

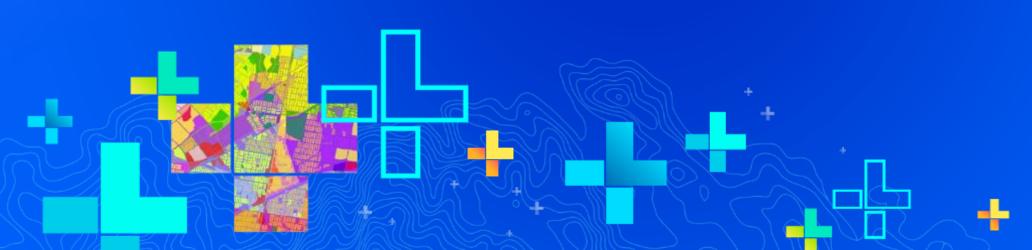
Spatial Data Mining I: Essentials of Cluster Analysis

Ankita Bakshi

Alberto Nieto

Flora Vale

esriurl.com/spatialstats

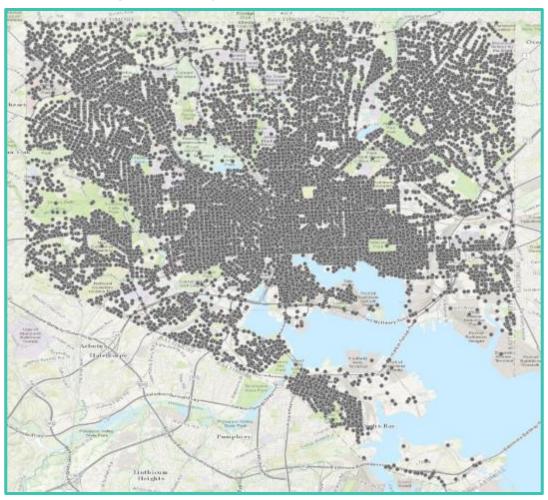


SEE WHAT OTHERS CAN'T

Subjectivity of Maps

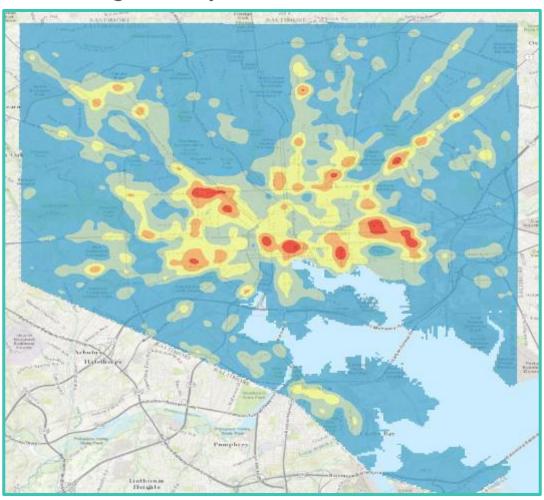
The map as data

High Priority 911 Calls in Baltimore



The map as data

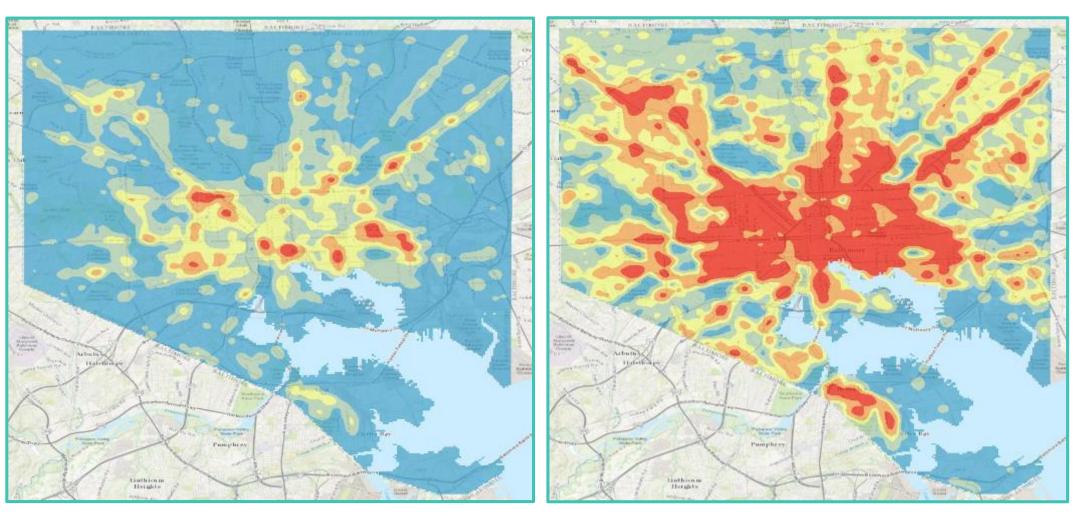
High Priority 911 Calls in Baltimore



Where are the hot spots? Where is the variation greater?

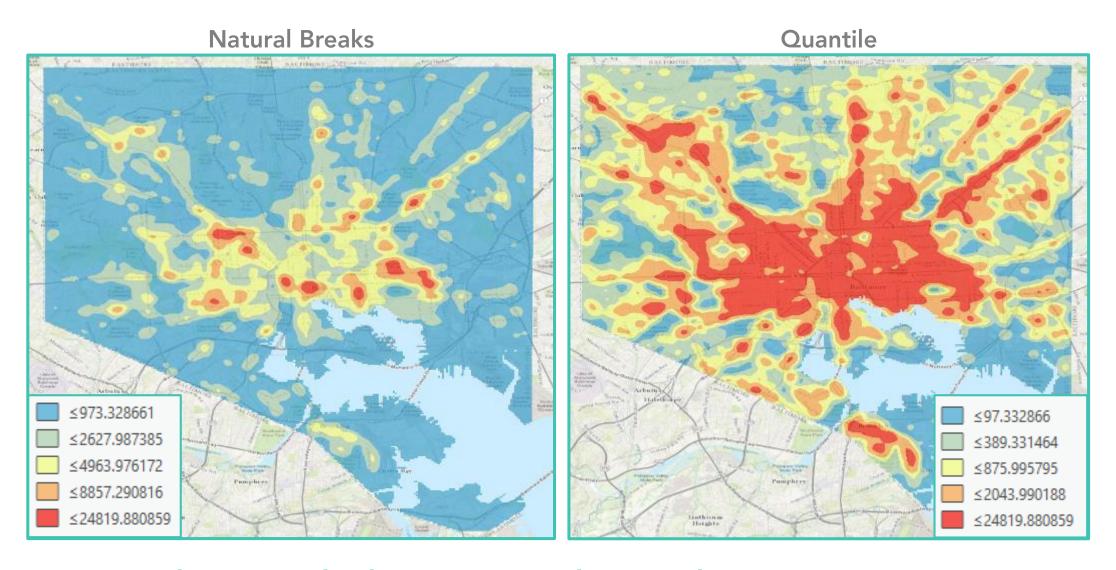
The map as data

High Priority 911 Calls in Baltimore



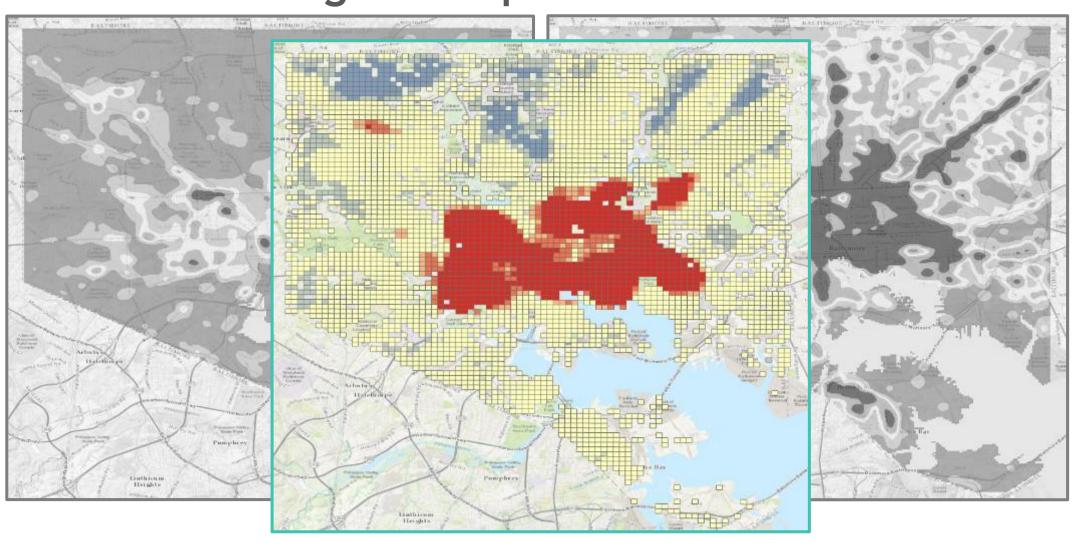
Where are the hot spots? Where is the variation greater?

The subjectivity of visual pattern analysis



Where are the hot spots? Where is the variation greater?

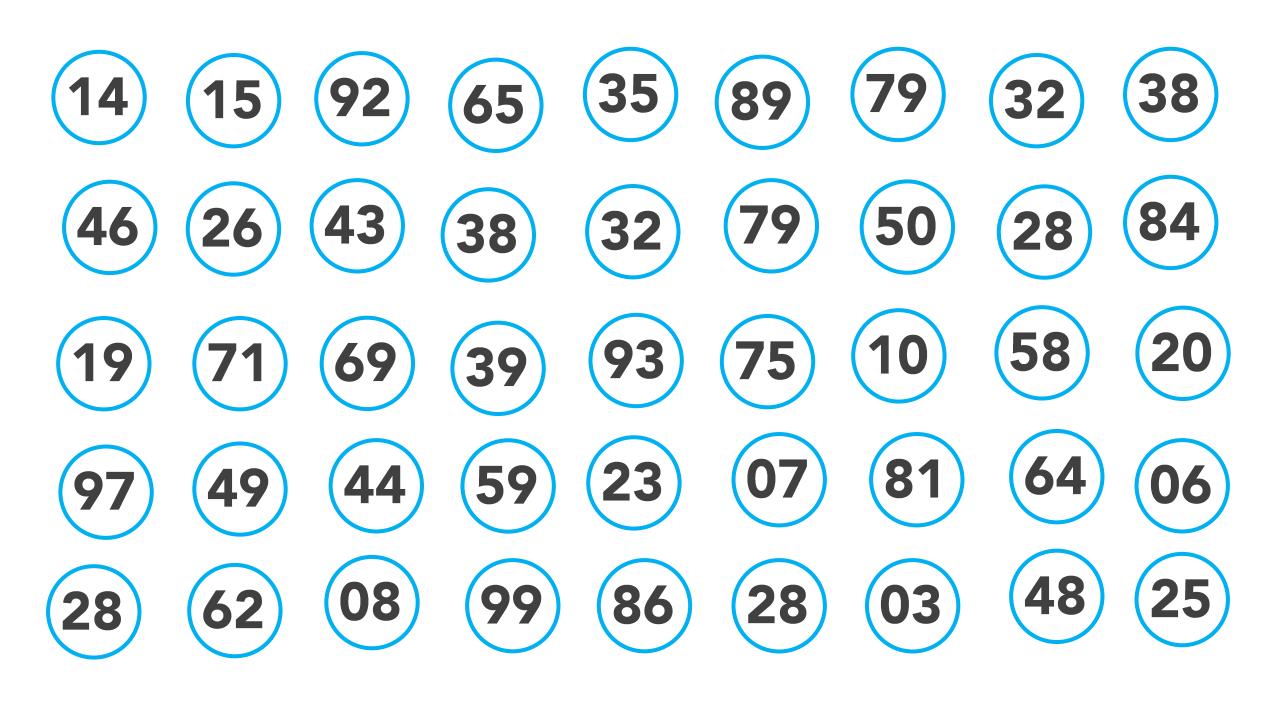
Minimizing the subjectivity Turning the map into information

