## Geographic Data Science -Lecture V

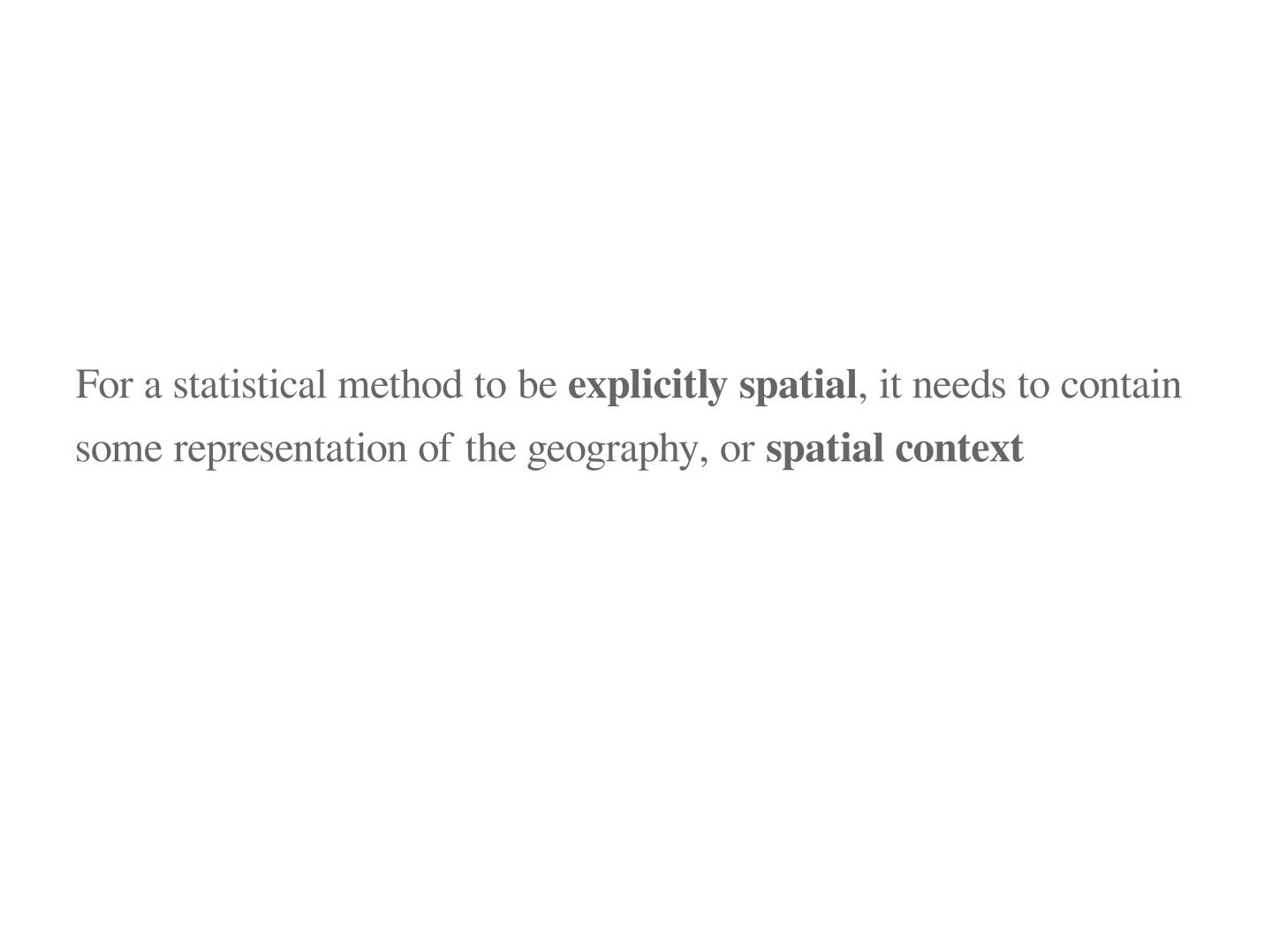
Space, formally

Dani Arribas-Bel

### Today

- The need to represent space formally
- Spatial weights matrices
  - What
  - Why
  - Types
- The spatial lag
- The Moran Plot

## Space, formally



For a statistical method to be **explicitly spatial**, it needs to contain some representation of the geography, or **spatial context**One of the most common ways is through **Spatial Weights Matrices** 

- (Geo)Visualization: translating numbers into a (visual) language that the human brain "speaks better"
- Spatial Weights Matrices: translating geography into a (numerical) language that a computer "speaks better".

