

Student ID No. ....

# ECONOMICS DEPT

## ASSESSED TEST

For Internal Students of  
Royal Holloway

**COURSE UNIT: EC5040**

**TITLE: Econometrics (Mid-Term Exam 1)**

**Date of Test 12<sup>th</sup> November 2008**

**Time Allowed: 1 hour**

***Instructions to candidates:***

**ANSWER BOTH QUESTIONS**

**WRITE ALL YOUR ANSWERS (INCLUDING ROUGH WORKING) ON THIS ANSWER BOOK**

**STATISTICAL TABLES ARE PROVIDED**

**SILENT NON-PROGRAMMABLE CALCULATORS MAY BE USED**

**DO NOT TURN OVER UNTIL TOLD TO BEGIN**

## Mid-Term Test No. 1 2008/09 – Answers

1. Given the general linear model

$$y = X\beta + u$$

where  $y$  is an  $n \times 1$  vector of observations on the dependent variable,  $X$  is an  $n \times k$  matrix of observations on a set of explanatory variables,  $\beta$  is a  $k \times 1$  vector of parameters and  $u$  is an  $n \times 1$  vector of residuals

a) Derive, from first principles, an expression for the ordinary least squares (OLS) estimate of  $\beta$

(8 marks)

*Minimising the sum of squared residuals implies*

$$\text{Min}_{\beta} \hat{u}'\hat{u} = \begin{bmatrix} \hat{u}_1 & \hat{u}_2 & \dots & \hat{u}_n \end{bmatrix} \begin{bmatrix} \hat{u}_1 \\ \hat{u}_2 \\ \vdots \\ \hat{u}_N \end{bmatrix} = \hat{u}_1^2 + \hat{u}_2^2 + \dots + \hat{u}_N^2$$

$$= (y - X\hat{\beta})'(y - X\hat{\beta}) = y'y - \hat{\beta}'X'y - y'X\hat{\beta} + \hat{\beta}'X'X\hat{\beta}$$

*Since all terms are scalars ( $1 \times 1$ ) can add middle two terms (one is transpose of the other)*

$$= y'y - 2\hat{\beta}'X'y + \hat{\beta}'X'X\hat{\beta}$$

*F.O.C. minimum*

$$\frac{\partial \hat{u}'\hat{u}}{\partial \hat{\beta}} = -2X'y + 2X'X\hat{\beta} = 0$$

which gives  $k$  normal equations  $X'X\hat{\beta} = X'y$

and the  $k$  variable OLS solution  $\hat{\beta} = (X'X)^{-1}X'y$

b) Derive an expression for the variance of the OLS estimator

(6 marks)

$$\begin{aligned} \text{Var}(\hat{\beta}) &= E \left[ (\hat{\beta} - E(\hat{\beta}))(\hat{\beta} - E(\hat{\beta}))' \right] \\ \text{Var}(\hat{\beta}) &= E \left[ (\hat{\beta} - \beta)(\hat{\beta} - \beta)' \right] \\ &= E \left[ (X'X)^{-1}X'uu'X(X'X)^{-1} \right] \\ &= (X'X)^{-1}X'E(uu')X(X'X)^{-1} \\ &= (X'X)^{-1}X'\sigma^2I X(X'X)^{-1} \\ &= \sigma^2(X'X)^{-1} \end{aligned}$$

c) Show that if the Gauss-Markov conditions are satisfied then OLS has the minimum variance of all linear unbiased estimators

(15 marks)

Consider another linear unbiased estimator  $\tilde{\beta} = Cy$

If  $\tilde{\beta}$  is unbiased then  $E(\tilde{\beta}) = E(Cy) = E[CX\beta + Cu] = \beta$

Hence  $CX=I$  and  $\tilde{\beta} = \beta + Cu$

$$\begin{aligned} \text{So } \text{Var}(\tilde{\beta}) &= E\left[(\tilde{\beta} - \beta)(\tilde{\beta} - \beta)'\right] \\ &= E[Cuu'C'] \\ &= \sigma^2 CC' \end{aligned}$$

Let  $D$  be the difference between the OLS and alternative estimated explanatory component ie

$$D = C - (X'X)^{-1}X'$$

So

$$\text{Var}(\tilde{\beta}) = \sigma^2 \left[ (D + (X'X)^{-1}X')(D + (X'X)^{-1}X)'\right]$$

