
Spatial Weights

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Outline

- Connectivity in Space
- Spatial Weights
- Practical Issues
- Spatial Lag Operator

Spatial Connectivity

Why Spatial Weights?

- Spatial Correlation
 - $\text{Cov}[y_i, y_h] \neq 0$, for $i \neq h$
- Structure of Correlation
 - which i, h interact?
 - N observations to estimate $N(N-1)/2$ interactions
 - impose **structure** in terms of what are the “**neighbors**” for each location

Spatial Arrangement

- Need to Impose **Structure** on the Extent of Spatial Interaction
- Neighborhood View
 - define **neighborhood set** $N(i)$ for each location i
 - **spatial weights matrix**
- Pairs View
 - order pairs of locations $i-j$ in function of separating distance
 - **semivariogram** (geostatistics)

Neighborhood Set

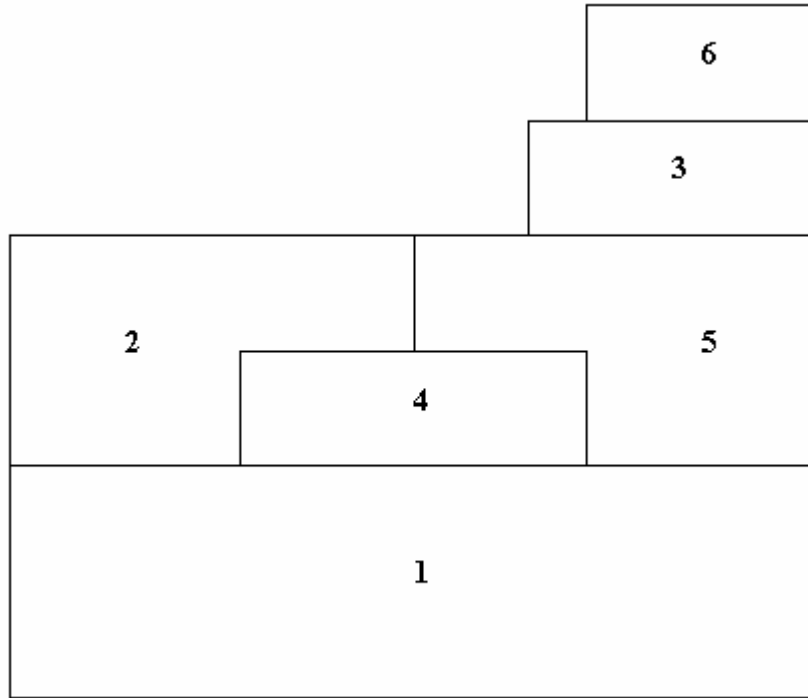
- Geographic-Cartographic Contiguity (GIS)
 - common boundary = contiguity
 - » common border, common vertex
 - distance band = isotropy
 - interaction border length and distance
- Spatial Interaction
 - distance decay, gravity, entropy
 - scale dependent, identification problems

Neighborhood Set (continued)

