

EXERCISE 2

Spatial Statistics: Spatial Pattern Analysis



Introduction

This exercise introduces methods to evaluate and statistically validate features such as identifying statistically significant clustering in the spatial distribution of your data; identifying spatial clustering/dispersion for a set of geographic features over a range of distances; and measuring clustering or dispersion of features based on feature locations and attribute values.

Objectives

- Learn to use the Analyzing Patterns Toolset
- Understand statistical significance and confidence intervals
- Exploring several methods to evaluate and statistically validate if features, or the values associated with features, form clustered, dispersed, or random spatial patterns

Prerequisites

- ArcGIS Pro installed on your machine



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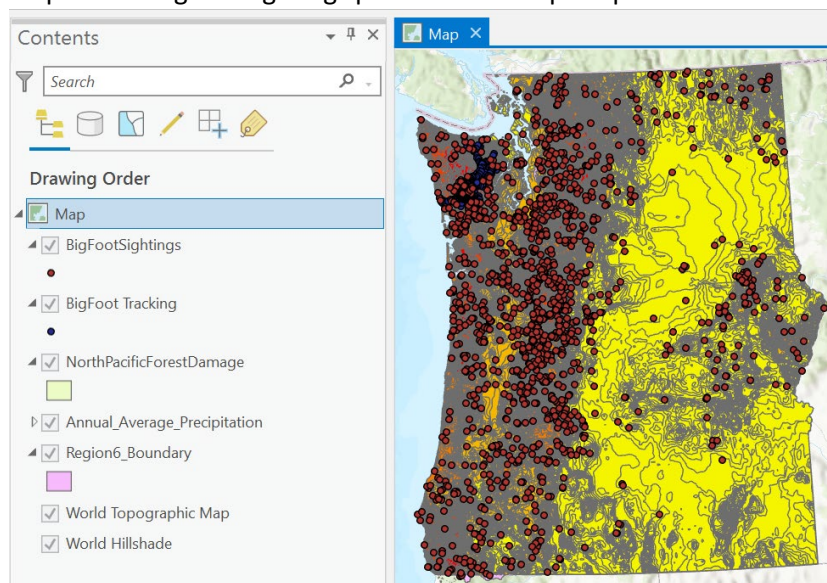
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Part 1: Explore the Project Data

A. Open the Exercise 2 project file

1. Open Windows Explorer and navigate to your course folder
....\ArcGISProSpatialStatistics\Data\Exercise_2
 - i. Open the **Exercise_2** ArcGIS Project File.
2. Inspect the BigFootSightings points in the map for patterns.



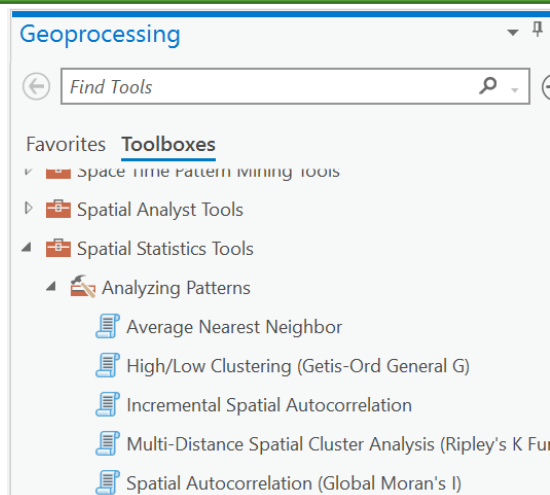
Do any specific spatial patterns such as clustering or dispersion exist in the Bigfoot sighting points? If so, does the clustering/dispersion have statistical significance? We can answer these questions with tools from the Analyzing Patterns toolset.

Part 2: Identify Spatial Patterns that Exist in the Data

The Analyzing Patterns toolset provides tools to help identify overall patterns in the data and identify statistically significant clustering or dispersion.

A. Open the analyzing patterns toolset

1. In the Geoprocessing pane, click the **Toolboxes** tab.
2. Expand the **Spatial Statistics** toolbox then expand the **Analyzing Patterns** toolset.



