



GE2215 Lecture 11 Spatial Statistical Analysis

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Recap: Local operations

- Local operations are cell-by-cell operations
- Local operations can be operated on single or multiple rasters
 - Local operations on a single raster
 - Reclassification: one-to-one mapping and one-to-many mapping
 - Local operations with multiple rasters
 - Mathematical operations: +, -, ×, \div
 - Summary statistics: maximum, minimum, range, sum, mean, median, and standard deviation, which apply to numeric data only
 - Tabulation of measures: majority, minority, combine and number of unique values, which apply to both numeric and categorical data



Recap: Focal operations

- A focal operation, also called a neighbourhood operation, involves a focal cell and a set of its surrounding cells.
- A neighbourhood is defined by shape and size
- Focal operations are usually performed on a single layer
 - Summary statistics: maximum, minimum, range, sum, mean, median, and standard deviation
 - Tabulation of measures: majority, minority, and variety



Recap: Zonal operations

- A zonal operation works with groups of cells of same values or like features.
 These groups are called zones
- Zones may be contiguous or non-contiguous





Recap: Global operations

- Global operations are similar to zonal operations whereby the entire raster dataset's extent represent a single zone
- Global operations apply a bulk change to all cells in a raster



Recap: Global operations

- Euclidean distance: calculating the closest distance away from a source
- Allocation: The cell value in an allocation raster corresponds to the closest source cell for the cell
- Direction: The cell value in a direction raster corresponds to the direction in degrees that the cell is from the closest source cell

| 1.0 | 2 | 1.0 | 2.0 | |
|-----|-----|-----|-----|--|
| 1.4 | 1.0 | 1.4 | 2.2 | |
| 1.0 | 1.4 | 2.2 | 2.8 | |
| 1 | 1.0 | 2.0 | 3.0 | |

| 2 | 2 | 2 | 2 |
|---|---|---|---|
| 2 | 2 | 2 | 2 |
| 1 | 1 | 1 | 2 |
| 1 | 1 | 1 | 1 |

| 90 | 2 | 270 | 270 | |
|-----|-----|-----|-----|--|
| 45 | 360 | 315 | 287 | |
| 180 | 225 | 243 | 315 | |
| 1 | 270 | 270 | 270 | |



Recap: Raster data clip

- There are two methods to clip a raster
 - Specify an analysis mask (clip by mask)
 - Specify the minimum and maximum x-, y-coordinates (clip by rectangle)









(a) Input raster (b) An analysis mask (c) Clip-by-mask result (e) Clip-by-rectangle result Slides for education purpose only



Recaps: Map algebra

- Map algebra basically involves doing math with maps
- The key difference with Algebra is that it only applies to rasters
- Map algebra can be local, focal, zonal and global
- Commonly used mathematical functions
 - Arithmetic operations
 - Statistical operations
 - Relational operations
 - Trigonometric operations
 - Exponential and logarithmic operations



Outline

- Basics of Statistics
 - Descriptive Standard Statistics
 - Inferential Standard Statistics
- Introducing Spatial Statistics
 - Descriptive Centrographic Statistics
 - Inferential Point Pattern Analysis

– Spatial Autocorrelation



Standard Statistical Analysis

- Two parts of Standard Statistics (A quick view)
 - **1. Descriptive statistics**
 - Obtain <u>summary</u> measures to <u>describe</u> a set of data
 - Three types of measures:
 - <u>Central tendency</u>: mean, median values
 - <u>Variability</u>: standard deviation
 - Frequency distributions



Normal Distribution



Uniform Distribution



Skewed Distribution



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Standard Statistical Analysis

- Two parts of Standard Statistics (A quick review)
 2. Inferential statistics
 - Make <u>inferences</u> from a sample about a population
 - Some definitions
 - **Population**: <u>all</u> occurrences
 - **Sample**: a <u>subset</u> of the population
 - Parameters
 - Statistics

An example: Are left-handed people more intelligent than right-handed people?





