

PS #7 (Due 6 March, 2019)

1. Consider the following regression model with a measurement error problem:

$$y_i = 1 + x_i^* + z_i + u_i$$

where

$$\begin{pmatrix} x_i^* \\ z_i \\ u_i \end{pmatrix} \sim N \left( \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 & \rho & 0 \\ \rho & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \right)$$

and  $x_i = x_i^* + v_i$  where  $v_i \sim N(0, \sigma_v^2)$  where  $\sigma_v^2 = 1$  and  $\rho = 0$ .

- Generate 1000 i.i.d. data from the model above and regress  $y_i$  on the constant term,  $x_i$  and  $z_i$  to verify that there is the attenuation bias using R and Python. State a way you can simulate data to verify the direction of inconsistency on the coefficient on  $x_i$  proved in class.
  - Is the coefficient on  $z_i$  consistently estimated? Explain your answer.
  - Theoretically, what will happen to the size of inconsistency on the OLS estimator of the coefficients on  $x_i$  and  $z_i$ , if  $\sigma_v^2$  increases? Verify this by simulating the data with  $\sigma_v^2 = 2$ .
  - What will happen to the size of inconsistency if  $\rho$  increases? Verify this by simulating the data with  $\rho = 0.5$ .
2. Consider the following simultaneous equations model

$$\begin{aligned} y_i^S &= \beta_1^S + \beta_2^S p_i + u_i^S \\ y_i^D &= \beta_1^D + \beta_2^D p_i + u_i^D \end{aligned}$$

where

$$\begin{pmatrix} u_i^S \\ u_i^D \end{pmatrix} \sim N \left( \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \sigma_S^2 & 0 \\ 0 & \sigma_D^2 \end{pmatrix} \right)$$

and  $p_i$  is determined via  $y_i^S = y_i^D$ . Let  $y_i = y_i^S = y_i^D$ .

- What is the equilibrium  $p_i$  in terms of  $u_i^S$  and  $u_i^D$ ? State clearly the conditions under which the solution exists.
  - Obtain the probability limit of the OLS estimator when observed  $y_i$  is regressed on the constant term and  $p_i$  under the standard assumptions.
  - Discuss conditions under which the probability limit of the OLS estimator of the coefficient on  $p_i$  equals  $\beta_2^S$  or  $\beta_2^D$ .
3. Consider the following linear regression model:

$$y_i = 1 + x_i + z_i + u_i$$

where

$$\begin{pmatrix} x_i \\ z_i \\ u_i \end{pmatrix} \sim N \left( \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 & \rho & 0 \\ \rho & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \right)$$

where  $\rho = 0.1$ . Assume that the sampling is i.i.d.

- Explain why this model satisfies all the assumptions we made on the linear regression model including homoskedasticity.

- (b) Using R and Python, generate 1000 i.i.d. data from the model above and select observations that satisfy  $x_i > 0$ . Using the selected observations, regress  $y_i$  on the constant term,  $x_i$  and  $z_i$  to verify, by simulation, that this type of sample selection still yields a consistent estimator.
  - (c) Explain why (b) is the case under correct model specification. Explain why this result is misleading when the model is misspecified.
  - (d) Using R and Python, generate 1000 i.i.d. data from the model above and select observations that satisfy  $y_i > \epsilon_i$ , where  $\epsilon_i$  is the standard normal random variable. Using the selected observations, regress  $y_i$  on the constant term,  $x_i$  and  $z_i$  to verify, by simulation, that this type of sample selection yields an inconsistent estimator.
  - (e) Explain why (d) is the case.
4. Consider the following linear regression model:
- $$\begin{aligned} y_t &= 1 + 0.3y_{t-1} + u_t \\ u_t &= 0.5u_{t-1} + v_t \end{aligned}$$
- where  $v_t$  is i.i.d and  $y_0, u_0$  and  $v_t$  for any  $t$  are mutually independent. Assume that  $y_0, u_0$ , and  $v_t$  all have the standard normal distribution.
- (a) Using R and Python, generate 1000 observations from the model above. Using the generated observations, regress  $y_t$  on the constant term and  $y_{t-1}$  to verify, by simulation, that the OLS estimator is inconsistent under this model.
  - (b) Explain why (a) is the case.
  - (c) Show by simulation that if there is no serial correlation, then the OLS estimator is consistent. Explain why.
5. Consider the same model as in problem 3 without the sample selection problem.
- (a) Show by simulation, using R and Python, that if we use the OLS estimator to estimate the coefficient on  $x_i$  by regressing  $y_i$  on a constant term and  $x_i$ , the estimator is inconsistent.
  - (b) Explain the direction and the size of the inconsistency by using auxiliary regression analysis.
  - (c) Do the same by setting  $\rho = 0$ . Explain the result in this case.
6. Suppose you have two omitted variables and two included variables in addition to the constant term. Extend the auxiliary regression analysis to discuss the directions of inconsistencies.