The tools in the Analyzing Patterns toolset are based on inferential statistics and start with the null hypothesis that your features, or the values associated with your features, exhibit a spatially random pattern. They then compute a p-value identifying the probability of whether the null hypothesis is correct. Calculating a probability is important when you need a high level of confidence in a particular decision.

The Analyzing Patterns tools answer questions such as, "Are the features in the dataset, or the values associated with the features in the dataset, spatially clustered?" and "Is the clustering becoming more or less intense over time?"

B. Execute the Average Nearest Neighbor tool to substantiate clustering or dispersion of BigFoot Sightings.

The Average Nearest Neighbor tool identifies statistically significant clustering or dispersion by calculating a nearest neighbor index based on the average distance from each feature to its nearest neighboring feature.

1. From the *Analyzing Patterns* toolset, click **Average Nearest Neighbor** to open the tool.
 - i. Input Feature Class: **BigFootSightings**.
 - ii. Distance Method: **Euclidian**. This is a straight-line distance. Manhattan takes into consideration buildings or other obstacles
 - iii. Generate Report: **checked**.
 - iv. Click **Run**.
 - v. At the bottom of the tool window, click **View Details** and notice five values in the window that appears: **Nearest Neighbor Index**, **z-score**, **p-value**, **Expected Mean Distance**, and **Observed Mean Distance**.

Average Nearest Neighbor (Spatial Statistics Tools)

Completed.

Started: Today at 4:45:17 PM
Completed: Today at 4:45:33 PM
Elapsed Time: 16 Seconds

Errors and warnings

Parameters

| | |
|------------------------|--|
| Input Feature Class | BigFootSightings |
| Distance Method | EUCLIDEAN_DISTANCE |
| Generate Report | GENERATE_REPORT |
| Area | |
| Nearest Neighbor Index | 0.591783 |
| z-score | -29.624556 |
| p-value | 0 |
| Expected Mean Distance | 9283.86194 |
| Observed Mean Distance | 5494.036221 |
| Report File | C:\Users\markhammond\Desktop ArcGIS_Pro_Spatial_Statistics\Exercise_2 NearestNeighbor_Result_10072_20200_.html |

Geoprocessing

Average Nearest Neighbor

Parameters Environments

Input Feature Class
BigFootSightings

Distance Method
Euclidean

☒ Generate Report

Area

Run

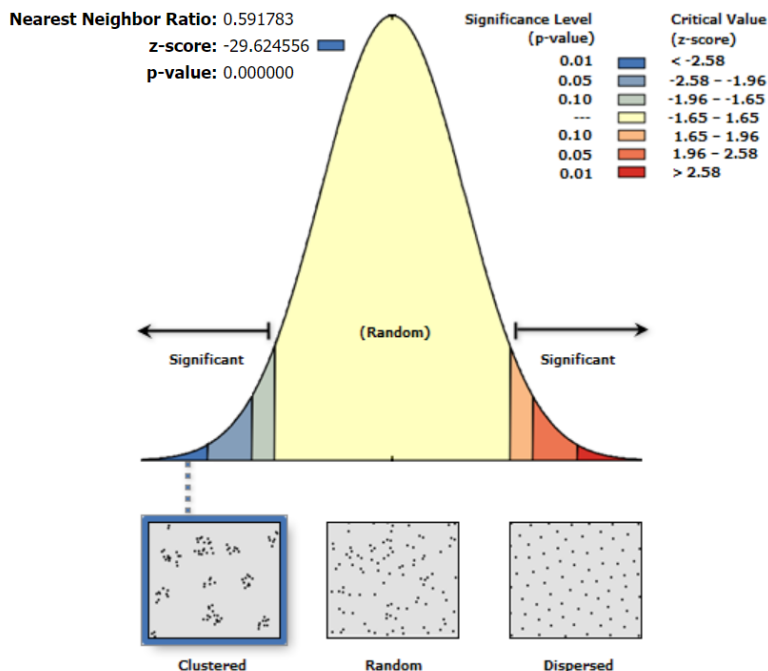
Average Nearest Neighbor completed.
View Details Open History

STATISTICS 101: The Z-score and P-value results are measures of statistical significance which tell you whether or not to reject the null hypothesis. For the Average Nearest Neighbor statistic, the null hypothesis states that features are randomly distributed.

