

ECON 7204: Problem set 3 (due date: 19 May 2022). Return your main file and the code file in the appropriate dropbox on MyUni.

Question 1 (IV estimation– 60 marks).

Consider the following linear IV regression model

$$y_t = \beta_0 + x_t \beta + u_t, t = 1, 2, \dots, T \quad (1)$$

$$x_t = \pi_0 + z_{1t} \pi_1 + z_{2t} \pi_2 + v_t, t = 1, \dots, T, \quad (2)$$

where $(z_{1t}, z_{2t}, u_t, v_t)' \stackrel{i.i.d.}{\sim} N(0, \Omega)$, and $\Omega = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0.8 \\ 0 & 0 & 0.8 & 1 \end{bmatrix}$ for all t . The true values of β_0 , β and π_0 are

set at $\beta_0 = \beta = \pi_0 = 1$. Let $x_1 = \text{ones}(T, 1)$ (a T -dimensional column vector of ones) and define $Z = [x_1 \ z_1 \ z_2]$ (a $T \times 3$ matrix of instruments). The OLS and 2SLS estimators of β are then given by $\hat{\beta} = (x' M_{x_1} x)^{-1} x' M_{x_1} y$, and $\tilde{\beta} = (x' (M_{x_1} - M_Z) x)^{-1} x' (M_{x_1} - M_Z) y$, where $M_{x_1} = I - x_1 (x_1' x_1)^{-1} x_1'$ and $M_Z = I - Z (Z' Z)^{-1} Z'$.

1. Is x_t exogenous in (1)-(2)? Why? ?
2. Are z_{1t} and z_{2t} valid IVs in (1)-(2)? Why? z_{1t} & z_{2t}
3. Suppose first that $\pi_1 = 5$, $\pi_2 = 10$ (strong IVs). For sample sizes $T = 50$ and $T = 500$:

- (a) Generate $r = 1, 2, \dots, R = 1000$ pseudo samples of size T for y and x using the above information. For each pseudo sample, compute the OLS and 2SLS estimators, $\hat{\beta}$ and $\tilde{\beta}$, of β ; their corresponding t -statistics, $t_{\hat{\beta}}$ and $\tilde{t}_{\tilde{\beta}}$, for $H_0 : \beta = 1$, as well as the Sargan's overidentifying restrictions statistic J_S . You will form a collection of $R = 1000$ estimators; $\hat{\beta}^{(1)}, \hat{\beta}^{(2)}, \dots, \hat{\beta}^{(R)}$ and $\tilde{\beta}^{(1)}, \tilde{\beta}^{(2)}, \dots, \tilde{\beta}^{(R)}$, $R = 1000$ t -statistics; $t_{\hat{\beta}}^{(1)}, t_{\hat{\beta}}^{(2)}, \dots, t_{\hat{\beta}}^{(R)}$ and $\tilde{t}_{\tilde{\beta}}^{(1)}, \tilde{t}_{\tilde{\beta}}^{(2)}, \dots, \tilde{t}_{\tilde{\beta}}^{(R)}$, and $R = 1000$ J_S -statistics; $J_S^{(1)}, J_S^{(2)}, \dots, J_S^{(R)}$.

Hint: For the J_S statistic, define $x_1 = \text{ones}(T, 1)$ (a T -dimensional column vector of ones) and $Z = [x_1 \ z_1 \ z_2] : T \times 3$. Then, J_S has the form $J_S = T g_T' W_T g_T$, where $W_T = \left(\frac{z' M_{x_1} z}{T} \right)^{-1}$, $g_T = \frac{1}{T} z' M_{x_1} (y - x \tilde{\beta})$, and $M_{x_1} = I - x_1 (x_1' x_1)^{-1} x_1'$; so as $J_S \xrightarrow{d} \chi^2(1)$ when the exclusion restrictions hold.

- (b) Plot the histograms of the $R = 1000$ estimates of: $\hat{\beta}$ and $\tilde{\beta}$, $t_{\hat{\beta}}$ and $\tilde{t}_{\tilde{\beta}}$, and J_S . Comment on the results.
- (c) Compute the estimated bias as $\widehat{Bias}(\beta) = \frac{1}{R} \sum_{r=1}^R (\hat{\beta}^{(r)} - \beta) = \frac{1}{R} \sum_{r=1}^R \hat{\beta}^{(r)} - \beta$ and $\widetilde{Bias}(\beta) = \frac{1}{R} \sum_{r=1}^R (\tilde{\beta}^{(r)} - \beta) = \frac{1}{R} \sum_{r=1}^R \tilde{\beta}^{(r)} - \beta$. Comment on the results.