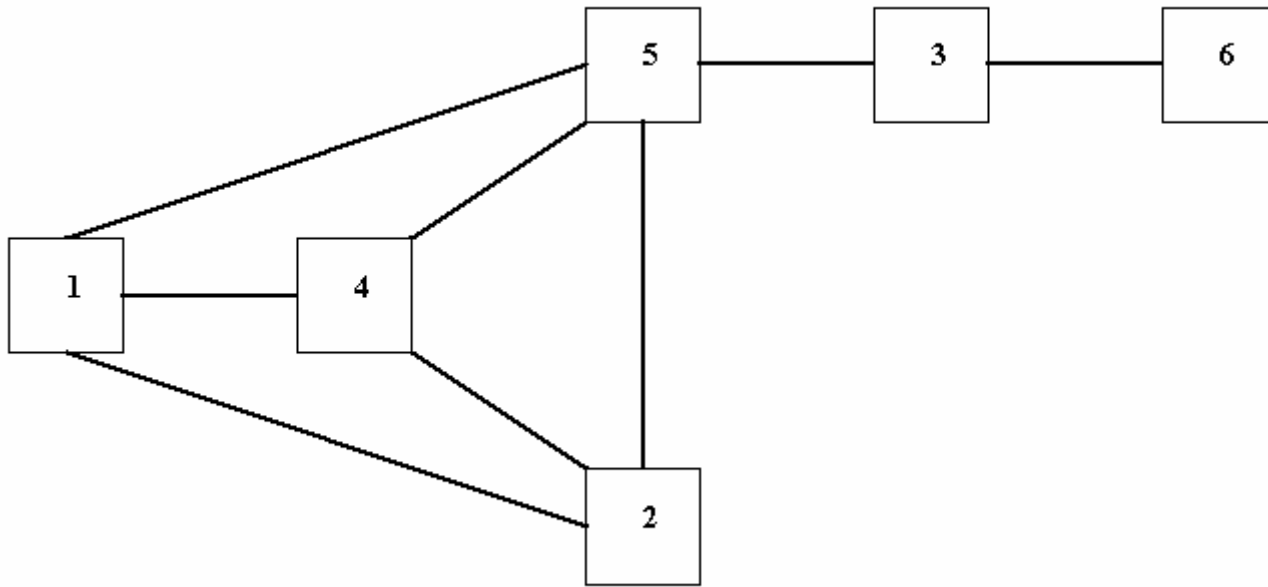
➤ Socio-Economic Distance

- multidimensional distance based on socio-economic indicators
 - » Euclidean, Mahalanobis
 - » example: income, ethnicity, industrial structure, trade flows, migration flows
- problem with endogeneity
 - » variables for distance same as in model
- zero distances
 - » $1/(z_i - z_j)$ when $z_i = z_j$

Spatial Weights



Example: $N=6$
contiguity = common boundary



contiguity as a graph
link between nodes = contiguity

0	1	0	1	1	0
1	0	0	1	1	0
0	0	0	0	1	1
1	1	0	0	1	0
1	1	1	1	0	0
0	0	1	0	0	0

First Order Contiguity Matrix

Spatial Weights Matrix

➤ Definition

- N by N **positive** matrix W , with elements w_{ij}

➤ Simplest Form: Binary Contiguity

» $w_{ij} = 1$ for i and j “neighbors”
(e.g. $d_{ij} < \text{critical distance}$)

» $w_{ij} = 0$ otherwise,

» $w_{ii} = 0$ by convention

How to Define Weights

- Contiguity
 - common boundary
- Distance
 - distance band
 - k-nearest neighbors
- General
 - social distance
 - complex distance decay functions

