Lecture 8 Multivariate Responses

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System of Equations

General setup

$$y_{ij} = x'_{ij}\beta_i + u_{ij}, \quad i = 1, \dots, n, \quad j = 1, \dots, m$$

- ▶ Two important examples
 - Zellner's (1962) seemingly unrelated regression (SUR): small # of units n, large # of observations m (e.g. time)
 - panel data model: large # of units n, small # of periods m = T

The Road Ahead...

- ► SUR model
- ▶ Panel data model

SUR Model

Setup

$$\underbrace{\begin{bmatrix} y_{1j} \\ \vdots \\ y_{nj} \end{bmatrix}}_{y_j} = \underbrace{\begin{bmatrix} x'_{1j} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & x'_{nj} \end{bmatrix}}_{X_j} \underbrace{\begin{bmatrix} \beta_1 \\ \vdots \\ \beta_n \end{bmatrix}}_{\beta} + \underbrace{\begin{bmatrix} u_{1j} \\ \vdots \\ u_{nj} \end{bmatrix}}_{u_j}, \qquad j = 1, \dots, m$$

▶ Likelihood function under $u_j|X \sim_{i.i.d.} \mathbb{N}(0, \Sigma)$

$$f(y|\beta,\Sigma) \propto \frac{1}{|\Sigma|^{m/2}} \exp \left[-\frac{1}{2} \sum_{j=1}^{m} (y_j - X_j \beta)' \Sigma^{-1} (y_j - X_j \beta) \right]$$

► Equivalent to single-equation OLS when (i) $\beta_i = \beta$ (same regressors) or (ii) $Cov(u_{sj}, u_{tj}) = 0$ for $s \neq t$ (truly unrelated)

Gibbs Algorithm

Conditionally conjugate prior

$$\beta \sim \mathbb{N}(\beta_0, B_0), \qquad \Sigma^{-1} \sim \mathbb{W}(\nu_0, V_0)$$
 (Wishart distribution)

▶ Gibbs sampler for $\pi(\beta, \Sigma^{-1}|y)$

$$\beta | y, \Sigma^{-1} \sim \mathbb{N}(\beta_1, B_1)$$

 $\Sigma^{-1} | y, \beta \sim \mathbb{W}(\nu_1, V_1)$

where (trick: $tr(A_{p \times q}B_{q \times p}) = tr(BA)$)

$$B_{1} = \left(\sum X_{j}' \Sigma^{-1} X_{j} + B_{0}^{-1}\right)^{-1}$$

$$\beta_{1} = B_{1} \left(\sum X_{j}' \Sigma^{-1} y_{j} + B_{0}^{-1} \beta_{0}\right)$$

$$\nu_{1} = \nu_{0} + m$$

$$V_{1} = \left(V_{0}^{-1} + \sum (y_{j} - X_{j}\beta)(y_{j} - X_{j}\beta)'\right)^{-1}$$

Python Pseudo Code

To be added...

Panel Data Model

Setup

$$\underbrace{\begin{bmatrix} y_{i1} \\ \vdots \\ y_{iT} \end{bmatrix}}_{y_i} = \underbrace{\begin{bmatrix} x'_{i1} \\ \vdots \\ x'_{iT} \end{bmatrix}}_{X_i} \beta + \underbrace{\begin{bmatrix} w'_{i1} \\ \vdots \\ w'_{iT} \end{bmatrix}}_{W_i} b_i + \underbrace{\begin{bmatrix} u_{i1} \\ \vdots \\ u_{iT} \end{bmatrix}}_{u_i}, \qquad i = 1, \dots, n$$

▶ Likelihood function under $u_i|X, W \sim_{i.i.d.} \mathbb{N}(0, h^{-1}I_T)$

$$f(y|\beta,b,h) \propto h^{nT/2} \exp \left[-\frac{h}{2} \sum_{i=1}^{n} (y_i - X_i \beta - W_i b_i)' (y_i - X_i \beta - W_i b_i) \right]$$

where β = fixed effect, b_i = random effect/heterogeneity

Conditionally conjugate prior

$$\beta \sim \mathbb{N}(\beta_0, B_0), \ h \sim \mathbb{G}(\alpha_0/2, \delta_0/2), \ b_i|D \sim \mathbb{N}(0, D), \ D^{-1} \sim \mathbb{W}(\nu_0, D_0)$$

Gibbs Algorithm

▶ Gibbs sampler for $\pi(h, D, (\beta, b)|y)$

$$h|y, \beta, b, D \sim \mathbb{G}(\alpha_1/2, \delta_1/2), \quad D^{-1}|y, \beta, h, b =_d D^{-1}|b \sim \mathbb{W}(\nu_1, D_1)$$

 $\beta, b|y, h, D : \quad b_i|y, \beta, D, h \sim \mathbb{N}(b_{1i}, D_{1i}), \quad \beta|y, h, D \sim \mathbb{N}(\beta_1, B_1)$

where $((\beta, b)$ in one block)

$$\delta_{1} = \delta_{0} + \sum_{i} (y_{i} - X_{i}\beta - W_{i}b_{i})'(y_{i} - X_{i}\beta - W_{i}b_{i})
D_{1} = \left(D_{0}^{-1} + \sum_{i} b_{i}b_{i}'\right)^{-1}
D_{1i} = \left(hW_{i}'W_{i} + D^{-1}\right)^{-1}
b_{1i} = D_{1i}[hW_{i}'(y_{i} - X_{i}\beta)]
B_{1i} = W_{i}DW_{i}' + h^{-1}I_{T}
B_{1} = \left(\sum_{i} X_{i}'B_{1i}^{-1}X_{i} + B_{0}^{-1}\right)^{-1}
\beta_{1} = B_{1}\left(\sum_{i} X_{i}'B_{1i}^{-1}y_{i} + B_{0}^{-1}\beta_{0}\right)$$

Python Pseudo Code

To be added...

Readings

Zellner (1962), "An Efficient Method of Estimating Seemingly Unrelated Regression Equations and Tests for Aggregation Bias," Journal of the American Statistical Association