

THE AUSTRALIAN NATIONAL UNIVERSITY

First Semester Final Examination– June, 2020

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.
- Work on your answers only between 3:00pm–5:00pm!
- Upload your file at 5:00pm sharp!
- 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!
- Make reasonable assumptions where necessary.
- Justify all steps that are not obvious.
- Good luck!

Beginning of Exam Questions

1. [1 mark total] Write the following statement by hand:

I hereby declare

- to uphold the principles of academic integrity, as defined in the University Academic Misconduct Rules;
- that your work in the final exam in no part involves copying, cheating, collusion, fabrication, plagiarism or recycling.

Important

Do NOT work on your exam after 5:00pm!
Scan and upload immediately at 5:00pm!

2. [20 marks total]

Consider the conditional expectation function $E(Y_i|X_i) = g(X_i, \beta)$ where

- X_i is a scalar
 - Y_i is binary: $Y_i \in \{0, 1\}$
- (a) [10 marks] Assume $g(X_i, \beta) = X_i\beta$. Propose an estimator for β that is unbiased, consistent, and efficient. Prove that it is unbiased and consistent. Derive its asymptotic distribution.
- (b) Assume $g(X_i, \beta) = F(X_i\beta)$ for some differentiable function F that is bounded between zero and one.
- i. [2 marks] Show that $Y_i = F(X_i\beta) + e_i$ such that $E(e_i|X_i) = 0$.
 - ii. [2 marks] Conditional on X_i , the error e_i can take on two values

$$e_i = \begin{cases} 1 - F(X_i\beta) & \text{if } Y_i = 1 \\ -F(X_i\beta) & \text{if } Y_i = 0 \end{cases}$$

Show that $\text{Var}(e_i|X_i) = F(X_i\beta) \cdot (1 - F(X_i\beta))$.

(Notice that $\Pr(Y_i = 1|X_i) = E(Y_i|X_i) = F(X_i\beta)$.)

iii. [4 marks] Consider the following least squares estimator:

$$\hat{\beta} := \underset{b}{\operatorname{argmin}} \sum_{i=1}^N \frac{(Y_i - F(X_i b))^2}{\omega_i},$$

where $\omega_i := F(X_i\beta) \cdot (1 - F(X_i\beta))$.

Why are we dividing by ω_i in the objective function? Is $\hat{\beta}$ a feasible estimator? If not, suggest a feasible estimator.

iv. [2 marks] Derive the first order condition for the minimization problem in part iii. Does the first order condition look familiar?

3. [20 marks total - 5 marks each]

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.)

(a) If $E(X^2) < \infty$ and $E(Y^2) < \infty$, then $|E(XY)| < \infty$.

(b) Let Y_1, \dots, Y_N be independent random variables, each with pdf

$$f(y) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}(y - \mu)^2\right).$$

The maximizer of the joint log-likelihood function is equal to \bar{Y} .

(c) In the model $Y_i = X_i\beta + e_i$ where X_i is a scalar and $E(e_iX_i) = 0$ you want to test the null hypothesis $\beta = 0$. You suggest the following hypothesis test: draw a number randomly from $\{1, 2, \dots, 10\}$ and reject the null if that number equals 4.

That hypothesis test has a size of 5%.

(d) Let Y_i be independent random variables (with $i = 1, 2, \dots$), each with pdf

$$f(y) = \begin{cases} \frac{1}{c} & \text{for } y \in (0, c) \\ 0 & \text{elsewhere} \end{cases}$$

for some $0 < c < \infty$. Then Y_1, Y_2, \dots is bounded in probability.

Important

Do NOT work on your exam after 5:00pm!
Scan and upload immediately at 5:00pm!

4. [20 marks total]

Consider the moment conditions $E(\psi(W_i, \theta_0)) = 0$, where W_i represents the variables of your model. The function $\psi(W_i, \cdot)$ is twice continuously differentiable. Let $\dim \psi(W_i, \theta_0) = \dim \theta_0 = K \times 1$.