Contiguity – Regular Grid

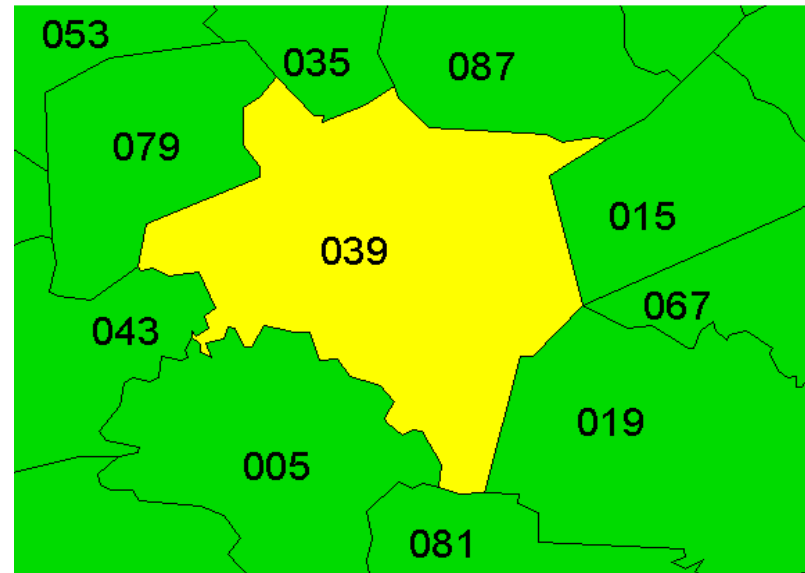
➤ Regular Grid

- rook
 - » 2, 4, 6, 8
- bishop
 - » 1, 3, 7, 9
- queen
 - » both

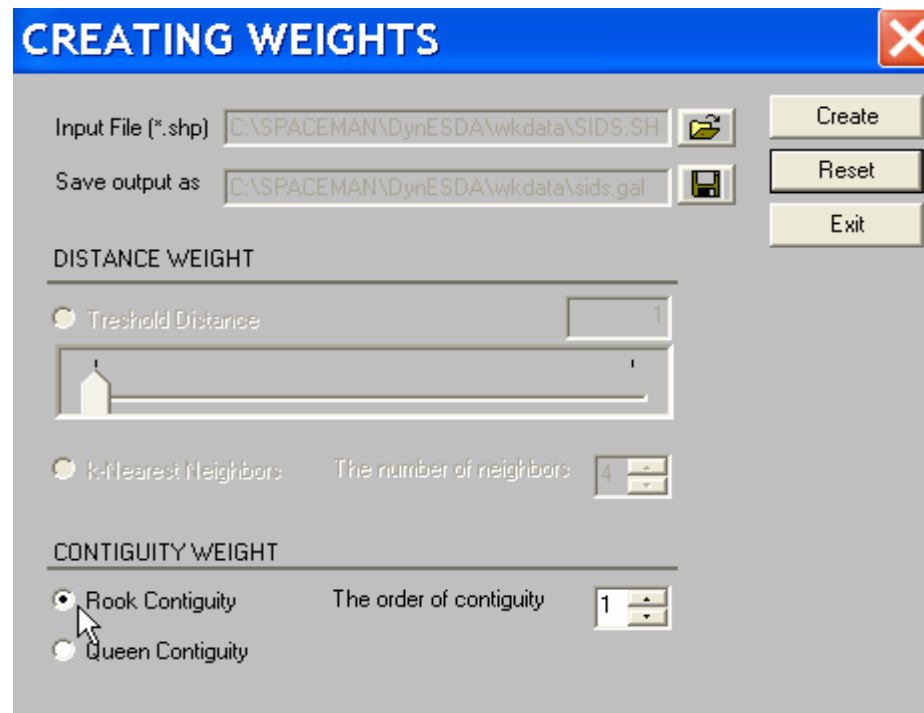
1	2	3
4	5	6
7	8	9

Contiguity – Irregular Units

- Irregular Units
 - common border
 - » rook
 - common vertex
 - » 039 and 067
 - » queen




Contiguity Weights in DynESDA2



Distance Based Weights

DISTANCE WEIGHT


Threshold Distance



A horizontal slider control with a yellow house-shaped knob positioned at approximately 15% of the total range. Above the slider are 15 vertical tick marks.


DISTANCE WEIGHT

Threshold Distance



A horizontal slider control with a yellow house-shaped knob positioned at the far left end of the range.

k-Nearest Neighbors The number of neighbors



A horizontal slider control with a yellow house-shaped knob positioned at the far left end of the range. A mouse cursor is visible over the right side of the slider.

General Spatial Weights

➤ Cliff-Ord Weights

- w_{ij} to reflect potential spatial interaction between i and j

- $w_{ij} = [d_{ij}]^{-a} \cdot [b_{ij}]^b$

» with

d_{ij} as distance between i and j

b_{ij} as share of common boundary between i and j in perimeter of i

General Spatial Weights

(continued)

- Weights May Contain Parameters
 - inverse distance weights
 - » $w_{ij} = 1 / d_{ij}^{\alpha}$
 - estimated from data or chosen a priori
 - » in practice: second power (gravity model)
 - identification problems in nonlinear weights
 - » interaction is multiplicative: $\rho \cdot w_{ij} = \rho (1 / d_{ij}^{\alpha})$
 - » parameters ρ and α not separately identified

General Spatial Weights

(continued)

- K-Nearest Neighbor Weights
 - k neighbors, irrespective of actual distance
 - “warps” space
- Economic Weights (Case)
 - block structure, state effect
 - » $w_{ij} = 1$ for all i, j in “block”
 - economic distance $|r_i - r_j|$, weight = $1/|r_i - r_j|$
 - » e.g., r = total employment