The Observed Mean Distance is the actual distance between each point and the closest neighboring point and the Expected Mean Distance is the average distance between neighbors in a hypothetical random distribution. The Nearest Neighbor Index is expressed as the ratio of the Observed Mean Distance to the Expected Mean Distance. If the index is less than 1, the pattern exhibits clustering; if the index is greater than 1, the trend leans toward dispersion.

vi. Click the **Report File** link near the bottom of the View Results window.

Average Nearest Neighbor Summary



INTERPRETATION: the report states very clearly that: Given the z-score of -29.62, there is less than 1% likelihood that this clustered pattern could be the result of random chance. We are 99% confident that the clustering in this dataset is "NOT" random.

2. Close the report.

C. Use the High/Low Clustering (Getis-Ord General G*) tool to verify clustering of BigFoot Sightings increases in areas with more rainfall.

Now that we have established a definitive clustering of Bigfoot sightings; let's determine if clustering also exist in sightings relative to the amount of annual rainfall for that region.

1. Analyze the **Annual_Average_Precipitation** layer in relation to the **BigFootSightings** points in the map. The points appear to be clustered in areas that have greater rainfall averages. We can use the High/Low Clustering (Getis-Ord General G*) tool to verify this belief.

Prior to performing the analysis, we must prepare the data by joining the attributes of the **Annual_Average_Precipitation** layer to the **BigFootSightings** layer based on their spatial relationship. The target features (**BigFootSightings**) and the attributes from the join features (**Annual_Average_Precipitation**) will be written to a new output feature class.

2. In the Contents pane, right-click **BigFootSightings** then click **Joins and Relates** and select **Spatial Join**.
3. In the Spatial Join window set the following parameters:

| BigFootSightings_Precipitation X | | | | | | | | | | |
|----------------------------------|-------|------------|--------|----------|-----------|-----------|--------|------|-------|--|
| Field: | | Selection: | | | | | | | | |
| ET_FID | Class | Year | Number | Date | Latitude | Longitude | Weight | Case | RANGE | |
| 1 | B | 2008 | 1 | 1/1/2008 | -121.6885 | 45.23679 | 2 | 6 | 55 | |
| 2 | B | 2008 | 1 | 1/1/2008 | -118.2287 | 48.7375 | 2 | 6 | 23 | |
| 3 | B | 2008 | 1 | 1/1/2008 | -123.3687 | 46.99385 | 2 | 6 | 69 | |
| 4 | B | 2008 | 1 | 1/1/2008 | -122.0428 | 47.49865 | 2 | 6 | 71 | |
| 5 | B | 2008 | 1 | 1/1/2008 | -121.8429 | 47.07024 | 2 | 6 | 89 | |
| 6 | B | 2008 | 1 | 1/1/2008 | -121.796 | 45.57957 | 2 | 6 | 105 | |
| 7 | B | 2008 | 1 | 1/1/2008 | -121.6423 | 44.05426 | 2 | 6 | 59 | |

Now that we have added precipitation values to the BigFoot Sightings layer, we can begin our analysis.

5. From the *Analyzing Patterns* toolset, click **High/Low Clustering (Getis-Ord General G*)** to open the tool.

The High/Low Clustering tool measures the concentration of high or low values for a given study area. The null hypothesis for the High/Low Clustering (General G) statistic states there is no spatial clustering of feature values.

6. In the High/Low Clustering tool window set the following parameters:
 - i. Input Feature Class: **BigFootSightings_Precipitation**.
 - ii. Input Field: **Range**.
 - iii. Generate Report: **checked**.
 - iv. Accept the other defaults and click **Run**.

