

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 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.

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

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

$$= \alpha_1 \alpha_2 y_2 + \alpha_2 e_1 + \beta_1 x_1 + \beta_2 x_2 + e_2$$

$$\Rightarrow y_2 = \frac{\beta_1}{1 - \alpha_1 \alpha_2} x_1 + \frac{\beta_2}{1 - \alpha_1 \alpha_2} x_2 + \frac{\alpha_2 e_1 + e_2}{1 - \alpha_1 \alpha_2}$$

$$= \pi_1 x_1 + \pi_2 x_2 + v_2$$

$$\text{cov}(y_2, e_1 | X) = E \left[\left(\frac{\beta_1}{1 - \alpha_1 \alpha_2} x_1 + \frac{\beta_2}{1 - \alpha_1 \alpha_2} x_2 + \frac{\alpha_2 e_1 + e_2}{1 - \alpha_1 \alpha_2} \right) e_1 | X \right]$$

$$= E \left[\frac{\alpha_2 e_1 + e_2}{1 - \alpha_1 \alpha_2} \cdot e_1 | X \right]$$

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

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

b.

Both equation (1) and (2) have an endogenous variable on the right-hand side. OLS is biased and inconsistent.

c.

$M=2$ Endogenous: y_1, y_2 Exogenous: x_1, x_2

equation (1) 2 variables are omitted \therefore unidentified

equation (2) 0 variable is omitted \therefore unidentified

d.

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

$$\Rightarrow N^{-1} \sum x_{1i} v_i = 0$$

$$\therefore E(x_{1i} v_i | X) = 0 \quad \text{and} \quad E(x_{2i} v_i | X) = 0$$

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

$$E \left(x_{1i} \left(\frac{e_2 + \alpha_2 e_1}{1 - \alpha_1 \alpha_2} \right) | X \right)$$

$$= E \left(x_{1i} \left(\frac{e_2}{1 - \alpha_1 \alpha_2} \right) | X \right) + E \left(x_{1i} \left(\frac{\alpha_2 e_1}{1 - \alpha_1 \alpha_2} \right) | X \right) = 0 + 0 = 0$$

e.

$$S(\pi_1, \pi_2 | y_1, X) = \sum (y_2 - \pi_1 x_1 - \pi_2 x_2)^2$$

$$\frac{\partial S(\pi_1, \pi_2 | y_1, X)}{\partial \pi_1} = -2 \sum (y_2 - \pi_1 x_1 - \pi_2 x_2) x_1 = 0$$

$$\frac{\partial S(\pi_1, \pi_2 | y_1, X)}{\partial \pi_2} = -2 \sum (y_2 - \pi_1 x_1 - \pi_2 x_2) x_2 = 0$$

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

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

f.

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

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

$$3 - \pi_1 - 0 = 0$$

$$\Rightarrow \hat{\pi}_1 = 3$$

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

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

$$4 - 0 - \pi_2 = 0$$

$$\Rightarrow \hat{\pi}_2 = 4$$

g.

$$\sum \hat{y}_{2i} (y_{1i} - \alpha_1 y_{2i}) = 0$$

$$\frac{1}{N} \sum (\pi_1 x_{1i} + \pi_2 x_{2i}) (y_{1i} - \alpha_1 y_{2i}) = 0$$

$$E(\hat{y}_2 e_1 | X) = E((\pi_1 x_1 + \pi_2 x_2)(y_1 - \alpha_1 y_2) | X) = 0$$

$$\hat{\pi}_1 \rightarrow \pi_1 \quad \hat{\pi}_2 \rightarrow \pi_2$$

$$\sum (\hat{\pi}_1 x_{1i} + \hat{\pi}_2 x_{2i}) (y_{1i} - \alpha_1 y_{2i}) = 0$$

$$\sum \hat{y}_{2i} (y_{1i} - \alpha_1 y_{2i}) = 0$$

$$\sum \hat{y}_{2i} y_{1i} = \alpha_1 \sum \hat{y}_{2i} y_{2i} \Rightarrow \hat{\alpha}_{1,IV} = \frac{\sum \hat{y}_{2i} y_{1i}}{\sum \hat{y}_{2i} y_{2i}}$$

$$\hat{\alpha}_{1,IV} = \frac{\sum (\hat{\pi}_1 x_{1i} + \hat{\pi}_2 x_{2i}) y_{1i}}{\sum (\hat{\pi}_1 x_{1i} + \hat{\pi}_2 x_{2i}) y_{2i}} = \frac{3 \cdot 2 + 4 \cdot 3}{3 \cdot 3 + 4 \cdot 4} = \frac{18}{25}$$

h.

