

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

$$y_1 = \alpha_1 y_2 + e_1 \quad \text{--- (1)}$$

$$y_2 = \pi_1 x_1 + \pi_2 x_2 + v_2 \quad \text{--- (2)}$$

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 π_1 and π_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 $\sum \hat{y}_{2i} (y_{1i} - \alpha_1 y_{2i}) = 0$ as a valid basis for consistently estimating α_1 . Obtain the IV estimate of α_1 .
- Find the 2SLS estimate of α_1 by applying OLS to $y_1 = \alpha_1 \hat{y}_2 + e_1^*$. Compare your answer to that in part (g).

a.

$$\text{将 (2) 代入 (1) : } y_1 = \alpha_1 (\alpha_2 y_2 + e_2) + \beta_1 x_1 + \beta_2 x_2 + e_1$$

$$\Rightarrow (1 - \alpha_1 \alpha_2) y_1 = \beta_1 x_1 + \beta_2 x_2 + \alpha_2 e_2 + e_1$$

$$\Rightarrow y_1 = \pi_1 x_1 + \pi_2 x_2 + v_1$$

$$\pi_1 = \frac{\beta_1}{1 - \alpha_1 \alpha_2}, \quad \pi_2 = \frac{\beta_2}{1 - \alpha_1 \alpha_2}, \quad v_1 = \frac{\alpha_2 e_2 + e_1}{1 - \alpha_1 \alpha_2}$$

" v_1 含有 e_1 , 即 y_1 和 e_1 相關

b.

第一式 (1) : " y_1 和 e_1 相關 " $\hat{\alpha}_1$ 不具一致性

第二式 (2) : " y_2 和 e_2 相關 " $\hat{\alpha}_2, \hat{\beta}_1, \hat{\beta}_2$ 不具一致性

reduced form : " 僅含外生变量 x " $\hat{\alpha}_1, \hat{\alpha}_2$ 具一致性

c.

有 2 個結構方程，至少要省略 2-1 = 1 個外生變數

①：無外生變數，省略 1 個外生變數 \Rightarrow identified

②： x_1, x_2 省略 0 個外生變數 \Rightarrow not identified

\Rightarrow 僅第一式可被識別

d.

" x_1, x_2 為外生且和 v_2 不相關， $E(x_{\bar{j}}, v_2) = 0$

\therefore 可得一致性估計量

e.

若 $y_2 = \pi_1 x_1 + \pi_2 x_2 + v_2$ 最小平方法誤差

$$S = \sum (y_{2i} - \pi_1 x_{1i} - \pi_2 x_{2i})^2$$

若 π_1 及 π_2 偏微為零 即得到 (d) 之條件

\Rightarrow MOM 估計值 = OLS 估計值

f.

已知 $E x_{ij} = 1$ ， $\sum x_{ij} = 1$ ， $\sum x_{i1} x_{i2} = 0$ ， $\sum x_{i1} y_{ri} = 3$ 。

$$\sum x_{ir} y_{ri} = 4$$

$$\Rightarrow \hat{\pi}_1 = \frac{\sum x_{i1} y_{ri}}{\sum x_{i1}^2} = \frac{3}{1} = 3, \quad \hat{\pi}_2 = \frac{\sum x_{i2} y_{ri}}{\sum x_{i2}^2} = \frac{4}{1} = 4$$

8. 外生IV^① 和內生变量^② 相關^③ 和結構誤差^④ 不相關
即可達成一致性

$$\text{plim } \hat{\alpha}_1 = \frac{E(\hat{y}_r y_1)}{E(\hat{y}_r y_r)} = \frac{E(\hat{y}_r (\alpha_1 + y_r e_1))}{E(\hat{y}_r y_r)} = \alpha_1 + \frac{E(y_r e_1)}{E(y_r y_r)} = \alpha_1$$

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

$$\sum \hat{y}_{ri} y_{1i} = 3 \times 2 + 4 \times 3 = 18, \sum \hat{y}_{ri} y_{ri} = 3 \times 3 + 4 \times 4 = 25$$

$$\Rightarrow \hat{\alpha}_1 = \frac{18}{25} = 0.7 \checkmark$$

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

$$\hat{\alpha}_1 = \frac{\sum \hat{y}_{ri} y_{1i}}{\sum y_{ri} y_{ri}} = \frac{18}{25} = 0.7 \checkmark$$

和LS之結果相同，在單內生变量便使用同一組IV時。

IV之結果 = LS之結果

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 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, $Q = \theta_1 + \theta_2 W + v_2$ and $P = \pi_1 + \pi_2 W + v_1$, 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.

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

$$\Rightarrow p = \pi_1 + \pi_2 w + v_1$$

$$\pi_1 = \frac{\beta_1 - \alpha_1}{\alpha_2 - \beta_2}, \quad \pi_2 = \frac{\beta_3}{\alpha_2 - \beta_2}, \quad v_1 = \frac{e_s - e_d}{\alpha_2 - \beta_2}$$

代回需求式：

$$Q = \alpha_1 + \alpha_2 (\pi_1 + \pi_2 w + v_1) + e_d = \theta_1 + \theta_2 w + v_2$$

$$\therefore \theta_1 = \alpha_1 + \alpha_2 \pi_1, \quad \theta_2 = \alpha_2 \pi_2, \quad v_2 = \alpha_2 v_1 + e_d$$

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

$$\theta_1 = \alpha_1 + \alpha_2 \pi_1 \Rightarrow \alpha_1 \text{ not identified}$$

但 $\beta_1, \beta_2, \beta_3$ 僅以 π_1, π_2, α_2 圖是一條關係，無法唯一

解出 \Rightarrow not identified

ii. 需求方程式 (α_1, α_2) 被識別，供給方程式未被識別

c.

$$\hat{\alpha}_2 = \frac{\hat{\theta}_2}{\pi_2} = \frac{0.5}{1} = 0.5$$

$$\hat{\alpha}_1 = \hat{\theta}_1 - \hat{\alpha}_2 \pi_1 = 5 - 0.5 \times 2.4 = 3.8$$

d.

第一階段: $\hat{p}_i = \pi_i + w_i$

第二階段: $\hat{q} = \hat{\alpha}_1 + \hat{\alpha}_2 \hat{p}$

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

和(c)之結果相同，驗證判斷良好且僅一內生變數時。

ILS 之結果 = 2SLS 之結果

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_{it} , wages of workers earned in the private sector. Call the parameters π_1, π_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. 系統共有 M 個，必要條件：每條方程式至少要省略 $M - 1$ 個外生變數。Klein I 已滿足，故至少弱識別

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

c. 合所有外生變數為 Z_t

$$W_t^P = \pi_0 + \pi_1 Z_{1t} + \pi_2 Z_{2t} + \dots + v_t$$

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：將所有右側外生變數（ $\pi_0, \pi_1, \dots, \pi_M$ ）以及那外生變數 Z_t 作 OLS，得到 \hat{Y}_t

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

e.

係數估計值：相同

七值 / 標準誤：不一定是 \Rightarrow 若手動第二步直接食用 OLS

標準誤而未調整第一步估計
誤差的不確定性，則標準誤
會被低估，七值會偏大。專門
的 2SLS 指令會給正確的漸近
夏異係數。