

**PhD/MA. Econometrics Examination**  
**January 13., 2010**

**Total Time: 8 hours**

**MA students are required to answer from A and B.**

**PhD students are required to answer from A, B, and C.**

**PART A**  
**(ANSWER ANY TWO)**

**Question 1.**

- a. Is  $f(y) = (1/2)y(1-y)$   $0 \leq y \leq 2$  a probability density function?
- b. The cumulative distribution function for random variable  $Y$  is given by  $F(y) = y/5$   $0 \leq y \leq 5$ . Find the probability density function for  $Y$ .
- c. If  $P(B) = .05$ ,  $P(A|B) = .80$ , and  $P(A|\bar{B}) = .50$  find  $P(B|A)$ .

**Question 2.**

- a.  $U$  is a random variable with probability density function  $f(u) = 2e^{-2u}$   $0 < u < \infty$ . Let  $g(U) = e^{-U}$ , what is  $E[g(U)]$ ?
- b. If  $E(Y_1) = 2$ ,  $E(Y_2) = 1$ ,  $V(Y_1) = \frac{1}{2}$ ,  $V(Y_2) = 1$ , and  $Cov(Y_1, Y_2) = -\frac{1}{2}$  find the mean and variance of the random variable  $U = 2Y_1 - Y_2$ .
- c. Is it possible that the two random variables  $Y_1$  and  $Y_2$ , have standard deviations  $\sigma_{Y_1} = 2.5$ ,  $\sigma_{Y_2} = .9$ , respectively,  $cov(Y_1, Y_2) = 2.7$  and correlation  $\rho = -1.2$ ? Explain.

**Question 3.**

- a. Find the maximum likelihood estimator of  $\beta$  where  $Y_i$  is independently and identically distributed with probability density function (the exponential density):

$$f(y) = \frac{1}{\beta} e^{-y/\beta} \quad y > 0.$$

- b. Let  $\hat{u}_i$  be the OLS residuals for the model  $y_i = \beta + u_i$ . Given that  $\frac{1}{\sigma^2} \sum_{i=1}^T \hat{u}_i^2 \sim \chi^2(n-1)$  and

$\tilde{\sigma}^2 = \frac{1}{n} \sum_{i=1}^T \hat{u}_i^2$  show that  $E(\tilde{\sigma}^2) \neq \sigma^2$ , i.e.  $\tilde{\sigma}^2$  is biased. Use the result to create an unbiased estimator of  $\sigma^2$  denoted by  $\hat{\sigma}^2$ .

- c. Normality is often assumed in applied work. State the *central limit theorem* and discuss how this result is useful.

**PART B**  
**(ANSWER ANY TWO)**

**Question 4.**

A. Consider the following stock-adjustment model:

$$\ln Y(t)^* = a_0 + a_1 \ln X(t) + U(t) \quad (1)$$

where  $Y(t)^*$  is desired inventory of cars in billions of \$;  $X(t)$  is the oil price (\$ per gallon)

The stock-adjustment mechanism is as follows:

$$\frac{Y(t)}{Y(t-1)} = \left( \frac{Y(t)^*}{Y(t-1)} \right)^d \quad 0 < d < 1 \quad (2)$$

(Hint: take the log of the equation (2) to make it operational.)

Combine equation (1) and (2) to come up with a Koyck type distributed lag model and discuss the estimation method. What is the interpretation of  $d$ ?

B. Given the following distributed lag type model for the annual car inventory data (1980 – 2004):

$$\ln Y(t) = \beta_0 + \beta_1 \ln Y(t-1) + \beta_2 \ln X(t) + V(t)$$

Where  $Y(t)$  = Inventory of cars in billions of \$;  $X(t)$  = oil price (\$ per gallon).

The following estimates were found:  $\hat{\beta}_0 = 1.34$ ,  $\hat{\beta}_1 = .62$ ,  $\hat{\beta}_2 = .42$ .

- Calculate the long-run multiplier effect of oil price shock on car inventory.
- Calculate the half life value and interpret the result.
- Calculate the impact multipliers at lag 0, 1, and 2.

**Question 5.**

Given the following simple slope-only model:  $y(t) = \beta x(t) + u(t)$

where  $u(t)$  is heteroskedastic, i.e.,  $\sigma^2(t) = \exp(g_0 z(t))$ ,  $\exp$  = exponent, and  $g_0$  is a scalar parameter.

- Show that the OLS estimator of  $\beta$  is still unbiased.
- Derive the variance of  $\hat{\beta}$ .
- Assuming the above heteroskedastic form, present the Var-Cov(U) matrix, and present its transformation matrix, P.
- Present the weighted GLS iterative method. Show all the iterative steps.
- What is difference between GLS versus OLS?

**Question 6.**

Consider the following linear model in matrix notation

$$\mathbf{Y} = \mathbf{X}\beta + \mathbf{U}$$

- Use the method of your choice --OLS or MLE—to derive the estimator of  $\beta$ .
- Now derive the Var-Cov( $\mathbf{U}$ )
- Demonstrate the Gauss-Markov theorem that  $\hat{\beta}_{OLS}$  is BLUE.
- Derive the Var-Cov matrix of the OLS estimator of  $\beta$  (either use OLS method or MLE-hessian matrix method)

**PART C**  
**(ANSWER TWO QUESTIONS)**

**Question 7.**

Calculate the autocorrelation functions up to 3 lags for the following model:

$$y_t = \phi_1 y_{t-1} + e_t$$

- Plot the autocorrelation functions for the following values, each separately
  - $\phi_1 = .7$
  - $\phi_1 = -.8$
- What is the difference between the unit root versus the long-memory models.
- Explain the meaning of cointegration with examples.
- Explain the meaning of the spurious regression phenomenon with examples.

**Question 8.**

Given the following nonlinear consumption function:

$$y(t) = \left( \frac{\alpha}{1 + \exp(\beta x(t))} \right) + u(t)$$

Where  $\alpha$ , and  $\beta$  are unknown parameters.

- You may express this function generically:  $y(t) = f(\alpha, \beta, x(t)) + u(t)$  and show how you might estimate using the method of non-linear least squares OR the Newton-Raphson (NR) method of estimating it using a distribution of your choice.
- If you were to test the unbiasedness of the estimators (of  $\alpha$  and  $\beta$ ) of the two methods – non-linear least squares versus the NR method—how would you set up a Monte Carlo experiment. (Show all the steps in details.)

**Question 9.**

Consider the following structural model

$$y_1(t) = \alpha_0 + \alpha_1 x_1(t) + \alpha_2 y_2(t) + \alpha_3 y_1(t-1) + u_1(t)$$

$$y_2(t) = \beta_0 + \beta_1 x_1(t) + \beta_2 x_2(t) + u_2(t)$$

- List the endogenous variables. List the predetermined variables.
- Write the above model in a general matrix notation:

$$\mathbf{Y}\Gamma + \mathbf{X}\Delta + \mathbf{U} = \mathbf{0}$$

- Using the general method, test the identification condition for these equations – one equation at a time.
- Derive the Var-Cov( $\mathbf{U}$ ).
- Present the 3SLS method of estimation.
- Consider a generic equation as follows to represent the first equation of the above model.

$$\mathbf{y}_1 = \mathbf{Y}_1\gamma_1 + \mathbf{X}_1\delta_1 + \mathbf{u}_1$$

Now, show that the 2SLS estimator of this equation is consistent.

$$\mathbf{y}_1 = \mathbf{Z}_1\beta_1 + \mathbf{u}_1$$