, 131) = 12.29
                                           Prob > F   = 0.0000
                                           R-squared  = 0.2703
                                           Root MSE   = 37592

```

```

-----+-----
      |           Robust
educ |   Coef.  Std. Err.   t  P>|t|  [95% Conf. Interval]
-----+-----
children | 12940.61 3906.446   3.31  0.001  5212.731  20668.49
level | 83.4779 21.75108   3.84  0.000  40.44908  126.5067
income | 544.5716 122.1927   4.46  0.000  302.8452  786.2979
_cons | -39470.79 11640.38  -3.39  0.001  -62498.25 -16443.33
-----+-----

```

#### READING

```
. reg read children level income, robust
```

```

Regression with robust standard errors      Number of obs = 138
                                           F( 3, 134) = 95.91
                                           Prob > F   = 0.0000
                                           R-squared  = 0.6504
                                           Root MSE   = 12276

```

```

-----+-----
      |           Robust
read |   Coef.  Std. Err.   t  P>|t|  [95% Conf. Interval]
-----+-----
children | 819.8321 824.8185   0.99  0.322  -811.5151  2451.179
level | -29.4431 10.14498  -2.90  0.004  -49.5081  -9.378096
income | 596.8813 58.99029  10.12  0.000  480.2087  713.5538
_cons | 4735.709 2389.274   1.98  0.050  10.14082  9461.277
-----+-----

```

OTHER

. reg other children level income, robust

Regression with robust standard errors      Number of obs = 138  
F( 3, 134) = 100.38  
Prob > F = 0.0000  
R-squared = 0.7707  
Root MSE = 24284

---

	Robust					
other	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
children	-18128.44	2803.47	-6.47	0.000	-23673.22	-12583.67
level	100.7372	12.68	7.94	0.000	75.65838	125.816
income	1392.624	91.47012	15.22	0.000	1211.712	1573.536
_cons	10166.47	7091.092	1.43	0.154	-3858.471	24191.42

---

NON-CONSUMPTION

. reg non\_cons children level income, robust

Regression with robust standard errors      Number of obs = 138  
F( 3, 134) = 168.00  
Prob > F = 0.0000  
R-squared = 0.8647  
Root MSE = 23644

---

	Robust					
non_cons	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
children	-4596.395	2490.715	-1.85	0.067	-9522.595	329.8041
level	10.29091	13.00911	0.79	0.430	-15.43885	36.02067
income	2153.589	99.66076	21.61	0.000	1956.477	2350.701
_cons	-21633.67	7732.84	-2.80	0.006	-36927.88	-6339.466

---