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 - Spatial Autocorrelation



What Is Spatial Statistics

- Definition of Spatial Statistics
 - The field of study concerning statistical methods that use <u>space</u> and <u>spatial</u> <u>data</u>
 - <u>Space</u> and <u>location</u> matters
- Spatial data
 - Point: bus stops, grocery stores
 - Polyline: transportation networks
 - Polygon: parks, campus
 - Continuous data: temperature, precipitation



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What Is Spatial Statistics

- Similarly, spatial statistics also have two parts:
 - **Descriptive** spatial statistics: Centrographic statistics
 - Inferential spatial statistics: e.g., Point Pattern Analysis, Spatial Autocorrelation





Positive spatial autocorrelation

Nearby cases are similar

| · · · · · · | | | |
|-------------|--|--|--|
| | | | |
| | | | |

No spatial autocorrelation

Neighbor cases have no particular relationship

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|---|--|--|--|
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| ſ | | | |
| | | | |
| | | | |

Negative spatial autocorrelation

Nearby cases are dissimilar



- Review (Refreshing your memory) of Statistics
 - Descriptive Standard Statistics
 - Inferential Standard Statistics

- Introducing Spatial Statistics
 - Descriptive Centrographic Statistics
 - Inferential Point Pattern Analysis
 - Spatial Autocorrelation



Descriptive Spatial Statistics: Centrographic Statistics

- Measures of Centrality
 - Centroid (Polygon data)
 - Weighted mean center





Descriptive Spatial Statistics: Centrographic **Statistics** A perfect circle here.

- Measures of Variance
 - Standard Distance
 - Measure the <u>degree</u> to which features are concentrated or dispersed around the geometric mean center
 - Does not capture the shape of distribution
 - Standard Deviational Ellipse (SDE)
 - Describe the <u>direction</u> of the dispersion
 - A typical method for measuring the activity space



https://gistbok.ucgis.org/bok-topics/point-pattern-analysis



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Inferential Spatial Statistics: Point Pattern Analysis

- What is the purpose of Point Pattern Analysis?
 - Test hypotheses about spatial patterns of points
- What do we mean by spatial patterns?
 - Mainly three types of spatial patterns
- Why spatial patterns matter?
 - Often, we cannot observe the <u>spatial process</u>, so we have to <u>infer</u> the process by observing the <u>spatial patterns</u>
 - Nonrandom \leftarrow Spatial processes
 - First order effect
 - Second order effect



Three types of spatial patterns



First Order Effect

- Diseases cluster (e.g., cancer, cold, coronavirus)
- First order effect describes Spatial Heterogeneity
 - Variation in the study area (uneven distribution)
 - The probability of a spatial phenomenon occurring at each location varies throughout a study region
 - Diseases cluster
 - Robberies cluster



Second Order Effect

- Diseases cluster (e.g., cancer, cold, coronavirus)
- Second order effect describes Spatial Interdependence
- The occurrence of an event at one location increases the probability of the same event occurrence at the neighboring location, sometimes resulting in clusters across a study region.
 - Interdependence between points themselves
 - Diseases cluster because people catch them from others who have the disease (e.g., cold)



Exploring the cause of a disease



- German launched 1,358 V-2 Rockets at London during World War II
- At first, its guidance systems were thought to be too primitive to hit specific targets
- However, according to the bomb damage maps, the impact sites seemed (visually) to be <u>clustered</u>
- If the rocket strikes were spatially clustered, the guidance systems must have been improved
- It was essential to assess if the impact sites were spatially clustered <u>quantitatively</u>



Distribution of V-2 Rocket Strikes on Central London, 1944

Exploratory Spatial Data Analysis (ESDA) method



- Approaches to point pattern analysis
 - <u>Density</u>-based approaches
 - <u>Distance</u>-based approaches
- Density-based approaches
 - Quadrat Analysis
 - Kernel Density Estimation

Quadrat Analysis

The biggest problem is selecting the suitable quadrat size (Modifiable Area

0 $^{\circ}$ Construct grids **Threshold values: Uniform** : Var/Mean ≈ 0 **Random** : Var/Mean ≈ 1 **Clustered** : Var/Mean >1