- (d) Compute the rejection frequencies of the t -test and J_S -test at 5% level, i.e., the quantity $Re_{OLS} = \frac{1}{R} \sum_{r=1}^R \mathbb{1}[|t_{\beta}^{(r)}| > 1.96]$, $Re_{2SLS} = \frac{1}{R} \sum_{r=1}^R \mathbb{1}[|\tilde{t}_{\beta}^{(r)}| > 1.96]$, $Re_{J_S} = \frac{1}{R} \sum_{r=1}^R \mathbb{1}[J_S^{(r)} > 3.84]$, where $\mathbb{1}[C] = 1$ if condition C holds and $\mathbb{1}[C] = 0$ otherwise. Comment on the results.
4. Suppose that $\pi_1 = 0.01$ and $\pi_2 = 10$ (i.e., z_1 is weak but z_2 is strong). For sample sizes $T = 50$ and $T = 500$, repeat (a)-(d) of question 3. Conclude.
5. Suppose that $\pi_1 = \pi_2 = 0.01$ (both z_1 and z_2 are weak). For sample sizes $T = 50$ and $T = 500$, repeat (a)-(d) of question 3. Conclude.
6. Suppose now that $\pi_1 = \pi_2 = 0$ (both z_1 and z_2 are irrelevant). For sample sizes $T = 50$ and $T = 500$, repeat (a)-(d) of question 3. Conclude.

QUESTION 2 (GMM estimation– 40 marks).

Consider the following population moment condition resulting from the consumption based asset pricing model:

$$g(x_{1t}, x_{2t}, z_t, \theta) = f_t(\theta) z_t$$

$$f_t(\theta) = \delta x_{1t}^{\gamma-1} x_{2t} - 1$$

$$\mathbb{E}[g(x_{1t}, x_{2t}, z_t, \theta)] = 0, \quad t = 1, \dots, T, \quad (3)$$

where $\theta = (\gamma, \delta)'$, $g(x_{1t}, x_{2t}, z_t, \theta) = f_t(\theta) \times z_t$, $f_t(\theta) = \delta x_{1t}^{\gamma-1} x_{2t} - 1$, $x_{1t} = c_{t+1}/c_t$, $x_{2t} = r_{t+1}/p_t$, $z_t = (1, x_{1,t-1}, x_{1,t-2}, x_{2,t-1}, x_{2,t-2})'$, c_t is aggregate per capita consumption, and r_{t+1} , p_t are respectively the gain in period $t+1$ and the price in period t on a NYSE index. The data is monthly and the sample period is 1960.1-1997.12. You are tasked to estimate $\theta = (\gamma, \delta)'$ using alternative GMM methods. The files cbapmewrdata.dat and cbapmewrinstr.dat contain data on (x_{1t}, x_{2t}) and z_t , respectively.

1. Efficient Two-Step GMM Estimation

- Write a code that computes the efficient two-step GMM estimates of γ and δ , as well as their resulting standard error estimates.
- Using the above data sets and your code, find the efficient two-step GMM estimates of γ and δ , and their standard error estimates. Comment on the results.
- Are the instruments in z satisfy the exclusion restrictions? Justify.

2. Iterative GMM Estimation

- Write a code that computes the iterative GMM estimates of γ and δ , as well as their resulting standard error estimates.
- Using the above data sets and your code, find the efficient iterative GMM estimates of γ and δ , and their standard error estimates (NB: initialize θ at the efficient two-step GMM estimate of the previous question). Compare your results with the efficient two-step GMM ones.

$$f_t(\theta) = \delta \left(\frac{c_{t+1}}{c_t} \right)^{\gamma-1} \left(\frac{r_{t+1}}{p_t} \right) - 1$$

(c) Are the instruments in z satisfy the exclusion restrictions? Why?

3. Conclude.

$$g(x_{1t}, x_{2t}, z_t, \theta) = \begin{pmatrix} \delta x_{1t} & r^{-1} x_{2t} & -1 \end{pmatrix} \times z_t$$

$$\theta = \begin{bmatrix} \delta \\ \delta \end{bmatrix}$$

$$x_{1t} = \frac{c_{t+1}}{c_t}$$

$$x_{2t} = \frac{r_{t+1}}{p_t}$$

$$z_t = \begin{pmatrix} 1, x_{1,t-1}, x_{1,t-2}, x_{2,t-1}, x_{2,t-2} \end{pmatrix}'$$