

d.

因 x_1, x_2 為外生且和 V_2 不相關, 故可得一致性估計。

$$\text{即 } E(x_j, V_2) = 0$$

e. 對 $y_2 = \pi_1 x_1 + \pi_2 x_2 + V_2$ 最小平方誤差

$$S = \sum (y_2 - \pi_1 x_1 - \pi_2 x_2)^2$$

對 π_1 及 π_2 偏微分即可得到 (d) 之條件

\Rightarrow MOM 估計值 = OLS

$$f. \hat{\pi}_1 = \frac{\sum x_{1i} y_{2i}}{\sum x_{1i}^2} = 3, \quad \hat{\pi}_2 = \frac{\sum x_{2i} y_{2i}}{\sum x_{2i}^2} = \frac{4}{1} = 4.$$

g. 只要 IV^① 和內生變數 y_2 相關^② 和結構誤差 e_1 不相關即可達成一致性。

$$\text{plim } \hat{\alpha}_1 = \frac{E(\hat{y}_2 y_1)}{E(\hat{y}_2 y_2)} = \frac{E[y_2(\alpha_1 y_2 + e_1)]}{E(\hat{y}_2 y_2)} = \alpha_1 + \frac{E(y_2 e_1)}{E(\hat{y}_2 y_2)} = \alpha_1$$

$\hat{\alpha}_1 \xrightarrow{P} \alpha_1$ 符合一致性。

$$3 \times 2 + 4 \times 3 = 18.$$

$$3 \times 3 + 4 \times 4 = 25$$

$$\Rightarrow \hat{\alpha}_1 = \frac{18}{25} = 0.72.$$

$$h. \sum y_{2i}^2 = 3 \sum x_{1i}^2 + 4 \sum x_{2i}^2 + 2 \times 3 \times 4 \sum x_{1i} x_{2i} = 9 + 16 + 0 = 25.$$

$$\alpha_{1, 2SLS} = \frac{\sum \hat{y}_{2i} y_{1i}}{\sum \hat{y}_{2i}^2} = \frac{18}{25} = 0.72.$$

和 (g) 結果相同

在單一內生變數且使用同一組 IV 時

IV 之結果 = 2SLS 結果。

$$11.1 \quad \begin{cases} y_1 = \alpha_1 y_2 + e_1 \\ y_2 = \alpha_2 y_1 + \beta_1 x_1 + \beta_2 x_2 + e_2 \end{cases}$$

$$(a) \quad y_2 = \alpha_2 (\alpha_1 y_2 + e_1) + \beta_1 x_1 + \beta_2 x_2 + e_2$$

$$\Rightarrow y_2 (1 - \alpha_1 \alpha_2) = \beta_1 x_1 + \beta_2 x_2 + (\alpha_2 e_1 + e_2)$$

$$y_2 = \left(\frac{\beta_1}{1 - \alpha_1 \alpha_2} \right) x_1 + \left(\frac{\beta_2}{1 - \alpha_1 \alpha_2} \right) x_2 + \left(\frac{\alpha_2 e_1 + e_2}{1 - \alpha_1 \alpha_2} \right) \Rightarrow \pi_1 x_1 + \pi_2 x_2 + v_2$$

$$\text{cov}(y_2, e_1 | x) = E(y_2, e_1 | x)$$

$$= E \left[\left(\frac{\beta_1}{1 - \alpha_1 \alpha_2} x_1 + \frac{\beta_2}{1 - \alpha_1 \alpha_2} x_2 + \frac{\alpha_2 e_1 + e_2}{1 - \alpha_1 \alpha_2} \right) e_1 | x \right]$$

$$= E \left[\left(\frac{\beta_1}{1 - \alpha_1 \alpha_2} x_1 e_1 \right) | x \right] + E \left[\left(\frac{\beta_2}{1 - \alpha_1 \alpha_2} x_2 e_1 \right) | x \right] + E \left[\left(\frac{\alpha_2 e_1 + e_2}{1 - \alpha_1 \alpha_2} e_1 \right) | x \right]$$

$$= 0 + 0 + E \left[\left(\frac{\alpha_2 e_1 + e_2}{1 - \alpha_1 \alpha_2} e_1 \right) | x \right]$$

$$\rightarrow \text{cov}(y_2, e_1 | x) = E(y_2, e_1 | x) = \frac{E(e_1 e_2 | x) + \alpha_2 E(e_1^2 | x)}{1 - \alpha_1 \alpha_2} = \frac{\alpha_2 \cdot 6_1^2}{1 - \alpha_1 \alpha_2}$$

$\text{cov}(y_2, e_1 | x)$ 不為 0, 除非 $\alpha_2 = 0$.