Geoprocessing

High/Low Clustering (Getis-Ord General G*)

Parameters Environments

Input Feature Class
BigFootSightings_Precipitation

Input Field
RANGE

☒ Generate Report

- v. Click **View Details** and notice the warning message that appears.

High/Low Clustering (Getis-Ord General G) (Spatial Statistics Tools)



⚠ Completed with warnings.

Started: Today at 2:16:55 PM

Completed: Today at 2:17:01 PM

Elapsed Time: 6 Seconds

▼ Errors and warnings

⚠ **WARNING 000853:** The default neighborhood search threshold was 62519.7903 Meters.

What does that warning message mean? This is a warning provided to notify you that a distance threshold was not used. When no value is specified for the distance threshold, the minimum distance to ensure every feature has at least one neighbor will be used.

vi. Click the **Report File** link.

High-Low Clustering Report

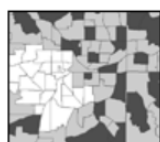
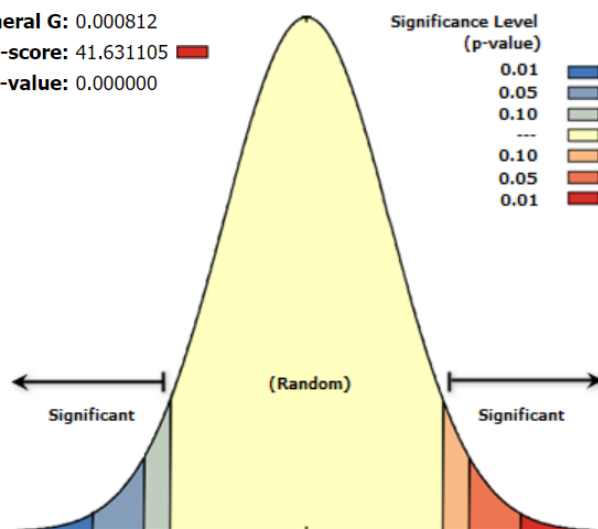
Observed General G: 0.000812

z-score: 41.631105

p-value: 0.000000

**Significance Level
(p-value)**

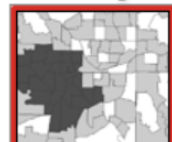
| Significance Level (p-value) | Critical Value (z-score) |
|---------------------------------|-----------------------------|
| 0.01 | < -2.58 |
| 0.05 | -2.58 - -1.96 |
| 0.10 | -1.96 - -1.65 |
| --- | -1.65 - 1.65 |
| 0.10 | 1.65 - 1.96 |
| 0.05 | 1.96 - 2.58 |
| 0.01 | > 2.58 |



Low-Clusters



Random



High-Clusters

Given the z-score of 41.63110496176197, there is a less than 1% likelihood that this high-clustered pattern could be the result of random chance.

INTERPRETATION: Our data is clustered indicating a correlation between BigFoot sightings and “high” rainfall. When the p-value returned by this tool is small (statistically significant), the null hypothesis can be rejected. If the null hypothesis is rejected, then the sign of the z-score becomes important. If the z-score is positive, the observed index is larger than the expected, indicating high attributes are clustered in the study area. If the z-score is negative, the observed index is smaller than the expected, indicating low attribute values are clustered in the study area. With these results, one might suggest, BigFoot sightings were “mistakenly” identified because of the rain or hazy view of the animal. Could they have been a bear standing on its hind legs? We will need to complete further analysis to decide.

D. Use the Spatial Autocorrelation (Moran’s I) tool to verify correlation between BigFoot Sightings and areas that are known to have black bears.

