



BA-BMECV2502U Econometrics

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Problem Set 5

Part 1: Simultaneous Equation Models

1. Use SMOKE.DTA for this exercise.

- (a) A model to estimate the effects of smoking on annual income (perhaps through lost work days due to illness, or productivity effects) is

$$\log(\text{income}) = \beta_0 + \beta_1 \text{cigs} + \beta_2 \text{educ} + \beta_3 \text{age} + \beta_4 \text{age}^2 + u,$$

where *cigs* is number of cigarettes smoked per day, on average. How do you interpret β_1 ?

- (b) To reflect the fact that cigarette consumption might be jointly determined with income, a demand for cigarettes equation is

$$\text{cigs} = \gamma_0 + \gamma_1 \log(\text{income}) + \gamma_2 \text{educ} + \gamma_3 \text{age} + \gamma_4 \text{age}^2 + \gamma_5 \log(\text{cigprice}) + \gamma_6 \text{restaurn} + u,$$

where *cigprice* is the price of a pack of cigarettes (in cents), and *restaurn* is a binary variable equal to unity if the person lives in a state with restaurant smoking restrictions. Assuming these are exogenous to the individual, what signs would you expect for γ_5 and γ_6 ?

- (c) Under what assumptions is the income equation from part (a) identified?
- (d) Estimate the reduced form for *cigs*. (Recall that this entails regressing *cigs* on all exogenous variables.) Are $\log(\text{cigprice})$ and *restaurn* significant in the reduced form?
- (e) Now, estimate the income equation by 2SLS. Discuss how the estimate of β_1 compares with the OLS estimate.
- (f) Do you think that cigarette prices and restaurant smoking restrictions are exogenous in the income equation?

2. Use the data *airfare.DTA*, but only for the year 1997.

- (a) A simple demand function for airline seats on routes is

$$\log(\text{passen}) = \beta_{10} + \alpha_1 \log(\text{fare}) + \beta_{11} \log(\text{dist}) + \beta_{12} [\log(\text{dist})]^2 + u,$$

where *passen* is the average number of passengers per day, *fare* is the average airfare, and *dist* is the route distance (in miles). If this is a demand function, what should be the sign of α_1 ?

- (b) Estimate the equation from part (a) by OLS. What is the estimated price elasticity?
- (c) Consider the variable *concen*, which is a measure of market concentration. (The share of business accounted for by the largest carrier.) Explain what we must assume to treat *concen* as exogenous in the demand equation.
- (d) Now assume that *concen* is exogenous to the demand equation. Estimate the reduced form for $\log(\text{fare})$ and confirm that *concen* has a positive partial effect on $\log(\text{fare})$.
- (e) Estimate the demand equation using IV. Now, what is the estimated price elasticity of demand? How does it compare with the OLS estimate?
- (f) Using the IV estimates, describe how demand for seats depends on route distance.

Part 2: Maximum Likelihood Estimation

3. If $f(\mathbf{y}|\mathbf{x}; \boldsymbol{\theta})$ is a correctly specified model for the density of \mathbf{y}_i given \mathbf{x}_i , does $\boldsymbol{\theta}_0$ solve $\max_{\boldsymbol{\theta} \in \Theta} E[f(y_i|\mathbf{x}_i; \boldsymbol{\theta})]$?
4. Suppose that for a random sample, $y_i|\mathbf{x}_i \sim N(m(\mathbf{x}_i, \boldsymbol{\beta}_0), \sigma_0^2)$, where $m(\mathbf{x}, \boldsymbol{\beta})$ is a function of the K -vector of explanatory variables \mathbf{x} and the $P \times 1$ parameter vector $\boldsymbol{\beta}$. Recall that $E(y_i|\mathbf{x}_i) = m(\mathbf{x}_i, \boldsymbol{\beta}_0)$ and $\text{var}(y_i|\mathbf{x}_i) = \sigma_0^2$.
 - (a) Write down the conditional log-likelihood function for observation i . Show that the CMLE of $\boldsymbol{\beta}_0$, $\hat{\boldsymbol{\beta}}$, solves the problem $\min_{\boldsymbol{\beta}} \sum_{i=1}^N [y_i - m(\mathbf{x}_i, \boldsymbol{\beta})]^2$. In other words, the CMLE for $\boldsymbol{\beta}_0$ is the so called nonlinear least squares estimator.
 - (b) Let $\boldsymbol{\theta} = (\boldsymbol{\beta}', \sigma^2)'$ denote the $(P + 1) \times 1$ vector of parameters. Find the score of the log-likelihood for a generic i . Show directly that $E[\mathbf{s}_i(\boldsymbol{\theta}_0)|\mathbf{x}_i] = \mathbf{0}$. What features of the normal distribution do you need to be correctly specified in order to show that the conditional expectation of the score is zero?
 - (c) Use the first-order condition to find $\hat{\sigma}^2$ in terms of $\hat{\boldsymbol{\beta}}$.
 - (d) Find the Hessian of the log-likelihood function with respect to $\boldsymbol{\theta}$.
 - (e) Show directly that $-E[\mathbf{H}_i(\boldsymbol{\theta}_0|\mathbf{x}_i)] = E[\mathbf{s}_i(\boldsymbol{\theta}_0)\mathbf{s}_i(\boldsymbol{\theta}_0)'|\mathbf{x}_i]$.
 - (f) Propose an estimated asymptotic variance of $\hat{\boldsymbol{\beta}}$, and explain how to obtain the asymptotic standard errors.

These problems are taken from the Wooldridge (2010) and Wooldridge (2020) textbooks.