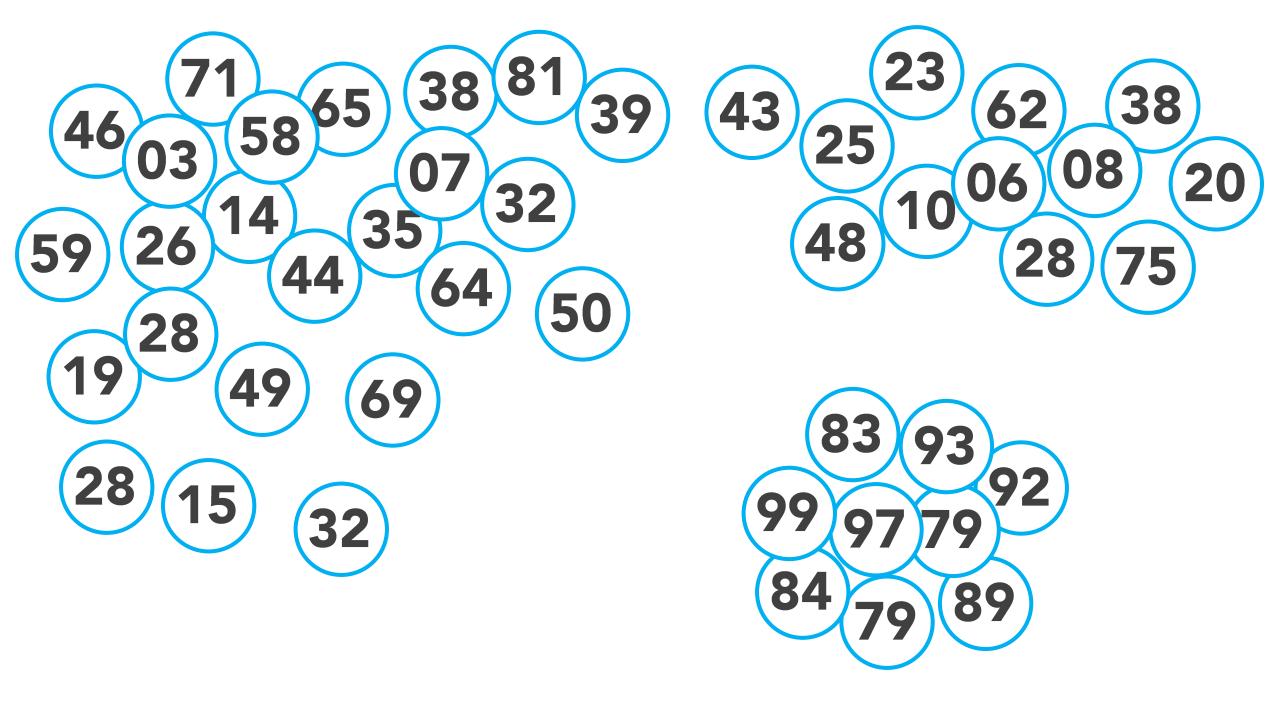


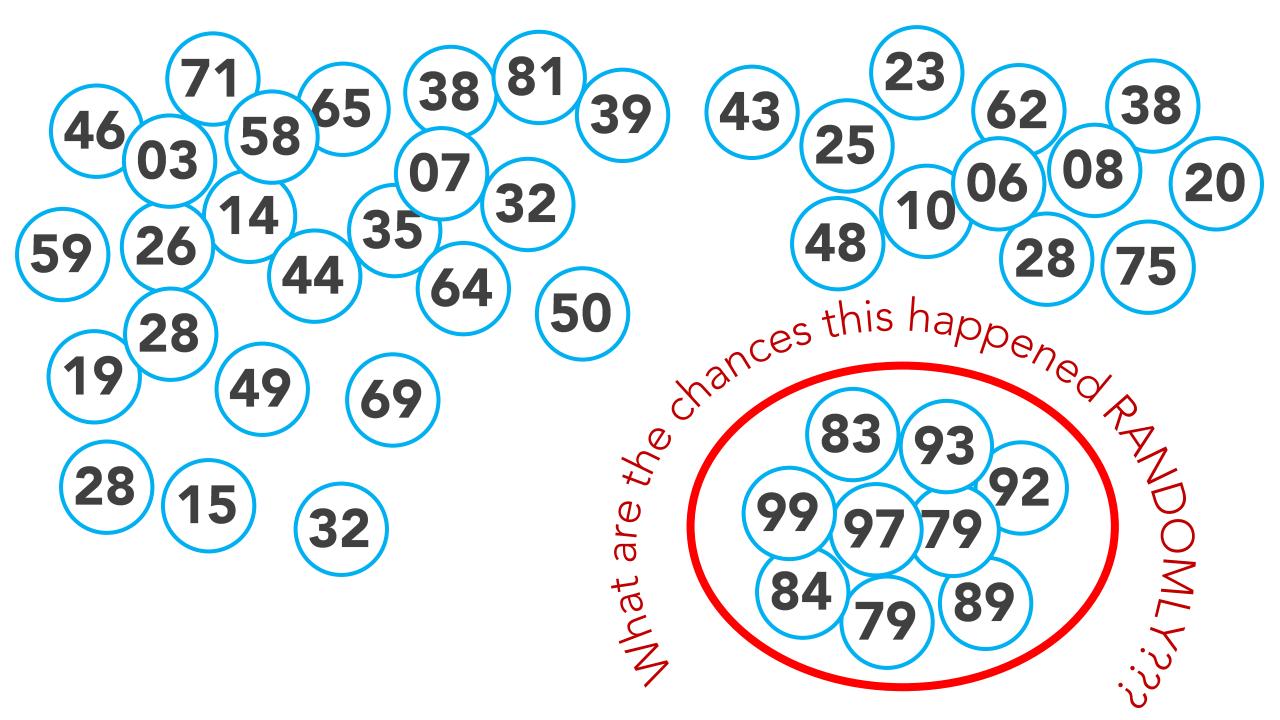
Inferential Statistics

Complete Spatial «RANDOMNESS

Is there a PATTERN?

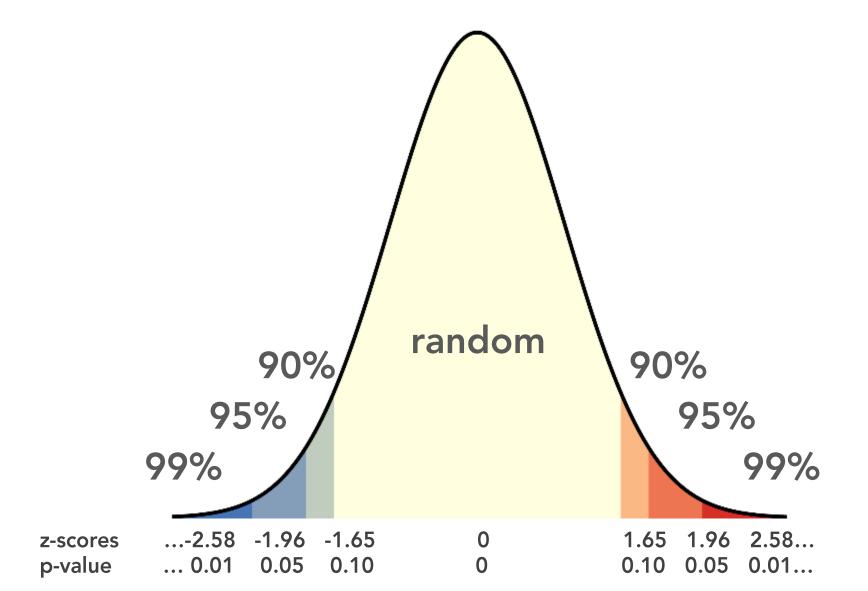


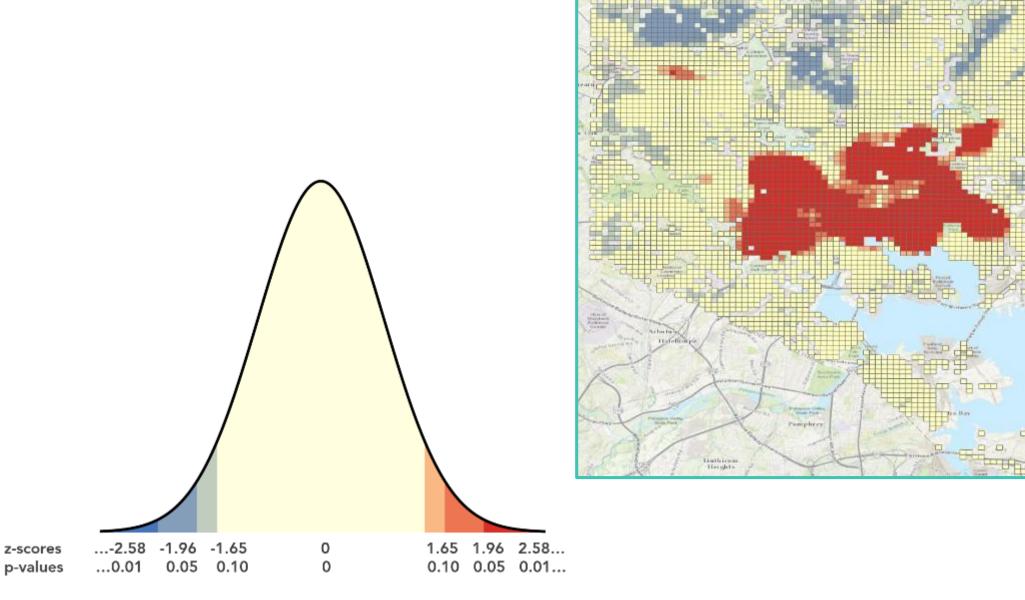




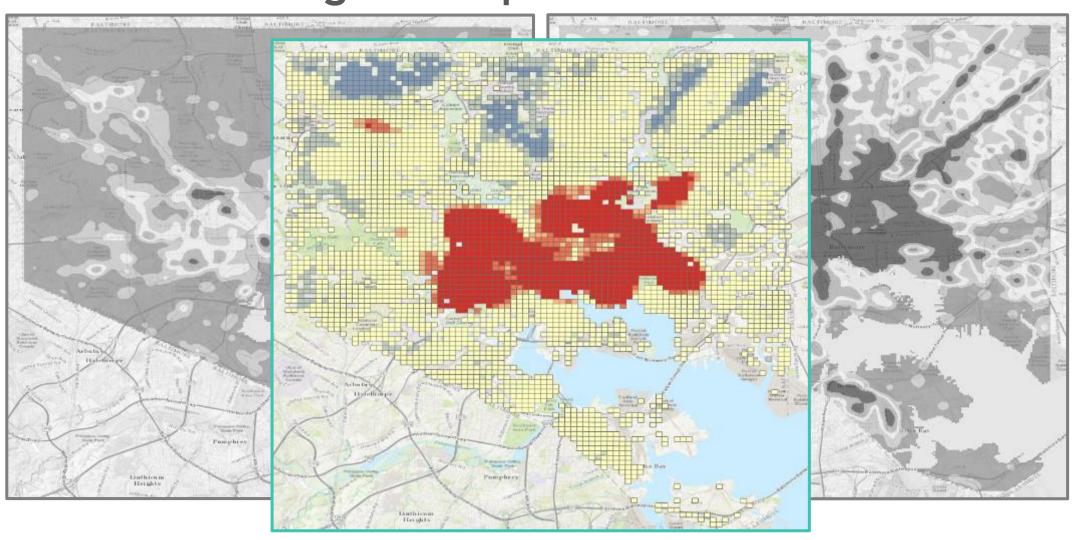
z-scores p-values

z-scores and p-values





Minimizing the subjectivity Turning the map into information

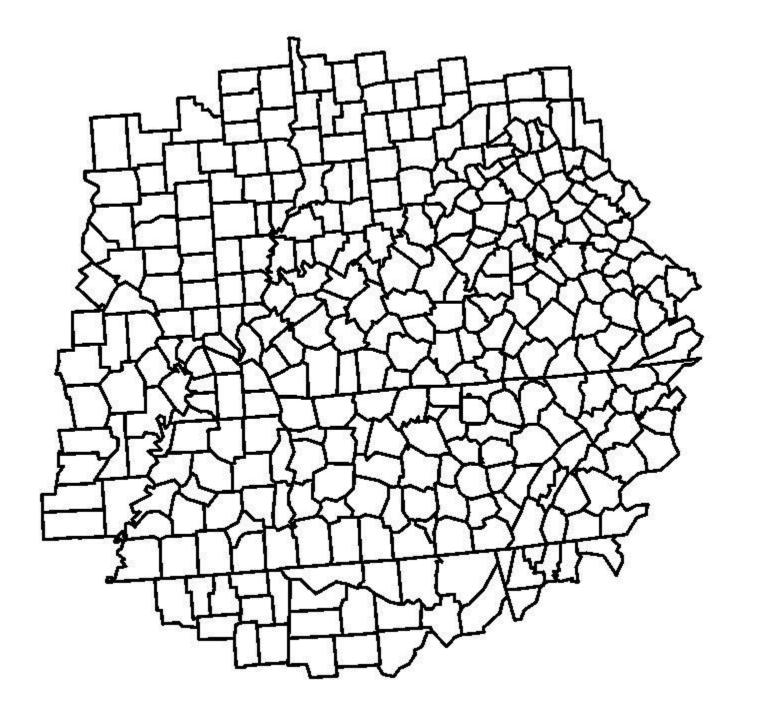


"...everything is related to everything else, but near things are more related than distant things."

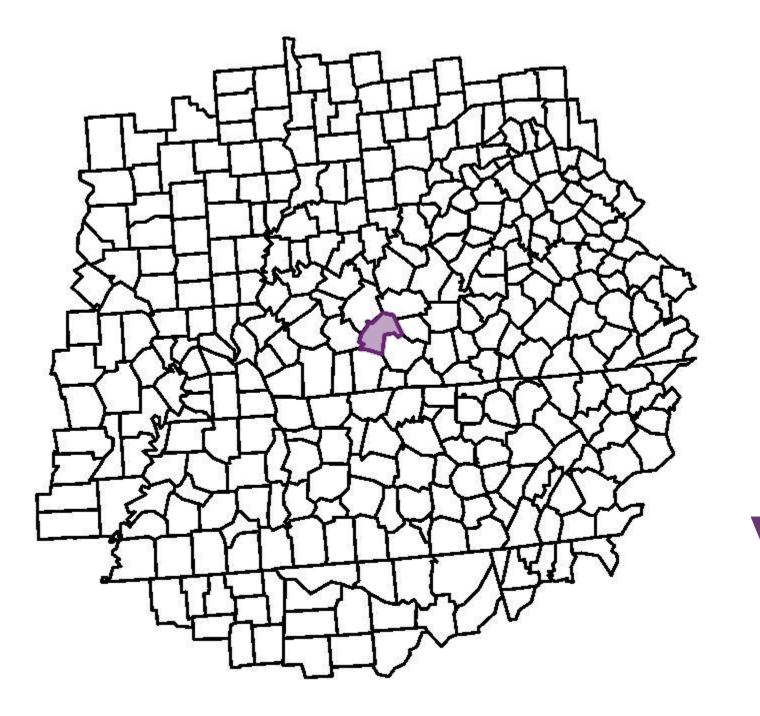
Hot Spot Analysis

given a set of weighted features, identifies statistically significant hot spots and cold spots using the Getis-Ord Gi* statistic