1. Examine the **NorthPacificForestDamage** layer. We need to repeat the steps from before to spatially join the attributes of the NorthPacificForestDamage layer to the BigFoot sightings layer.
2. In the Contents pane, right-click **BigFootSightings** then click **Joins and Relates** and select **Spatial Join**.
3. In the Spatial Join window set the following parameters:
 - i. Target Features: **BigFootSightings**
 - ii. Join Features: **NorthPacificForestDamage**
 - iii. Output Feature Class: **BigFootSightings_BlackBear**
 - iv. Join Operation: **Join one to one**
 - v. Keep All Target Features (**checked**)
 - vi. Match Option: **Closest**
 - vii. Distance Field Name: **Distance**
 - viii. Click **OK**
4. Open the attribute table for the new **BigFootSightings_BlackBear** layer and scroll through the attributes paying close attention to the **Distance** field.

| NorthPacificForestDamage BigFootSightings_BlackBear | | | | | |
|--|------------|----------|------------|---------------|--------|
| Field: | Selection: | | | | |
| | OBJECTID * | Shape * | Join_Count | Distance | TARGET |
| | 1 | Point ZM | 1 | 12044.425796 | |
| | 2 | Point ZM | 1 | 279823.802786 | |
| | 3 | Point ZM | 1 | 536.584444 | |
| | 4 | Point ZM | 1 | 4701.134575 | |
| | 5 | Point ZM | 1 | 3063.616459 | |
| | 6 | Point ZM | 1 | 4240.569794 | |

The Distance field (created during the join) gives the distance in meters between the target feature (BigFoot Sightings) and the closest join feature (Forest Damage). Because Black Bears typically travel up to 100 miles away from their home range, we only want to include points that are within that distance. To make this easier, we will create a new field to convert the distance from meters into miles.

5. In the top left corner of the BigFootSightings_BlackBear attribute table, click **Add Field**.

i. Field Name: **Dist_Miles**

ii. Data Type: **Float**

| | Visible | Read Only | Field Name | Alias | Data Type | |
|--|-------------------------------------|--------------------------|------------|---------|-----------|--|
| | <input checked="" type="checkbox"/> | <input type="checkbox"/> | AGNT_NM | AGNT_NM | Text | |
| | <input checked="" type="checkbox"/> | <input type="checkbox"/> | HOST | HOST | Text | |
| | <input checked="" type="checkbox"/> | <input type="checkbox"/> | FORTYPE | FORTYPE | Text | |
| | <input checked="" type="checkbox"/> | <input type="checkbox"/> | ACRES | ACRES | Double | |
| | <input checked="" type="checkbox"/> | <input type="checkbox"/> | Dist_Miles | | Float | |

6. Close the Fields table and click **Yes** to Save all changes.

7. Right-click **Dist_Miles** field header and select **Calculate Field**.

There are 1609 meters in a mile; we will create an expression in the Field Calculator to convert meters from the Distance field to miles in the new Dist_Miles field.

i. In the Field Calculator, enter **!Distance! / 1609**. *Hint: double-click Distance, click the divide (/) button, and then enter the numbers 1609.*

Calculate Field

This tool modifies the input data.

Input Table: BigFootSightings_BlackBear

Field Name (Existing or New): Dist_Miles

Expression Type: Python 3

Expression:

Fields: OBJECTID, Shape, Join_Count, Distance, TARGET_FID, Class, Year

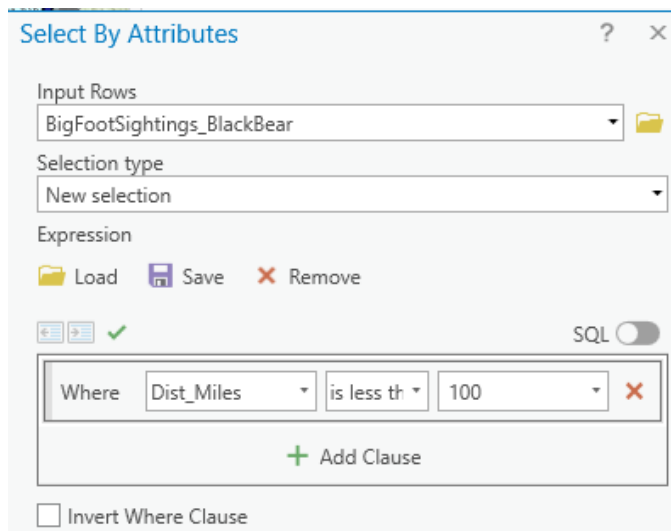
Helpers: .as_integer_ratio(), .capitalize(), .center(), .conjugate(), .count(), .decode(), .denominator()

