

# CSISS WORKSHOP

## **Introduction to Spatial Pattern Analysis in a GIS Environment**

Spatial Modeling: Spatial Parameters in  
Autoregressive Models and  
Geographically Weighted Regression

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# Typology of Spatial Econometric Models

- General Model:

$$Y = \theta W_1 Y + X\beta + \varepsilon$$

$\varepsilon = \kappa W_2 \varepsilon + \lambda$  with  $\lambda$  normal, 0 mean, and constant variance  $\Omega$  (i.e., variance is the same for every variable and covariance for every combination of variables is always 0)

# Typology of Spatial Econometric Models

$$Y = \theta W_1 Y + X\beta + \varepsilon \quad \text{and} \quad \varepsilon = \kappa W_2 \varepsilon + \lambda$$

set  $\theta=0, \kappa = 0$  ----- RESULT is  $Y = X\beta + \varepsilon$

set  $\kappa = 0$  ----- RESULT is  $Y = \theta W_1 Y + X\beta + \varepsilon$

set  $\theta=0$  ----- RESULT is  $Y = X\beta + (I - \kappa W_2)^{-1} \lambda$

also -----  $Y = \theta W_1 Y + X\beta + (I - \kappa W_2)^{-1} \lambda$

These are: linear regression model

spatial lag model

spatial disturbance model

spatial lag and disturbance model

# Typology of Spatial Econometric Models

- Also, vary variance and covariance assumptions. ( $\Omega$  can represent heteroscedasticity as well as homoscedasticity)
- Also, can create space-time models

# Testing the Models

- In any model, if error term is correlated OLS is inappropriate device to find parameters. Usual tests on parameters and  $R^2$  cannot be used.
- Use maximum likelihood approach (i.e., the parameters most likely to give you your data).

**Wald** test on parameters; Likelihood Ratio test on the goodness of the model; **La Grange Multiplier** test on residuals (non-spatial); **Moran's I** test on residuals (spatial)

# Example

- Crime in Columbus
- Bisection search for  $\theta$  parameter yields 0.4310.  
$$CR = 45.1 + 0.43 W_1 CR - 1.0 IN - 0.3 Ho$$
- Now, La Grange Test on Residuals (non-spatial) = OK;  
$$Z(I) = 0.65$$
- Conclusion: Not strong evidence of a spatial disturbance process after introduction of a spatially lagged dependent variable.

# Geographically Weighted Regression (GWR)

## “Geography of Parameter Space”

- Based on need to deal with heterogeneity in geographical space.
- $Y_i$  is explained by  $X_j$  within  $d$ .
- One equation for each  $Y_i$ .
- $X_j$  are weighted by distance from  $i$ .
- $d$  represents a kernel size.
- Repeat for all  $i$ .
- $Y_i = a_i + b_i \sum X_j [d/w] + e$  for all  $j$  within  $d$  of  $i$ ;  
 $w$  is a spatial weight

# Problems Associated with Spatial Pattern Analysis

## The Problems Help to Define the Field

- Scale (statistics that vary  $d$  -- LISA, G, K, ESDA)
- Zoning (use smallest units and then aggregate upward if necessary; use statistics that vary  $d$ ; ESDA)
- **Dependence** (SpaStats, Variogram, Spa. Econ.)
- Heterogeneity (statistics that vary  $d$ , GWR, Spa. Econ.)
- Boundaries (NN, K function)
- Missing Data (Kriging, TINs)
- Large Data Sets (all of the above)

**S p a t i a l   i s   S p e c i a l**