

CH 11 Q 1

11.1 Our aim is to estimate the parameters of the simultaneous equations model

$$y_1 = \alpha_1 y_2 + e_1$$

$$y_2 = \alpha_2 y_1 + \beta_1 x_1 + \beta_2 x_2 + e_2$$

We assume that  $x_1$  and  $x_2$  are exogenous and uncorrelated with the error terms  $e_1$  and  $e_2$ .

- Solve the two structural equations for the reduced-form equation for  $y_2$ , that is,  $y_2 = \pi_1 x_1 + \pi_2 x_2 + v_2$ . Express the reduced-form parameters in terms of the **structural parameters** and the reduced-form error in terms of the structural parameters and  $e_1$  and  $e_2$ . Show that  $y_2$  is correlated with  $e_1$ .
- Which equation parameters are consistently estimated using OLS? Explain.
- Which parameters are "identified," in the simultaneous equations sense? Explain your reasoning.

- To estimate the parameters of the reduced-form equation for  $y_2$  using the method of moments (MOM), which was introduced in Section 10.3, the two moment equations are

$$N^{-1} \sum x_{1i} (y_2 - \pi_1 x_{1i} - \pi_2 x_{2i}) = 0$$

$$N^{-1} \sum x_{2i} (y_2 - \pi_1 x_{1i} - \pi_2 x_{2i}) = 0$$

Explain why these two moment conditions are a valid basis for obtaining consistent estimators of the reduced-form parameters.

- Are the MOM estimators in part (d) the same as the OLS estimators? Form the sum of squared errors function for  $y_2 = \pi_1 x_1 + \pi_2 x_2 + v_2$  and find the first derivatives. Set these to zero and show that they are equivalent to the two equations in part (d).
- Using  $\sum x_{1i}^2 = 1$ ,  $\sum x_{2i}^2 = 1$ ,  $\sum x_{1i} x_{2i} = 0$ ,  $\sum x_{1i} y_{1i} = 2$ ,  $\sum x_{1i} y_{2i} = 3$ ,  $\sum x_{2i} y_{1i} = 3$ ,  $\sum x_{2i} y_{2i} = 4$ , and the two moment conditions in part (d) show that the MOM/OLS estimates of  $\pi_1$  and  $\pi_2$  are  $\hat{\pi}_1 = 3$  and  $\hat{\pi}_2 = 4$ .
- The fitted value  $\hat{y}_2 = \hat{\pi}_1 x_1 + \hat{\pi}_2 x_2$ . Explain why we can use the moment condition  $\sum \hat{y}_{2i} (y_{1i} - \alpha_1 \hat{y}_{2i}) = 0$  as a valid basis for consistently estimating  $\alpha_1$ . Obtain the IV estimate of  $\alpha_1$ .
- Find the 2SLS estimate of  $\alpha_1$  by applying OLS to  $y_1 = \alpha_1 \hat{y}_2 + e_1^*$ . Compare your answer to that in part (g).

$$(a) \quad y_1 = \alpha_1 y_2 + e_1 \quad (1)$$

$$y_2 = \alpha_2 y_1 + \beta_1 x_1 + \beta_2 x_2 + e_2 \quad (2)$$

$$\begin{aligned} \Rightarrow y_2 &= \alpha_2 (\alpha_1 y_2 + e_1) + \beta_1 x_1 + \beta_2 x_2 + e_2 \\ &= \alpha_1 \alpha_2 y_2 + \alpha_2 e_1 + \beta_1 x_1 + \beta_2 x_2 + e_2 \end{aligned}$$

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

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

$$\text{相关性: } \text{cov}(y_2, e_1 | x) = E(y_2 e_1 | x) - E(y_2 | x) E(e_1 | x)$$