$$\hat{\alpha}_{1,2SLS} = \frac{\sum \hat{y}_{i2} y_{i1}}{\sum \hat{y}_{i2}^2}$$

$$\hat{v}_2 = y_2 - \hat{y}_2 \Rightarrow \hat{y}_2 = y_2 - \hat{v}_2$$

$$\begin{aligned} \sum \hat{y}_{i2}^2 &= \sum \hat{y}_{i2} (y_2 - \hat{v}_2) = \sum \hat{y}_{i2} y_2 - \sum \hat{y}_{i2} \hat{v}_2 \\ &= \sum \hat{y}_{i2} y_2 - \sum (\hat{\pi}_{i1} x_{i1} + \hat{\pi}_{i2} x_{i2}) \hat{v}_2 \\ &= \sum \hat{y}_{i2} y_2 - \hat{\pi}_{i1} \sum x_{i1} \hat{v}_2 - \hat{\pi}_{i2} \sum x_{i2} \hat{v}_2 \\ &= \sum \hat{y}_{i2} y_2 \end{aligned}$$

$$\therefore \hat{\alpha}_{1,2SLS} = \frac{\sum \hat{y}_{i2} y_{i1}}{\sum \hat{y}_{i2}^2} = \frac{\sum \hat{y}_{i2} y_{i1}}{\sum \hat{y}_{i2} y_{i2}} = \hat{\alpha}_{1,1V}$$

11.16 Consider the following supply and demand model

Demand: $Q_i = \alpha_1 + \alpha_2 P_i + e_{di}$, 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, $\hat{Q} = \theta_1 + \theta_2 W + v_2$ 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.

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

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

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

$$= \pi_1 + \pi_2 W_i + v_1$$

$$Q_i = \alpha_1 + \alpha_2 P_i + e_{di}$$

$$= \alpha_1 + \alpha_2 \left(\frac{\beta_1 - \alpha_1}{\alpha_2 - \beta_2} + \frac{\beta_3}{\alpha_2 - \beta_2} W_i + \frac{e_{si} - e_{di}}{\alpha_2 - \beta_2} \right) + e_{di}$$

$$= \left(\alpha_1 + \alpha_2 \frac{\beta_1 - \alpha_1}{\alpha_2 - \beta_2} \right) + \frac{\alpha_2 \beta_3}{\alpha_2 - \beta_2} W_i + \left(\alpha_2 \frac{e_{si} - e_{di}}{\alpha_2 - \beta_2} + e_{di} \right)$$

$$= \theta_1 + \theta_2 W_i + v_2$$

b. $m=2$ endogenous: P, Q exogenous: W

Demand: 1 variable is omitted \Rightarrow unidentified

Supply: 2 variable is omitted \Rightarrow unidentified

c.

$$\hat{Q} = \theta_1 + \theta_2 W = 5 + 0.5W$$

$$\hat{P} = \pi_1 + \pi_2 W = 2.4 + W$$

$$\hat{Q} = \alpha_1 + \alpha_2 \hat{P}$$

$$5 + 0.5W = \alpha_1 + \alpha_2 (2.4 + W)$$

$$5 + 0.5W = \alpha_1 + \alpha_2 2.4 + \alpha_2 W$$

$$\therefore \begin{cases} \alpha_1 = 3.8 \\ \alpha_2 = 0.5 \end{cases}$$

d.

$$\hat{P} = 2.4 + W$$

$$\hat{Q} = \alpha_1 + \alpha_2 \hat{P}$$

$$\sum (\hat{P}_i - \bar{P})^2 = \sum \hat{P}_i^2 - n \bar{P}^2 = 4$$

$$\sum (\hat{P}_i - \bar{P})(Q_i - \bar{Q}) = 2$$

$$\hat{\alpha}_2 = \frac{2}{4} = 0.5$$

$$\hat{\alpha}_1 = \bar{Q} - \hat{\alpha}_2 \bar{P} = 3.8$$

$$\hat{Q} = 3.8 + 0.5 \cdot \hat{P}$$

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 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?

$$CN_t = \alpha_1 + \alpha_2 (W_{1t} + W_{2t}) + \alpha_3 P_t + \alpha_4 P_{t-1} + e_{1t} \quad (11.17)$$

$$I_t = \beta_1 + \beta_2 P_t + \beta_3 P_{t-1} + \beta_4 K_{t-1} + e_{2t} \quad (11.18)$$

$$W_{1t} = \gamma_1 + \gamma_2 E_t + \gamma_3 E_{t-1} + \gamma_4 TIME_t + e_{3t} \quad (11.19)$$

a. $M = 8$ 7 omitted variable at least total 16 variables

| | include variable | omits | |
|-------------|------------------|-------|--------------|
| Consumption | 6 | 10 | => satisfied |
| Investment | 5 | 11 | => satisfied |
| Wage | 5 | 11 | => satisfied |

b.

| | RHS endogenous variable | exclude exogenous | |
|-------------|-------------------------|-------------------|--------------|
| consumption | 2 | 5 | => satisfied |
| Investment | 1 | 5 | => satisfied |
| wage | 1 | 5 | => satisfied |

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. Obtain fitted values \hat{W}_{1t} from the estimated reduced form equation in (c) and similarly obtain \hat{P}_t . Create $W_t^* = \hat{W}_{1t} + W_{2t}$. Regress CN_t on W_t^* , \hat{P}_t and P_{t-1} plus a constant by OLS.

(e) The coefficient estimates will be the same. The t -value will not be because the standard errors in part (d) are not correct 2SLS standard errors.