An obvious analog estimator for θ_0 is $\hat{\theta}$ which solves $(1/N) \sum_{i=1}^N \psi(W_i, \hat{\theta}) = 0$.

You may assume throughout that $\hat{\theta} = \theta_0 + o_p(1)$.

To establish the asymptotic distribution of $\hat{\theta}$ you will use a mean value theorem approximation, like so:

$$0 = (1/N) \sum_{i=1}^N \psi(W_i, \theta_0) + \left(\frac{1}{N} \sum_{i=1}^N \frac{\partial \psi}{\partial \theta'}(W_i, \tilde{\theta}) \right) (\hat{\theta} - \theta_0).$$

(a) [2 marks] State the approximate distribution of $(1/\sqrt{N}) \sum_{i=1}^N \psi(W_i, \theta_0)$? (This should be brief.)

(b) [2 marks] What is the probability limit of $\frac{1}{N} \sum_{i=1}^N \frac{\partial \psi}{\partial \theta'}(W_i, \tilde{\theta})$?

(c) [2 marks] The above mean value approximation can be used to obtain the following asymptotic distribution:

$$\sqrt{N}(\hat{\theta} - \theta_0) \xrightarrow{d} N(0, PQP').$$

Determine P and Q here.

(Note: I am NOT asking you to derive the asymptotic distribution.)

Note: This exercise continues on next page ...

Important

Do NOT work on your exam after 5:00pm!
Scan and upload immediately at 5:00pm!

Now consider the model

$$\begin{aligned} Y_i &= \beta_0 X_i + e_i \\ X_i &= Z_i' \pi_0 + v_i, \end{aligned}$$

where $\dim(X_i) = 1 \times 1$ and $\dim Z_i = L \times 1$ with $L > 1$. Let

$$\begin{pmatrix} e_i \\ v_i \end{pmatrix} \Big| Z_i \sim N \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \sigma_e^2 & \sigma_{ev} \\ \sigma_{ev} & \sigma_v^2 \end{pmatrix} \right).$$

Let

$$\psi(W_i, \beta, \pi) := \begin{pmatrix} \psi_1(W_i, \pi) \\ \psi_2(W_i, \pi, \beta) \end{pmatrix},$$

where

$$\begin{aligned} \psi_1(W_i, \pi) &:= Z_i(X_i - Z_i' \pi) \\ \psi_2(W_i, \beta, \pi) &= (Z_i' \pi) \cdot (Y_i - \beta(Z_i' \pi)). \end{aligned}$$

For brevity we may also write $\psi(\beta, \pi)$ instead of $\psi(W_i, \beta, \pi)$.

(d) [2 marks] Let $u_i := Y_i - \beta_0(Z_i' \pi_0)$. Derive the distribution of $\begin{pmatrix} u_i \\ v_i \end{pmatrix} \Big| Z_i$.

(e) [1 marks] Let $\theta := (\pi', \beta)'$. Derive $\frac{\partial \psi}{\partial \theta'}(\beta_0, \pi_0)$.

(f) [3 marks] Using $\Gamma := E(Z_i Z_i')$, determine $E \left(\frac{\partial \psi}{\partial \theta'}(\beta_0, \pi_0) \right)$.

(g) [2 marks] Determine $\left(E \left(\frac{\partial \psi}{\partial \theta'}(\beta_0, \pi_0) \right) \right)^{-1}$.

Hint: make use of

$$\begin{pmatrix} A & 0 \\ C & D \end{pmatrix}^{-1} = \begin{pmatrix} A^{-1} & 0 \\ -D^{-1}CA^{-1} & D^{-1} \end{pmatrix}.$$

(h) [3 marks] Determine $E \left(\psi(\beta_0, \pi_0) \cdot \psi(\beta_0, \pi_0)' \right)$.

(i) [3 marks] Let $\hat{\beta}$ be the obvious analog estimator of β_0 . What is the asymptotic distribution of $\sqrt{N}(\hat{\beta} - \beta_0)$? Be specific about its asymptotic variance! (Do not provide the asymptotic variance of $\hat{\pi}$.)

End of Exam