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

$$= E \left[ \left( \frac{e_2 e_1 + \alpha_2 e_1^2}{1 - \alpha_1 \alpha_2} \right) \mid x \right]$$

$$= \frac{E(e_2 e_1 | x) + \alpha_2 E(e_1^2 | x)}{1 - \alpha_1 \alpha_2}$$

$$= \frac{\alpha_2}{(1 - \alpha_1 \alpha_2)} \sigma_1^2$$

b) 對聯立方程式  $y_1 = \alpha_1 y_2 + e_1$  , 用 OLS 估計不一致  
 $y_2 = \alpha_2 y_1 + \beta_1 x_1 + \beta_2 x_2 + e_2$

因為等式的右手邊含有內生變數  $(y_1, y_2)$  ,

而用 reduced form 的方程式則可用 OLS 一致地估計

c) 有  $M=2$  個方程式, 必須省略至少  $M-1$  個變數才能 identify  
 方程 (1) identified 因為  $x_1, x_2$  are omitted  
 方程 (2) is not identified

d) 假設  $x_1, x_2$  外生

$$E(x_{1i} v_{1i} | x) = E(x_{2i} v_{2i} | x) = 0$$

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

e) OLS

$$\frac{\partial S(\pi_1, \pi_2 | y, x)}{\partial \pi_1} = 2 \sum (y_2 - \pi_1 x_1 - \pi_2 x_2) x_1 = 0$$

$$\frac{\partial S(\pi_1, \pi_2 | y, x)}{\partial \pi_2} = 2 \sum (y_2 - \pi_1 x_1 - \pi_2 x_2) x_2 = 0$$

同除 2 再乘  $N^{-1}$  就與 MOM 相等

f)

$$N^{-1} \sum x_{1i} (y_2 - \pi_1 x_1 - \pi_2 x_2) = 0$$

$$N^{-1} \sum x_{2i} (y_2 - \pi_1 x_1 - \pi_2 x_2) = 0$$

$$\Rightarrow \sum x_{1i} y_{2i} - \pi_1 \sum x_{1i}^2 - \pi_2 \sum x_{1i} x_{2i} = 0$$

$$\sum x_{2i} y_{2i} - \pi_1 \sum x_{1i} x_{2i} - \pi_2 \sum x_{2i}^2 = 0$$

$$\Rightarrow 3 - \hat{\pi}_1 = 0 \quad \Rightarrow \quad \hat{\pi}_1 = 3$$

$$4 - \hat{\pi}_2 = 0 \quad \Rightarrow \quad \hat{\pi}_2 = 4$$

$$y_1 = \alpha_1 y_2 + e_1, \quad \hat{y}_2 = \hat{\pi}_1 x_1 + \hat{\pi}_2 x_2$$

$$\rightarrow \sum (\hat{\pi}_1 x_1 + \hat{\pi}_2 x_2) (y_1 - \alpha_1 y_2) = \sum \hat{y}_2 (y_1 - \alpha_1 y_2) = 0$$

$$\sum \hat{y}_2 y_1 - \alpha_1 \sum \hat{y}_2 y_2 = 0 \quad \rightarrow \hat{\alpha}_{1,2S} = \frac{\sum \hat{y}_2 y_1}{\sum \hat{y}_2 y_2}$$

$$\hat{\alpha}_{1,2S} = \frac{\sum \hat{y}_2 y_1}{\sum \hat{y}_2 y_2} = \frac{\sum (\hat{\pi}_1 x_1 + \hat{\pi}_2 x_2) y_1}{\sum (\hat{\pi}_1 x_1 + \hat{\pi}_2 x_2) y_2} = \frac{\hat{\pi}_1 \sum x_1 y_1 + \hat{\pi}_2 \sum x_2 y_1}{\hat{\pi}_1 \sum x_1 y_2 + \hat{\pi}_2 \sum x_2 y_2} = \frac{18}{25}$$

$$b) \quad \hat{\alpha}_{1,2SLS} = \frac{\sum \hat{y}_2 y_1}{\sum \hat{y}_2^2} \quad \text{also} \quad \hat{u}_1 = y_1 - \hat{y}_1 \quad \rightarrow \quad \hat{y}_1 = y_1 - \hat{u}_1$$

$$\text{因此} \quad \sum \hat{y}_2^2 = \sum \hat{y}_2 (y_2 - \hat{u}_2) = \sum \hat{y}_2 y_2 - \sum \hat{y}_2 \hat{u}_2 = \sum \hat{y}_2 y_2$$

$$\text{其中} \quad \sum \hat{y}_2 \hat{u}_2 = \sum (\hat{\pi}_1 x_1 + \hat{\pi}_2 x_2) \hat{u}_2 = \hat{\pi}_1 \sum x_1 \hat{u}_2 + \hat{\pi}_2 \sum x_2 \hat{u}_2 = 0$$