Determine spatial distribution

Calculate variance and mean, and create the variance-mean ratio

3

5

2

1

3

1

0

2

3

1

Variance

Var/Mean

Mean



- Approaches to point pattern analysis
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- Approaches to point pattern analysis
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- Approaches to point pattern analysis
 - <u>Density</u>-based approaches
 - <u>Distance</u>-based approaches
- Distance-based approaches
 - Nearest Neighbor Index (NNI)
 - G function
 - F function
 - Ripley's K function



- Distance-based approaches
 - Nearest Neighbor Index (NNI)
- Calculate NNI
 - 1. The mean of the distance <u>observed</u> between each point and its nearest neighbor
 - 2. The <u>expected</u> mean distance <u>if</u> the distribution is random

 $NNI = \frac{Observed mean distance}{Expected mean distance}$

Expected mean distance = $0.5 \sqrt{\frac{Numb}{Numb}}$

| | Point | Nearest Point | Distance |
|---|-------|------------------|----------|
| | 1 | 2 | 1 |
| | 2 | 3 | 0.1 |
| _ | 3 | 2 | 0.1 |
| ~ | ••• | ••• | |
| | 20 | 21 | 1 |
| | 21 | 19 | 0.4 |

Calculating the <u>nearest neighbor</u> distance

Threshold values:

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0

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° 0

Area

0

0

о

- Random pattern, NNI ≈ 1
- Clustered pattern, NNI ≈ 0
- Uniform pattern, NNI ≈ 2.149



- Outline
- Review (Refreshing your memory) of Statistics
 - Descriptive Standard Statistics
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Spatial Autocorrelation

- Spatial Autocorrelation (SA) is the most important concept in Spatial Statistics
- Definition of Spatial Autocorrelation
 - The <u>degree</u> to which characteristics at one location are similar (or dissimilar) to those nearby
- It is the confirmation of Tobler's first law of geography
 - Everything is related to everything else, but near things are more related than distant things





Spatial Autocorrelation

Differences and similarities between Point Pattern Analysis and SA

- > Point Pattern Analysis
 - ✓ Points only
 - ✓ Only location, no "magnitude" value



- > Spatial Autocorrelation
 - ✓ Points and polygons
 - ✓ Location and attribute variable

✤Income, GDP, rainfall, crime rate etc.





Spatial Autocorrelation

The name explains itself

Spatial:

Location-related,

On a map

Auto:

White: High value

Dark: Low value



Self (Interdependence)Positive Spatial AutocorrelationCorrelation:Similar values cluster together on a map

Degree of relative similarity

Source: Dr Dan Griffith

Negative Spatial Autocorrelation <u>Dissimilar</u> values cluster together on a map



Why Is Spatial Autocorrelation Important?

- Two reasons
- 1. It implies the existence of spatial process
 - Why are near-by things similar to each other?
 - Why do high-income people live "next door" to each other?
 - Why are rich cities surrounded by other rich cities?
 - All these **geographical** questions are related to **locations**, and can be explained by **Spatial Autocorrelation**

Other examples?

- 1. Nearby things are similar to each other
- 2. Nearby things are dissimilar to each other

Why Is Spatial Autocorrelation Important?

- Two reasons
- 1. Spatial Autocorrelation <u>invalidates many traditional statistical inference</u> <u>assumptions</u>
 - Traditional statistical inference often assumes that values of observations in each sample are **<u>independent</u>** of one another
 - However, if SA exists, the assumption is wrong
 - In this case, <u>spatial statistics</u> should be used



How Does SA Impact Standard Statistics?

• Correlation analysis between Income and Education by **Standard Statistics**



Sample collection locations





Measuring Spatial Autocorrelation

- Spatial Autocorrelation is measured based on the <u>Spatial Weight Matrix</u> (<u>SWM</u>)
- The <u>SWM</u> measures the relative location between each pair of points (or polygons)
- The <u>SWM</u> is a representation of the spatial structure of your data
- Two methods for measuring *Spatial Weight Matrix*
 - Weights based on <u>contiguity</u> **binary** (0, 1)
 - Weights based on <u>distance</u> **continuous** variable



