



AFRICAN ECONOMIC RESEARCH CONSORTIUM

Collaborative MA Programme in Economics for Anglophone Africa
(Except Nigeria)

JOINT FACILITY FOR ELECTIVES (JFE) 2010

JUNE – OCTOBER

ECONOMETRICS THEORY AND PRACTICE II

Second Semester: Final Examination

Duration: 3 Hours

Date: Monday, September 27, 2010

INSTRUCTIONS:

1. Answer **Question 1** and **ANY TWO (2)** from the other remaining questions.
 2. All questions carry equal weight.
 3. Credit will be given for orderly presentation of **relevant** materials.
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Question 1 (COMPULSORY)

- (a)
- (i) Discuss stationarity versus nonstationarity in panel data econometrics. [2 marks]
 - (ii) Given a typical panel data model, how would you test for the ADF unit roots in it. [Outline the steps involved in the ADF test] [4 marks]
 - (iii) What is panel cointegration? [2 marks]
 - (iv) Consider a panel regression of the form:
$$y_{it} = \alpha_i + \beta x_{it} + e_{it}$$
$$y_{it} = y_{it-1} + u_{it}$$
$$x_{it} = x_{it-1} + \varepsilon_{it}$$
$$t = 1, 2, \dots, T; i = 1, 2, \dots, N$$

How would you conduct the KAO [Engle-Granger Based] cointegration test? [3 marks]



(b)

- (i) What is the relationship between the hazard rate and the survivor rate functions [Mathematical derivation is important] **[6 marks]**

- (ii) Given the basic Cox proportional hazard model

$$\mu(t|x_i) = \mu_0(t) \exp(\beta_1 x_1 + \beta_2 x_2 + \dots + \beta_k x_k)$$

Why is this model considered as a semi-parametric model? Discuss the method of estimation of the parameters $\beta_1, \beta_2, \dots, \beta_k$ **[3 marks]**

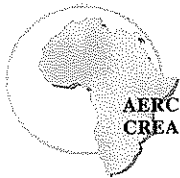
Question 2

Consider the following model:

$$y_{it} = \alpha_i + \beta x_{it} + u_{it}, \quad t = 1, \dots, T$$

The panel is balanced. There are N individuals. u_{it} is an idiosyncratic error term with the usual properties. α_i is the individual effect that does not vary overtime. y_{it} is the dependent variable and x_{it} is a scalar covariate.

- (a) Write down formulae for the pooled OLS, the Fixed Effects, and the between estimators of β . Denote these as $\hat{\beta}$, $\hat{\beta}_\omega$ and $\hat{\beta}_b$ respectively. **[6 marks]**
- (b) In the Stata output below someone is trying to compute estimates for the wage- gender relationship using the data taken from the 1991 and 1997 Waves of the British Household Panel Survey. The dependent variable is the log nominal weekly earnings ($\ln w$), personal identifier (pid) and the independent variable is a dummy for whether the individual is male (m). The sample size is 1685. The three 2 by 2 matrices report sample variance and covariances. Complete this person's calculations. Comment on your result. **[8 marks]**



```
. by pid: egen lw1=mean(lw)

. by pid: egen m1=mean(m)

. gen lw2=lw-lw1

. gen m2=m-m1

. li pid yr lw m lw1 m1 lw2 m2 in 1/20
```

