

Questions

1. Consider the following LIML model:

$$y_{i1} = Y'_{i1}\gamma + x'_{i1}\beta + u_{i1} \quad (1)$$

$$Y_{i1} = x'_i\Pi_1 + V_{i1} \quad (2)$$

where y_{i1} is a scalar, and Y_{i1} is a vector with second equation characterizing the reduced form of the model. u_{i1} is a scalar random variable and V_{i1} is a vector, where the vector (u_{i1}, V_{i1}) is mean 0 with variance matrix

$$\Sigma = \begin{pmatrix} \sigma_1^2 & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix}$$

Show the LIML estimator of γ is obtained by minimizing

$$\delta'W_1\delta/\delta'W\delta$$

where $\delta = (1, -\gamma)$ with respect to γ . Also show that the minimization of $\delta'W_1\delta - \delta'W\delta$ yields the 2SLS estimator of γ , where

$$W = [\mathbf{y}, \mathbf{Y}]'M[\mathbf{y}, \mathbf{Y}]$$

where the bold letters indicate stacking the individual elements of the dependent variables, and M is annihilator matrix of the matrix obtained by stacking x_i . Similarly,

$$W_1 = [\mathbf{y}, \mathbf{Y}]'M_1[\mathbf{y}, \mathbf{Y}]$$

where M_1 is the annihilator matrix of the matrix obtained by stacking x_{i1} .

2. Consider the model

$$y_i = Y_i\gamma + x'_i\beta + u_i \quad (3)$$

$$Y_i = X_i\Pi + V_i = \bar{Y}_i + V_i \quad (4)$$

where y_i and Y_i are a scalar and vector respectively, with variance of the mean 0 vector (u_i, V_i) is

$$\Sigma = \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix}$$

Let α denote the stacked parameter vector (γ, β) .

Derive the asymptotic variance of the (infeasible) estimator obtained by regressing y_i on \tilde{Y}_i , and compare this to the 2SLS estimator.

3. Consider the following 2 equation model:

$$y_{i1} = \gamma_1 y_{i2} + u_{i1} \quad (5)$$

$$y_{i2} = \gamma_2 y_{i1} + \beta_1 x_{i1} + \beta_2 x_{i2} + u_{i2} \quad (6)$$

Compute the LIML and 2SLS estimates of γ_1 given the following moment conditions:

$$E[x_i x_i'] = I_2,$$

$$E[y_i y_i'] = \begin{pmatrix} 7/2 & 2 \\ 2 & 3/2 \end{pmatrix}$$

$$E[x_i y_i] = \begin{pmatrix} 1 & 1 \\ 3/2 & 1/2 \end{pmatrix}$$

where $x_i = x_{i1}, x_{i2}$, $y_i = y_{i1}, y_{i2}$.

4. Let the sequence X_i be i.i.d., with bernouilli distribution with probability $X_i = 1$ denoted by p . Find the MLE of p and prove consistency and asymptotic normality of this estimator.
5. Let X_i be i.i.d with a uniform distribution between 0 and θ . Find the MLE of θ and establish consistency, by showing both the bias and the variance of the MLE converge to 0.
6. Consider the Logit model with $P(y_i = 1|x_i) = \Lambda(\beta_0 + \beta_1 x_i)$ where the scalar random variable x_i is binary, taking values 0 and 1.
 - (a) Show this model can be written as a linear probability model, $P(y_i = 1|x_i) = \gamma_0 + \gamma_1 x_i$ and derive γ_0 and γ_1 as functions of β_0 and β_1 .
 - (b) Show the MLE of γ_0 and γ_1 is equal to the least squares estimate of y_i on x_i with intercept.

7. For the MA(1) process defined by

$$y_t = \epsilon_t - \rho\epsilon_{t-1}$$

where ϵ_t is i.i.d. with mean 0 and variance σ^2 . Define

$$y_t^* = \epsilon_t^* - \rho^{-1}\epsilon_{t-1}^*$$

where ϵ_t^* are i.i.d. with mean 0 and variance $\rho\sigma^2$. Compare the autocovariances for y_t and y_t^* .

8. Consider the AR(2) process:

$$y_t = \rho_1 y_{t-1} + \rho_2 y_{t-2} + \epsilon_t$$

where ϵ_t is i.i.d, mean 0, variance σ^2 .

(a) Determine if the stability (stationarity) condition is satisfied if

i. $\rho_1 = \rho_2 = 1/2$

ii. $\rho_1 = \rho_2 = -1/2$

(b) Assuming ρ_1, ρ_2 satisfy the stability condition. Derive γ_0 and γ_1 .