



# Spatial Data Models

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# Introduction

**Table 2.1** Focal problems in spatial statistics and spatial econometrics

Spatial statistics	Spatial econometrics
Super population perspective (i.e., realizations from a theoretical population): model-based inference	
Properties of estimators	
Specification of geographic neighbourhood structure	
Modifiable areal unit problem (MAUP)	
Quantifying spatial autocorrelation	
Variable transformations: Box-Cox, Box-Tidwell	
Spatially adjusted statistical techniques	
Cluster detection: hot and cold spots; LISA statistics	
Distance as a covariate	
Bayesian hierarchical models	
Exploratory spatial data analysis	
Space-time modelling	
Sampling network structure: design-based inference	Constrained parameter estimation
Ecological fallacy	Optimization models
Map generalization: spatial interpolation	Endogenous versus exogenous variables
Missing spatial data imputation	Spatial complexity and spatial regimes
Auto- model specification: normal, Poisson, binomial	
Spatial structure as a covariate (spatial filtering)	
Bayesian smoothing of map values	
Error propagation	

# Introduction

- Spatial econometrics is a subfield of econometrics dealing with spatial effects among spatial unit (Elhorst, 2014).
- Spatial data are data that have a spatial component; it means that data are connected to a place in the Earth. Data usually collected with reference to location:
  - Location-administrative spatial units (states, districts, counties, etc.)
  - Functional regions (e.g. labour market regions)
  - Points in space (e.g. cities, municipalities, plants)

# Introduction

- The distinction between mainstream econometrics and spatial econometrics is an existence of spatial interaction effects.
- Spatial effects have two form (Anselin, 1988):
  - Spatial Dependence (Spatial Autocorrelation)
  - Spatial Heterogeneity (Spatial Structure)
    - non-constant error variances (spatial heteroskedasticity)
    - model coefficients (variable coefficients, spatial regimes).

# Introduction

- **Spatial dependence** reflects a situation where values observed at one location or region, say observation  $i$ , depend on the values of *neighboring* observations at nearby locations (LeSage & Pace, 2009).
- **Tobler's first law of geography:**  
"Everything is related to everything else, but near things are more related than distant things"
- **Spatial dependence** is associated with the notion of relative space (location):
  - Neighboring regions are expected to be more alike than arbitrary regions,
  - Spatial dependence is expected to diminish with increasing distance

# Introduction

- **Spatial error dependence** is often interpreted as a nuisance., which caused by:
  - Measurement error
  - Omitted variables
- In regression analysis, measurement error and omitted variables are accommodated in error term  $\varepsilon$ .  
The error term in region  $i$  and  $j$  are no more independent but correlate

$$\bullet \text{Cov}(\varepsilon_i, \varepsilon_j) = E(\varepsilon_i, \varepsilon_j) ; i \neq j \quad (1)$$

# Introduction

- The identically independently normal distribution of error term  $\varepsilon$ , can be wrote in matrix as:

$$\bullet \varepsilon \sim \text{IIDN}(\mathbf{o}, \sigma^2 \mathbf{I})$$

- with  $\mathbf{o}$  is a  $(p \times 1)$  vector mean,  $\sigma^2$  is variance error, and  $\mathbf{I}$  is  $(n \times n)$  identity matrix.
- When the error term are correlated the distribution of  $\varepsilon$  is

$$\bullet \varepsilon \sim \text{N}(\mathbf{o}, \Omega)$$

- with  $\Omega$  is a covariance matrix, which modeled with the aid of spatial weights, defined by distance or contiguity measures