| | pid | yr | lw | m | lw1 | m1 | lw2 | m2 |
|-----|----------|----|----------|---|----------|----|-----------|----|
| 1. | 10023526 | 1 | 5.273769 | 0 | 5.228222 | 0 | .045547 | 0 |
| 2. | 10023526 | 7 | 5.182675 | 0 | 5.228222 | 0 | -.0455465 | 0 |
| 3. | 10028005 | 1 | 5.848423 | 1 | 5.88939 | 1 | -.0409675 | 0 |
| 4. | 10028005 | 7 | 5.930357 | 1 | 5.88939 | 1 | .040967 | 0 |
| 5. | 10060111 | 1 | 5.790793 | 1 | 5.908262 | 1 | -.1174688 | 0 |
| 6. | 10060111 | 7 | 6.025732 | 1 | 5.908262 | 1 | .1174693 | 0 |
| 7. | 10061649 | 1 | 6.06943 | 1 | 6.117547 | 1 | -.0481167 | 0 |
| 8. | 10061649 | 7 | 6.165663 | 1 | 6.117547 | 1 | .0481162 | 0 |
| 9. | 10071687 | 1 | 5.481409 | 1 | 5.517569 | 1 | -.03616 | 0 |
| 10. | 10071687 | 7 | 5.553729 | 1 | 5.517569 | 1 | .0361605 | 0 |
| 11. | 10071717 | 1 | 5.394397 | 0 | 5.549475 | 0 | -.1550775 | 0 |
| 12. | 10071717 | 7 | 5.704552 | 0 | 5.549475 | 0 | .1550775 | 0 |
| 13. | 10080643 | 1 | 5.848423 | 0 | 5.701295 | 0 | .1471276 | 0 |
| 14. | 10080643 | 7 | 5.554167 | 0 | 5.701295 | 0 | -.1471281 | 0 |
| 15. | 10092986 | 1 | 5.442957 | 0 | 5.686657 | 0 | -.2436996 | 0 |
| 16. | 10092986 | 7 | 5.930357 | 0 | 5.686657 | 0 | .2437 | 0 |
| 17. | 10094083 | 1 | 6.620843 | 1 | 6.529266 | 1 | .0915766 | 0 |
| 18. | 10094083 | 7 | 6.437689 | 1 | 6.529266 | 1 | -.0915771 | 0 |
| 19. | 10127666 | 1 | 5.848423 | 0 | 5.949624 | 0 | -.1012015 | 0 |
| 20. | 10127666 | 7 | 6.050826 | 0 | 5.949624 | 0 | .101202 | 0 |



```
. corr lw m, cov
(obs=3370)
```

```
-----+-----
              lw      m
lw | .282251
m  | .071346 .233713
```

```
. corr lw1 m1, cov
(obs=3370)
```

```
-----+-----
              lw1     m1
lw1 | .221596
m1  | .071346 .233713
```

```
. corr lw2 m2, cov
(obs=3370)
```

```
-----+-----
              lw2     m2
lw2 | .060655
m2  | 0      0
```

- (c) The random effects estimator is found by running OLS on the following transformed equation:

$$y_{it} - \mu \bar{y}_i = (1 - \mu)\beta_0 + \beta(x_{it} - \mu \bar{x}_i) + (v_{it} - \mu \bar{v}_i),$$

where

$$\mu = 1 - \sqrt{\sigma_u^2 / (\sigma_u^2 + T\sigma_a^2)} = 1 - \sqrt{\theta},$$

and σ_u^2 and σ_a^2 are the population variances of u_{it} and a_i respectively.

- (i) Clearly $0 \leq \mu \leq 1$. Explain what happens when $\mu = 0$ and when $\mu = 1$.
- (ii) The person above discovers the Stata command for the Random Effects estimator, and generates the output below. Comment on the relationship between the estimate on the male dummy, and how close/far away from the three estimates computed in (b) above

[6 marks]



```
. xtreg lw m, re i(pid)
```

```
Random-effects GLS regression           Number of obs   =       3370
Group variable (i): pid                 Number of groups =       1685

R-sq:  within = .
      between = 0.0983
      overall  = 0.0772

Obs per group: min =         2
               avg  =        2.0
               max  =         2

Random effects u_i ~ Gaussian           Wald chi2(1)     =       183.45
corr(u_i, X) = 0 (assumed)             Prob > chi2      =        0.0000
```

| lw | Coef. | Std. Err. | z | P> z | [95% Conf. Interval] | |
|---------|-----------|-----------------------------------|--------|-------|----------------------|----------|
| m | .3052729 | .0225388 | 13.54 | 0.000 | .2610976 | .3494482 |
| _cons | 5.437063 | .0178597 | 304.43 | 0.000 | 5.402059 | 5.472067 |
| ----- | | | | | | |
| sigma_u | .37330476 | | | | | |
| sigma_e | .34824533 | | | | | |
| rho | .53468816 | (fraction of variance due to u_i) | | | | |