Contiguity-based SWM

| | BUKIT BATOK | bukit Panjang | bukit Timah | CLEME NTI | JURONG EAST | JURONG WEST | QUEEN STOWN |
|------------------|----------------|------------------|----------------|--------------|----------------|----------------|----------------|
| BUKIT BATOK | * | | | | | | |
| BUKIT PANJANG | 1 | * | | | | | |
| BUKIT TIMAH | 1 | 1 | * | | | | |
| CLEMENTI | 1 | 0 | 1 | * | | | |
| JURONG EAST | 1 | 0 | 0 | 1 | * | | |
| JURONG WEST | 0 | 0 | 0 | 0 | 1 | * | |
| QUEENSTOWN | | 0 | 1 | 1 | 0 | 0 | * |



Spatial Weight

Matrix

If in adjacency relation 1, else 0

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Distance-based SWM

| | BUKIT BATOK | BUKIT PANJANG | bukit Timah | CLEMENTI | JURONG EAST | JURONG WEST | QUEENSTOWN |
|------------------|----------------|------------------|----------------|----------|----------------|----------------|------------|
| BUKIT BATOK | * | | | | | | |
| BUKIT PANJANG | 0.42 | * | | | | | |
| BUKIT TIMAH | 0.20 | 0.22 | * | | | | |
| CLEMENTI | 0.23 | 0.18 | 0.28 | * | | | |
| JURONG EAST | 0.22 | 0.15 | 0.16 | 0.34 | * | | |
| JURONG WEST | 0.17 | 0.12 | 0.10 | 0.15 | 0.25 | * | |
| QUEENSTOWN | 0.12 | 0.12 | 0.22 | 0.24 | 0.15 | 0 | * |



The weight between two polygons is inversely proportional to the distance between them



Global Measures of Spatial Autocorrelation

• Moran's I is the most common measure of Spatial Autocorrelation







Global Measures of Spatial Autocorrelation

• <u>Geary's C</u> is similar to Moran's I





- Comparison of Moran's I and Geary's C
 - \succ In general, they result in <u>similar</u>

conclusions

Moran's I is often preferred due to the

greater general stability



Hot Spots and Cold Spots

- Four types of <u>Spatial Clustering Patterns</u>
 - Positive Spatial Autocorrelation
 - High-High clustering (<u>High</u> values cluster together, **Hot Spots**)
 - Low-Low clustering (Low values cluster together, Cold Spots)
 - Negative Spatial Autocorrelation
 - Low-High clustering (Low values surrounded by high values)
 - High-Low clustering (<u>High</u> values surrounded by <u>low</u> values)



- Both Moran's I and Geary's C can <u>only distinguish</u> positive or negative spatial autocorrelation
- <u>Cannot tell</u> if these are hot spots or cold spots



Getis-Ord General G-Statistic

- The general G statistic can distinguish between hot spots and cold spots
 - G is relatively <u>large</u> \rightarrow potential <u>hot spots</u>
 - G is relatively $\underline{low} \rightarrow potential \underline{cold \ spots}$





Arthur Getis



J. Keith Ord

- General G will **NOT** show <u>negative</u> spatial autocorrelation
- General G can <u>only</u> use <u>distance-based</u> Spatial Weight Matrix
- General G should <u>only</u> be calculated for <u>ratio scale</u> data
 - It allows you to compare two values using ratio
 - E.g., crime rates & birth rates



Local Indicators of Spatial Association (LISA)

- <u>Global measures</u> of Spatial Autocorrelation are a measure for the entire area
- <u>Local measures</u> are needed when we want to look at the SA at <u>each</u> location or <u>observation unit</u>
- Why local measures matter?
 - Global measures indicate <u>No</u> SA, but <u>local</u> SA exists
 - <u>Positive</u> and <u>negative</u> SA exist <u>simultaneously</u> on the same map
 - The above issues are caused by <u>spatial</u> <u>heterogeneity</u>





Local Indicators of Spatial Association (LISA)

- LISA was proposed by Luc Anselin in 1995, therefore, LISA is often called Anselin's LISA
 - Luc Anselin 1995. Local Indicators of Spatial Association-LISA. Geographical Analysis 27: 93 -115