Since  $CX = I = DX + (X'X)^{-1}X'X$  the  $DX = 0$

Cross product terms vanish and

$$\text{Var}(\tilde{\beta}) = \sigma^2 DD' + \sigma^2 (X'X)^{-1} = \sigma^2 DD' + \text{Var}(\hat{\beta})_{OLS}$$

ie variance of alternative estimator equals that of OLS plus a non-negative definite matrix (see problem set 0)

Hence OLS estimate has minimum variance property (BLUE – Best Linear Unbiased Estimator). Main reason for widespread use of OLS, will always provide estimators with smaller standard errors

e) Suppose that one of the independent variables is subject to a linear transformation, (multiplied by a constant  $\lambda$ ) such that  $Z = X\Lambda$  where  $\Lambda$  is a diagonal matrix containing the transformation constant. Show the effect of this transformation on the OLS estimates of the parameters

(10 marks)

Given  $y = Z\gamma + v$

OLS implies  $\hat{\gamma} = (Z'Z)^{-1}Z'y$

Sub. in  $Z = X\Lambda$

$$\hat{\gamma} = (\Lambda'X'X\Lambda)^{-1}\Lambda'X'y$$

Using rules on inverse of a matrix product

$$\hat{\gamma} = \Lambda^{-1}(X'X)^{-1}\Lambda^{-1}\Lambda'X'y$$

$$\hat{\gamma} = \Lambda^{-1}(X'X)^{-1}X'y$$

$$\hat{\gamma} = \Lambda^{-1}\hat{\beta}$$

If the variable to be transformed is  $X_j$  then the transformation matrix looks like

$$\Lambda = \begin{bmatrix} 1 & & & 0 \\ & 1 & & \\ & & \lambda_j & \\ 0 & & & 1 \end{bmatrix}$$

ie a diagonal matrix with ones down the main diagonal except for the  $j$ th element which contains the constant of multiplication for the  $j^{\text{th}}$  variable

Since the inverse of a diagonal matrix is also diagonal with the reciprocal of each original element on the new main diagonal then

$$\Lambda^{-1} = \begin{bmatrix} 1 & & & 0 \\ & 1 & & \\ & & 1/\lambda_j & \\ 0 & & & 1 \end{bmatrix}$$

So using the result in (3) it follows that then the corresponding regression coefficient is multiplied by  $1/a$  and all other coefficients are unchanged.

f) Show the consequences for OLS estimation of omitting relevant variables from your model specification

(11 marks)

True:  $y = X_1\beta_1 + X_2\beta_2 + e$

Estimate:  $y = X_1\beta_1 + u$

$$\hat{\beta}_1 = (X_1'X_1)^{-1}X_1'y$$

$$\hat{\beta}_1 = (X_1'X_1)^{-1}X_1'y + (X_1'X_1)^{-1}(X_1'X_2)\beta_2 + (X_1'X_1)^{-1}X_1'e$$

$$\hat{\beta}_1 = \hat{\beta}_1 + (X_1'X_1)^{-1}(X_1'X_2)\beta_2 + (X_1'X_1)^{-1}X_1'e$$

so

$$E(\hat{\beta}_1) = \beta_1 + (X_1'X_1)^{-1}(X_1'X_2)\beta_2 \neq \beta_1$$

OLS estimates of the coefficients on the set of  $X_1$  variables are biased in the presence of omitted variables

and sign of bias depends on

- the effect of the omitted variables on  $y$ ,  $\beta_2$ ,
- the covariance of  $X_1$  and  $X_2$

Not only is mean biased so is OLS estimates of parameter variances in an unknown way (If  $\sigma^2$  known variance estimate is biased down. But estimate of  $\sigma^2$ 's biased up, so hard to sign direction of bias)

2. The following regression output is taken from a regression of the log of hourly pay (*lnhpay*) on the years of work experience (*xper*), years of education (*yearsed*), years of job tenure, (*tenure*) a dummy variable for being female, (*female*).