Row-Standardized Spatial Weights

➤ Motivation

- averaging of neighboring values
= form of **spatial smoothing**
- spatial parameters comparable

➤ $w_{ij}^s = w_{ij} / \sum_j w_{ij}$

➤ $\sum_j w_{ij}^s = 1$

Practical Issues

Characteristics of Spatial Weights

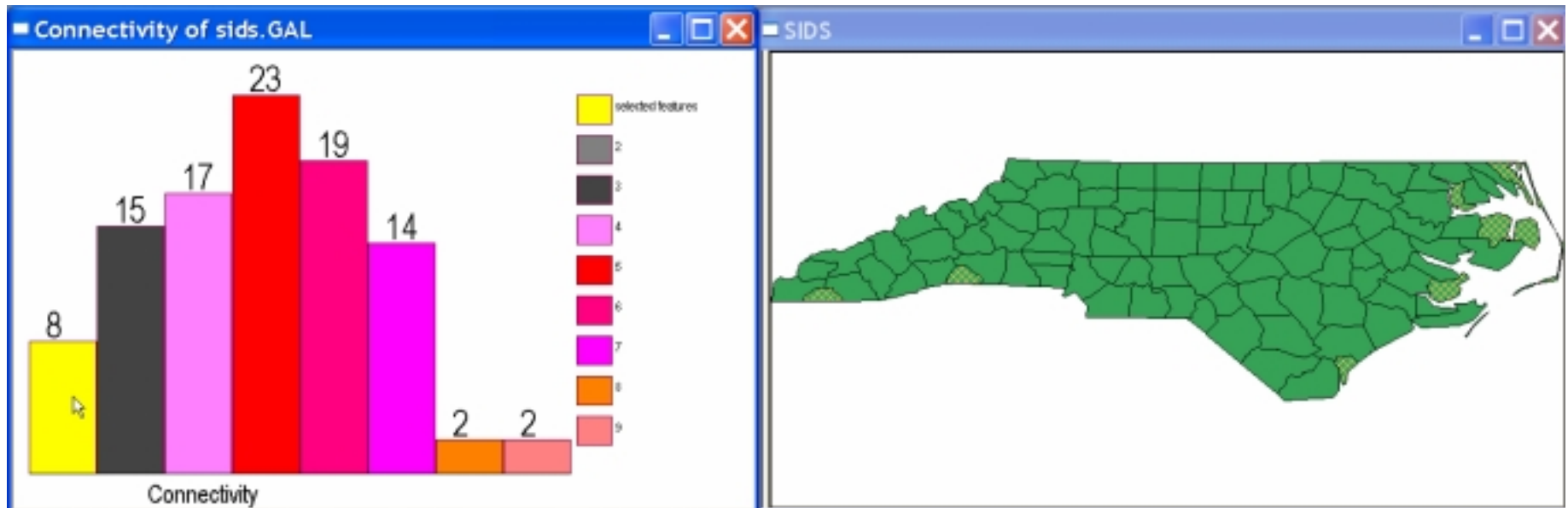
- Measures of Overall Connectedness
 - percent nonzero weights (sparseness)
 - average weight
 - average number of links
 - principal eigenvalue
- Location-Specific Measures
 - most/least connected observations
 - unconnected observations = islands

```
Connectivity characteristics of weights matrix wvhouse
Weights matrix stored in GAL format
Weights are row-standardized
Dimension:                55
# nonzero links:          248
% nonzero weights:        8.35017
Average weight:           0.221774
Average # links:          4.50909
With observations referred to by indicator variable FIPSNO
1 most connected observation(s) - with 8 link(s):
54039
2 least connected observation(s) - with 1 link(s):
54029 54037
```

```
Frequency Count for Weights File wvhouse
Connections      Frequency

1                2
2                6
3                7
4               10
5               14
6               10
7                5
8                1
```