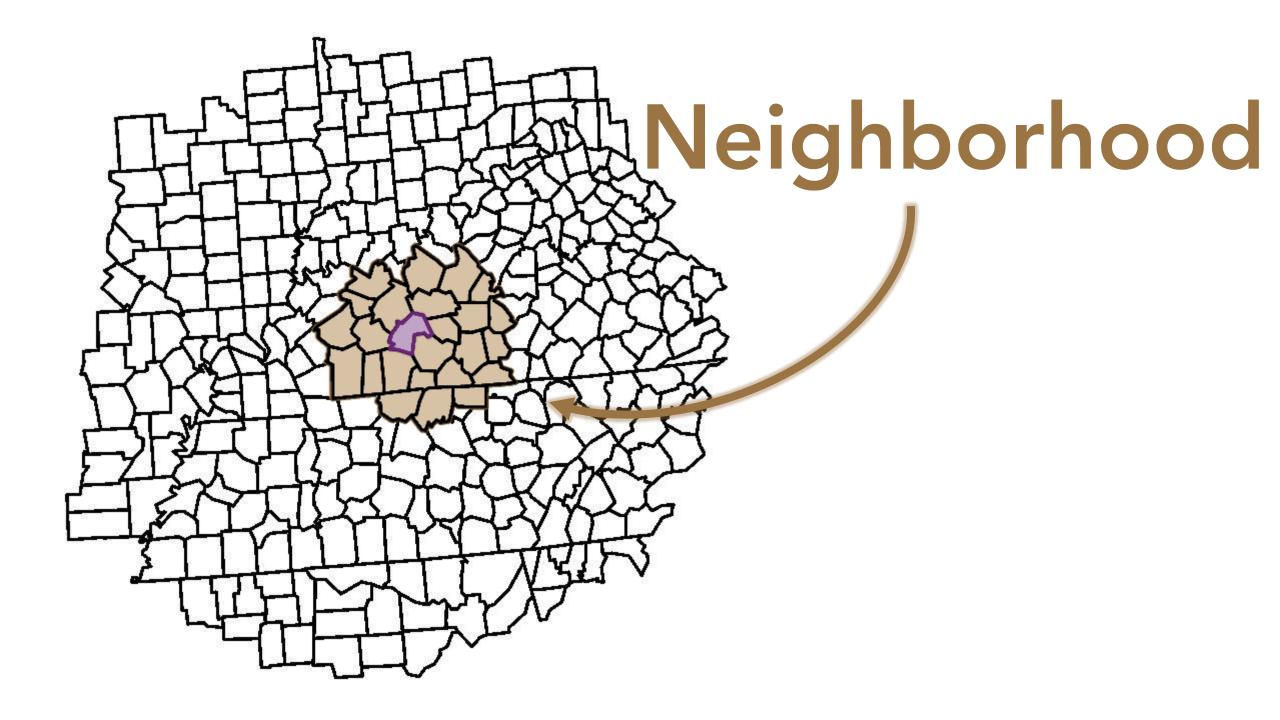
Polygons

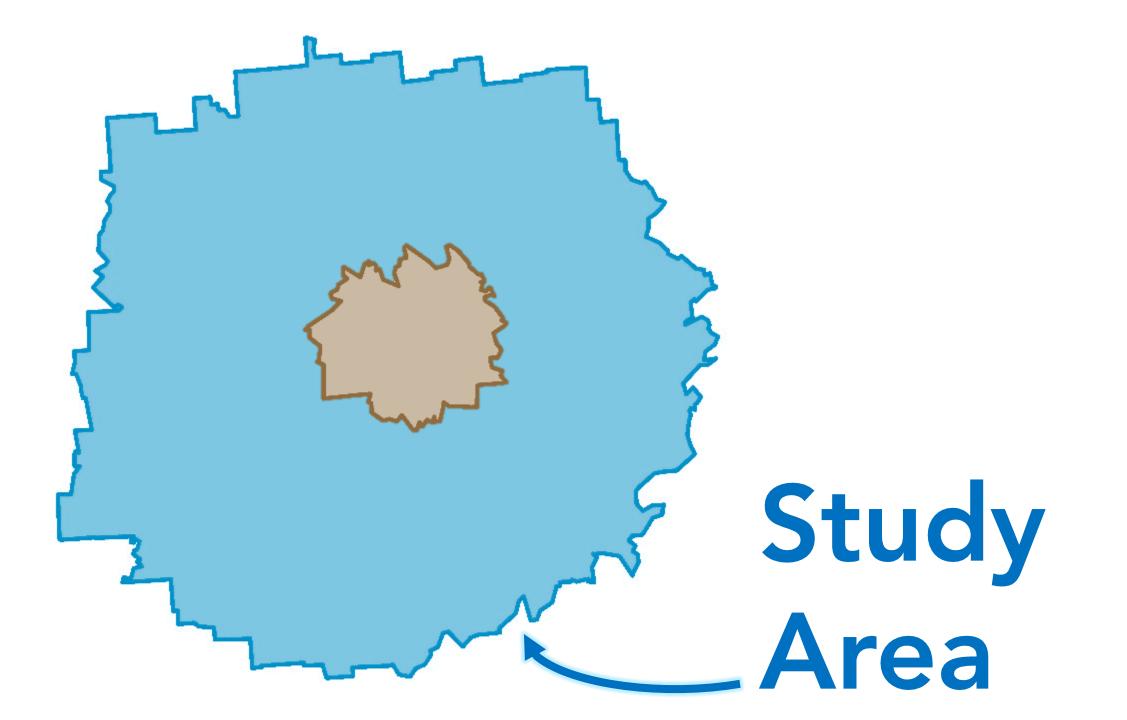


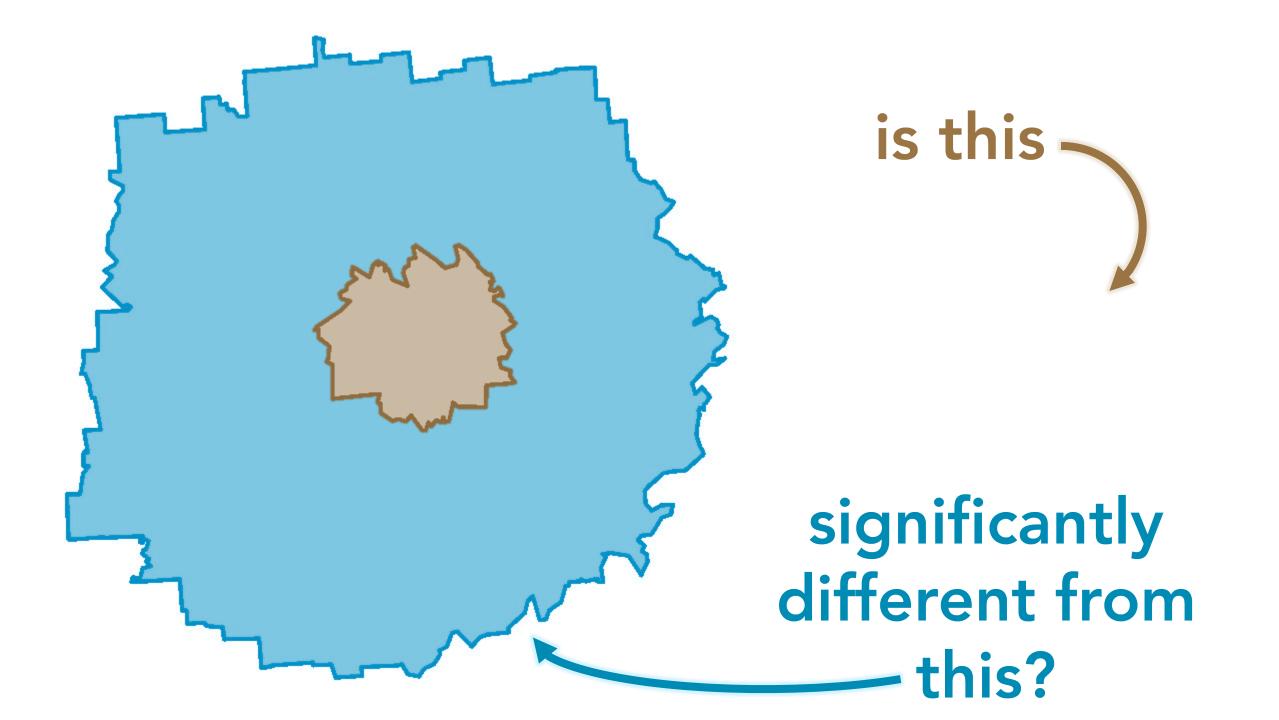


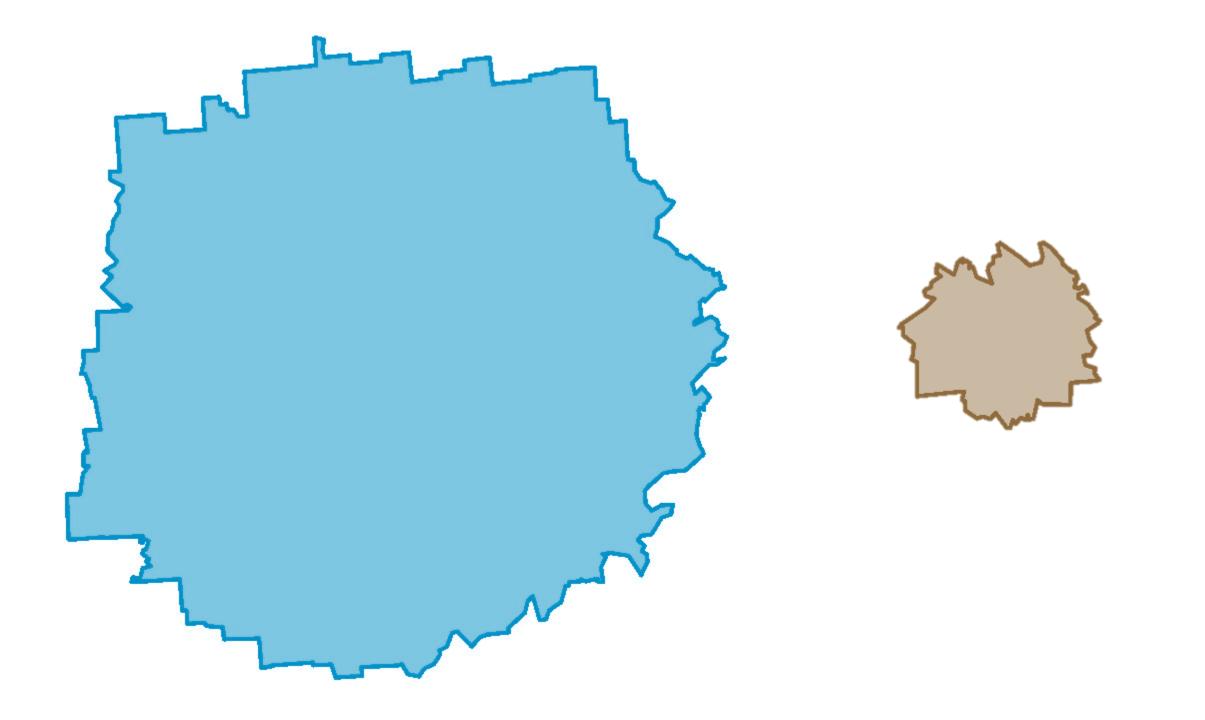


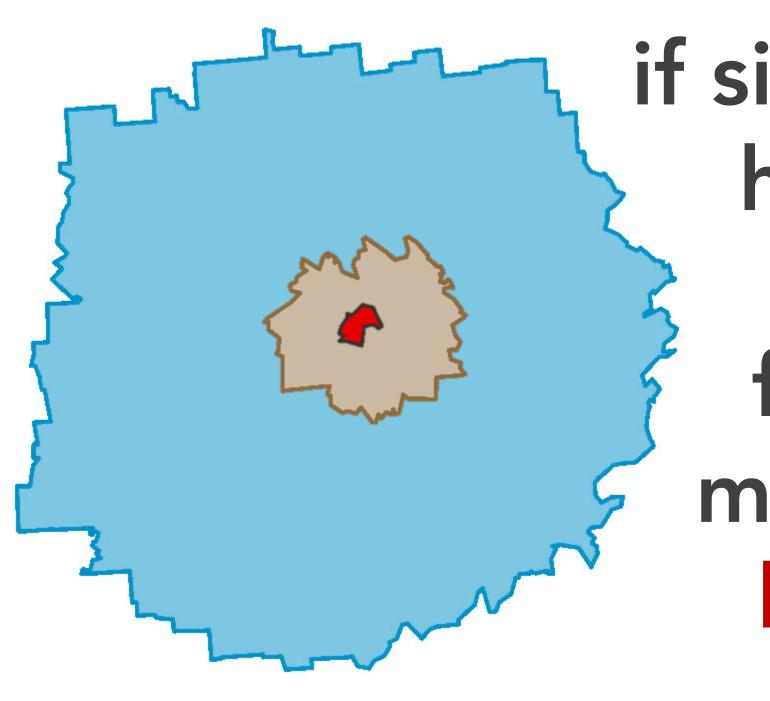
each feature has a value





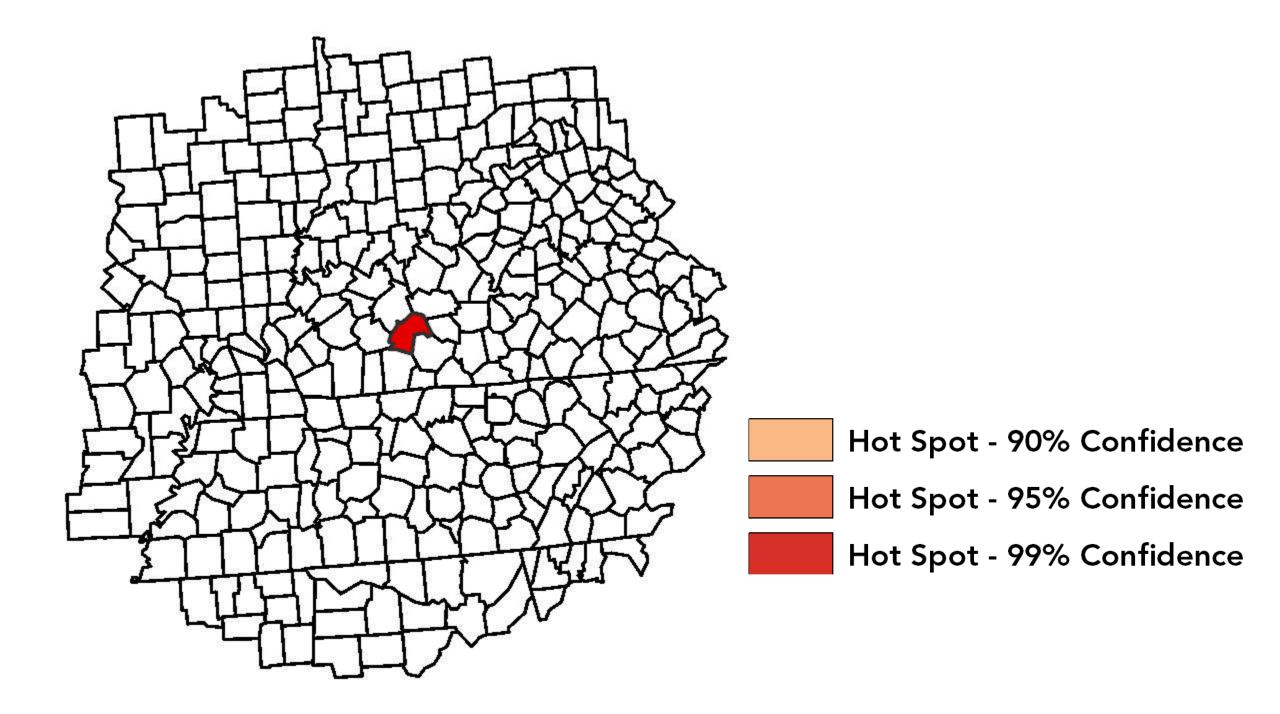