Core element in several spatial analysis techniques:

- Spatial autocorrelation
- Spatial clustering / geodemographics
- Spatial regression

## W as a formal representation of space



N x N positive matrix

#### $\overline{W}$

N x N positive matrix that contains spatial relations

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N x N positive matrix that contains **spatial relations** between all the observations in the sample

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$$w_{ij} = \left\{egin{array}{ll} x > 0 & ext{if } i ext{ and } j ext{ are neighbors} \ 0 & ext{otherwise} \end{array}
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 $w_{ii} = 0$  by convention

#### W

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... What is a neighbor???

- Next door
- Close
- In the same "place" as us
- •

- Next door  $\rightarrow$  Contiguity-based Ws
- Close
- In the same "place" as us
- •

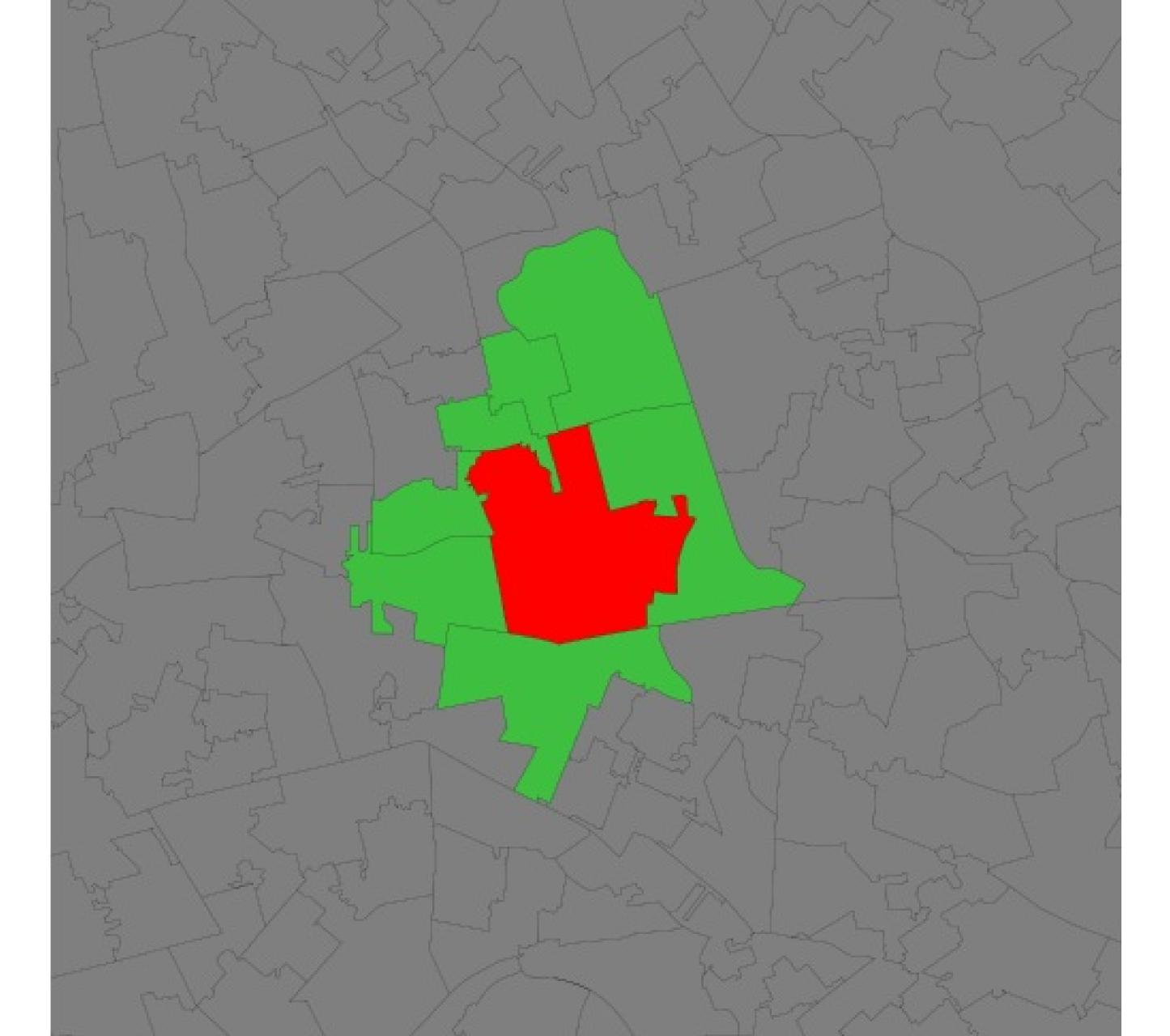
- Next door  $\rightarrow$  Contiguity-based Ws
- Close  $\rightarrow$  **Distance**-based Ws
- In the same "place" as us
- •

