

**Swiss Program for Beginning Doctoral Students in Economics 2003**

**Final Exam in Econometrics**

**Monday, February 23, 2004, 14.00h - 17.00h**

1. You are allowed to use all material that you want (lecture notes, books, etc.) with the exception of PC's.
2. Please **do not** mention your name on top of the pages, but use your identification number from the enclosed list. The reason is that the exams will be graded anonymously.
3. Please use **a pen** rather than a pencil so that your answers can be read without problems.
4. Please **write legibly**. Remember that your exams will be photocopied for grading.
5. Answers should be **concise and precise!** The space provided should be sufficient to answer each question.
6. Good luck!

ID-Number: \_\_\_\_\_

**Problem 1. (20 points)**

Suppose that  $y_t$  is a series that is available semiannually (that is, twice per year), once in the winter and once in the summer. Suppose that the semiannual seasonal process for the series is

$$y_t = 0.9y_{t-1} + \varepsilon_t$$

where  $\varepsilon_t$  is iid(0, 1).

**(a) (10 points)** Derive and plot the spectrum of  $y$ . Discuss how the seasonality in the process is evident in spectrum.

**(b) (10 points)** A researcher proposes to use  $x_t = 0.5(1+L)y_t$ , as a “seasonally adjusted” version of  $y$ . Compute the gain of the filter  $0.5(1+L)$ . Does this filter eliminate/and or attenuate the seasonality in  $y$ ? Explain.

**Problem 2. (15 points)**

Suppose

$$\begin{aligned}y_t &= \xi_t + w_t \\ \xi_t &= 0.8\xi_{t-1} + v_t\end{aligned}$$

where  $w_t$  is iid  $N(0,2)$ ,  $v_t$  is iid  $N(0,3)$  and  $\{w_t\}$  and  $\{v_t\}$  are independent. Suppose that  $\xi_{t-1} = 3.4$  and  $y_t = 4.1$ . Find  $E(\xi_t | y_t = 4.1, \xi_{t-1} = 3.4)$  and  $\text{var}(\xi_t | y_t = 4.1, \xi_{t-1} = 3.4)$ .

**Problem 3. (15 points)**

Suppose that  $y_t = \mu_1 + \varepsilon_t$ ,  $\varepsilon_t$  iid( $0, \sigma_1^2$ ) for  $t = 1, \dots, \tau$ , that  $y_t = \mu_2 + \varepsilon_t$ ,  $\varepsilon_t$  iid( $0, \sigma_2^2$ ) for  $t = \tau + 1, \dots, T$ , and  $E(\varepsilon_t^4) < \infty$  for all  $t$ .

**(a) (7 points)** Suppose that you knew  $\tau$ , and that both  $\tau$  and  $T - \tau$  are large. How would you test  $H_0: \mu_1 = \mu_2$  versus  $H_a: \mu_1 \neq \mu_2$ , using a test that allowed  $\sigma_1^2 \neq \sigma_2^2$ ?

**(b) (8 points)** Suppose that you knew  $\tau$ , and that both  $\tau$  and  $T-\tau$  are large. How would you test  $H_0: \sigma_1^2 = \sigma_2^2$  versus  $H_a: \sigma_1^2 \neq \sigma_2^2$  using a test that allowed  $\mu_1 \neq \mu_2$ ?

**Problem 4. (20 points)**

Suppose that  $y$  and  $x$  follow the process

$$y_t = \alpha x_{t-1} + \varepsilon_t$$

$$x_t = \phi x_{t-1} + v_t$$

where  $v_t = \varepsilon_t + e_t$  where  $\varepsilon_t$  and  $e_t$  are mutually independent iid(0,1) processes. Suppose that  $\phi < 1$ . Let  $\hat{\alpha}$  and  $\hat{\phi}$  denote the OLS estimators of  $\alpha$  and  $\phi$ .

**(a) (12 points)** Derive the asymptotic distribution of the  $2 \times 1$  vector  $(\hat{\alpha}, \hat{\phi})$ .

**(b) (8 points)** In a sample of 100 observations,  $\hat{\alpha} = 1.3$  and  $\hat{\phi} = 0.72$ . Derive a 95% confidence interval for  $\alpha$ .

**Problem 5. (20 points)**

Consider the same model as in question (2), but now assume that  $\phi = 1$ .