Some of the regression output has been obscured.

```
reg lnhpay xper yearsed tenure female
```

| Source   | SS         | df   | MS         |                 |        |  |
|----------|------------|------|------------|-----------------|--------|--|
| Model    |            | 4    |            | Number of obs = | 6005   |  |
| Residual | 1600.00000 | 6000 | .266666666 | F( 7, 6000) =   |        |  |
| Total    | 2400.00000 | 6004 | .399733510 | Prob > F =      |        |  |
|          |            |      |            | R-squared =     |        |  |
|          |            |      |            | Adj R-squared = |        |  |
|          |            |      |            | Root MSE =      | .51639 |  |

  

| lnhpay  | Coef.     | Std. Err. | t      | P> t  | [95% Conf. Interval] |           |
|---------|-----------|-----------|--------|-------|----------------------|-----------|
| xper    | .1100000  | .0900000  | 1.222  | 0.212 | -.0664000            | .2864000  |
| yearsed | .0700000  | .0030894  | 22.861 | 0.000 | .0645705             | .076683   |
| tenure  | -.0100000 | .0300000  | 0.333  | 0.515 | -.0688000            | .0488000  |
| female  | -.1948365 | .0723197  | -2.694 | 0.007 | -.3366081            | -.0530649 |
| _cons   | .7426208  | .0496538  | 14.956 | 0.000 | .6452822             | .8399593  |

a) Interpret the meaning of the coefficient on the female dummy variable

(8 marks)

*this is a "semi-log" equation so the impact of being female relative to being male (net of differences in mean values of control variables) is equal to the % difference /100 in hourly pay of being female relative to being male, (since in the continuous variable case  $d\ln w/d(x) = b_i = dw/dw(x) = \% \text{ change in } w /100$  with respect to a unit change in  $x$ )*

*However since the coefficient is a dummy variable this is only an approximation to the proportionate*

*difference and the true effect is closer to  $\exp(\hat{\beta}_{female}) - 1$*

*So other things equal women earn  $\exp(-.195) - 1 = .177 = 17.7\%$  less than men*

b) Find the estimate of  $R^2$  and hence test the hypothesis that the model as a whole is a good fit

(10 marks)

$$R^2 \text{ (the coefficient of determination)} = ESS/TSS = 1 - (RSS/TSS)$$

*From information in the regression output (highlighted in yellow)*

$$R^2 = ESS/TSS = 1 - (1600/2400) = 1 - .666 = .333$$

$$\text{Test of goodness of fit of the model is given by } F = \frac{ess/q}{RSS/N-k} = \frac{R^2/q}{(1-R^2)/N-k} \sim F[q, N-k]$$

$$\text{So } F = \frac{800/4}{1600/600} = \frac{0.333/4}{0.666/6000} \sim F[4, 6000]$$

$$= 750 \sim F[4, 6000]$$

*So estimated F is greater than 5% critical value ( $F(4, \infty) = 2.37$ ) so **reject null** that model as a whole has no explanatory power*

d) The variance/covariance matrix of the OLS parameter estimates (excluding the constant) is given by

|        | xper              | yrsted     | female     | tenure   |
|--------|-------------------|------------|------------|----------|
| xper   | .00810000         |            |            |          |
| yrsted | .00024251         | .00456114  |            |          |
| female | .00058968         | .00013298  | .13443127  |          |
| tenure | <b>-.00050000</b> | -.00670241 | -.00574375 | .0009000 |

Given this information test the hypothesis that the returns to experience (*xper*) equal the returns to job tenure (*tenure*) in the model above

(12 marks)

The test of equality of experience and tenure effects is given by a form of the *F* test

$$(\hat{\beta}_{xper} - \hat{\beta}_{tenure})^2 / \text{Var}(\hat{\beta}_{xper} - \hat{\beta}_{tenure}) \sim F(1, N-k)$$

which can check from the variance/covariance matrix of the OLS estimates since the square root of the  $i^{\text{th}}$  element on the main diagonal should equal the standard error on the  $i^{\text{th}}$  variable in the regression and the off diagonal terms are the covariances of the parameter estimates. The relevant covariance is highlighted (in green)

$$\text{Since } \text{Var}(\hat{\beta}_{xper} - \hat{\beta}_{tenure}) = \text{Var}(\hat{\beta}_{xper}) + \text{Var}(\hat{\beta}_{tenure}) - 2\text{Cov}(\hat{\beta}_{xper}, \hat{\beta}_{tenure})$$

$$\text{Var}(\hat{\beta}_{xper} - \hat{\beta}_{tenure}) = .0081 + .0009 - 2(-.0005) = .01$$