# Introduction

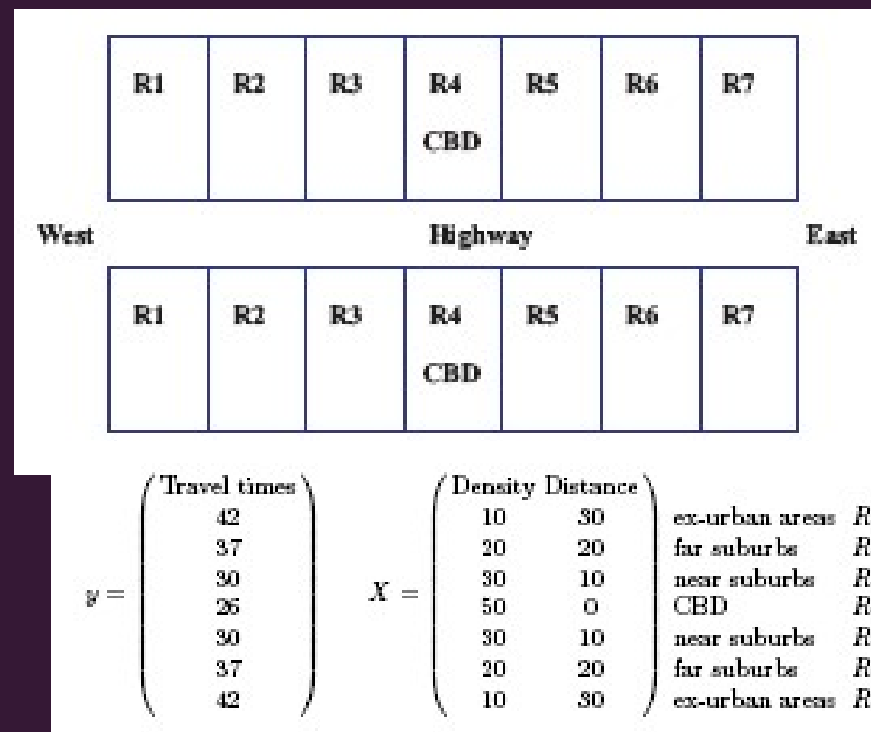
- **Substantive spatial** dependence reflects the existence of spatial interaction effect. The complex pattern of the interaction and dependence of socio-demographic and economic activity on the regional level to be a problem in modeling. In regional science, location and distance are important variables to be considered at work in human geography and market activity.
- The formal way to modeling the substantive spatial dependence the geo-referenced variable to be explained  $Y$ , is subject to the moment condition.

$$\bullet \text{Cov}(y_i, y_j) \neq 0; i \neq j \quad (1.2)$$

- **Geo-referenced variable  $Y$**  is values of  $Y$  are measured for spatial units (regions, districts, etc.)

# Introduction

- Example



# Spatial Weight Matrix

A spatial weights matrix is an  $n \times n$  positive symmetric matrix  $\mathbf{W}$  with element  $w_{ij}$  at location  $i, j$  for  $n$  locations. The values of  $w_{ij}$  or the weights for each pair of locations are assigned by some preset rules which define the spatial relations among locations and, therefore, determine the spatial autocorrelation statistics. By convention,  $w_{ii} = 0$  for the diagonal elements (Zhou X and Lin H, 2008)

- **Contiguity (neighborhood)**

- Contiguity reflects the relative location of one spatial unit to other regions in space. Neighbourhoodships of spatial units are usually established from a map.

- **Distance**

- The location in space represented by latitude and longitude is one source of information. This information allows to calculate distance between points in space. In regional science points in space may represent centers or cities of regions.

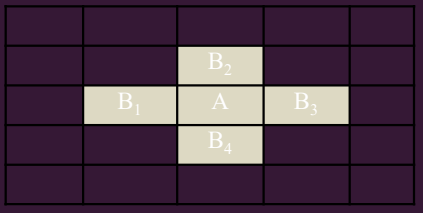
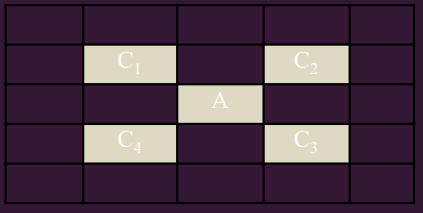
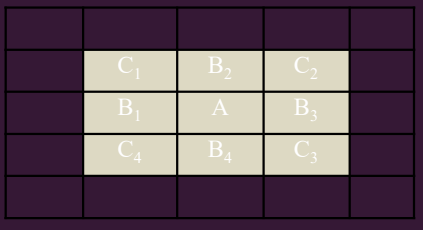
# Spatial weight matrix