Luc Anselin

Local indicators of spatial association—LISA

L Anselin - Geographical analysis, 1995 - Wiley Online Library ... where A is a global **indicator** of **spatial association** and 7 is a scale factor. In other words, the sum of the **local indicators** is proportional to a global **indicator**. For the latter, a statement ... ☆ Save 奶 Cite Cited by 16906 Related articles All 9 versions ≫

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Home > Directory > Luc Anselin



Luc Anselin

Stein-Freiler Distinguished Service Professor of Sociology and the College; Director, Center for Spatial Data Science; Committee on Quantitative Methods in Social, Behavioral, and Health Science; Senior Fellow, NORC

Luc Anselin is a native of Belgium, where he did undergraduate work and a master's degree in economics and econometrics at the Free University of Brussels (VUB). His PhD is from Cornell University in the interdisciplinary field of Regional Science.

Anselin comes to the University of Chicago from Arizona State University where he was a Regents' Professor and held the Walter Isard Chair. He was the founding Director of the School of Geographical Sciences and Urban Planning. He also started and directed the GeoDa Center for Geospatial Analysis and Computation. Before ASU, he was a Professor at the University of Illinois, Urbana-Champaign, where he directed the

https://sociology.uchicago.edu/directory/Luc-Anselin



Identifying Hot and Cold Spots with LISA

- Global measures can <u>only</u> tell us there are <u>Hot or Cold</u> Spots in the whole area
- But LISA can tell us <u>where</u> the Hot and Cold Spots are



Cold Spots



Local Indicators of Spatial Association (LISA)

- Moran's I, Geary's C, and Getis-Ord General G are the global measures, and they all have their <u>local versions</u>
 - Local Moran's I (<u>https://pro.arcgis.com/en/pro-app/latest/tool-reference/spatial-statistics/cluster-and-outlier-analysis-anselin-local-moran-s.htm</u>)
 - Local Geary's C
 - Local Getis-Ord G (<u>https://pro.arcgis.com/en/pro-app/latest/tool-reference/spatial-statistics/hot-spot-analysis.htm</u>)



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ICOUNT

1.000 - 5.800

5.800 - 10.60

10.60 - 15.40

15.40 - 20.20

20.20 - 25.00

Demo: Identifying Hot Spots

• Exploring 911Call data using Hot Spot Analysis



Where are the Hot **Spots** of 911Calls

Distributions and numbers of 911Calls in Portland, Oregon



Demo: Identifying Hot Spots

Red: Hot Spots

Blue: Cold Spots

Use IDW to estimate values between points

Slides for education purpose only

Gi_Bin

- Cold Spot 99% Confidence
- Cold Spot 95% Confidence
- Cold Spot 90% Confidence
- Not Significant
- Hot Spot 90% Confidence
- Hot Spot 95% Confidence
- Hot Spot 99% Confidence



Demo: Identifying Hot Spots



Red indicates potential hot spots

Blue indicates potential cold spots

- What does the map imply?
- How can we apply the Hot Spot map?

Slides for education purpose only

IDW Interpolation



Demo: Identifying Hot Spots

Where should the Response Stations be located?



Please watch the full demo here: https://storymaps.arc gis.com/stories/4231 3404667341c3a6b33 ce5f73573e6

The relevant data and lab handout are available upon request



Software for Spatial Statistics

- ArcGIS (licensed)
- QGIS
- GeoDA (free)
 - Designed by Luc Anselin, Arizona University
- R software (free)
- Python
 - PySAL: Python Spatial Analysis Library





- Descriptive and Inferential Standard Statistics
- What is spatial statistics
 - Descriptive and Inferential Spatial Statistics
 - Spatial data: point, polyline, polygon and continuous data
- Point pattern analysis
 - Spatial patterns: Random, Uniform and Clustered distributions
 - Spatial processes: <u>First</u> and <u>second</u> order effects (Spatial heterogeneity and interdepedence)



- Spatial autocorrelation
 - Global measures: Global Moran's I, Geary's C, and Getis-Ord general G
 - Spatial Weight Matrix: contiguity-based and distance-based
 - Local Indicators of Spatial Association (LISA)
 - Identifying Hot Spots and Cold Spots



THANK YOU