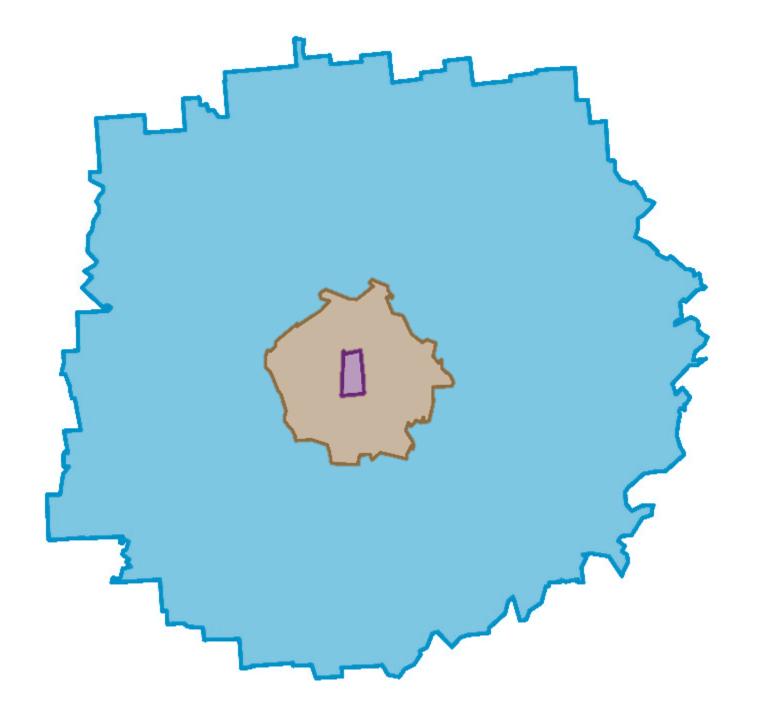


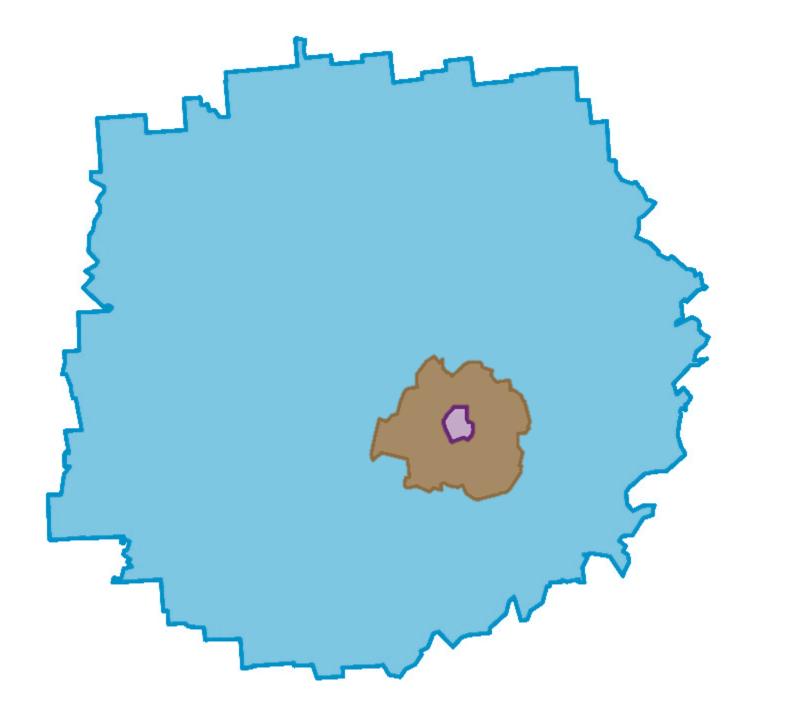


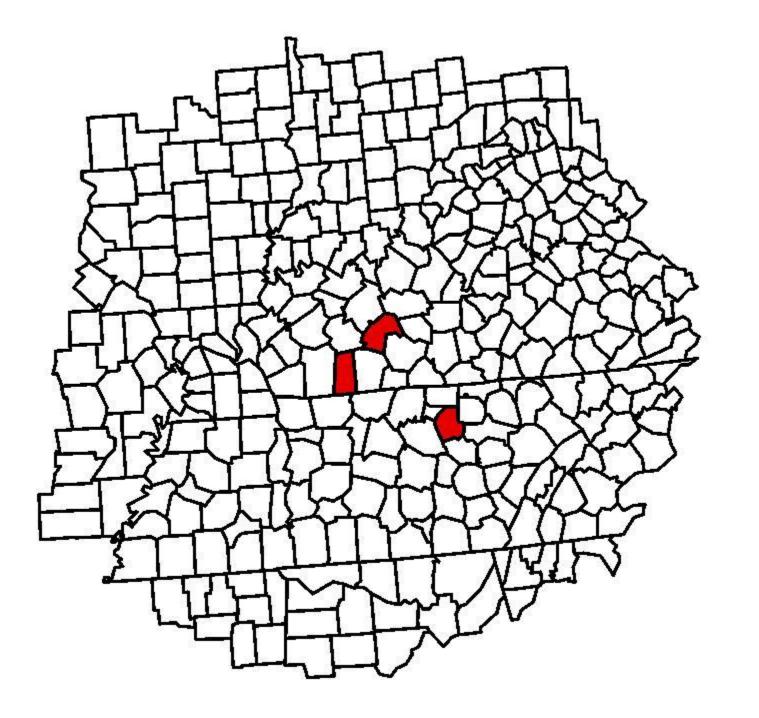
if significantly higher...

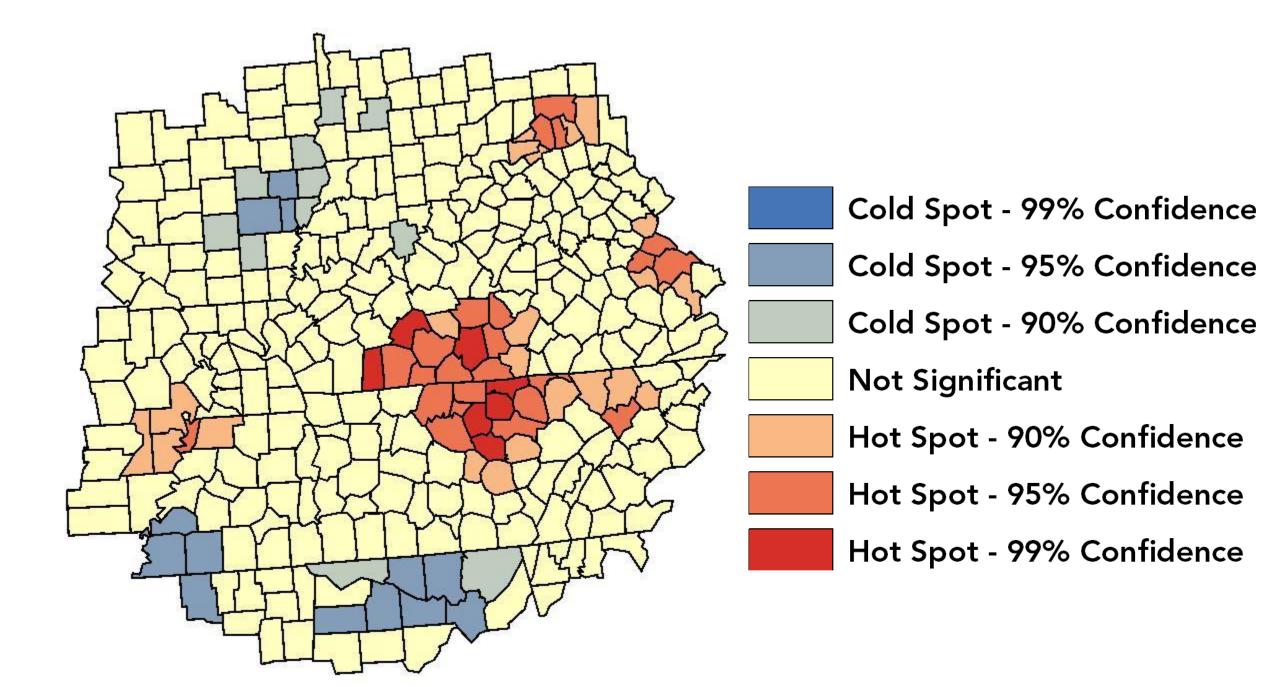
feature is marked as a hot spot!











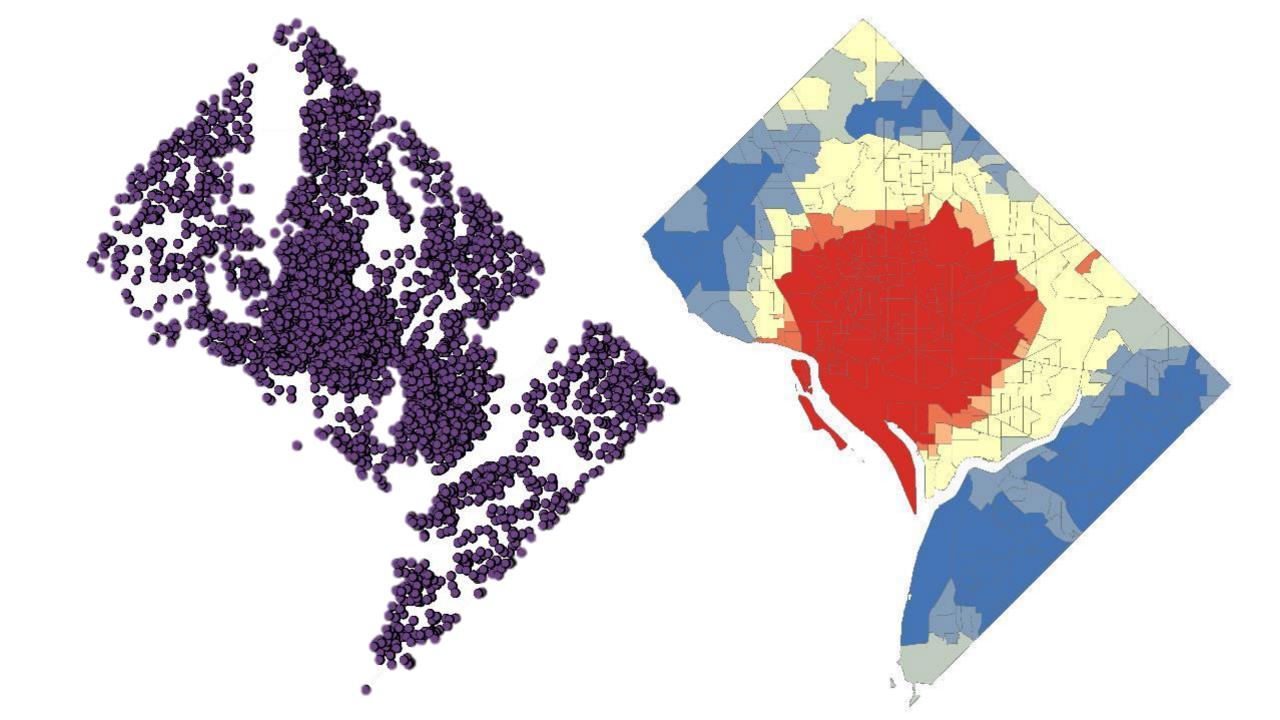
...how do we know if it's SIGNIFICANTLY different???

