$$y_1 = \alpha_1 y_2 + e_1$$

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

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

- a. 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 .
- b. Which equation parameters are consistently estimated using OLS? Explain.
- c. Which parameters are "identified," in the simultaneous equations sense? Explain your reasoning.
- d. 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

$$N^{-1} \sum x_{i1} (y_2 - \pi_1 x_{i1} - \pi_2 x_{i2}) = 0$$

$$N^{-1} \sum x_{i2} (y_2 - \pi_1 x_{i1} - \pi_2 x_{i2}) = 0$$

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

- e. 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 $\sum x_{i1}^2 = 1$, $\sum x_{i2}^2 = 1$, $\sum x_{i1}x_{i2} = 0$, $\sum x_{i1}y_{1i} = 2$, $\sum x_{i1}y_{2i} = 3$, $\sum x_{i2}y_{1i} = 3$, $\sum x_{i2}y_{2i} = 4$, and the two moment conditions in part (d) show that the MOM/OLS estimates of π_1 and π_2 are $\hat{\pi}_1 = 3 \text{ and } \hat{\pi}_2 = 4.$
- The fitted value $\hat{y}_2 = \hat{\pi}_1 x_1 + \hat{\pi}_2 x_2$. Explain why we can use the moment condition $\sum \hat{y}_{i2}(y_{i1} - \alpha_1 y_{i2}) = 0$ as a valid basis for consistently estimating α_1 . Obtain the IV estimate
- **h.** 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

$$\exists \int_{z} = \left(\frac{\beta_{1}}{1-\alpha_{1}\alpha_{2}}\right) \times_{1} + \left(\frac{\beta_{2}}{1-\alpha_{1}\alpha_{2}}\right) \times_{2} + \left(\frac{\alpha_{2}e_{1}+e_{2}}{1-\alpha_{1}\alpha_{2}}\right)$$

$$(ov(y_2, \theta_1|x)) = E(y_2, \theta_1|x) = E\left[\left(\frac{\rho_1}{1-\alpha_1\alpha_2}X_1 + \frac{\rho_2}{1-\alpha_1\alpha_2}X_2 + \frac{\alpha_2\theta_1 te_2}{1-\alpha_1\alpha_2}\right)e_1 \mid x\right]$$

$$= \overline{E} \left[\left(\frac{\alpha_{1}}{1-\alpha_{1}\alpha_{2}} \times |e_{1}\right) \middle| X \right] + \overline{E} \left[\left(\frac{\alpha_{2}}{1-\alpha_{1}\alpha_{2}} \times |e_{1}\right) \middle| X \right] + \overline{E} \left[\left(\frac{\alpha_{2}\alpha_{1}+e_{2}}{1-\alpha_{1}\alpha_{2}} \cdot |e_{1}\right) \middle| X \right]$$

$$\frac{1}{2} \cos(y_1, e_1 \mid x) = E(y_1, e_1 \mid x) = \frac{\bar{t}(e_1 e_2 \mid x) + q_2 \bar{t}(e_1^2 \mid x)}{1 - q_1 q_2} = \frac{q_2 \cdot 6_1^2}{1 - q_1 q_2}$$

$$\Rightarrow cov(y^2, e_1 \mid x) = E(y, e_1 \mid x) = \frac{c(e_1e_2 \mid x) + \alpha_2 E(e_1 \mid x)}{1 - \alpha_1 \alpha_2} = \frac{\alpha_2 \cdot e_1}{1 - \alpha_1 \alpha_2}$$

