

## Problem Set 2

Due: Monday, April 24, 2006

### Question 1:

This question is based on the multivariate regression model of Lecture note 4. Chamberlain(1984). The model is as follows: for  $i = 1, \dots, N$ ,

$$\begin{aligned} y_{i1} &= \pi_1' x_i + u_{i1}, \\ &\vdots \\ y_{iT} &= \pi_T' x_i + u_{iT}, \end{aligned}$$

where  $E(u_{it} | x_i) = 0$ ,  $\text{Cov}(u_{it}, u_{js}) = \sigma_{st}$  if  $i = j$  and  $\text{Cov}(u_{it}, u_{js}) = 0$  if  $i \neq j$ , and

$$X = \begin{pmatrix} x_1' \\ \vdots \\ x_N' \end{pmatrix} = \begin{pmatrix} x_{11} & \cdots & x_{1K} \\ \vdots & \ddots & \vdots \\ x_{N1} & \cdots & x_{NK} \end{pmatrix}.$$

a. Assume that the relevant matrices are conformable and show the following Kronecker product properties:

1.  $(A \otimes B)(C \otimes D) = AC \otimes BD$
2.  $(A \otimes B)' = A' \otimes B'$
3. If  $A$  and  $B$  are non-singular, then  $(A \otimes B)^{-1} = A^{-1} \otimes B^{-1}$ .

b. Using the Kronecker product notation, we can write the multivariate regression model as:

$$y = (I \otimes X)\pi + u, \quad E(u | X) = 0, \quad \text{Var}(u | X) = \Sigma \otimes I_N,$$

where  $\Sigma$  is a  $T \times T$  matrix with the  $s, t$  elements  $\sigma_{st}$ .

1. Show that the GLS estimator of  $\pi$  is given by

$$\hat{\pi} = \left( I \otimes (X'X)^{-1} X' \right) y.$$

2. Show that

$$\text{Var}(\hat{\pi} | X) = \Sigma \otimes (X'X)^{-1}.$$

3. Suppose that there are linear restrictions on  $\pi$  of the form  $\pi = G\theta$ , where  $\theta$  is an unrestricted parameter vector and  $G$  is a known matrix. Show that the GLS estimator for  $\theta$  can be written as

$$\hat{\theta} = (G' A^{-1} G)^{-1} G' A^{-1} \hat{\pi},$$

where  $A = \text{Var}(\hat{\pi} | X)$ .

**Question 2 (question 6 in Greene, page 376):**

Consider the model given by

$$\begin{aligned}y_{1i} &= \alpha_1 + \beta x_i + \varepsilon_{1i}, \\y_{2i} &= \alpha_2 + \varepsilon_{2i},\end{aligned}$$

for  $i = 1, \dots, n$ . Assume that

$$\varepsilon_i = (\varepsilon_{1i}, \varepsilon_{2i})' \sim (0, \Sigma) \quad \text{and} \quad \Sigma = \begin{pmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{pmatrix}.$$

Prove that the GLS estimate applied to the system leads to the OLS estimates for  $\alpha_1$  and  $\alpha_2$ , but to a mixture of two least-squares slopes from the regressions of  $y_1$  on  $x$ , and  $y_2$  on  $x$ . For simplicity of the algebra, assume that the mean of the  $x_i$ 's is zero, i.e.,  $\bar{x} = 0$ .

**Question 3 (question 1 in Greene, page 422):**

Consider the two-equation model:

$$\begin{aligned}y_{1i} &= \gamma_1 y_{2i} + \beta_{11} x_{1i} + \beta_{21} x_{2i} + \beta_{31} x_{3i} + \varepsilon_{1i}, \\y_{2i} &= \gamma_2 y_{1i} + \beta_{12} x_{1i} + \beta_{22} x_{2i} + \beta_{32} x_{3i} + \varepsilon_{2i},\end{aligned}$$

where

$$\varepsilon_i = (\varepsilon_{1i}, \varepsilon_{2i})' \sim (0, \Sigma) \quad \text{and} \quad \Sigma = \begin{pmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{pmatrix}.$$

- a. Verify that neither equations is identified.
- b. Establish whether or not the stated restrictions are sufficient for identification (or partial identification) of the model and briefly justify your answers:

1.  $\beta_{21} = \beta_{32} = 0$ .
2.  $\beta_{12} = \beta_{22} = 0$ .
3.  $\gamma_1 = 0$ .
4.  $\gamma_1 = \gamma_2$  and  $\beta_{32} = 0$ .
5.  $\sigma_{12} = 0$  and  $\beta_{31} = 0$ .
6.  $\gamma_1 = 0$  and  $\sigma_{12} = 0$ .
7.  $\beta_{21} + \beta_{22} = 1$ .
8.  $\sigma_{12} = 0$  and  $\beta_{21} = \beta_{22} = \beta_{31} = \beta_{32} = 0$ .
9.  $\sigma_{12} = 0$  and  $\beta_{11} = \beta_{21} = \beta_{22} = \beta_{31} = \beta_{32} = 0$ .

**Question 4 (question 3 in Greene, page 423):**

Check the identifiability of the parameters of the following model:

$$\begin{aligned} [y_1, y_2, y_3, y_4] & \begin{bmatrix} 1 & \gamma_{12} & 0 & 0 \\ \gamma_{21} & 1 & \gamma_{23} & \gamma_{24} \\ 0 & \gamma_{32} & 1 & \gamma_{34} \\ \gamma_{41} & \gamma_{42} & 0 & 1 \end{bmatrix} = \\ [x_1, x_2, x_3, x_4, x_5] & \begin{bmatrix} 0 & \beta_{12} & \beta_{13} & \beta_{14} \\ \beta_{21} & 1 & 0 & \beta_{24} \\ \beta_{31} & \beta_{32} & \beta_{33} & 0 \\ 0 & 0 & \beta_{43} & \beta_{44} \\ 0 & \beta_{52} & 0 & 0 \end{bmatrix} + [\varepsilon_1, \varepsilon_2, \varepsilon_3, \varepsilon_4]. \end{aligned}$$