## 1. Spatial Contiguity Matrix

The spatial contiguity matrix  $W^*$  is a binary  $n \times n$  matrix whose elements of the matrix  $w_{ij}^*$  are 0 or 1. The element  $w_{ij}^*$  is equal to one if the regions  $i$  and  $j$  are neighbors and 0 otherwise.

$$w_{ij}^* = \begin{cases} 1 & \text{if } i \text{ and } j \text{ are neighbours} \\ 0 & \text{otherwise} \end{cases} \quad (2.4)$$

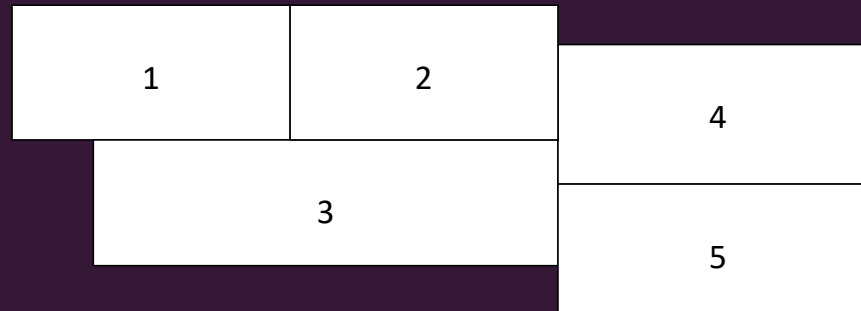
**In a regular grid**, the contiguity matrix can be distinguished in three different types with the game of chess analogy, i.e Rook contiguity, bishop contiguity, and queen contiguity

Types	Form
<p><b>Rock contiguity</b></p> <p>A spatial unit is a neighbour of another unit if both areas share a common edge (side). Unit B<sub>1</sub>, B<sub>2</sub>, B<sub>3</sub> and B<sub>4</sub> are neighbor of unit A</p>	 <p>The diagram shows a 5x5 grid. Unit A is in the center (row 3, column 3). Units B<sub>1</sub>, B<sub>2</sub>, B<sub>3</sub>, and B<sub>4</sub> are located at (row 3, column 2), (row 2, column 3), (row 3, column 4), and (row 4, column 3) respectively. All these units share a common edge with unit A.</p>
<p><b>Bishop contiguity</b></p> <p>A spatial unit is a neighbor of another unit if both areas share a common vertex. Unit C<sub>1</sub>, C<sub>2</sub>, C<sub>3</sub> and C<sub>4</sub> are neighbor of unit A</p>	 <p>The diagram shows a 5x5 grid. Unit A is in the center (row 3, column 3). Units C<sub>1</sub>, C<sub>2</sub>, C<sub>3</sub>, and C<sub>4</sub> are located at (row 2, column 2), (row 2, column 4), (row 4, column 2), and (row 4, column 4) respectively. All these units share a common vertex with unit A.</p>
<p><b>Queen contiguity:</b></p> <p>A spatial unit is a neighbor of another unit if both areas share a common edge or vertex. Unit B<sub>1</sub>, B<sub>2</sub>, B<sub>3</sub>, B<sub>4</sub>, C<sub>1</sub>, C<sub>2</sub>, C<sub>3</sub>, and C<sub>4</sub> are neighbor of unit A</p>	 <p>The diagram shows a 5x5 grid. Unit A is in the center (row 3, column 3). Units B<sub>1</sub>, B<sub>2</sub>, B<sub>3</sub>, and B<sub>4</sub> are at (row 3, column 2), (row 2, column 3), (row 3, column 4), and (row 4, column 3). Units C<sub>1</sub>, C<sub>2</sub>, C<sub>3</sub>, and C<sub>4</sub> are at (row 2, column 2), (row 2, column 4), (row 4, column 2), and (row 4, column 4). All these units share a common edge or vertex with unit A.</p>

# Spatial Weight Matrix

A common border (not vertex) is usually used to define the neighborhood in irregular grid.

