

ECONOMETRICS COMPREHENSIVE EXAM  
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YOU MUST ANSWER QUESTION 1.

1. Consider the following model

$$y_i = \beta_1 + \beta_2 x_{i2} + \beta_3 x_{i3} + \varepsilon_i, \quad i = 1, \dots, N,$$

where  $(y_i, x_{i2}, x_{i3})$  are observed and have finite moments, and  $\varepsilon_i$  is an unobserved error term. Suppose this model is estimated by ordinary least squares. Denote the OLS estimator by  $b$ .

- a. What are the *essential* conditions required for unbiasedness of  $b$ ? What are the *essential* conditions required for consistency of  $b$ ? Explain the difference between unbiasedness and consistency.
- b. Show how the conditions for consistency can be written as moment conditions. Explain how a method of moments estimator can be derived from these moment conditions. Is the resulting estimator any different from the OLS one?

Now suppose that  $\text{cov}\{\varepsilon_i, x_{i3}\} \neq 0$ .

- c. Give an example of cases where one can expect a nonzero correlation between a regressor,  $x_{i3}$ , and the error  $\varepsilon_i$ . In this case, is it possible to still make appropriate inferences based on the OLS estimator, while adjusting the standard errors appropriately?
- d. Explain how an instrumental variable,  $z_i$ , say, leads to a new moment condition and, consequently, an alternative estimator for  $\beta$ . Why does this alternative estimator lead to a smaller  $R^2$  than the OLS one? What does this say of the  $R^2$  as a measure for the adequacy of the model?
- e. Why can we not choose  $z_i = x_{i2}$  as an instrument for  $x_{i3}$ , even if  $E\{x_{i2} \varepsilon_i\} = 0$ ? Would it be possible to use  $x_{i2}^2$  as an instrument for  $x_{i3}$ ?

**ANSWER EITHER QUESTION 2 or 3**

2. Let  $y_i$  denote the number of times individual  $i$  buys tobacco in a given month. Suppose a random sample of  $N$  individual is available, for which we observe values  $0, 1, 2, 3, \dots$ . Let  $x_i$  be an observed characteristic of these individuals (e.g., gender). If we assume that, for given  $x_i$ ,  $y_i$  has a Poisson distribution with parameter  $\lambda_i = \exp\{\beta_1 + \beta_2 x_i\}$ , the probability mass function of  $y_i$  conditional upon  $x_i$  is given by

$$P\{y_i = y | x_i\} = \frac{e^{-\lambda_i} \lambda_i^y}{y!}.$$

- a. Write down the log-likelihood function (LLF) for this so-called Poisson regression model.
- b. Derive the score contributions (first derivatives of the LLF with respect to the parameters). Using that the Poisson distribution implies that  $E\{y_i | x_i\} = \lambda_i$ , show that the score contributions have expectation zero.
3. Consider the partitioned regression model  $y = X_1\beta_1 + X_2\beta_2 + u$ .

- a. Prove that the subvector  $\hat{\beta}_1$  of the ordinary least-squares regression estimate  $\hat{\beta} = (X'X)^{-1}X'y$ , where  $X = [X_1 \ X_2]$  is the set of coefficients obtained when the residuals from a regression of  $y$  on  $X_2$  alone are regressed on the set of residuals obtained when each column of  $X_1$  is regressed on  $X_2$ .
- b. Show that the ordinary least-squares estimator of  $\beta$  in the model  $y = \alpha \iota + X\beta + u$  is equal to the ordinary least-squares estimator of  $\beta$  in  $y - \bar{y}\iota = [X - \iota\bar{x}]\beta + \varepsilon$ , where  $\alpha$  is the intercept,  $\iota = [1, \dots, 1]'$ ,  $\bar{y} = n^{-1} \sum_{i=1}^n y_i$ , and  $\bar{x} = [\bar{x}_1, \dots, \bar{x}_k]$  with  $\bar{x}_j = n^{-1} \sum_{i=1}^n x_{ji}$ ;  $j = 1, \dots, k$ .
- c. What is the likely effect upon the ordinary least-squares estimate of  $\beta_1$  in the model  $y = X_1\beta_1 + X_2\beta_2 + u$  of omitting the variables in  $X_2$ ?

**ANSWER EITHER QUESTION 4 or 5.**

4. Let  $Y = X\beta + \varepsilon$ , where  $Y$  is a  $n \times 1$  dependent variable vector,  $X$  is a  $n \times k$  matrix of independent variables,  $\beta$  is a vector of coefficients, and  $\varepsilon$  is a  $n \times 1$  disturbance vector. Assume that errors terms are not homoscedastic.

- i) Derive  $V(b)$  where  $b$  is the least squares estimator of  $\beta$ .
- ii) Write the estimator of  $V(b)$  suggested by H. White.
- iii) Describe the White's test of heteroscedasticity, defining all your notations.

5. Let  $\ln L(x; \theta)$  denote the log likelihood function where  $x$  is a  $n \times 1$  iid random vector and  $\theta$  is a  $1 \times 1$  parameter.

- i) Under regularity conditions, prove  $\text{Var } \partial \ln L(x; \theta) / \partial \theta = - E \partial^2 \ln L(x; \theta) / \partial \theta^2$ .
- ii) Show three ways of estimating  $\text{Var } \partial \ln L(x; \theta) / \partial \theta$ .

**YOU MUST ANSWER QUESTION 6.**

6. To evaluate the benefit from a social program, the following model is employed:

$$\begin{aligned} y_{1i} &= X_i \beta_1 + u_{1i} \text{ (for program participants)} \\ y_{2i} &= X_i \beta_2 + u_{2i} \text{ (for nonparticipants)} \\ I_i &= Z_i \gamma + \varepsilon_i \text{ (participation decision function)} \end{aligned}$$

The observed  $y_i$  is defined as

$$\begin{aligned} y_i &= y_{1i} \text{ iff } I_i > 0 \\ y_i &= y_{2i} \text{ iff } I_i \leq 0 \end{aligned}$$

Assume  $(u_{1i}, u_{2i}, \varepsilon_i)$  is normally distributed with

$$\text{Cov}(u_{1i}, u_{2i}, \varepsilon_i) = \begin{bmatrix} \sigma_{11} & \sigma_{12} & \sigma_{1\varepsilon} \\ \sigma_{12} & \sigma_{22} & \sigma_{2\varepsilon} \\ \sigma_{1\varepsilon} & \sigma_{2\varepsilon} & 1 \end{bmatrix}$$

- i) Find  $E(y_{1i} | I_i > 0)$  and  $E(y_{2i} | I_i > 0)$
- ii) Explain the statement “under self-selection those individuals who have a comparative advantage with the program will be joining the program and thus will benefit from it than would a randomly selected individual with the same characteristics.”
- iii) Prove the statement in ii).