

# THE AUSTRALIAN NATIONAL UNIVERSITY

*First Semester Final Examination– June, 2019*

## **Advanced Econometrics I**

**(EMET 4314/8014)**

*Reading Time: 0 Minutes*  
*Writing Time: 120 Minutes*

### INSTRUCTIONS

- Answer all 4 questions of this handout
- Write on paper or an electronic device (such as an Ipad). Handwritten answers only! Do not type anything.
- If written on a piece of paper, scan your work and upload to Wattle.
- If using an electronic device, upload your file to Wattle.
- Provide complete, self-contained, and correct answers!
- Always show your work!
- Good luck!

## EXAM QUESTIONS

1. [ 15 marks ]

Let  $Y_i = X_i' \beta + u_i$ , where  $E(u_i | X_i) = 0$ . Notice that  $\dim X_i = K \times 1$ .

You know that the solution to  $\hat{\beta}^{\text{OLS}} := \underset{b \in \mathbb{R}^K}{\operatorname{argmin}} \sum_{i=1}^N (Y_i - X_i' b)^2$  is

$$\hat{\beta}^{\text{OLS}} = \left( \sum_{i=1}^N X_i X_i' \right)^{-1} \sum_{i=1}^N X_i Y_i.$$

Note: In your derivations for parts (a), (b), and (c), make use of the stochastic order symbols  $o_p(1)$  and  $O_p(1)$  wherever you can.

- Derive the probability limit of  $\hat{\beta}^{\text{OLS}}$ .
- Derive the asymptotic distribution of  $\sqrt{N}(\hat{\beta}^{\text{OLS}} - \beta)$  assuming homoskedasticity.
- Suggest a consistent estimator for the asymptotic variance of  $\sqrt{N}(\hat{\beta}^{\text{OLS}} - \beta)$  under homoskedasticity. Prove that it is consistent.

2. [ 15 marks ]

Are the following statements true or false? Provide a complete yet short explanation. Use mathematical derivations where necessary.

(Note: you will not receive any credit without providing a correct explanation.)

- $\operatorname{Cov}(X, Y) = \operatorname{Cov}(X, E(Y|X))$ .
- In the linear model, omitting explanatory variables does not necessarily lead to bias in the OLS estimator.
- Any maximum likelihood estimator is also an M-estimator.
- You have a random sample  $Y_1, \dots, Y_N$  with  $N > 3$  and your goal is to estimate  $E(Y_3)$ . Consider the estimator  $\hat{\theta} := Y_3$ . That estimator is unbiased and efficient.
- In a model with heterogeneous slope coefficients, we defined

$$\text{LATE} := \frac{E(\beta_{1i} \cdot \pi_{1i})}{E(\pi_{1i})}.$$

It can be shown that  $\text{LATE} = \text{ATE}$ .

3. [ 15 marks ]

You have a random sample  $Y_1, \dots, Y_N$  drawn from a *Poisson* distribution. The probability mass function of  $Y_i$  is

$$f_Y(y|\theta) = \frac{\theta^y \exp(-\theta)}{y!}, \quad \text{with } y = 0, 1, 2, \dots$$

It is known that  $E(Y_i) = \operatorname{Var}(Y_i) = \theta$ .

- (a) Derive the likelihood function  $\mathcal{L}(\theta|y_1, \dots, y_N)$ .
- (b) Derive the log likelihood function  $L(\theta|y_1, \dots, y_N)$ .
- (c) Derive the score function  $\partial \ln f_Y(y|\theta) / \partial \theta$ .
- (d) Derive the maximum likelihood estimator  $\hat{\theta}^{\text{ML}}$  of  $\theta$ .
- (e) Derive  $E(\hat{\theta}^{\text{ML}})$ .
- (f) Derive  $\text{Var}(\hat{\theta}^{\text{ML}})$ .
- (g) Derive the Cramér Rao bound.
- (h) Is  $\hat{\theta}^{\text{ML}}$  efficient?

4. [ 15 marks ]

(a) [ 5 marks ]

Consider Heckman's sample selection model

$$Y_i^* = X_i' \beta + e_i$$

$$D_i = 1 \cdot (Z_i' \gamma + v_i > 0),$$

where, most generally,  $X_i \neq Z_i$ .

You do not observe  $Y_i^*$ . Instead, you observe a random sample  $(D_i, X_i, Y_i, Z_i)$  for  $i = 1, \dots, N$ , where

$$Y_i := \begin{cases} Y_i^* & \text{if } D_i = 1 \\ \text{unobserved} & \text{if } D_i = 0, \end{cases}$$

and your goal is to estimate  $\beta$ .

A naïve approach here would be to simply regress  $Y_i$  on  $X_i$ , ignoring possible sample selection. Let  $\hat{\beta}^{\text{OLS}}$  be the estimator in a regression of  $Y_i$  on  $X_i$ . Prove or disprove:  $E(\hat{\beta}^{\text{OLS}} | D_i = 1, X_i, Z_i) = \beta$ .

To obtain useful results, you should impose additional assumptions in your derivation. These additional assumptions should be based on the discussion of the sample selection model in the lecture.

[Question 4 continues on next page]

(b) [ 10 marks ]

Consider the *censored regression model*

$$Y_i^* = X_i' \beta + e_i,$$

where the errors  $e_i$  are iid draws from  $N(0, 1)$ . Alternatively, this model is called the *standard Tobit model* or *Type 1 Tobit model*.

You do not observe  $Y_i^*$ . Instead, you observe a random sample  $(X_i, Y_i)$  for  $i = 1, \dots, N$ , where

$$Y_i := \begin{cases} Y_i^* & \text{if } Y_i^* \geq 0 \\ 0 & \text{if } Y_i^* < 0, \end{cases}$$

and your goal is to estimate  $\beta$  via maximum likelihood.

Write down the log likelihood function  $L(\beta | x_1, \dots, x_N, y_1, \dots, y_N)$ .

End of Exam

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