Getis-Ord Gi* Statistic

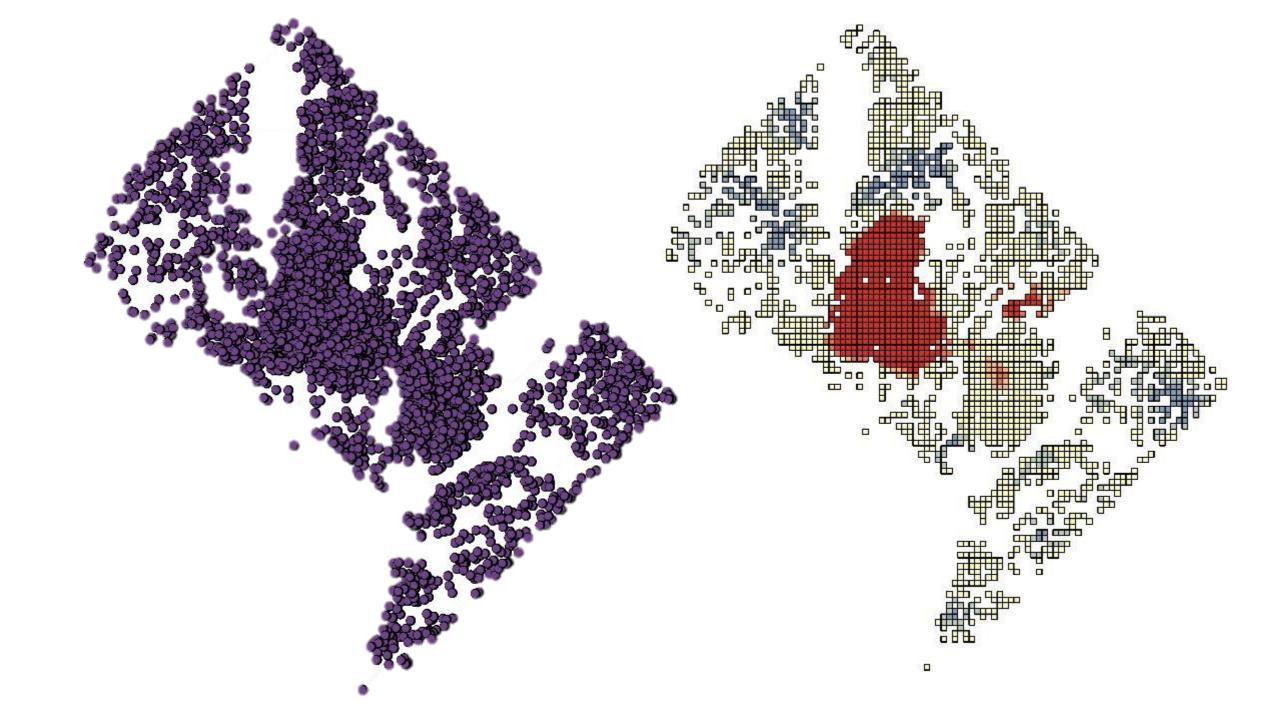
Points











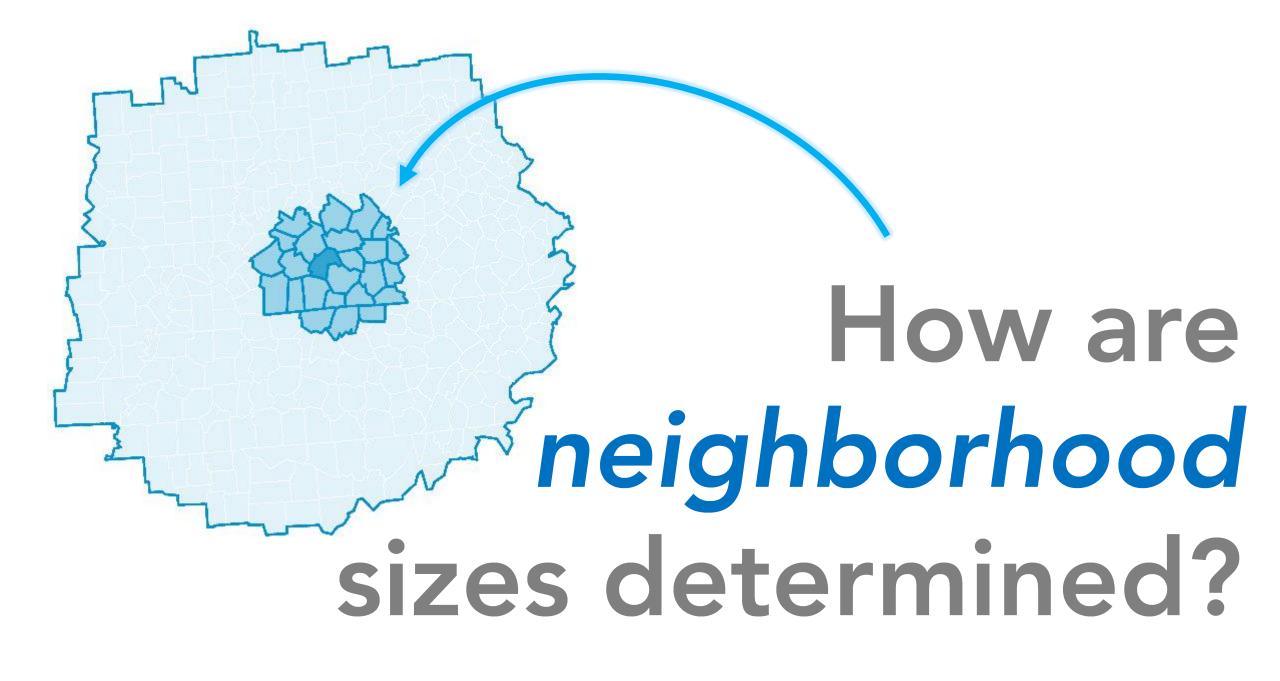


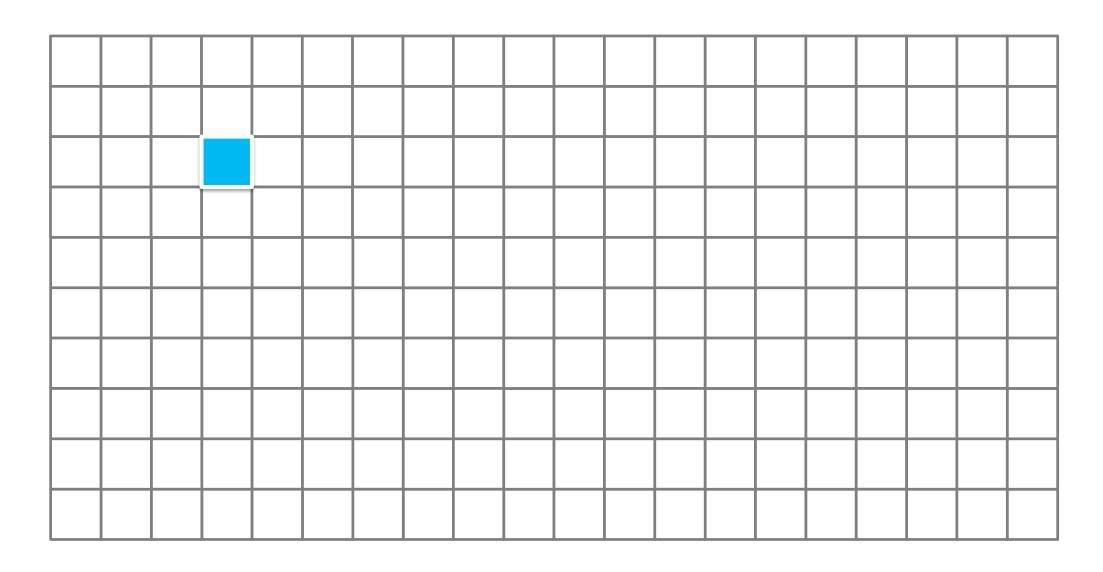


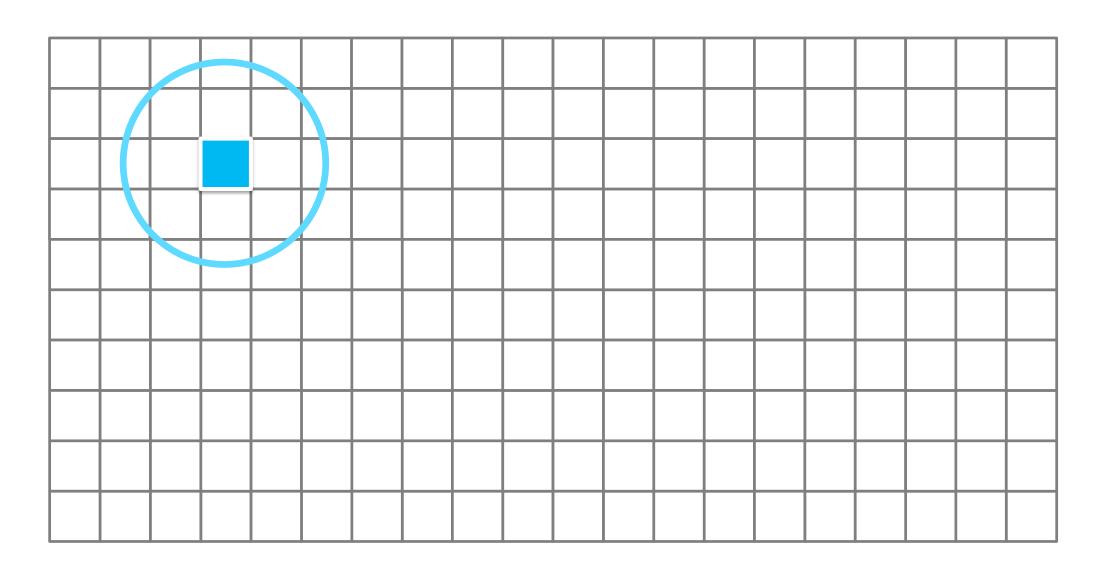
"Where are the Hot Spots?"

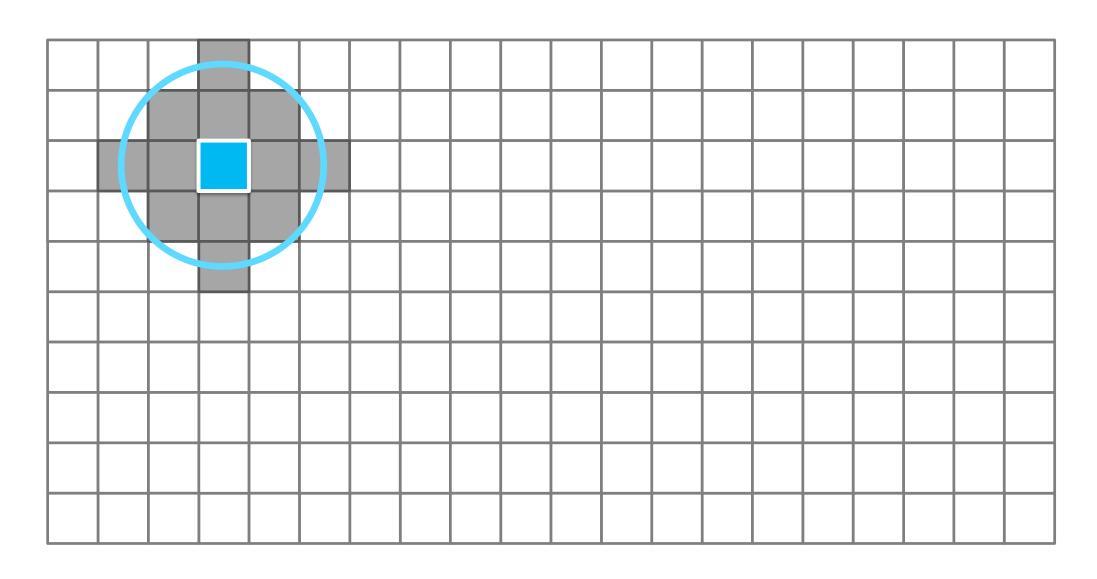
not (necessarily)
the same as

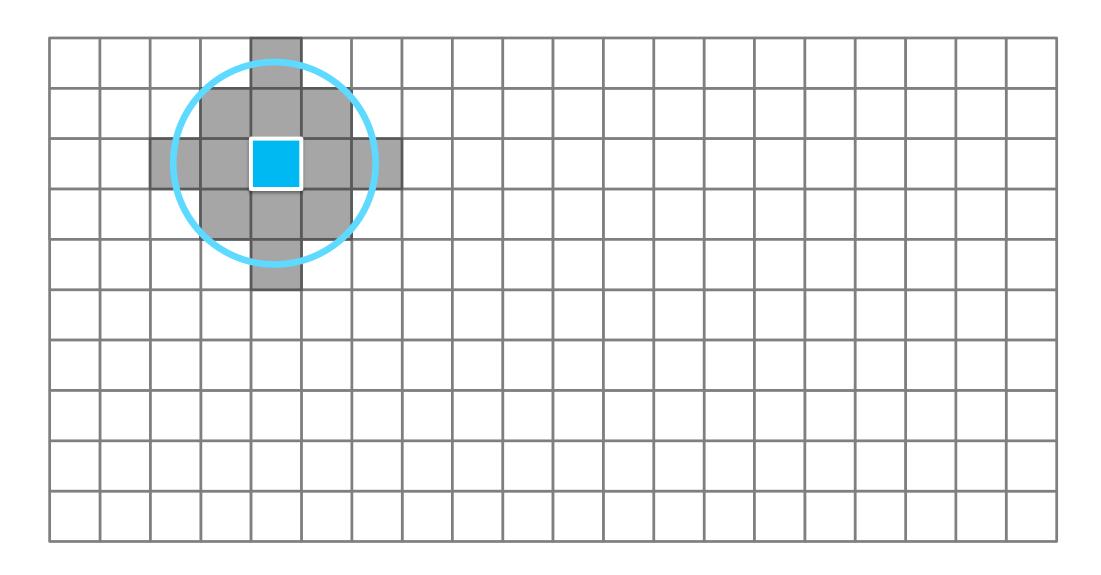
"Where are the highest values?"