**(a) (10 points)** Show that  $x$  and  $y$  are cointegrated. What is the cointegrating vector?

**(b) (10 points)** Derive the asymptotic distribution of  $\hat{\alpha}$ .

**Problem 6. (20 points)**

This problem is concerned with bounding treatment effects. Suppose that  $D$  is a random variable that indicates whether an individual has been “treated”. If  $D = 1$ , the individual has received treatment and we observe a random variable  $Y_1$ . If  $D = 0$ , the individual has not received treatment and we observe a random variable  $Y_0$ . We do not observe  $Y_0$  if  $D = 1$  and we do not observe  $Y_1$  if  $D = 0$ . This is the standard notation in this literature.

The parameter of interest is the average treatment effect on the treated,  $E[Y_1 - Y_0 | D = 1]$ .

- (a) (10 points) Suppose that it is known that  $0 \leq Y_0 \leq 100$  and  $0 \leq Y_1 \leq 100$ . Use this to construct bounds on  $E[Y_1 - Y_0 | D = 1]$ .

- (b) (10 points) Suppose that it is known that  $0 \leq Y_0 \leq Y_1 \leq 100$ . Use this to construct bounds on  $E[Y_1 - Y_0 | D = 1]$ .

**Problem 7. (30 points)**

Suppose that you have a random sample of  $(y_i, x_i)$  of size  $n$  from

$$y_i = \log \left( (x_i' \beta)^2 + 1 \right) + \varepsilon_i$$

where  $E[\varepsilon_i | x_i] = 0$ . Assume that all relevant moments exist.

- (a) (10 points) Find the asymptotic distribution of the nonlinear least squares estimator that minimizes

$$\sum_{i=1}^n \left( y_i - \log \left( (x_i' b)^2 + 1 \right) \right)^2$$

(b) (10 points) How would you estimate the asymptotic variance of the nonlinear least squares estimator?

- (c) (10 points) Now assume that you know that  $V[\varepsilon_i | x_i] = \exp(x_i' \alpha)$  and that you estimate  $\alpha$  by minimizing

$$\sum_{i=1}^n (e_i - \exp(x_i' a))^2$$

where  $e_i = y_i - \log\left(\left(x_i' \widehat{\beta}_{NLS}\right)^2 + 1\right)$ . Briefly discuss how you would find the asymptotic distribution of this estimator.

**Problem 8. (20 points)**

Consider observations of  $(y_{it}, x_{it})$  from the linear panel data model

$$y_{i,t} = x_{it}'\beta + \gamma y_{i,t-1} + \alpha_i + \varepsilon_{it}, \quad t = 1, \dots, T, \quad i = 1, \dots, N$$

where  $\alpha_i$  is an unobserved individual-specific effect. No assumption is made on the relationship between  $\alpha_i$  and  $x_{it}$ . (Note that you do not observe  $y_{i,0}$ ).

(a) (10 points) Suppose that the dimensionality of  $x_{it}$  is 1 and

$$E[\varepsilon_{is}x_{it}] = 0 \quad \text{for all } t \leq s \quad (1)$$

What is the minimum  $T$  such that  $\beta$  are identified? Is the model over-identified for that  $T$ ? Explain and explicitly state any additional “regularity conditions” that you assume.

(b) (10 points) Suppose that the dimensionality of  $x_{it}$  is 2 and

$$E[\varepsilon_{is}x_{it}] = 0 \quad \text{for all } t \leq s + 1 \quad (2)$$

(loosely speaking, we allow feedback from today's  $\varepsilon$  to future values of  $x$ , but we are willing to assume that it takes two periods for this to happen). What is the minimum  $T$  such that  $\beta$  are identified? Is the model over-identified for that  $T$ ? Explain and explicitly state any additional "regularity conditions" that you assume.

**Problem 9. (20 points)**

You have a sample of  $n$  independent observations. For each observation, you have a discrete variable,  $y$ , which takes values 0 and 1. You also have a regressor,  $x$ , and you estimate a logit model in order to characterize the relationship between  $y$  and  $x$  (you use using  $x$  and a constant as explanatory variables). Let  $\alpha$  be the constant and let  $\beta$  be the coefficient on  $x$ .

Suppose you estimate  $\alpha$  to be 1 and  $\beta$  to be  $-1$ . Suppose further that you have estimated the covariance matrix of  $\hat{\alpha}$  and  $\hat{\beta}$  to be  $\begin{pmatrix} 0.04 & 0.02 \\ 0.02 & 0.04 \end{pmatrix}$

(a) (4 points) What would be your estimate of  $P(y = 1|x = 1)$ ?

(b) (6 points) Test whether  $P(y = 1|x = 1) = 0.45$  at a 5% level of significance.

- (c) (10 points) Explain how you would construct a 95% confidence interval for the marginal effect  $\frac{dP(y=1|x=a)}{da}$ . (The confidence interval will depend on  $a$ . You are not expected to calculate it numerically).