

Fall 2005 UCLA Department of Economics  
Written Qualifying Examination in Quantitative Methods

Instructions:

Answer **ALL** questions in Parts I, II, and III

Use a separate answer book for each Part.

You have four hours to complete the exam.

Calculators and other electronic devices are not allowed.

**First Year Comprehensive Exam in Econometrics**  
**Part I**

1. (10 pt.) Consider the linear regression model

$$y = X\beta + \epsilon$$

where  $y$ ,  $X$ ,  $\beta$ ,  $\epsilon$  are  $n \times 1$ ,  $n \times k$ ,  $k \times 1$ ,  $n \times 1$  matrices, and  $\epsilon \sim N(0, \sigma^2 I_n)$ . We will assume that  $X$  is nonstochastic with full column rank. Let

$$\hat{\beta} = (X'X)^{-1} X'y = \beta + (X'X)^{-1} X'\epsilon$$

denote the OLS estimator, and let

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

denote the residual vector. Let  $c$  denote an arbitrary nonstochastic  $k \times 1$  vector.

- (a) (2 pt.) Show that

$$\hat{\beta} - \beta \sim N(0, \sigma^2 (X'X)^{-1})$$

- (b) (3 pt.) Show that  $\hat{\beta}$  and  $e'e$  are independent of each other.

- (c) (2 pt.) Show that

$$\frac{e'e}{\sigma^2} = \left(\frac{\epsilon}{\sigma}\right)' P \left(\frac{\epsilon}{\sigma}\right) \sim \chi^2(n-k)$$

- (d) (3 pt.) Show that

$$c'\hat{\beta} \pm t_{.975}(n-1) \sqrt{s^2 c' (X'X)^{-1} c}$$

where

$$s^2 = \frac{e'e}{n-k}$$

is a valid 95% confidence interval for  $c'\beta$ .

2. (10 pt.) Let  $Y_n$  be a statistic such that  $\lim_{n \rightarrow \infty} E[Y_n] = \theta$  and  $\lim_{n \rightarrow \infty} \text{Var}(Y_n) = 0$ . Prove that  $Y_n$  is a consistent estimator of  $\theta$ . Hint: According to the Markov Inequality, we have

$$\Pr[|X| \geq c] \leq \frac{E[|X|^k]}{c^k}$$

for a random variable  $X$  and a nonstochastic constant  $c$ .

3. (10 pt.) Prove that, if  $X_1, \dots, X_n$  are independent Poisson random variables with means  $\mu_i$ , then  $\sum_{i=1}^n X_i$  is Poisson with mean  $\sum_{i=1}^n \mu_i$ . Hint: The MGF  $M(t)$  of Poisson with mean  $m$  is equal to  $\exp[m(e^t - 1)]$ .
4. (10 pt.) Show that a sequence of random variables  $Y_n$  converges in probability to a (nonstochastic) constant  $c$  if and only if it converges in distribution to a limiting distribution degenerate at  $c$ .

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**PART II**

**Problem 1:** Consider a sample of  $n$  i.i.d. observations on  $(Y_t, X_t)$  from the model

$$Y_t = X_t\beta + \varepsilon_t \quad t = 1, \dots, n,$$

where  $\varepsilon_t$  is independent of  $X_t$  for all  $t$ .

- (a) Provide conditions for the OLS estimator of  $\beta$  to be consistent.
- (b) Derive the asymptotic distribution of the OLS estimator and provide a consistent estimator of its asymptotic covariance matrix.
- (c) Describe how you would test the hypothesis that  $\gamma(\beta) = 0$  for some known  $p$ -dimensional function  $\gamma$  with matrix of first derivatives  $\Gamma$  of full row-rank  $p$ .

Make all necessary moment assumptions to prove your claims. Make sure to prove your claims and state all necessary theorems.

**Problem 2:** Define the tobit model. Provide primitive conditions for consistency and asymptotic normality of the ML estimator and derive its asymptotic distribution.

**First Year Comprehensive Exam in Econometrics**  
**Part III**

**Question 1)** True/Questionable/False? (No points are given for just stating true/questionable or false. The explanation is what counts.)

- (a) In a linear regression model with iid data, HAC estimation is consistent, only if the bandwidth  $S_T$  grows to infinity when  $T \rightarrow \infty$ .
- (b) With an MSE loss function, the optimal linear forecast of  $Y_t$  based on  $(Y_{t-1}, \dots, Y_1)$  does not depend on third and higher moments of the data *only if* the data are Gaussian.
- (c) In a linear regression model with omitted regressors, OLS estimation is inconsistent.
- (d) If the errors in a linear regression have a Cauchy distribution then the OLS estimator is better than the LAD estimator in terms of asymptotic relative efficiency.
- (e) The linear difference equation  $y_t = \alpha y_{t-2} + u_t$  where  $u_t \sim iid N(0, \sigma^2)$  has a causal stationary solution if  $|\alpha| < 1$ . (Note,  $y_t$  is not an AR(1) here.)

**Question 2)**

- (a) Is there a stationary solution  $y_t$  to the scalar difference equation  $y_t = y_{t-1} + \varepsilon_t$  (where  $\varepsilon_t$  is white noise with zero mean and variance  $\sigma^2 > 0$ ; Note: we do not assume  $y_0 \equiv 0$ ). If yes, provide a solution, if no, prove your claim.
- (b) If an AR(1) process is observed every other time period (i.e. instead of  $y_1, y_2, y_3, \dots$  you observe  $y_1, y_3, y_5, \dots$ ), is the observed process still AR(1)? If yes, express the parameters of the new process in terms of those of the original process, if no, prove your claim.

**Question 3)** Consider the linear simultaneous equations model

$$\begin{aligned} Y &= X\beta + \varepsilon, \\ X &= Z\Pi + v, \end{aligned}$$

where  $Y$  is  $n \times 1$ ,  $X$  is  $n \times d$ . The observations are i.i.d. and you can assume conditional homoskedasticity. The sample size  $n$  is larger than  $d$ . Interest focuses on estimation of  $\beta$ .

Consider 2SLS estimators  $\hat{\beta}_i$  ( $i = 1, 2$ ) based on instruments  $Z_i$  of dimensions  $n \times k_i$ , where  $k_1 < k_2$  and  $Z_1$  consists of the first  $k_1$  columns of  $Z_2$ . Verify that the asymptotic variance of  $\hat{\beta}_2$  is smaller than the one of  $\hat{\beta}_1$  (in the positive definite sense).

**Question 4)** You want to test  $H_0 : \theta = 0$  against  $H_1 : \theta > 0$ . The test for  $H_0$  is to reject if  $T_n = n^{1/2}\hat{\theta}/s(\hat{\theta}) > c$ , where  $c$  is picked so that the type I error is  $\alpha$  (the data are iid and as usually  $\hat{\theta}$  is a root- $n$  consistent estimator of  $\theta$  and  $s(\hat{\theta})$  is a consistent estimator of the standard deviation). You do this as follows. Using the non-parametric bootstrap, you generate  $B$  bootstrap samples, calculate the estimates  $\hat{\theta}^*$  on these samples and then calculate  $T_n^* := n^{1/2}\hat{\theta}^*/s(\hat{\theta}^*)$   $B$  times. Let  $q_n^*(1 - \alpha)$  denote the  $100(1 - \alpha)\%$  quantile of the empirical distribution of  $T_n^*$ . You reject  $H_0$  iff  $T_n > q_n^*(1 - \alpha)$ . Discuss the power properties of this test.