- Next door  $\rightarrow$  Contiguity-based Ws
- Close  $\rightarrow$  **Distance**-based Ws
- In the same "place" as us  $\rightarrow$  **Block** weights
- •

## Contiguity-based weights

Sharing boundaries to any extent

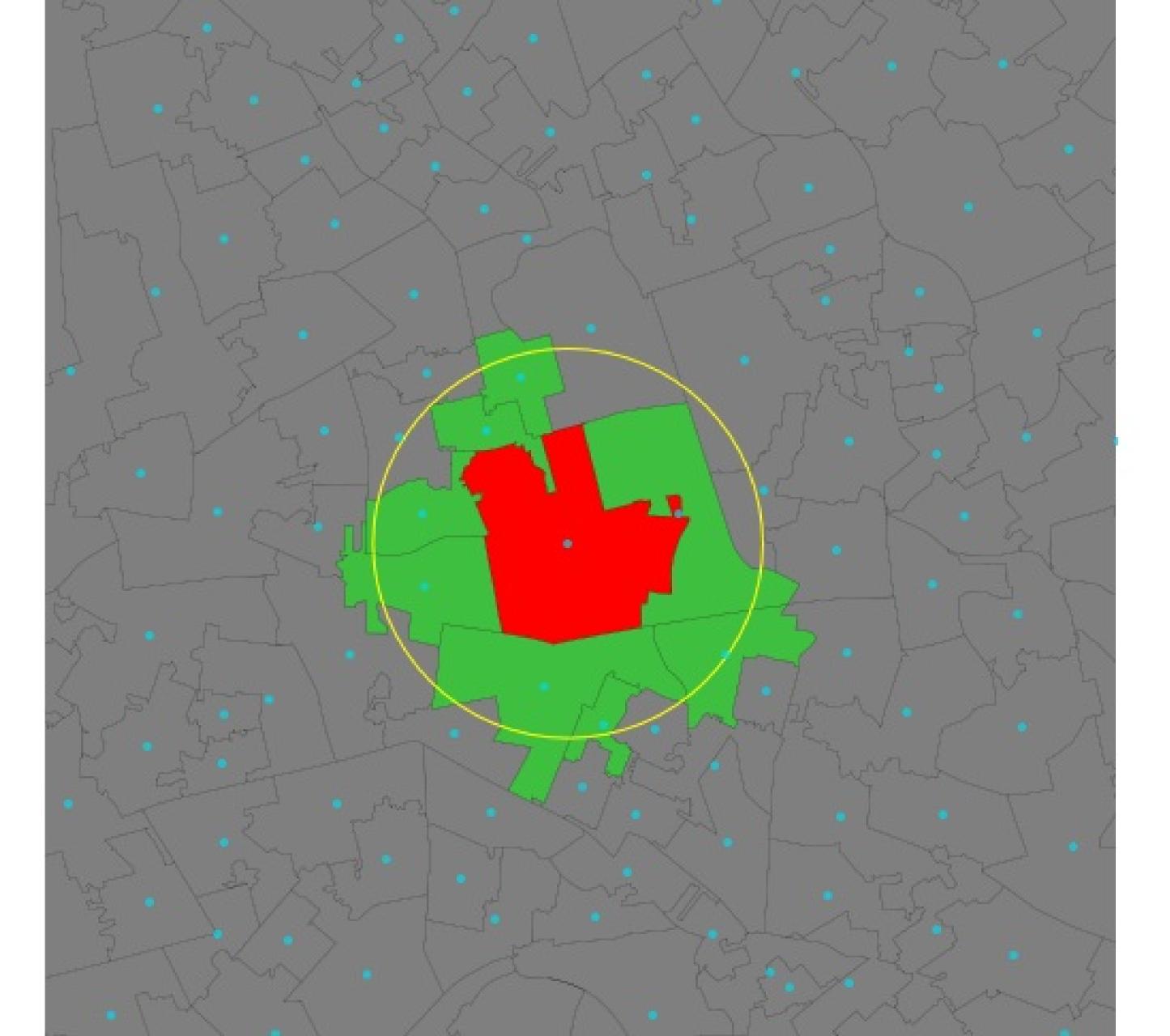
- Rook
- Queen
- •