Figure: Irregular arrangement of spatial unit



Contiguity matrix

$$W^*_{5 \times 5} = \begin{bmatrix} 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 \\ 1 & 1 & 0 & 1 & 1 \\ 0 & 1 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 & 0 \end{bmatrix}$$

# Spatial weight matrix

## Standardized Weight Matrix

### Row-standardization

For ease of interpretation, the weight matrix is often standardized such that the elements of a row sum to one. The elements of a row-standardized weights matrix thus equal to:

$$w_{ij} = \frac{w_{ij}^*}{\sum_{j=1}^n w_{ij}^*} \quad (2.5)$$

Standardized contiguity matrix for the irregular grid:

$$W = \begin{bmatrix} 0 & 1/2 & 1/2 & 0 & 0 \\ 1/3 & 0 & 1/3 & 1/3 & 0 \\ 1/4 & 1/4 & 0 & 1/4 & 1/4 \\ 0 & 1/3 & 1/3 & 0 & 1/3 \\ 0 & 0 & 1/2 & 1/2 & 0 \end{bmatrix}$$

# Spatial Weight Matrix

This ensure that all weight are between 0 and 1 and facilitates the interpretation of operations with the weight matrix as an averaging of the neighboring value. It also ensures that the spatial parameter in many spatial stochastic process are comparable between models. For a row-standardized weight matrix, the largest Eigen value is always +1, which facilitates the interpretation of the autoregressive coefficient as "Correlation"

**Effect of row-standardization with X is geo-referenced variable**

$$\begin{bmatrix} 0 & 1/2 & 1/2 & 0 & 0 \\ 1/3 & 0 & 1/3 & 1/3 & 0 \\ 1/4 & 1/4 & 0 & 1/4 & 1/4 \\ 0 & 1/3 & 1/3 & 0 & 1/3 \\ 0 & 0 & 1/2 & 1/2 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = \begin{bmatrix} (x_2 + x_3)/2 \\ (x_1 + x_3 + x_4)/3 \\ (x_1 + x_2 + x_4 + x_5)/4 \\ (x_2 + x_3 + x_5)/3 \\ (x_3 + x_5)/2 \end{bmatrix} = \begin{bmatrix} \bar{x}_1^N \\ \bar{x}_2^N \\ \bar{x}_3^N \\ \bar{x}_4^N \\ \bar{x}_5^N \end{bmatrix}$$

# Spatial weight matrix

## 2. Distance-based spatial weight matrix

It is assumed that spatial interaction will decline increasing distance due to increasing geographical impediments. Nearer regions have a greater potential influence.

### 1. Power function

$$w_{ij}^* = d_{ij}^{-\alpha} \quad (2.6)$$

Where

$\alpha$ : power parameter

$\alpha=1$ : inverse distance

$\alpha=2$ : quadratic inverse distance (gravity model of spatial interaction)

For spatial units outside a critical distance cut-off  $d_{\max}$  the weights may be set equal to 0.

The distance  $d_{ij}$  are usually measured between the centers of the region. Using the latitude and longitude coordinates the shortest distance between two centers are given by the great circle distance.

$$d = r * \arccos[\sin(lat_1) * \sin(lat_2) + \cos(lat_1) * \cos(lat_2) * \cos(lat_2 - lat_1)]$$

radius of the earth:  $r = 6387.7$  kilometers

# Spatial weight matrix

## 2. Negative exponential function

The weight formulation of negative exponential function is

$$w_{ij}^* = e^{-\beta d_{ij}} \quad (2.7)$$

$$\gamma \bar{d} = 1 - e^{-\beta \bar{d}} \quad (2.8)$$

$\bar{d}$  : average distance between immediate neighboring regions over the whole cross-section.