CH 11 Q16

11.16 Consider the following supply and demand model

$$\text{Demand: } Q_i = \alpha_1 + \alpha_2 P_i + e_{di}, \quad \text{Supply: } Q_i = \beta_1 + \beta_2 P_i + \beta_3 W_i + e_{si}$$

where  $Q$  is the quantity,  $P$  is the price, and  $W$  is the wage rate, which is assumed exogenous. Data on these variables are in Table 11.7.

| TABLE 11.7 |     |     | Data for<br>Exercise 11.16 |
|------------|-----|-----|----------------------------|
| $Q$        | $P$ | $W$ |                            |
| 4          | 2   | 2   |                            |
| 6          | 4   | 3   |                            |
| 9          | 3   | 1   |                            |
| 3          | 5   | 1   |                            |
| 8          | 8   | 3   |                            |

- Derive the algebraic form of the reduced-form equations,  $\hat{Q} = \theta_1 + \theta_2 W + v_2$  and  $\hat{P} = \pi_1 + \pi_2 W + v_1$ , expressing the reduced-form parameters in terms of the structural parameters.
- Which structural parameters can you solve for from the results in part (a)? Which equation is "identified"?
- The estimated reduced-form equations are  $\hat{Q} = 5 + 0.5W$  and  $\hat{P} = 2.4 + 1W$ . Solve for the identified structural parameters. This is the method of **indirect least squares**.
- Obtain the fitted values from the reduced-form equation for  $P$ , and apply 2SLS to obtain estimates of the demand equation.

$$a) \quad \begin{cases} Q_i = \alpha_1 + \alpha_2 P_i + e_{di} \\ Q_i = \beta_1 + \beta_2 P_i + \beta_3 W_i + e_{si} \end{cases}$$

$$\alpha_1 + \alpha_2 P_i + e_{di} = \beta_1 + \beta_2 P_i + \beta_3 W_i + e_{si}$$

$$\rightarrow \alpha_2 P_i - \beta_2 P_i = \beta_1 - \alpha_1 + \beta_3 W_i + e_{si} - e_{di}$$

$$\rightarrow P_i = \frac{\beta_1 - \alpha_1}{\alpha_2 - \beta_2} + \frac{\beta_3}{\alpha_2 - \beta_2} W_i + \frac{e_{si} - e_{di}}{\alpha_2 - \beta_2} \quad (1)$$

$$\begin{aligned} Q_i &= \alpha_1 + \alpha_2 \left( \frac{\beta_1 - \alpha_1}{\alpha_2 - \beta_2} + \frac{\beta_3}{\alpha_2 - \beta_2} W_i + \frac{e_{si} - e_{di}}{\alpha_2 - \beta_2} \right) + e_{di} \\ &= \frac{\alpha_2 \beta_1 - \alpha_1 \beta_2}{\alpha_2 - \beta_2} + \frac{\alpha_2 \beta_3}{\alpha_2 - \beta_2} W_i + \frac{\alpha_2 e_{si} - \beta_2 e_{di}}{\alpha_2 - \beta_2} \quad (2) \end{aligned}$$

b)  $M=2$  , 至少  $M-1=1$  個 variable 被 omitted

方程 (1) 省略  $w_1$  , identified

方程 (2) not identified

c)

$$\begin{cases} \hat{Q} = 5 + 0.5w \\ \hat{p} = 2.4 + 1w \end{cases} \Rightarrow \begin{matrix} \theta_1 = 5, \theta_2 = 0.5 \\ \pi_1 = 2.4, \pi_2 = 1 \end{matrix}$$