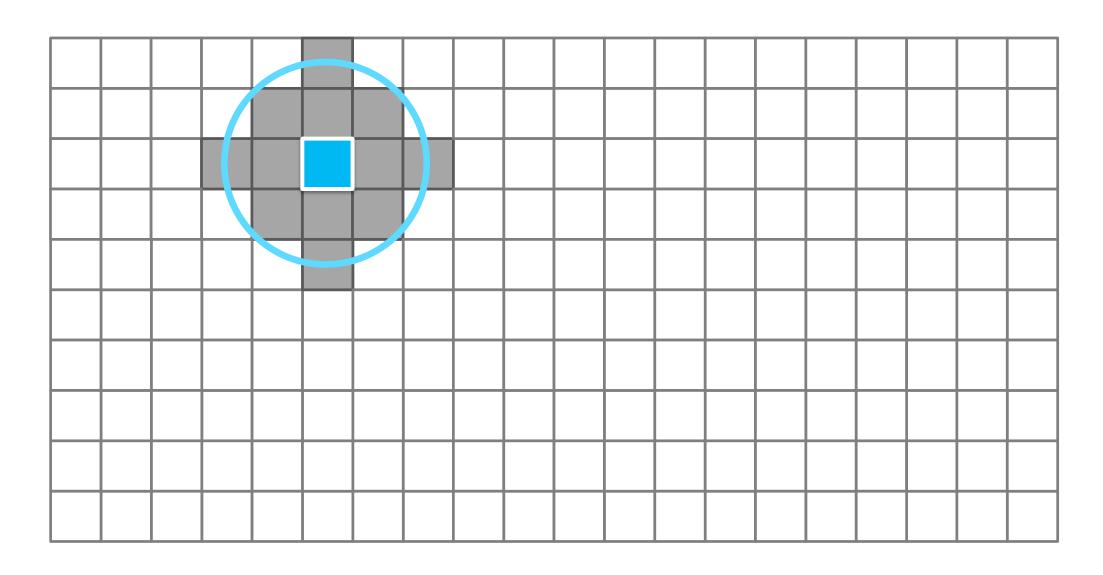


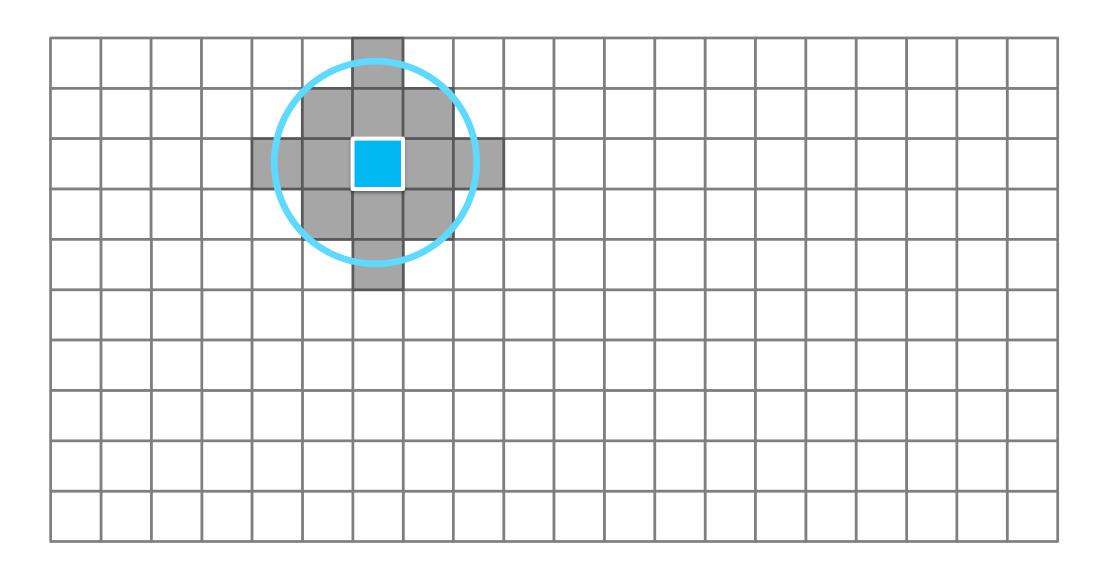




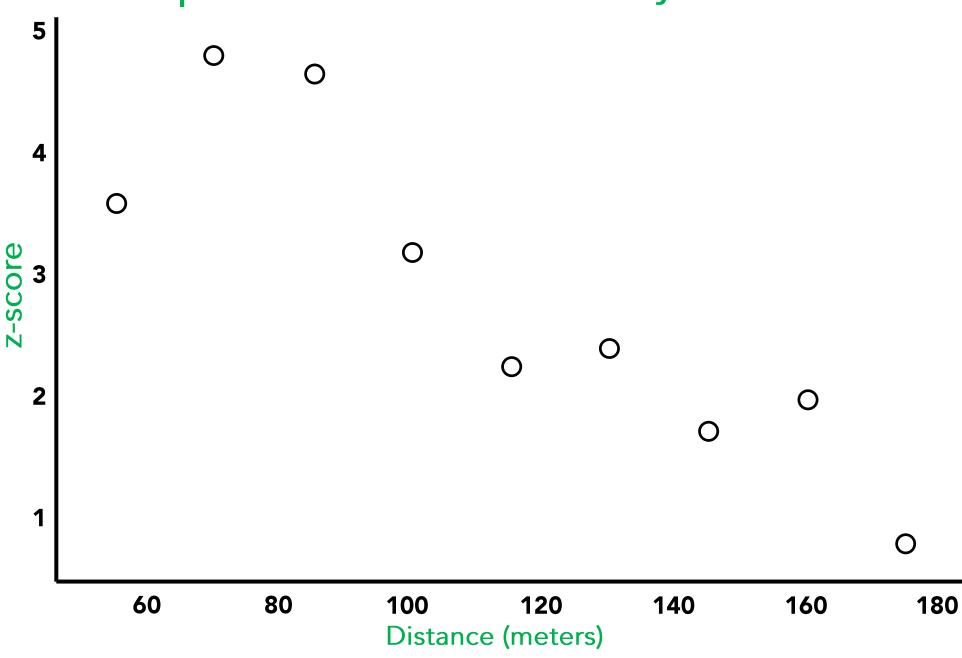




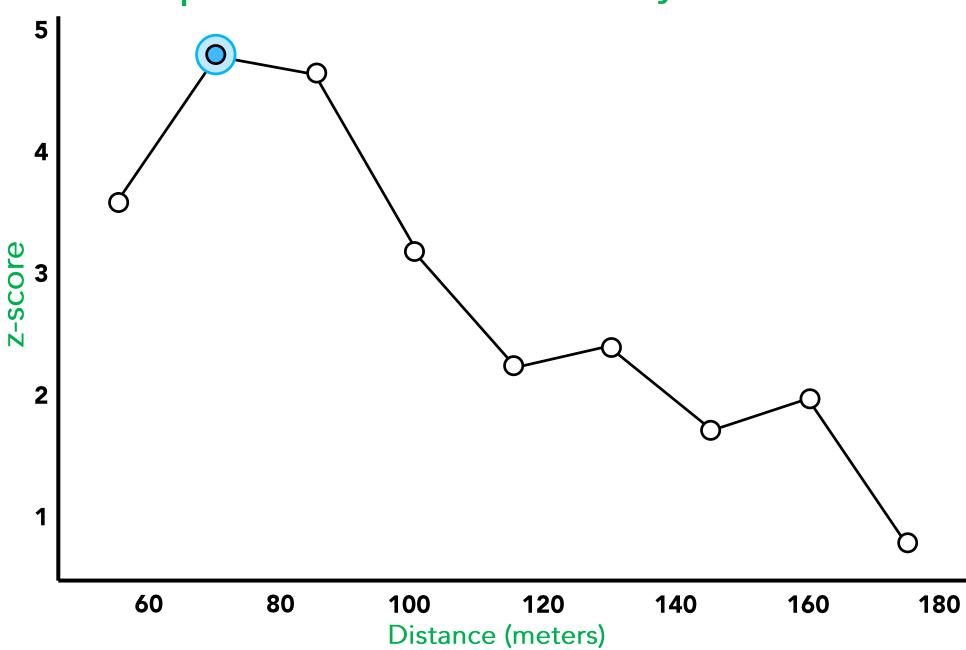




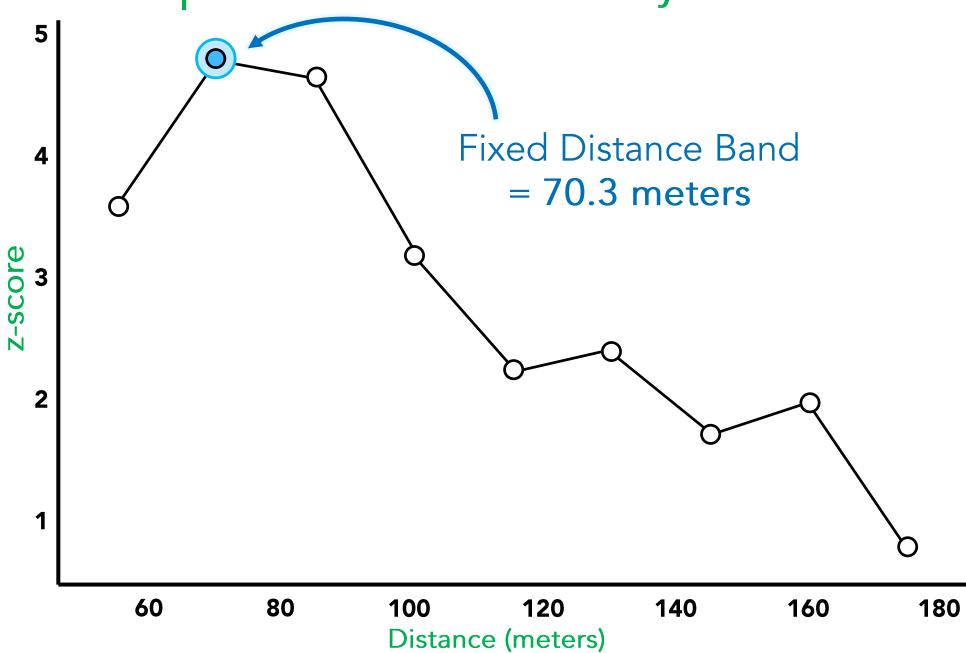
Spatial Autocorrelation by Distance

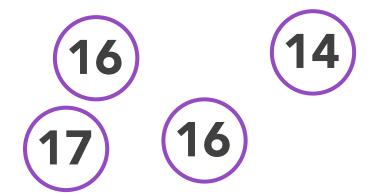


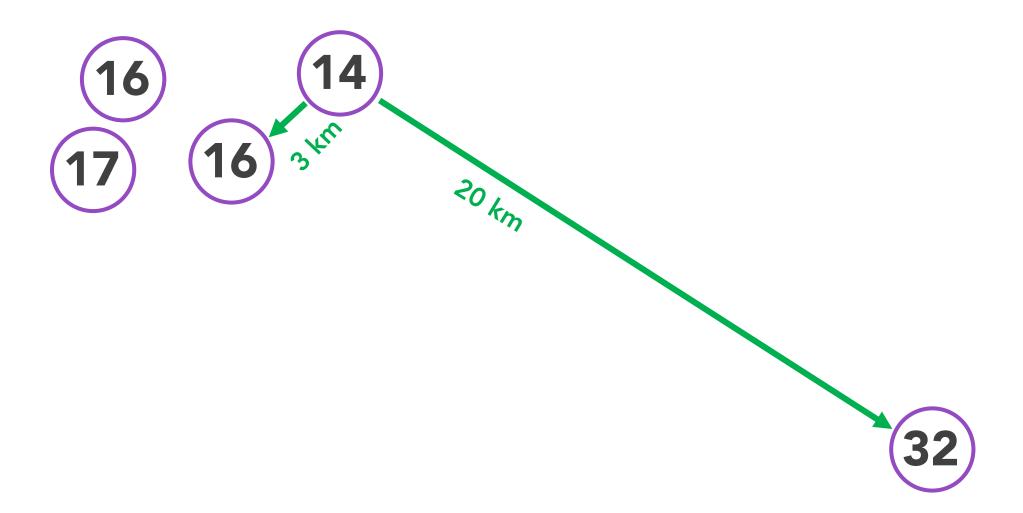
Spatial Autocorrelation by Distance

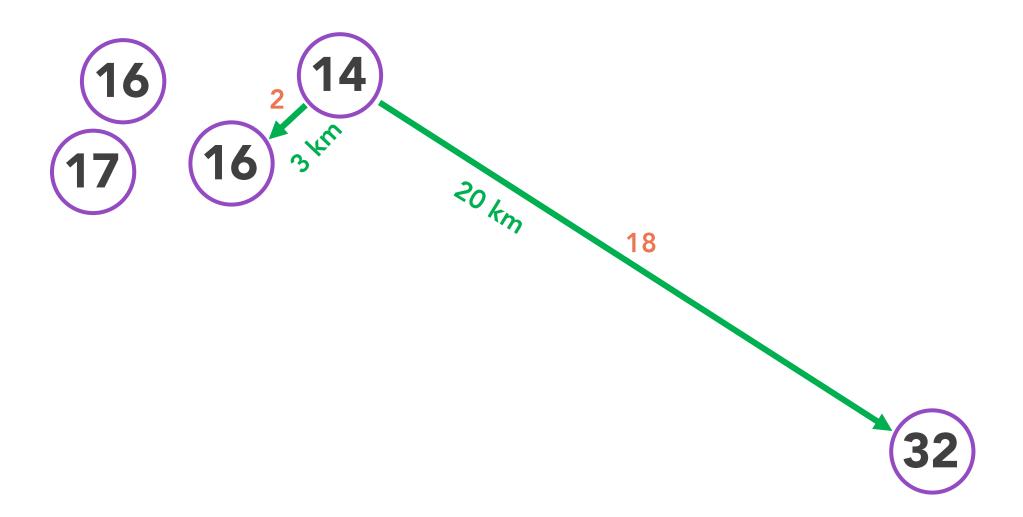


Spatial Autocorrelation by Distance

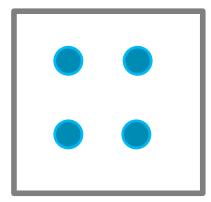




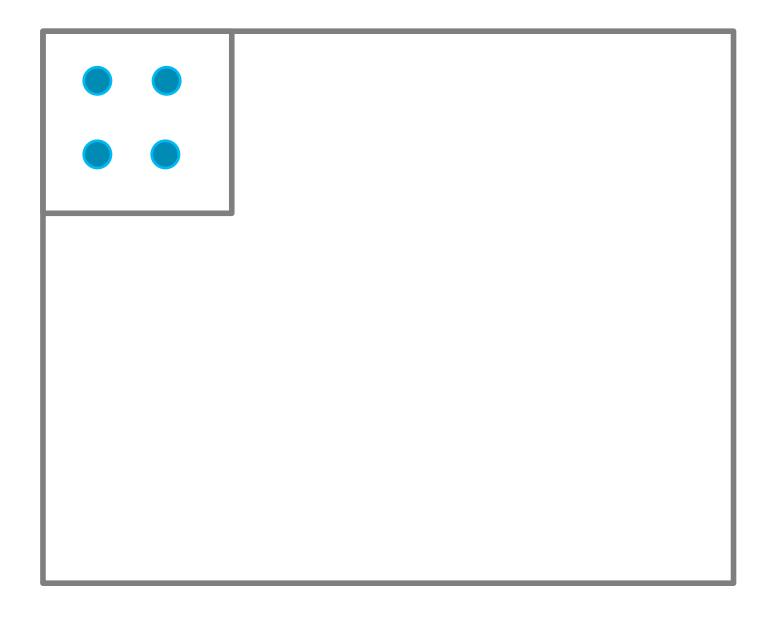




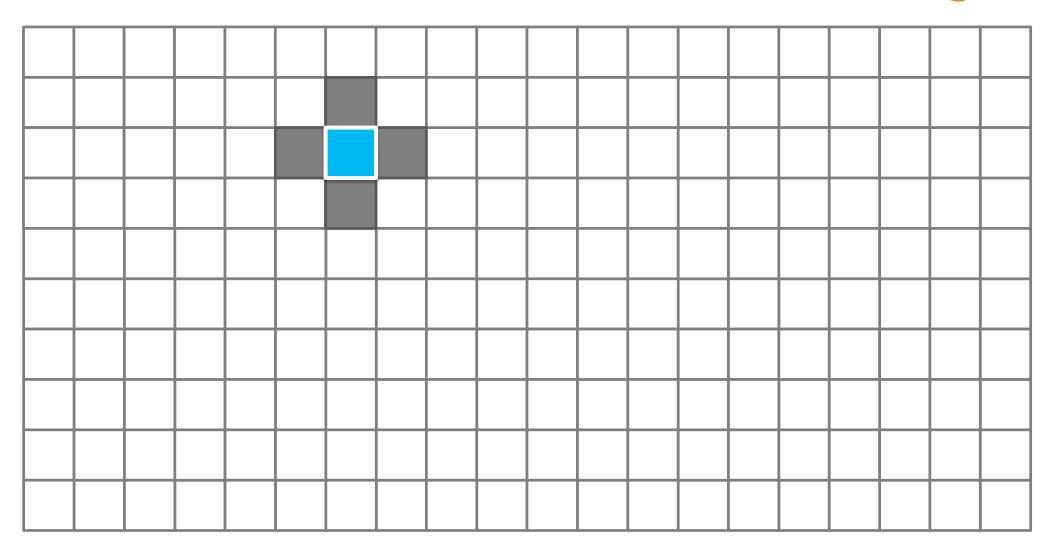
dispersed



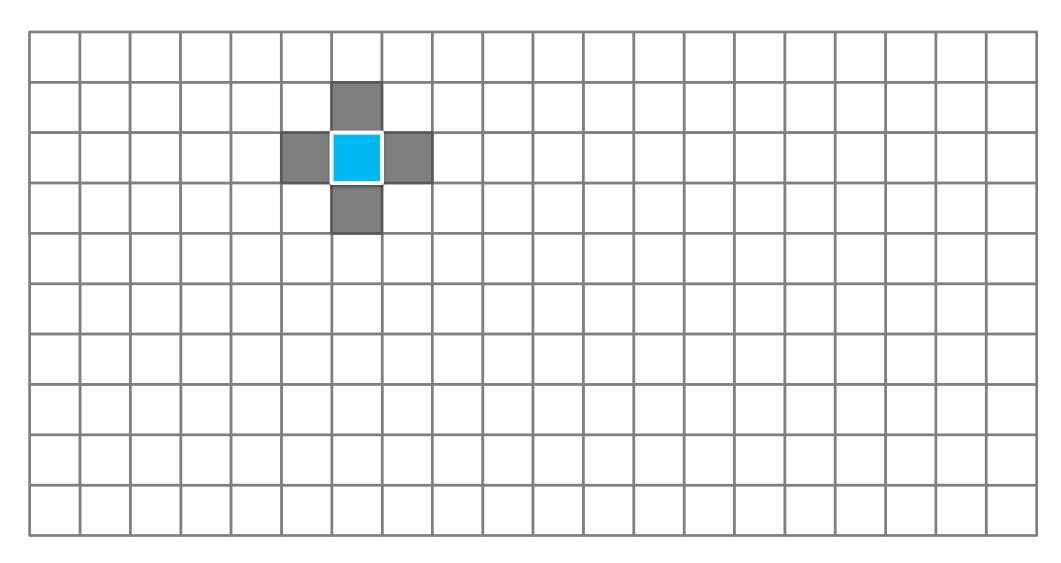
clustered



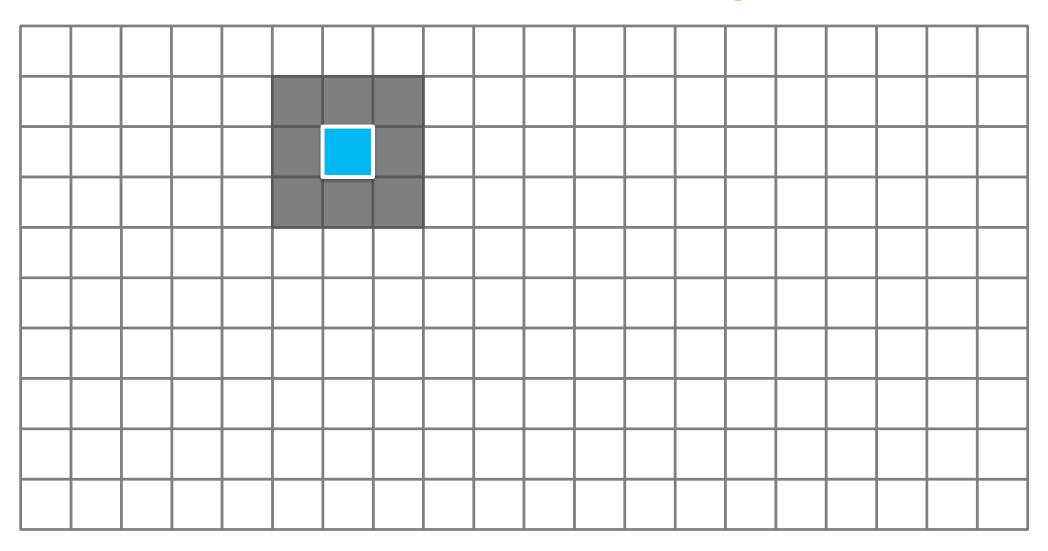
Edges



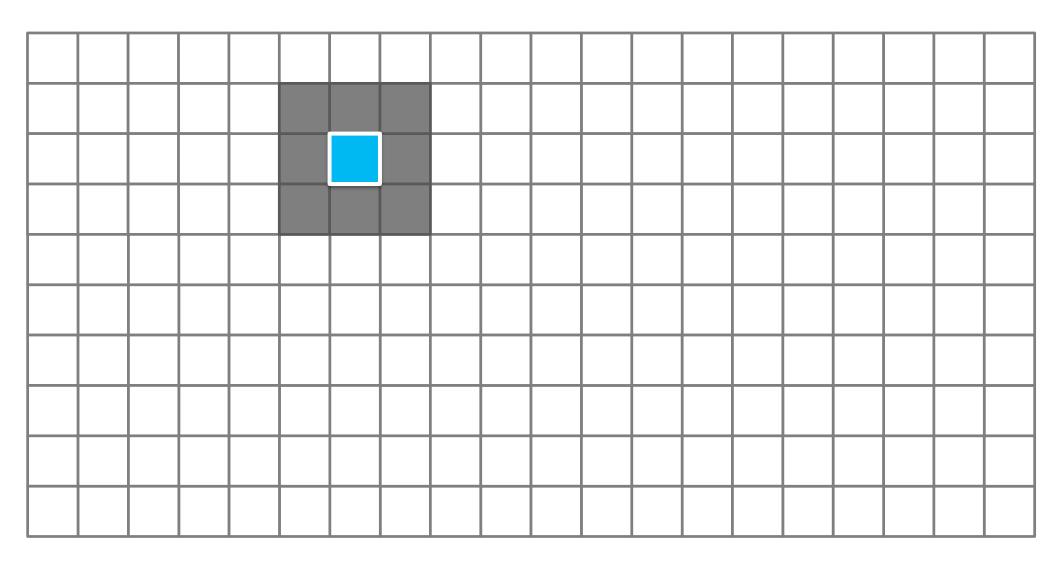
Rook's Case



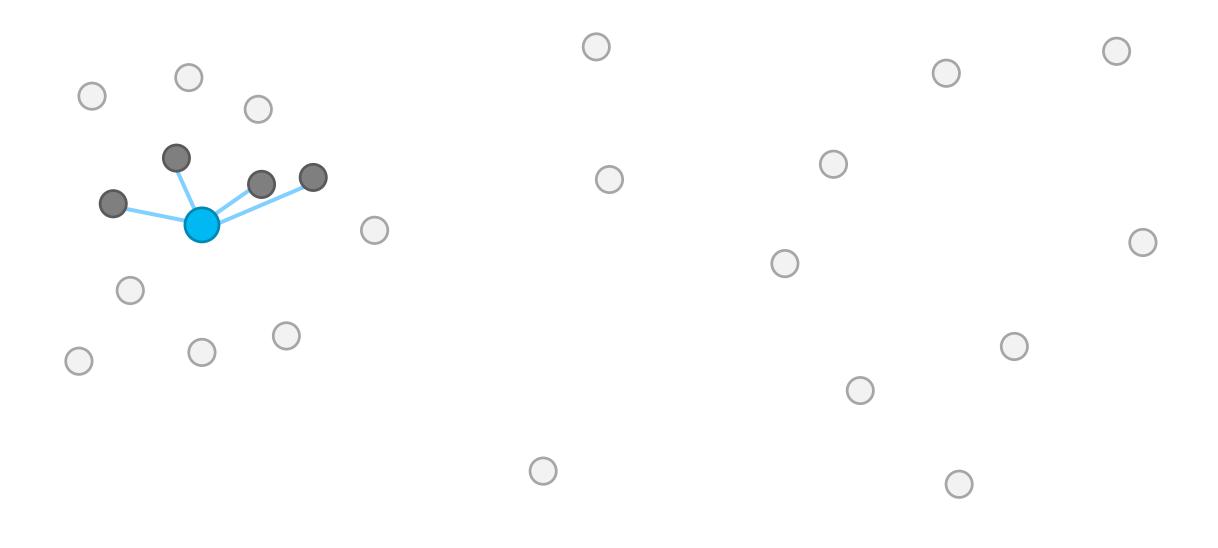
Edges/Corners