Question 3

We are interested in explaining unemployment. We believe that the probability of being unemployed is affected by gender, age and education level. Let y_i denote unemployment ($y_i=1$ if a person i is unemployed and 0 otherwise), let x_{1i} denote gender ($x_{1i}=1$ if person i is male and 0 otherwise), let x_{2i} denote age (x_{2i} is the age of person i , continuously measured) and let x_{3i} denote education level ($x_{3i}=1$ if person i has a university degree and otherwise). We have data on these variables for $n = 7867$ individuals taken from Kenya Household Panel Survey. For the purpose of explaining the probability of being unemployed, we consider a binary choice model:

$$y_i^* = x_i' \beta + \varepsilon_i, \quad \varepsilon_i \text{ iid}, \quad i = 1, 2, \dots, n$$

$$y_i = \begin{cases} 1 & y_i^* \geq 0 \\ 0 & y_i^* < 0 \end{cases}$$

Let $F(\cdot)$ denote the cdf of ε ,

- (a) Give an expression for log likelihood function of the logit model. [5 marks]



- (b) The binary choice model is now estimated both by logit and probit (see stata output in appendix A). How do you interpret the coefficient estimates of β from the logit and probit? [4 marks]
- (c) Using the logit estimates, what is the difference in the predicted probabilities of being unemployed between men and women? [6 marks]
- (d) How much does the probability of being unemployed increase/decrease with age. Comment on your findings? [5 marks]

Question 4

The Belgian government is contemplating increasing the tax on tobacco in order to lower the incidence of smoking. A consultant to the government therefore wishes to examine whether such a tax increase will actually lower the demand for tobacco. He knows that in order to assess this he needs to take both the price and income effect into account. He has an estimate of the price elasticity for tobacco and he turns to you for assistance in estimating the income elasticity. The data he has available for estimating the income elasticity is the Belgian Household Budget Survey 1995-1996, which contains information on household, how many adults live in the household as well as the age class of the head of the household for $n = 2724$ households. He has estimated a linear regression model on this data by running OLS of the budget share for tobacco (*btobacco*) on log total expenditure (*lnx*), number of children in the household (*nkids*, *nkids2*), number of adults in the household (*nadults*) and age (*age*). The stata output for this regression is given in appendix B. Denoting the budget share of tobacco by ω and the log total expenditure by $\ln x$, the income elasticity e for tobacco can be calculated by the formula

$$e = \frac{1}{\omega} \frac{\partial \omega}{\partial \ln x} + 1$$

- (a) Calculate the income elasticity e_{OLS} resulting from the OLS. [6 marks]
- (b) As an alternative to the OLS, you suggest to model tobacco expenditures by a tobit model. Why would a tobit model be more appropriate for analyzing this data? Write down the appropriate tobit model. [5 marks]
- (c) The Stata output for the estimation of the tobit model is given in appendix B. Calculate the income elasticity e_{TOBIT} resulting from the tobit. Comment on your findings for e_{OLS} and e_{TOBIT} . [9 marks]



Question 5

There are many factors that influence elections. One such factor that has received considerable attention is the impact of campaign expenditures on election outcomes. The following equation describes the percentage of the vote (pctvote) received by a candidate (measured on a 0% - 100% scale):

$$pctvote = \beta_0 + \beta_1 \log(exp_cond) + \beta_2 \log(exp_opp) + \beta_3 Party + u$$

where $\log(exp_cond)$ is the log of the candidate's own expenditures, $\log(exp_opp)$ is the log of the candidate's opponent's expenditures (with expenditures measured in thousands of dollars), and Party is the political party of the candidate (1 if Democrat, 0 if Republican).