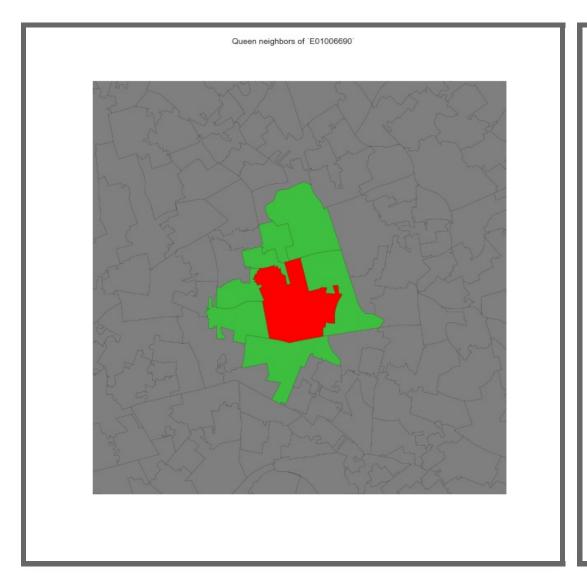


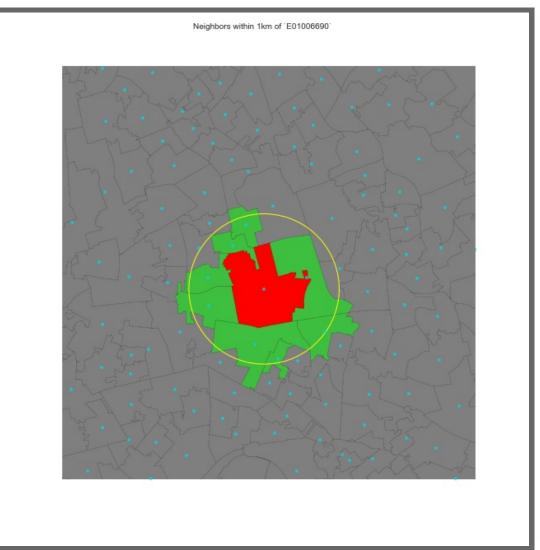
#### Distance-based weights

Weight is (inversely) proportional to distance between observations

- Inverse distance (threshold)
- KNN (fixed number of neighbors)
- •





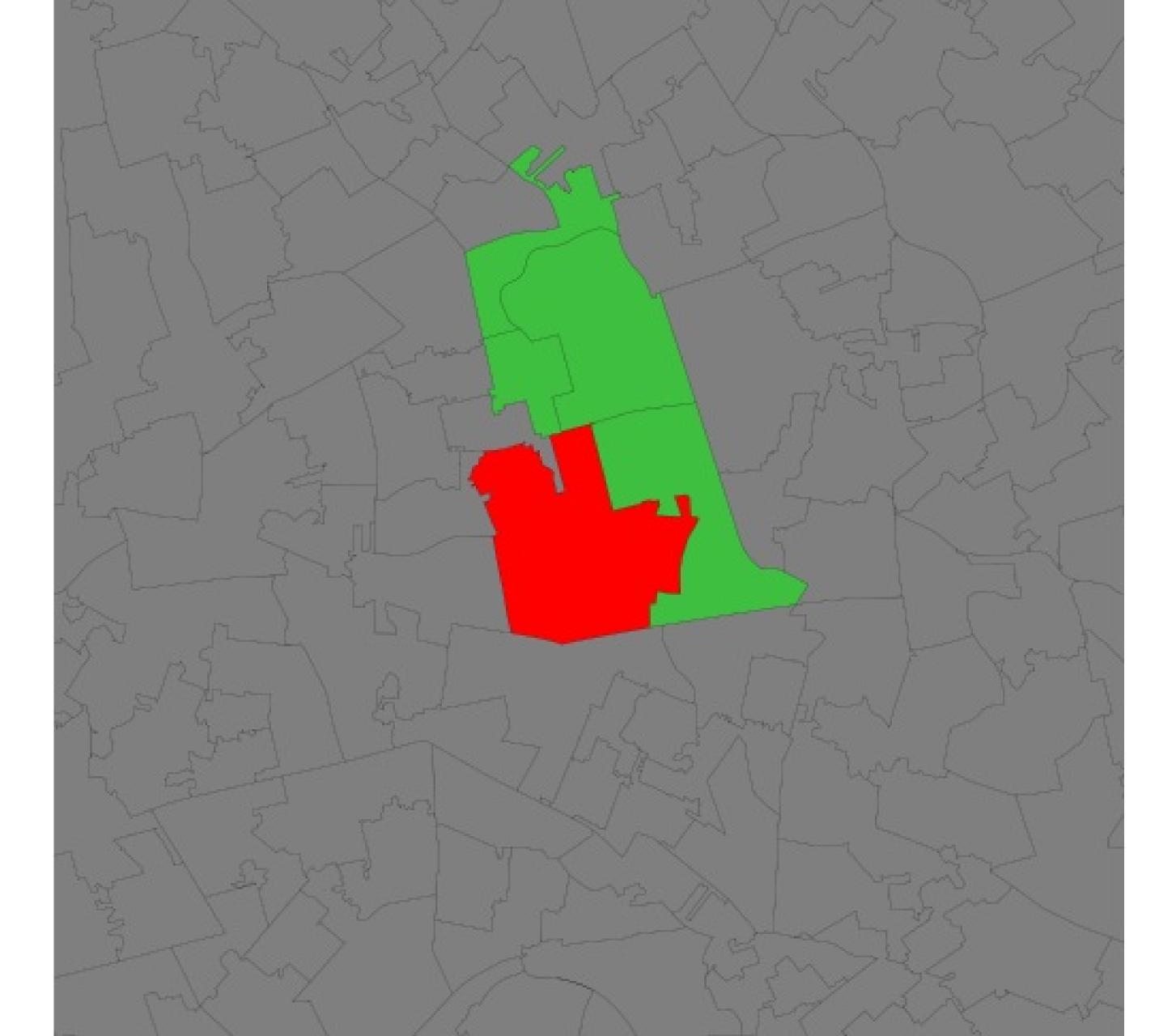


#### Block weights

Weights are assigned based on discretionary rules loosely related to geography

For example:

- LSOAs into MSOAs
- Post-codes within city boundaries
- Counties within states
- •



## Other types of weights

- Combinations of the above
- Kernel
- Statistically-derived
- ...