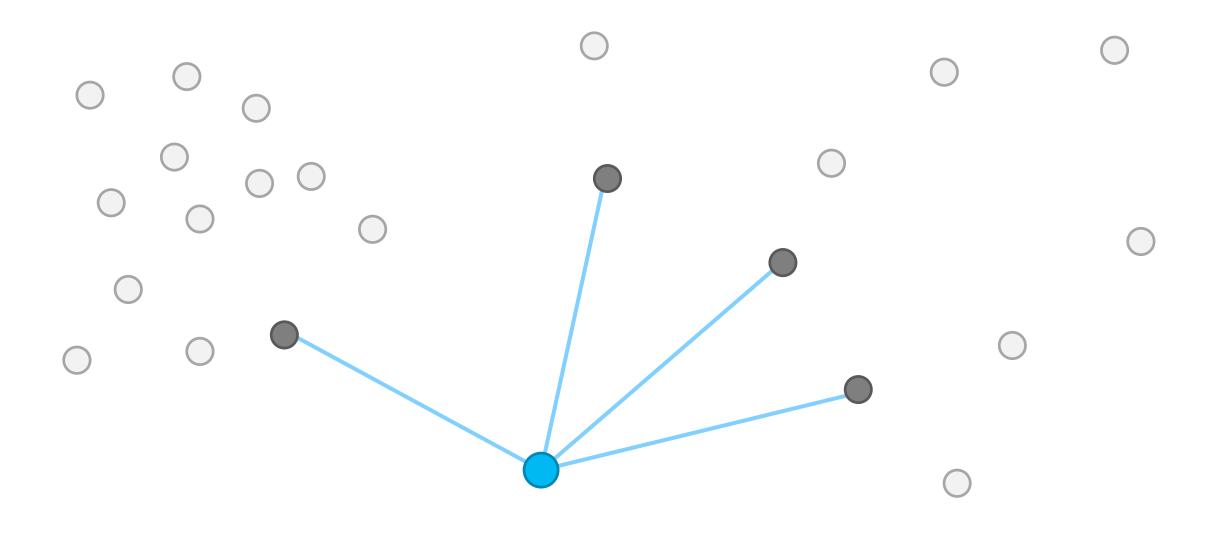
Queen's Case



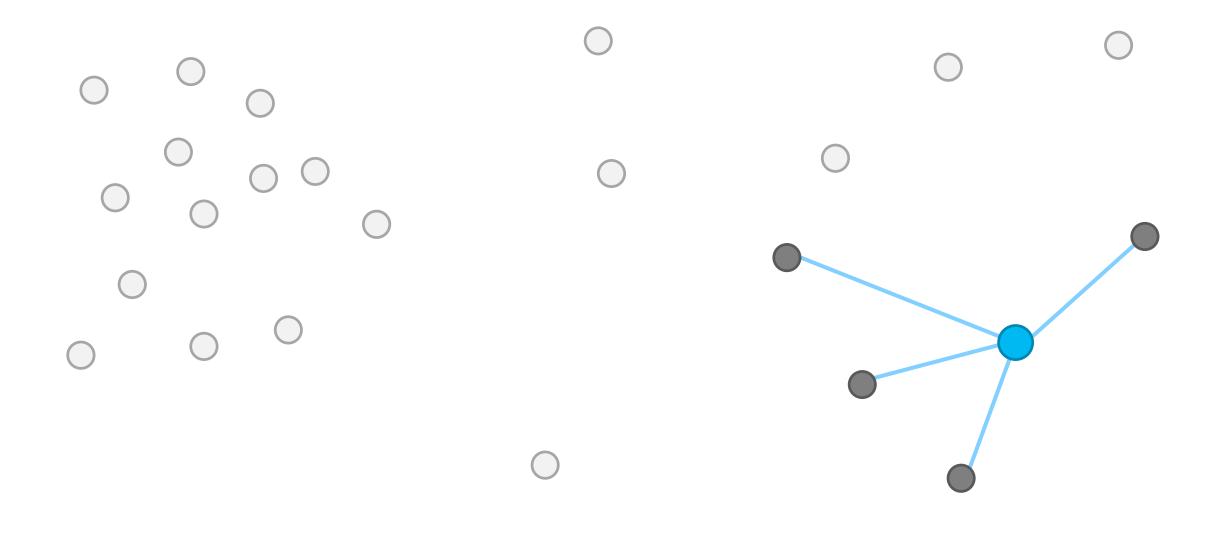
K Nearest Neighbors



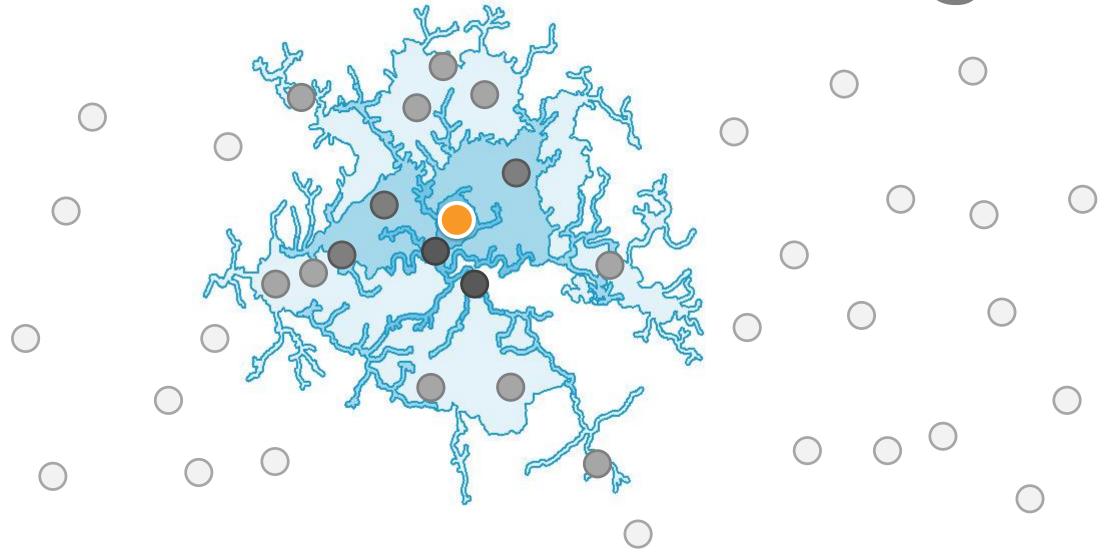
K Nearest Neighbors



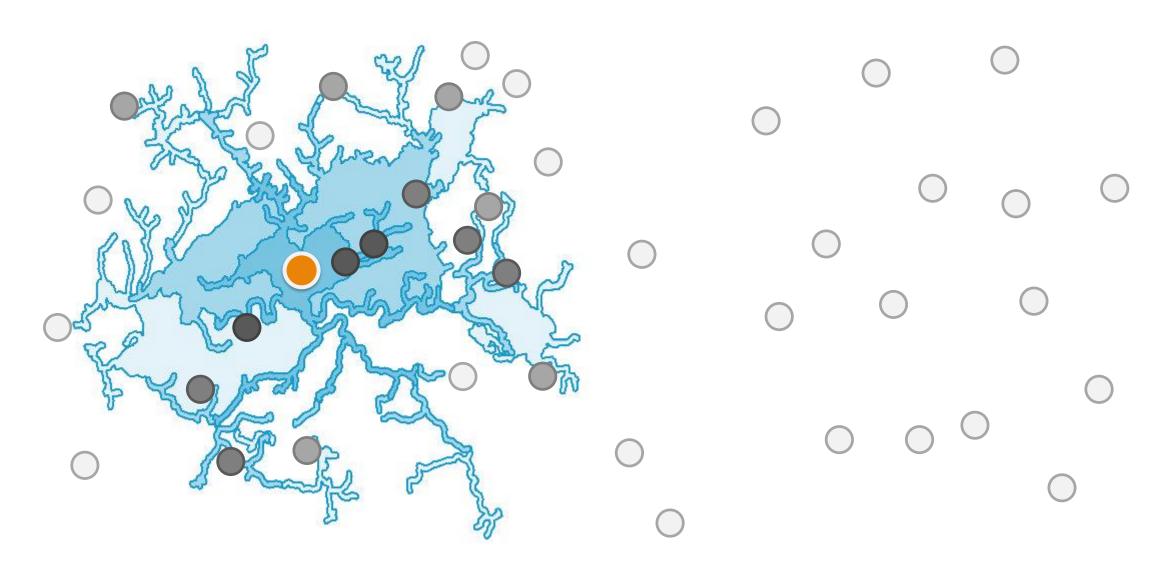
K Nearest Neighbors



Network Spatial Weights



Network Spatial Weights

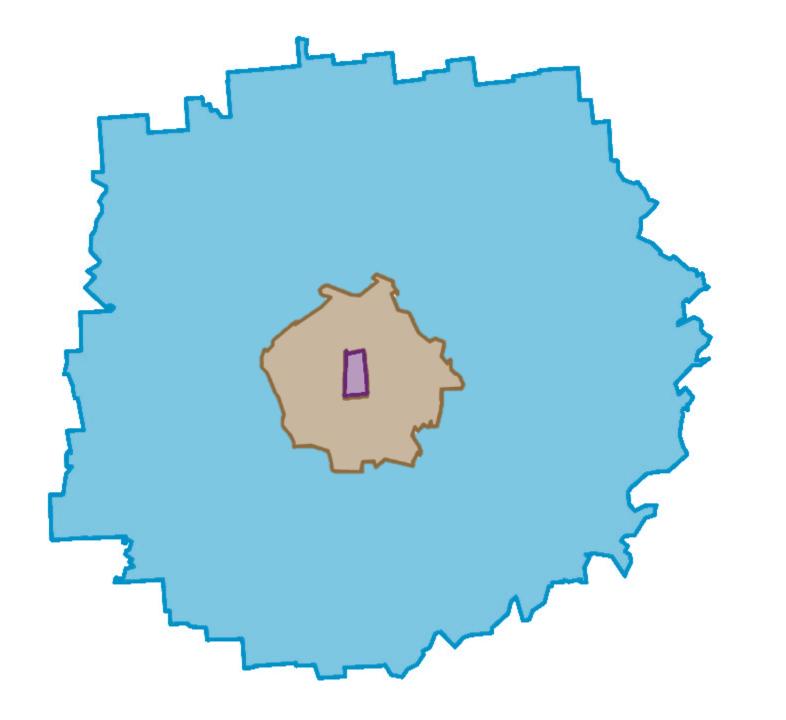


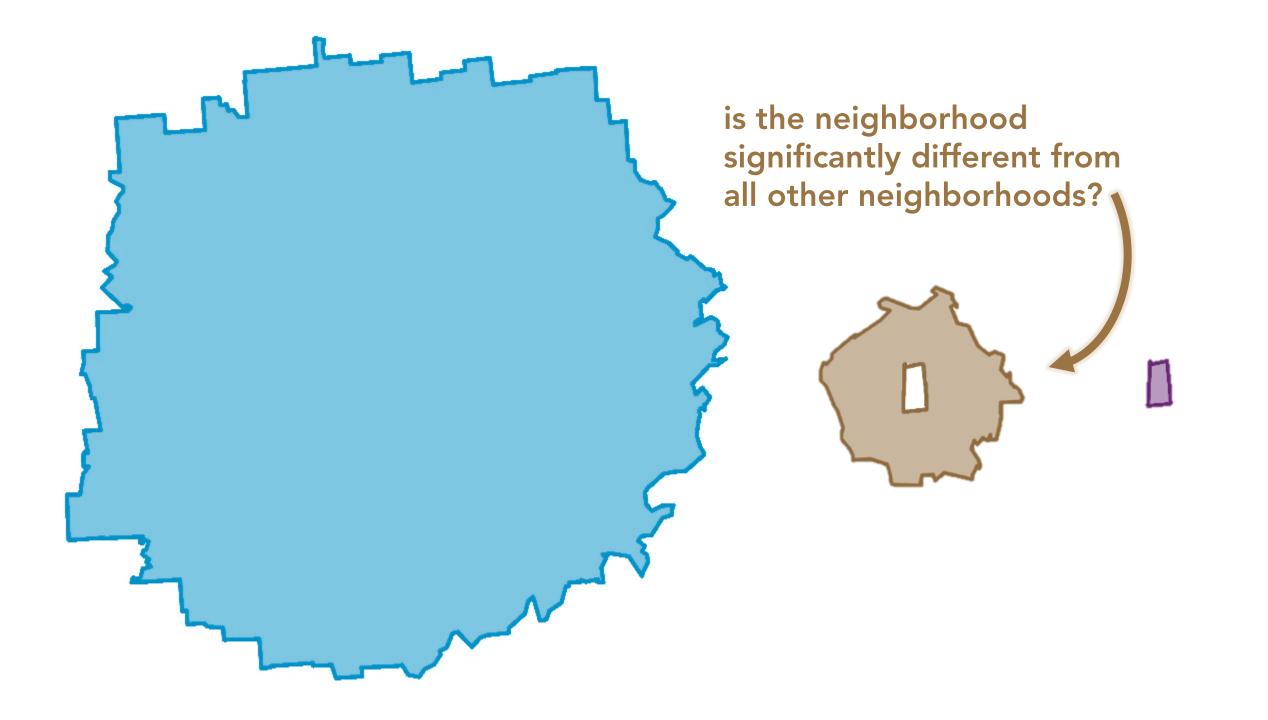
demo

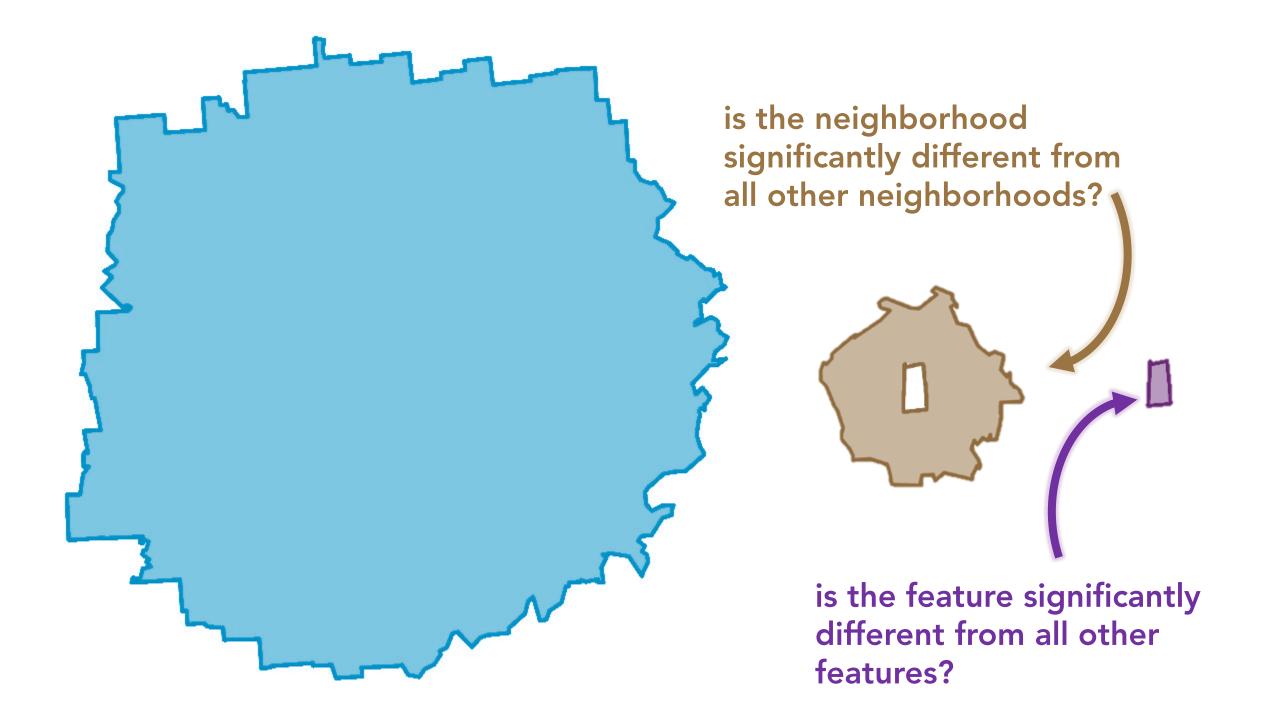


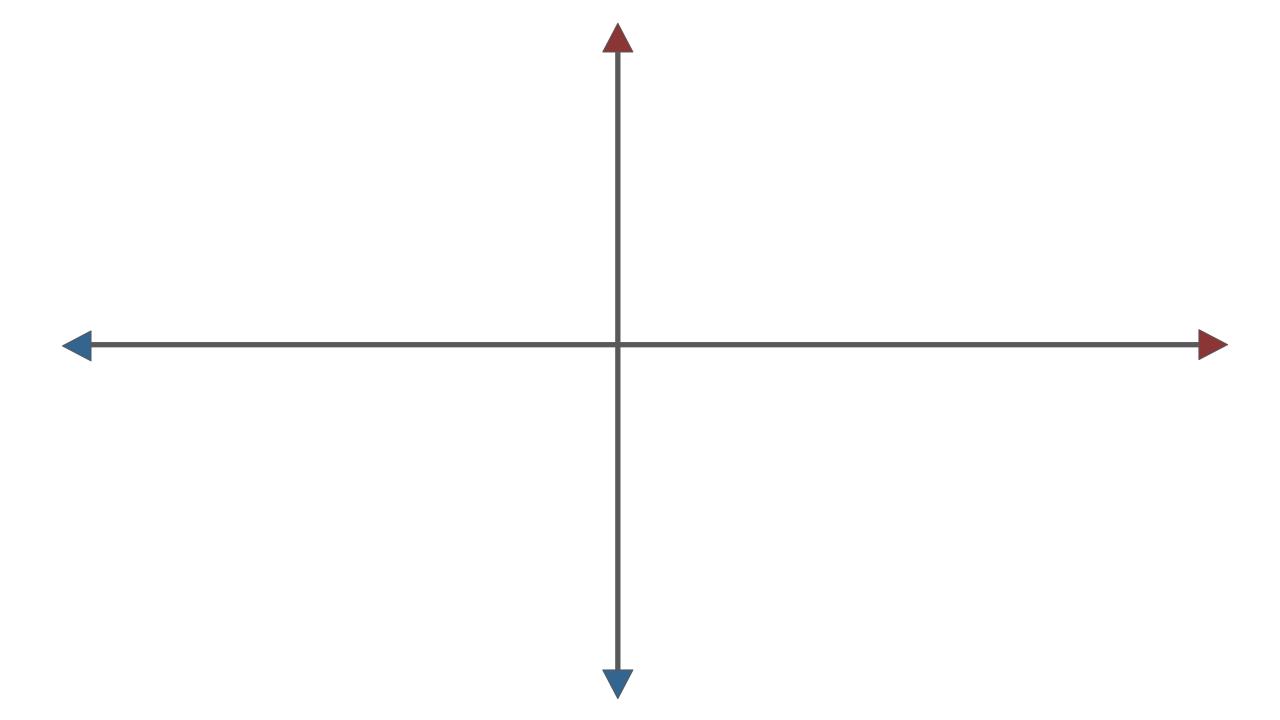
Cluster and Outlier Analysis

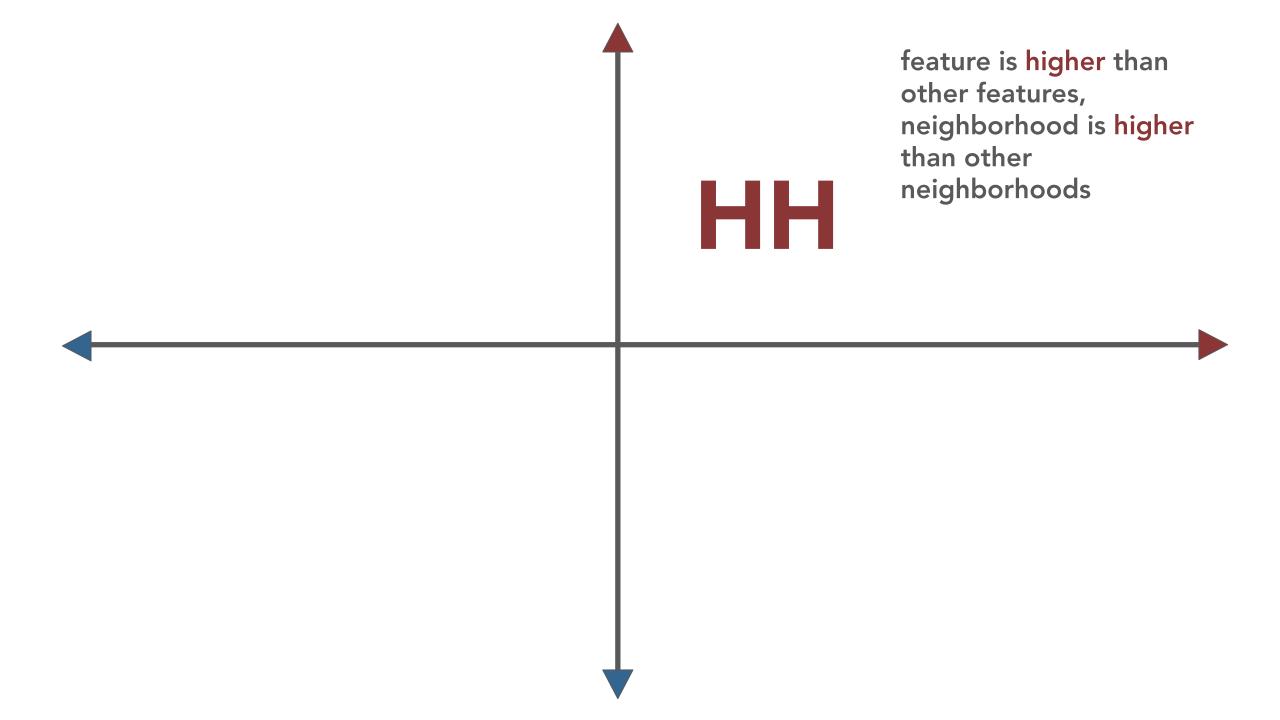
given a set of weighted features, identifies statistically significant hot spots, cold spots, and spatial outliers using the Anselin Local Moran's I statistic













feature is higher than other features, neighborhood is higher than other neighborhoods

feature is lower than other features, neighborhood is lower than other neighborhoods