Weights Characteristics in DynESDA2

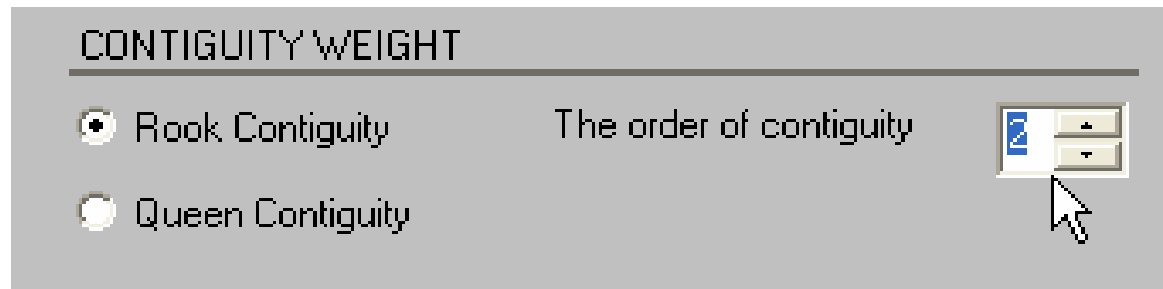


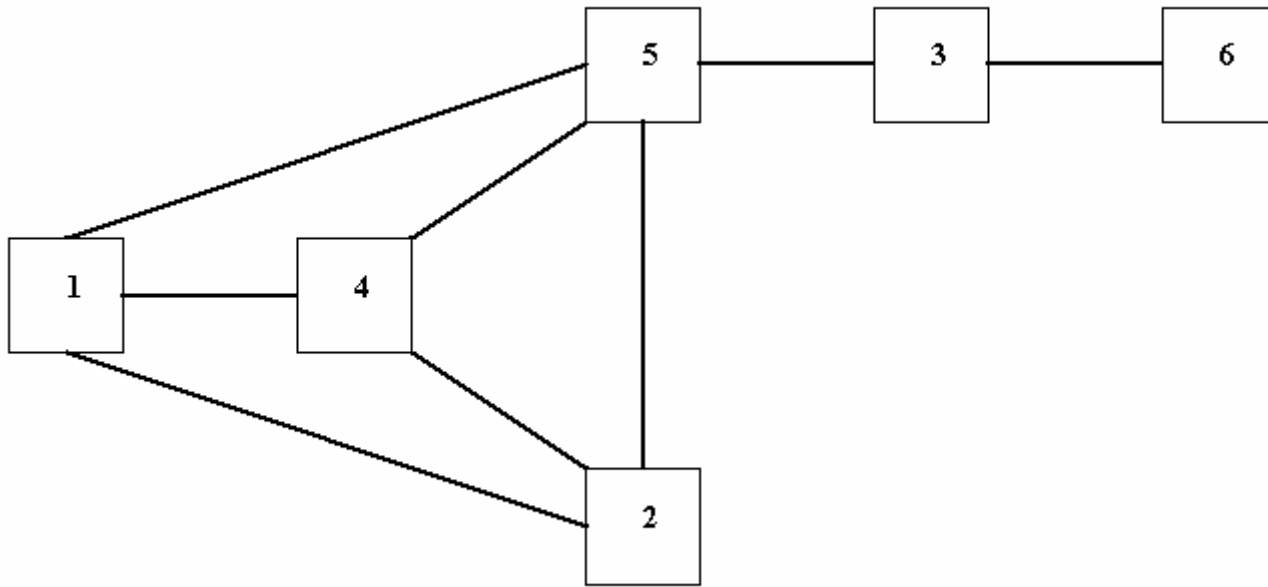
Higher Order Contiguity

➤ Recursive Definition

- contiguous of order k is **first order** contiguous to order $k-1$
 - » 2nd is first order contiguous to first

➤ Remove Circularity and Redundancy





contiguity as a graph
link between nodes = contiguity

Circularity and Redundancy in Higher Order Weights

- Powering of the Weights Matrix
 - standard approach invalid for contiguity
- Removing Circularity and Redundancy
 - sparse network representation of weights
 - modified Dijkstra algorithm to identify number of steps between nearest neighbors (Anselin and Smirnov)
 - number of steps = order of contiguity