(b) 因為等式一和二有 endogenous variable

so the OLS is biased and inconsistent.

而 reduce form 具有一致性

(c) 有 2 條結構方程, 至少要省略 $2-1=1$ 個外生變數.

equation (1): 無外生變數, 省略了 2 個外生變數 \Rightarrow identified

equation (2): x_1, x_2 省略了 0 個外生變數 \Rightarrow unidentified.

11.16.

a.

$$\alpha_1 + \alpha_2 P = \beta_1 + \beta_2 P + \beta_3 W$$

$$\Rightarrow P = \pi_1 + \pi_2 W + V_1$$

$$\pi_1 = \frac{\beta_1 - \alpha_1}{\alpha_2 - \beta_2}, \pi_2 = \frac{\beta_3}{\alpha_2 - \beta_2}, V_1 = \frac{e_3 - e_d}{\alpha_2 - \beta_2}$$

代回需求式

$$Q = \alpha_1 + \alpha_2 (\pi_1 + \pi_2 W + V_1) + e_d = \theta_1 + \theta_2 W + V_2$$

$$\therefore \theta_1 = \alpha_1 + \alpha_2 \pi_1, \theta_2 = \alpha_2 \pi_2, V_2 = \alpha_2 V_1 + e_d$$

b. $\pi_2 = \frac{\beta_3}{\alpha_2 - \beta_2}, \theta_2 = \alpha_2 \pi_2 \Rightarrow \alpha_2 = \frac{\theta_2}{\pi_2}$ identified

$\theta_1 = \alpha_1 + \alpha_2 \pi_1 \Rightarrow \alpha_1$ 亦 identified

但 $\beta_1, \beta_2, \beta_3$ 僅以 π_1, π_2, α 圈定一條關係，無法唯一

\Rightarrow unidentified

c. $\hat{\alpha}_2 = \frac{\hat{\theta}_2}{\hat{\pi}_2} = \frac{0.5}{1} = 0.5$

$$\hat{\alpha}_1 = \hat{\theta}_1 - \hat{\alpha}_2 \hat{\pi}_1 = 5 - 0.5 \times 2.4 = 3.8$$

d. 第一階段: $\hat{P}_1 = 2.4 + W_1$

第二階段: $\hat{Q} = \hat{\alpha}_1 + \hat{\alpha}_2 \hat{P}$

$$\begin{pmatrix} \hat{\alpha}_1 \\ \hat{\alpha}_2 \end{pmatrix} = (X'X)^{-1} X'Q \Rightarrow \begin{matrix} \hat{\alpha}_1 = 3.8 \\ \hat{\alpha}_2 = 0.5 \end{matrix}$$

和(c)之結果相同，驗證 identified 良好且僅一內生變數時，

ILS 結果 = 2SLS 結果

a. 系統共有 M 條, 必要條件: 每條方程式至少要省略 $M-1$ 個外生變數, Klein I 皆滿足, 至少 weak identification.

b. 每條方程式被排除的外生變數數目 $>$ 該方程式右側內生變數數目 $>$ Klein I 方程式皆可估.

c. 令所有外生變數為 Z_t

$$w_t^p = \pi_0 + \pi_1 z_{1t} + \pi_2 z_{2t} + \dots + u_t$$

π 係數可用 OLS 拿到 fitted value \hat{w}_t^p , 供第二階段使用.

d. 第一步: 對所有右側外生變數 (ex. 所得, 工資) 以全部外生 Z_t 為 IV 作 OLS 得到 \hat{y}_t

第二步: 以原方程的因變數對第一步 fitted values 和所有外生變數重新作 OLS, 即得 2SLS 係數