- (a) Using data on 173 congressional races for the U.S. House of Representatives in the 1992 election, the following equation was estimated:

$$\hat{pctvote} = 51.13 + 6.30 \log(exp_cond) - 6.67 \log(exp_opp) + 1.21 Party$$

$$R^2 = 0.786$$

$$SSR = 10351.2$$

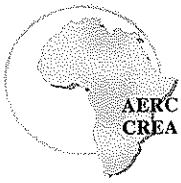
How do you interpret the coefficients on $\log(exp_cond)$ and $\log(exp_opp)$? How do you interpret the coefficient on Party? **[5 marks]**

- (b) The following standard errors (and covariance) were obtained from the computer regression output:

| Estimated Standard errors | |
|----------------------------|----------|
| S.E. (β_0) | 2.90 |
| S.E. (β_1) | 0.37 |
| S.E. (β_2) | 0.39 |
| S.E. (β_3) | 1.34 |
| Cov (β_1, β_2) | -0.00057 |

Test the hypothesis that political party has no effect on the percentage of the vote a candidate receives. (Notes: Be sure to state the null and alternative hypotheses. Use a 5% significance level for a two-sided test. As the degrees of freedom are greater than 150, you can use the critical values from a standard normal distribution (i.e., ± 1.96). You can test the hypothesis using either a t-statistic or a confidence interval.) **[5 marks]**

- (c) It appears from the estimated regression that the coefficients on $\log(exp_cond)$ and $\log(exp_opp)$ are of equal magnitudes and opposite signs. Test the hypothesis $H_0: \beta_1 = -\beta_2$. (Note: This hypothesis also means that it is only the difference in log expenditures between the two candidates that matters.) **[5 marks]**
- (d) Your colleague does not believe that campaign expenditures matter in an election, and therefore estimates the following regression:



$$\hat{pctvote} = 45.70 + 8.65\text{Party}$$

$$R^2 = 0.066$$

$$SSR = 45258.1$$

Test the joint hypothesis that campaign expenditures do not matter ($H_0: \beta_1 = \beta_2 = 0$). (The critical value for the test is 3.02.) **[5 marks]**



APPENDICES

Appendix A

```

obs:          7,867          BHPS annual panel 1991-2002
vars:          4            13 Oct 2005 07:21
size:         86,537 (99.8% of memory free)

```

| variable name | storage type | display format | value label | variable label |
|---------------|--------------|----------------|-------------|--|
| age | byte | %8.0g | aage | age at date of interview |
| unemp | byte | %9.0g | | unemployed |
| male | byte | %9.0g | | =1 if individual is male |
| degree | float | %9.0g | | =1 if individual has university degree |

summarize male age degree

| Variable | Obs | Mean | Std. Dev. | Min | Max |
|----------|------|----------|-----------|-----|-----|
| unemp | 7867 | .0526249 | .2232977 | 0 | 1 |
| male | 7867 | .5215457 | .4995673 | 0 | 1 |
| age | 7867 | 38.84912 | 11.85259 | 16 | 65 |
| degree | 7867 | .1552053 | .3621233 | 0 | 1 |



. logit unemp male age degree

Iteration 0: log likelihood = -1621.9607
 Iteration 1: log likelihood = -1574.1991
 Iteration 2: log likelihood = -1571.2739
 Iteration 3: log likelihood = -1571.2507
 Iteration 4: log likelihood = -1571.2507

Logistic regression

Number of obs = 7867
 LR chi2(3) = 101.42
 Prob > chi2 = 0.0000
 Pseudo R2 = 0.0313

Log likelihood = -1571.2507

| unemp | Coef. | Std. Err. | z | P> z | [95% Conf. Interval] | |
|--------|-----------|-----------|-------|-------|----------------------|-----------|
| male | .2148666 | .1025754 | 2.09 | 0.036 | .0138225 | .4159107 |
| age | -.0361777 | .0044211 | -8.18 | 0.000 | -.0448429 | -.0275125 |
| degree | -.9472709 | .1926279 | -4.92 | 0.000 | -1.324815 | -.5697272 |
| _cons | -1.574088 | .1721039 | -9.15 | 0.000 | -1.911406 | -1.236771 |

. probit unemp male age degree

Iteration 0: log likelihood = -1621.9607
 Iteration 1: log likelihood = -1574.2058
 Iteration 2: log likelihood = -1573.3143
 Iteration 3: log likelihood = -1573.3125

Probit regression

Number of obs = 7867
 LR chi2(3) = 97.30
 Prob > chi2 = 0.0000
 Pseudo R2 = 0.0300