0	1	0	1	1	0
1	0	0	1	1	0
0	0	0	0	1	1
1	1	0	0	1	0
1	1	1	1	0	0
0	0	1	0	0	0

First Order Contiguity Matrix

3	2	1	2	2	0
2	3	1	2	2	0
1	1	2	1	0	0
2	2	1	3	2	0
2	2	0	2	4	1
0	0	0	0	1	1

Second Power of First Order Contiguity

0	0	1	0	0	0
0	0	1	0	0	0
1	1	0	1	0	0
0	0	1	0	0	0
0	0	0	0	0	1
0	0	0	0	1	0

Correct Second Order Contiguity Matrix

-1	1	2	1	1	3
1	-1	2	1	1	3
2	2	-1	2	1	1
1	1	2	-1	1	3
1	1	1	1	-1	2
3	3	1	3	2	-1

Matrix Constructed by Bottom-Up Algorithm

Spatial Lag Operator

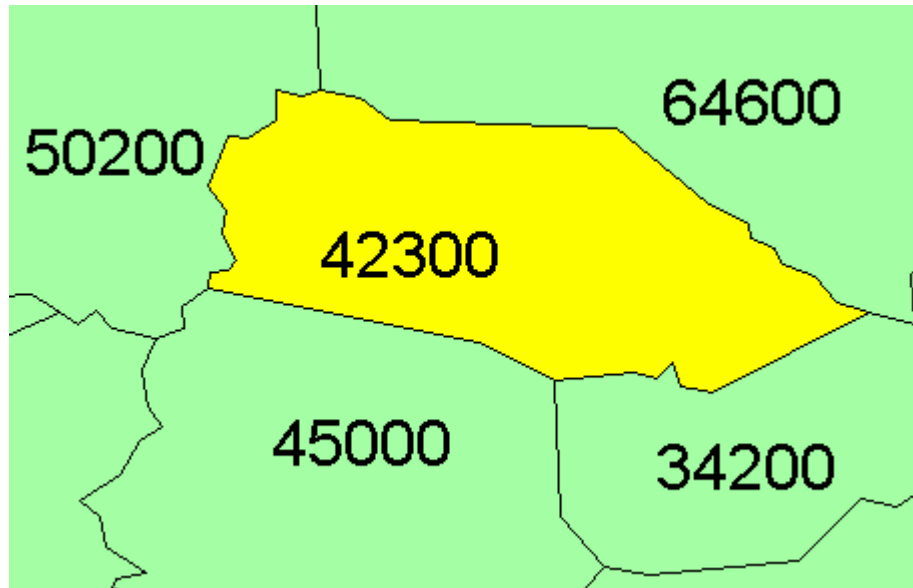
Spatial Shift

- No Direct Counterpart to Time Series Shift Operator
 - time series: $L_k = y_{t-k}$
 - spatial series: which h are shifted by “k” from location i?
 - » on regular lattice: east, west, north, south
 - » $(i - 1, j)$ $(i + 1, j)$ $(i, j - 1)$ $(i, j + 1)$
 - arbitrary for irregular lattice
 - » different number of neighbors by observation

Spatial Lag Operator

➤ Distributed Lag

- row-standardized weights $\sum_j w_{ij} = 1$
- spatial lag is weighted average of neighboring values
 - » $\sum_j w_{ij} \cdot y_j$, for each i
 - » vector Wy
 - » spatial lag does not contain y_i
- spatial lag is a smoother
 - » **not** a window average



value $y_i = \$42,300$

4 neighbors

values for neighbors: \$50,200,
\$64,600, \$45,000, \$34,200

spatial lag = $(1/4)\$50,200 +$
 $(1/4)\$64,600 + (1/4)\$45,000 +$
 $(1/4)\$34,200 = \$48,500$

Interpretation of Spatial Lag

- **Linear Association = Spatial Autocorrelation**
 - comparison of value of y at i to average of values at neighboring locations
 - » y_i and $(Wy)_i$ similar = **positive** spatial autocorrelation (**high-high, low-low**)
 - » y_i and $(Wy)_i$ dissimilar = **negative** spatial autocorrelation (**low-high, high-low**)