See Anselin & Rey (2014) for an in-detail discussion.

#### How much of a neighbor?

No neighbors receive zero weight:  $w_{ij} = 0$ 

Neighbors, it depends, wij can be:

- One  $w_{ij} = 1 \rightarrow \mathbf{Binary}$
- Some proportion  $(0 < w_{ij} < 1, \text{ continuous})$  which can be a function of:
  - Distance
  - Strength of interaction (e.g. commuting flows, trade, etc.)
  - **...**

#### Choice of W

Should be based on and reflect the underlying channels of interaction for the question at hand.

#### Examples:

- Processes propagated by inmediate contact (e.g. disease contagion) → Contiguity weights
- Accessibility → Distance weights
- Effects of county differences in laws  $\rightarrow$  Block weights

# Do your own (contiguity) weights time!

1	2	3
4	5	6
7	8	9

1	2	3		
4	5	6		
7	8	9		

1	2	3	4	5	6	7	8	9
0	1	0	1	0	0	0	0	0
1	0	1	0	1	0	0	0	0
0	1	0	0	0	1	0	0	0
1	0	0	0	1	0	1	0	0
0	1	0	1	0	1	0	1	0
0	0	1	0	1	0	0	0	1
0	0	0	1	0	0	0	1	0
0	0	0	0	1	0	1	0	1
0	0	0	0	0	1	0	1	0
	0 1 0 0 0	0 1 0 0 1 0 0 0 0 0 0 0	0 1 0 1 0 1 0 1 0 1 0 0 0 1 0 0 0 1 0 0 0	0     1     0     1       1     0     1     0       0     1     0     0       1     0     0     0       0     1     0     1       0     0     1     0       0     0     0     1       0     0     0     0       0     0     0     0	0     1     0     1     0       1     0     1     0     1       0     1     0     0     0       1     0     0     0     1       0     1     0     1     0       0     0     1     0     1       0     0     0     1     0       0     0     0     0     1       0     0     0     0     1	0     1     0     1     0     0       1     0     1     0     1     0       0     1     0     0     0     1       1     0     0     0     1     0       0     1     0     1     0     1       0     0     1     0     1     0       0     0     0     1     0     0       0     0     0     1     0     0       0     0     0     0     1     0	0     1     0     1     0     0       1     0     1     0     1     0     0       0     1     0     0     0     1     0       1     0     0     0     1     0     1       0     1     0     1     0     1     0       0     0     1     0     1     0     0       0     0     0     1     0     0     0       0     0     0     0     1     0     1       0     0     0     0     1     0     1	0     1     0     1     0     0     0     0       1     0     1     0     1     0     0     0       0     1     0     0     0     1     0     0       1     0     0     0     1     0     1     0       0     1     0     1     0     1     0     1       0     0     1     0     1     0     0     0       0     0     0     0     1     0     1     0       0     0     0     0     1     0     1     0       0     0     0     0     1     0     1     0

#### Standardization

In some applications (e.g. spatial autocorrelation) it is common to standardize W

The most widely used standardization is **row-based**: divide every element by the sum of the row:

$$ar{w_{ij}} = rac{w_{ij}}{w_{i.}}$$

where  $w_i$  is the sum of a row.