$$\text{then } F = \frac{(.11 - -.01)^2}{.01} = 1.44 \sim F[1, 6000]$$

Since 95% critical value = 3.84 then  $F < F_{\text{critical}}$

So can **not** reject null that coefficients are equal (the relatively large standard errors ensure that the confidence intervals overlap)

d) Consider a simple model of 204 observations split equally into two sub-samples such that

$$y_i = a_1 + b_1 X_i + u_i \quad i=1..N_1 \quad \text{in sub-sample 1}$$

and

$$y_i = a_2 + b_2 X_i + u_i \quad i=N_1+1..N \quad \text{in sub-sample 2}$$

Suppose that  $RSS_1 = 8$  and  $RSS_2 = 2$  and that the RSS from the pooled regression is 12. Test the hypothesis of no structural change across the two sub-samples at the 5% level.

(10 marks)

The unrestricted form of the model (intercepts and the slopes vary in two periods) in (partitioned) matrix form is given by

$$y = \begin{bmatrix} y_1 \\ \dots \\ y_2 \end{bmatrix} = \begin{bmatrix} X_1 & : & 0 \\ \dots & \dots & \dots \\ 0 & : & X_2 \end{bmatrix} \begin{bmatrix} a_1 \\ b_1 \\ a_2 \\ b_2 \end{bmatrix} + \begin{bmatrix} u_1 \\ \dots \\ u_2 \end{bmatrix} = X\beta + u \quad (1)$$

where  $X_1$  is an  $N_1$  by 2 matrix of observations from the 1<sup>st</sup> sub-sample and  $X_2$  is an  $N_2$  by 2 matrix of observations from the 1<sup>st</sup> sub-sample with  $N = N_1 + N_2$

ie stacking the data from the second period below that of the observations from the 1st period in a way that allows the coefficients to differ between the periods

Compare this with estimates from the restricted (pooled) model based on

$$y = \begin{bmatrix} y_1 \\ \dots \\ y_2 \end{bmatrix} = \begin{bmatrix} i X_1 \\ i X_2 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} + \begin{bmatrix} u_1 \\ \dots \\ u_2 \end{bmatrix} = X\beta + u$$

To test formally use

$$F = \frac{(RSS_{restricted} - RSS_{unrestricted})/q}{RSS_{unrestricted} / N - k} \sim F[q, N - k]$$

which in this case becomes the Chow test

$$F = \frac{(RSS_{restricted} - RSS_1 + RSS_2)/q}{RSS_1 + RSS_2 / N - 2k} \sim F[q, N - 2k]$$

(remember that there are 4 parameters in the unrestricted model so  $k=4$  and  $q=2$  restrictions)

$$\text{hence } \hat{F} = \frac{(12 - (8 + 2))/2}{(8 + 2)/204 - 2 \cdot 2} = 20$$

From Tables the 5% critical value given the degrees of freedom  $F^{0.05}[2, \infty] = 3.0$

$\hat{F} > F_{critical}$  so reject null (of no structural change)

d) Outline the form of a technique that could be used to test for functional form error (10 marks)

Either

Ramsey RESET Test

- if model is good fit then addition of extra variables should not be statistically significant

rather than add higher order terms of original variables a more parsimonious alternative is to use fact that

$$\hat{y} = X \hat{\beta}$$

so predicted values are linear function of all the  $X$  variables (weighted by their estimated coefficients)

and hence  $(\hat{y})^j = (X \hat{\beta})^j$

are linear functions of higher powers of all the  $X$  variables

$$y = X\beta + \delta_2 \hat{y}^2 + \delta_3 \hat{y}^3 + \dots + \delta_j \hat{y}^j + u$$

and test null  $H_0: \delta_2 = \delta_3 = \dots = \delta_j = 0$

If estimated  $F$  value greater than critical value reject null that functional form is acceptable.

OR

### **LM Test of Omitted Variables**

1. Run restricted regression (no higher order terms)

2. save residuals

3. Regress residuals on unrestricted model (containing higher order values of  $X$  (or the  $\hat{y}_j$ ) - the auxiliary regression

Can show

$$NR^2_{aux} \overset{a}{\sim} \chi^2_{(No.ofrestrictions)}$$

If estimated Chi-squared value greater than critical value reject null that functional form is acceptable.