$\gamma$  : transformed distance decay parameter

It is assumed that spatial interaction such as commuting, migration or interregional trade is exposed to the frictional effects of geographical distance. With increasing distance from region  $i$  these geographical impediments gain in strength, so that the decline of spatial effects become more and more pronounced.

Distance corresponding to a decrease of spatial effects of  $\gamma 100\%$

$$d_\gamma = \frac{\ln(1 - \gamma)}{\beta}; \quad 0 < \gamma < 1 \quad (2.9)$$

# The Spatial Econometrics taxonomy

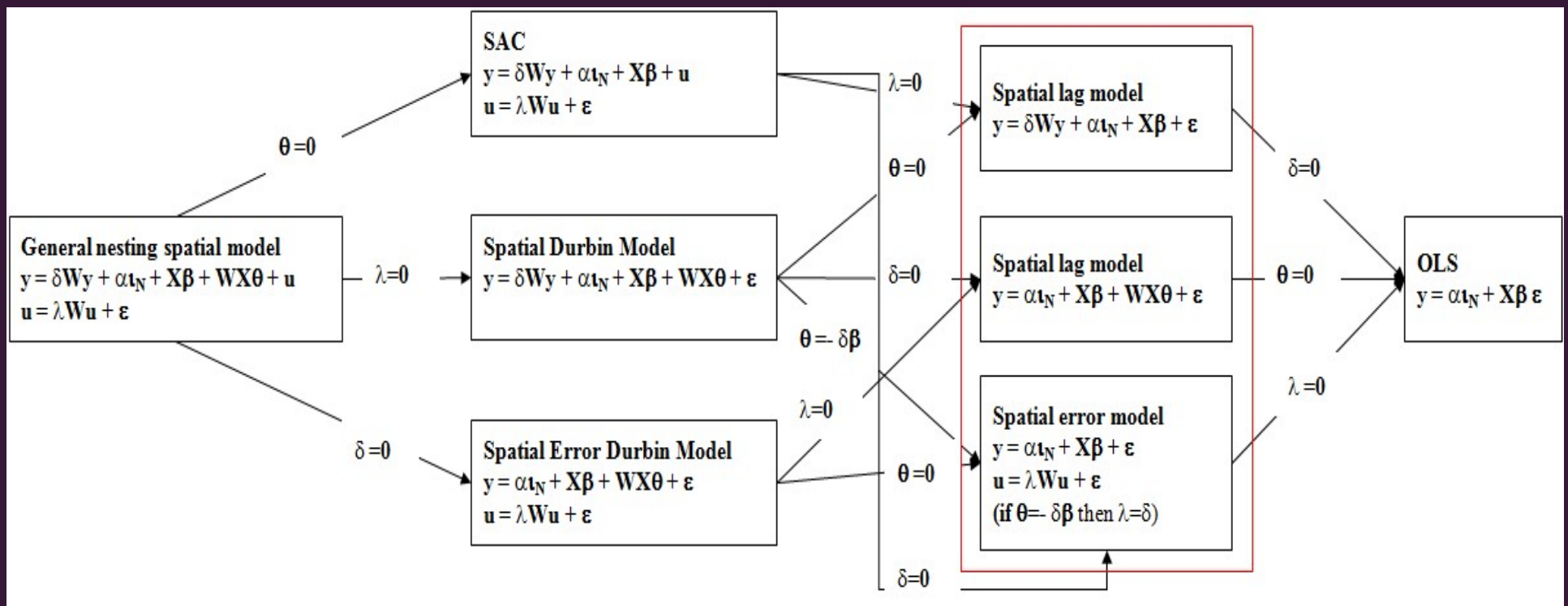


Figure 5.1 The relationships between different spatial dependence models for cross-section data (source Halleck Vega and Elhorst 2012)