feature is higher than other features, neighborhood is lower than other neighborhoods





feature is higher than other features, neighborhood is higher than other neighborhoods

feature is lower than other features, neighborhood is lower than other neighborhoods



feature is higher than other features, neighborhood is lower than other neighborhoods





feature is higher than other features, neighborhood is higher than other neighborhoods

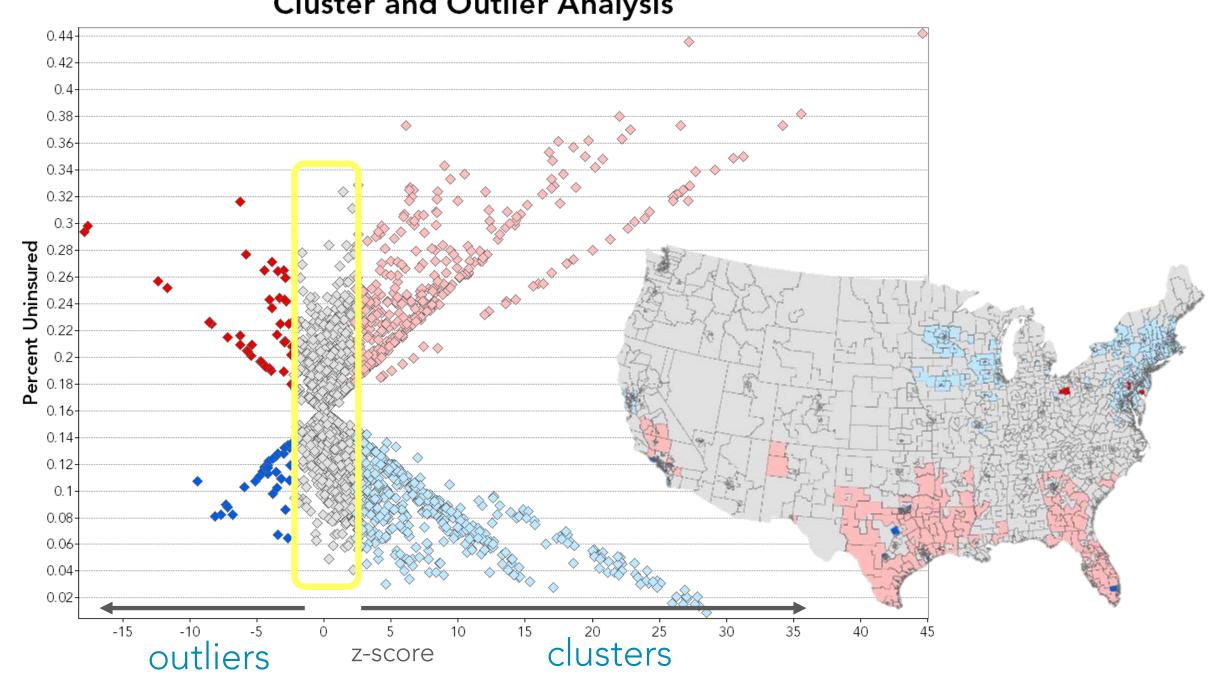
feature is lower than other features, neighborhood is lower than other neighborhoods





feature is lower than other features, neighborhood is higher than other neighborhoods

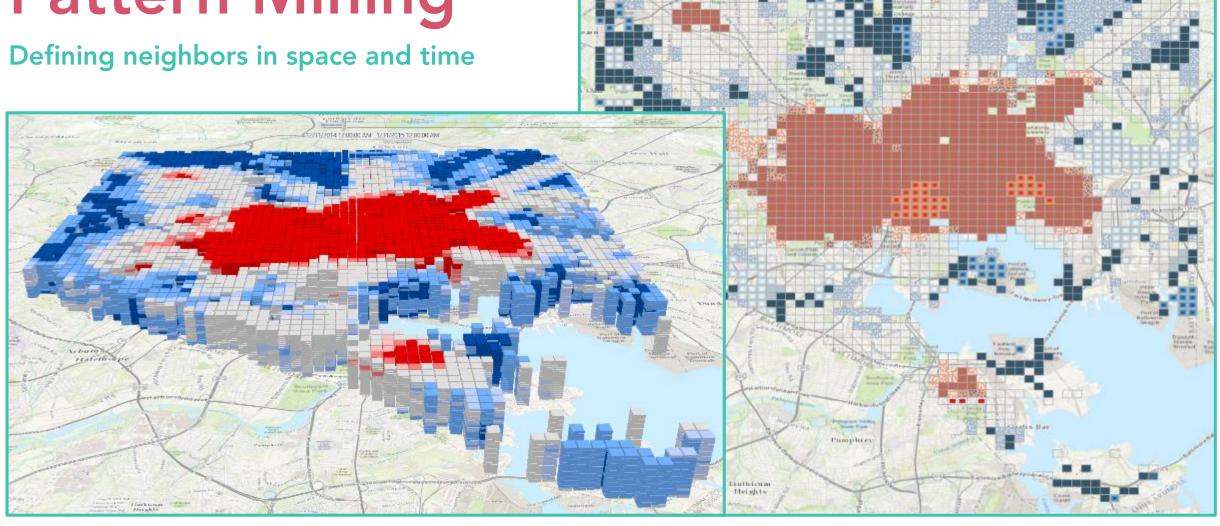
Cluster and Outlier Analysis



demo



Space Time Pattern Mining





esriurl.com/spatialstats