Log likelihood = -1573.3125

| unemp | Coef. | Std. Err. | z | P> z | [95% Conf. Interval] | |
|--------|-----------|-----------|--------|-------|----------------------|-----------|
| male | .0960036 | .0479677 | 2.00 | 0.045 | .0019885 | .1900186 |
| age | -.0159835 | .0019968 | -8.00 | 0.000 | -.0198971 | -.0120698 |
| degree | -.4218795 | .0816118 | -5.17 | 0.000 | -.5818358 | -.2619233 |
| _cons | -1.028603 | .0810051 | -12.70 | 0.000 | -1.18737 | -.8698358 |



Appendix B

```
. reg btobacco lnx age nadults nkids2 nkids
```

| Source | SS | df | MS | Number of obs = | 2724 |
|----------|------------|------|------------|-----------------|--------|
| Model | .116758246 | 5 | .023351649 | F(5, 2718) = | 40.32 |
| Residual | 1.5741102 | 2718 | .000579143 | Prob > F = | 0.0000 |
| | | | | R-squared = | 0.0691 |
| | | | | Adj R-squared = | 0.0673 |
| Total | 1.69086845 | 2723 | .000620958 | Root MSE = | .02407 |

| btobacco | Coef. | Std. Err. | t | P> t | [95% Conf. Interval] | |
|----------|-----------|-----------|--------|-------|----------------------|-----------|
| lnx | -.0141745 | .0011452 | -12.38 | 0.000 | -.0164202 | -.0119289 |
| age | -.0025072 | .0003865 | -6.49 | 0.000 | -.0032651 | -.0017493 |
| nadults | .0027508 | .0006524 | 4.22 | 0.000 | .0014716 | .0040301 |
| nkids2 | -.0047776 | .0022332 | -2.14 | 0.032 | -.0091565 | -.0003987 |
| nkids | .001168 | .0005623 | 2.08 | 0.038 | .0000654 | .0022705 |
| _cons | .2069766 | .0151311 | 13.68 | 0.000 | .1773069 | .2366462 |

```
. predict wreg, xb
```

```
. summarize wreg
```

| Variable | Obs | Mean | Std. Dev. | Min | Max |
|----------|------|----------|-----------|-----------|---------|
| wreg | 2724 | .0321908 | .0062637 | .00128599 | .060071 |

```
. tab d2
```

| | | | |
|-------------|-------|---------|--------|
| dummy=1 if | | | |
| tobacco | | | |
| expenditure | | | |
| >0 | Freq. | Percent | Cum. |
| 0 | 1,688 | 61.97 | 61.97 |
| 1 | 1,036 | 38.03 | 100.00 |
| Total | 2,724 | 100.00 | |



```
. tobit btobacco lnx age nadults nkids2 nkids , ll(0)
```

Tobit regression

```
Number of obs   =      2724
LR chi2(5)      =      145.58
Prob > chi2     =      0.0000
Pseudo R2      =     -0.1081
```

Log likelihood = 746.40082

| btobacco | Coef. | Std. Err. | t | P> t | [95% Conf. Interval] | |
|----------|-----------|-----------|-------|-------|----------------------|-----------|
| lnx | -.0256124 | .0027221 | -9.41 | 0.000 | -.03095 | -.0202748 |
| age | -.006387 | .0009186 | -6.95 | 0.000 | -.0081882 | -.0045858 |
| nadults | .0076941 | .001545 | 4.98 | 0.000 | .0046645 | .0107237 |
| nkids2 | -.0135256 | .0054335 | -2.49 | 0.013 | -.0241798 | -.0028714 |
| nkids | .0029758 | .0012966 | 2.29 | 0.022 | .0004333 | .0055183 |
| _cons | .334203 | .0357935 | 9.34 | 0.000 | .2640178 | .4043883 |
| /sigma | .0483493 | .0011926 | | | .0460108 | .0506879 |

```
Obs. summary:      1688 left-censored observations at btobacco<=0
                   1036 uncensored observations
                   0 right-censored observations
```

```
. predict wtobit, e(0,1)
```

```
. summarize wtobit
```

| Variable | Obs | Mean | Std. Dev. | Min | Max |
|----------|------|----------|-----------|----------|----------|
| wtobit | 2724 | .0334894 | .0040911 | .0230732 | .0575392 |