
Spatial Econometrics: Final Exam

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1. Consider the following SAC model with heteroskedastic errors:

$$\mathbf{y} = \rho \mathbf{W}_1 \mathbf{y} + \mathbf{X} \boldsymbol{\beta} + \mathbf{u} \quad (1)$$

$$\mathbf{u} = \lambda \mathbf{W}_2 \mathbf{u} + \boldsymbol{\varepsilon} \quad (2)$$

$$\boldsymbol{\varepsilon} \sim N(\mathbf{0}, \boldsymbol{\Omega}) \quad (3)$$

The matrix $\boldsymbol{\Omega}$ is the variance-covariance matrix of the error terms, which is assumed to be known a priori. For example, we can assume that:

$$\text{Var}(\epsilon_i) = \sigma_i^2 = \mathbf{z}_i^\top \boldsymbol{\alpha} \quad (4)$$

or

$$\text{Var}(\epsilon_i) = \sigma_i^2 = \exp(\mathbf{z}_i^\top \boldsymbol{\alpha}) \quad (5)$$

or more general,

$$\text{Var}(\epsilon_i) = \sigma_i^2 = \mathbf{h}(\mathbf{z}_i^\top \boldsymbol{\alpha}) \quad (6)$$

where $\mathbf{h}(\cdot)$ is any function, \mathbf{z}_i is a vector of covariates for each spatial unit, and $\boldsymbol{\alpha}$ is a vector of parameters with element $\alpha_p, p = 0, 1, \dots, P$. Therefore, the diagonal elements of the error covariance matrix $\boldsymbol{\Omega}$ are:

$$\boldsymbol{\Omega}_{ii} = \mathbf{h}_i(\mathbf{z}_i^\top \boldsymbol{\alpha}), \quad \mathbf{h}_i > 0 \quad (7)$$

Note that the model has $2 + K + P$ unknown parameters:

$$\boldsymbol{\theta} = (\rho, \boldsymbol{\beta}^\top, \lambda, \boldsymbol{\alpha}^\top)^\top. \quad (8)$$

- a) Find the Log-likelihood function. Show every step.
 - b) Find the first order conditions. Show all your work.
2. Explain and derive the LR test for the SLM model. Show all your work.
 3. Explain how to estimate a SEM model using the Method of Moments and the Spatial FGLS. Show all your work.

Some important results are the followings:

$$\frac{\partial(\rho\mathbf{W})}{\partial\rho} = \mathbf{W} \quad (9)$$

$$\begin{aligned} \frac{\partial\mathbf{A}}{\partial\rho} &= \frac{\partial(\mathbf{I} - \rho\mathbf{W})}{\partial\rho} \\ &= \frac{\partial\mathbf{I}}{\partial\rho} - \frac{\partial\rho\mathbf{W}}{\rho} \\ &= -\mathbf{W} \end{aligned} \quad (10)$$

$$\frac{\partial \log |\mathbf{A}|}{\partial\rho} = \text{tr}(\mathbf{A}^{-1}\partial\mathbf{A}/\partial\rho) = \text{tr}[\mathbf{A}^{-1}(-\mathbf{W})] \quad (11)$$

Let $\boldsymbol{\varepsilon} = \mathbf{A}\mathbf{y} - \mathbf{X}\boldsymbol{\beta}$, then:

$$\frac{\partial\boldsymbol{\varepsilon}}{\partial\rho} = \frac{\partial(\mathbf{A}\mathbf{y} - \mathbf{X}\boldsymbol{\beta})}{\partial\rho} = -\mathbf{W}\mathbf{y} \quad (12)$$

$$\frac{\partial\boldsymbol{\varepsilon}^\top\boldsymbol{\varepsilon}}{\partial\rho} = \boldsymbol{\varepsilon}^\top(\partial\boldsymbol{\varepsilon}/\partial\rho) + (\partial\boldsymbol{\varepsilon}^\top/\partial\rho)\boldsymbol{\varepsilon} = 2\boldsymbol{\varepsilon}^\top(\partial\boldsymbol{\varepsilon}/\partial\rho) = 2\boldsymbol{\varepsilon}^\top(-\mathbf{W})\mathbf{y} \quad (13)$$

$$\frac{\partial\mathbf{A}^{-1}}{\partial\rho} = -\mathbf{A}^{-1}(\partial\mathbf{A}/\partial\rho)\mathbf{A}^{-1} = \mathbf{A}^{-1}\mathbf{W}\mathbf{A}^{-1} \quad (14)$$

$$\frac{\partial \text{tr}(\mathbf{A}^{-1}\mathbf{W})}{\partial\rho} = \text{tr}(\partial\mathbf{A}^{-1}\mathbf{W}/\partial\rho) \quad (15)$$