$Q = \alpha_1 + \alpha_2 p + e_d$  代入 reduced form 的  $p$

$$\Rightarrow Q = \alpha_1 + \alpha_2 (2.4 + 1w) + e_d = (\alpha_1 + 2.4\alpha_2) + \alpha_2 w + e_d$$

比較係數  $\hat{Q} = 5 + 0.5w$

$$\begin{matrix} \alpha_1 + 2.4\alpha_2 = 5 & (1) \\ \alpha_2 = 0.5 & (2) \end{matrix} \Rightarrow \begin{matrix} \hat{\alpha}_1 = 3.8 \\ \hat{\alpha}_2 = 0.5 \end{matrix}$$

d)  $\hat{p} = 2.4 + 1w$  ,  $Q_i = \alpha_1 + \alpha_2 p_i + e_{di}$

| w         | 2   | 3   | 1   | 1   | 3   |
|-----------|-----|-----|-----|-----|-----|
| $\hat{p}$ | 4.4 | 5.4 | 3.4 | 3.4 | 5.4 |
| Q         | 4   | 6   | 9   | 3   | 8   |

$$Q = \alpha_1 + \alpha_2 \hat{p} + e_i$$

$$X = \begin{bmatrix} 1 & 4.4 \\ 1 & 5.4 \\ 1 & 3.4 \\ 1 & 3.4 \\ 1 & 5.4 \end{bmatrix} , Y = \begin{bmatrix} 4 \\ 6 \\ 9 \\ 3 \\ 8 \end{bmatrix}$$

$$\hat{\beta} = (X^T X)^{-1} X^T Y = \begin{bmatrix} 3.8 \\ 0.5 \end{bmatrix}$$

$$Q_i = 3.8 + \frac{1}{2} p_i$$

## 11.17 Example 11.3 introduces Klein's Model I.

- Do we have an adequate number of IVs to estimate each equation? Check the necessary condition for the identification of each equation. The necessary condition for identification is that in a system of  $M$  equations at least  $M - 1$  variables must be omitted from each equation.
- An equivalent identification condition is that the number of excluded exogenous variables from the equation must be at least as large as the number of included right-hand side endogenous variables. Check that this condition is satisfied for each equation.
- Write down in econometric notation the first-stage equation, the reduced form, for  $W_{1t}$ , wages of workers earned in the private sector. Call the parameters  $\pi_1, \pi_2, \dots$
- Describe the two regression steps of 2SLS estimation of the consumption function. This is not a question about a computer software command.
- Does following the steps in part (d) produce regression results that are identical to the 2SLS estimates provided by software specifically designed for 2SLS estimation? In particular, will the  $t$ -values be the same?

(a)  $M = 8$  equations, 須 omit  $M - 1 = 7$  個變數

總共有 16 個變數

Consumption equation : 省略 10 個變數

Investment equation : 省略 11 個變數

Wage equation : 省略 11 個變數

→ 皆 identified

(b) Consumption equation : 2 個內生變數, 排除 5 個外生

Investment equation : 1 個內生, 排除 5 個外生

Wage equation : 1 個內生, 排除 5 個外生

(c) 
$$W_{1t} = \pi_1 + \pi_2 G_t + \pi_3 W_{2t} + \pi_4 TX_t + \pi_5 TIME_t + \pi_6 P_{t-1} + \pi_7 K_{t-1} + \pi_8 E_{t-1} + V$$

(d) 從 reduced form equation 取得預測值  $\hat{W}_{1t}$ , 用相同方法

取得  $\hat{P}_t$ . Create  $W_{1t}^* = \hat{W}_{1t} + W_{2t}$ ,

接著估計  $CN_t = \alpha_1 + \alpha_2 W_{1t}^* + \alpha_3 \hat{P}_t + \alpha_4 P_{t-1} + u_t$

(e) 估計係數會相同, 但  $t$  值不同.

因為 (d) 中的 SE 不是正確的 2SLS SE