bigsed and inconsistene

(d) E(xn, Vn | x) = E(xuz Viz | x) = 0 > E(Xink (diei + ez) | x] = E[a1 e, Xik | x] + E[1 a1 a2 ez Xik | x] = 0 (U) min 2 (92 - 101 x, - 12 x2) 2 2] [U, Thin] = 2 [42 - [U, x] - NE X2) x (-x,) =0 2 1 102T偏 1 次 : 2 と 1 42 - ル、メノ - なと X2) *(- X2) =0 4 12, ×N-1 so they are equivalent to the two equations. Cf.) N Exu, 1/2-TU, Xiv | - Tuc x h 2) =0 =) Exu, y2-TU, Exi, 2 - TUZ Exi, Xiv 2 =0 N E Xiz (42 - TV, Xu - TV - Xx =)=0 => E Xi= 92 - TV, E Xi, Xiz - TV = E Xiz =0 $\frac{1}{2} \int_{-\infty}^{\infty} \frac{1}{|x|} |x| - |x| \times |x| = 0$ $\frac{1}{2} \int_{-\infty}^{\infty} \frac{1}{|x|} |x| - |x| \times |x| = 0$ $\frac{1}{2} \int_{-\infty}^{\infty} \frac{1}{|x|} |x| - |x| \times |x| = 0$ $\frac{1}{2} \int_{-\infty}^{\infty} \frac{1}{|x|} |x| - |x| \times |x| = 0$ $\frac{1}{2} \int_{-\infty}^{\infty} \frac{1}{|x|} |x| - |x| \times |x| = 0$ proment condition $(y_2, e_1) = \sum \hat{y_2} (y_1 - \alpha, y_2) = 0 \implies \sum \hat{y_2} y_1 - \alpha, \sum \hat{y_2} y_2 = 0 \implies \alpha_1 = \sum \hat{y_2} y_1$ 9 di = \frac{\(\bar{\varphi}_1 \times_1 \times_1 \\ \frac{\varphi}{\varphi_2 \times_2 \\ \frac{\varphi}{\varphi_2 \times_2 \\ \frac{\varphi}{\varphi_2 \times_2 \\ \frac{\varphi}{\varphi_2 \\ \varphi_2 \\ \varphi_2 \\ \varphi_2 \\ \frac{\varphi}{\varphi_2 \\ \varphi_2 \\ \varphi = 3xx T4x3 = 18 3x3 T4x9 = 45 ch.) [] a1 54 = \(\frac{\frac{1}{2} \hat{j} \frac{1}{2}}{\frac{1}{2}} \frac{1}{2} \frac\ To prove a 12525 = a, (by moment condition) we need to prove $\Sigma \hat{y}_z^2 = \Sigma \hat{y}_1, \hat{y}_2$ Σ g² = ε g² (g² - ν²) = ε ĝ g² - ε g² ν² - ε ĝ y² - ε ĝ ν² - ε ĝ ν² - ε ĝ y² σ () 解釋(蒙執和誤差無問)

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.

TABLI	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

- a. Derive the algebraic form of the reduced-form equations, $Q = \theta_1 + \theta_2 W + v_2$ and $P = \pi_1 + \pi_2 W + \nu_1$, expressing the reduced-form parameters in terms of the structural parameters.
- b. 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.
- d. Obtain the fitted values from the reduced-form equation for P, and apply 2SLS to obtain estimates

a, reduced-torm

Demand = Supply

=)
$$P_N = \frac{\beta_1 - \alpha_1}{\alpha_2 - \beta_2} + \frac{\beta_3}{\alpha_2 - \beta_2}$$
 $W_N + \frac{e s_N - e q_N}{\alpha_2 - \beta_2}$

Quedit de
$$\left(\frac{\alpha_1-\alpha_1}{\alpha_2-\beta_2} + \frac{\beta_3}{\alpha_2-\beta_2}\right)$$
 $\frac{+e_{3n}-c_{4n}}{\alpha_2-\beta_2}$ + e iv

$$\widehat{d}_{2} = \frac{\sum (\widehat{p}_{N} - \widehat{p}) (0x - \overline{p})}{\sum (\widehat{p}_{N} - \widehat{p})^{2}} = \frac{-3 + 3 + 2}{7} = \frac{1}{2}$$

$$= \frac{-3+3+2}{4} = \frac{1}{2}$$

0-0

P=4.9

Q=6

P- P

11.17 Example 11.3 introduces Klein's Model I.

- a. 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.
- b. 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 $\pi_1, \pi_2,...$
- d. Describe the two regression steps of 2SLS estimation of the consumption function. This is not a question about a computer software command.
- e. 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?
- P_{t-1} . Thus, the consumption function is

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

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

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

=) all satisfied necessary condition for identification

The consumption tenction inculdes 6 variables and omits

11 The

II The wage JUNCTION

variables and exludes 5 exogenous variables Junction: 2PMS endogenous

Investment function: 1 RHJ

wage function: 1 RHS

investment function

=) all satisfied

01 WITE = TUI + TUZ G t + TUZ WZZ T TV4 TX t T TUX TIME + TUB Per + TUZ KOJ + TUB ECI +V

d. Get fitted value wit from (v), and using the same method Pe, create wit = wix + were

Regress (Ne by OLS.