## The spatial lag

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Product of a spatial weights matrix W and a given variably Y

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Product of a spatial weights matrix W and a given variably Y

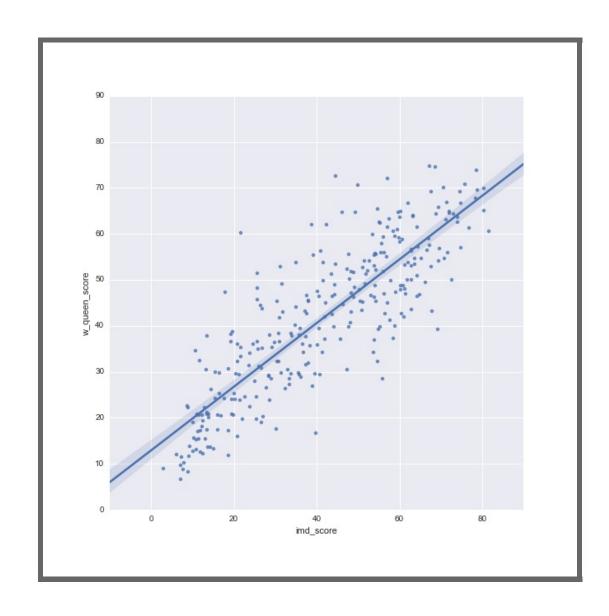
$$Y_{Sl} = WY$$

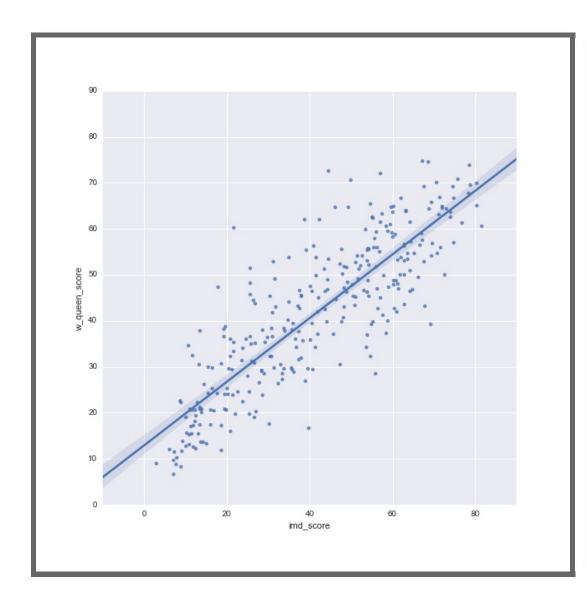
$$y_{Sl} - i = \sum_{j} w_{ij} y_{j}$$

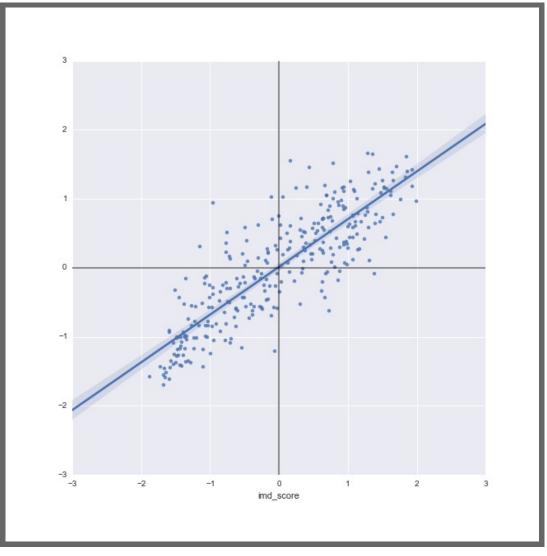
- Measure that captures the behaviour of a variable in the neighborhood of a given observation *i*.
- If W is **standardized**, the spatial lag is the *average value of the variable in the neighborhood*

- Common way to introduce space formally in a statistical framework
- Heavily used in both ESDA and spatial regression to delineate neighborhoods. Examples:
  - Moran's I
  - LISAs
  - Spatial models (lag, error...)

- Graphical device that displays a variable on the horizontal axis against its spatial lag on the vertical one
- Usually, variables are standardized  $(\frac{y-mean(y)}{std(y)})$ , which divides the space into **quadrants**
- Tool to start exploring spatial autocorrelation







## Recapitulation

#### Recapitulation

- Spatial Weights matrices: matrix encapsulation of space
- Different types for different cases
- Useful in many contexts, like the spatial lag and Moran plot, but also many other things!



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