Insert Values: *

=

!Distance! / 1609

- ii. Click **Apply**.
 - iii. Close the Calculate Field window.
8. Because we only want to include sightings within 100 miles of damaged forest, we will create a query to select only those features meeting that requirement. In the Attribute Table, click the **Select by Attributes** button.
9. In the Select by Attributes window create a **New expression**.
 - i. Set the following query: *Where **Dist_Miles** is less than or equal to 100.*



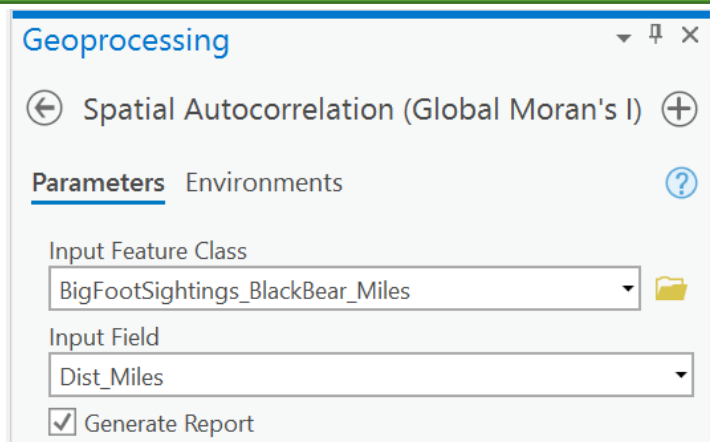
- ii. Click **Apply** then close the Select by Attributes window. Only the features that meet the 100 mile distance are selected.
10. Export the selected features to a new layer. In the Contents pane, right-click **BigFoot_Sightings_BlackBear** layer, click **Data**, then click **Export Features**.
 - i. In the Export Features window, name the output file **BigFoot_Sightings_BlackBear_Miles** and click **OK**.

LET'S REVIEW: We joined the attributes of the NorthPacificForestDamage layer to the BigFoot sightings layer, converted the distance field from meters to miles. Queried and selected distances of 100 miles or less and converted that to a new layer. Whew! We are now ready to begin our analysis.

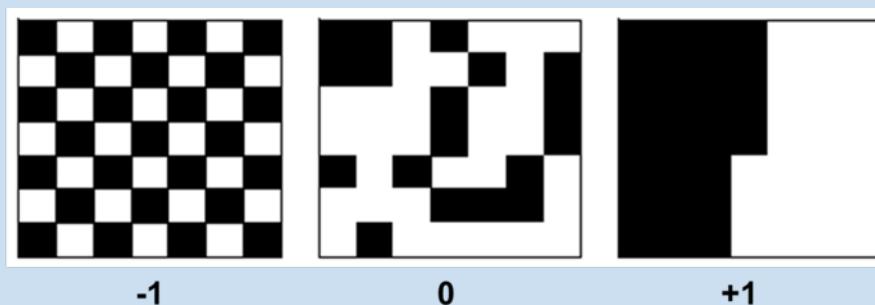
11. From the *Analyzing Patterns* toolset, click **Spatial Autocorrelation (Global Moran's I)** to open the tool.

The Spatial Autocorrelation (Global Moran's I) tool measures clustering or dispersion based on feature locations and attribute values. The null hypothesis for the Spatial Autocorrelation (Moran's I) tool statistic states the attribute being analyzed is randomly distributed among the features in the study area.

12. Set the following parameters:
 - i. Input Feature Class: **BigFoot_Sightings_BlackBear_Miles**.
 - ii. Input Field: **Dist_Miles**.
 - iii. Generate Report: **checked**.
 - iv. Accept the other defaults and click **Run**.



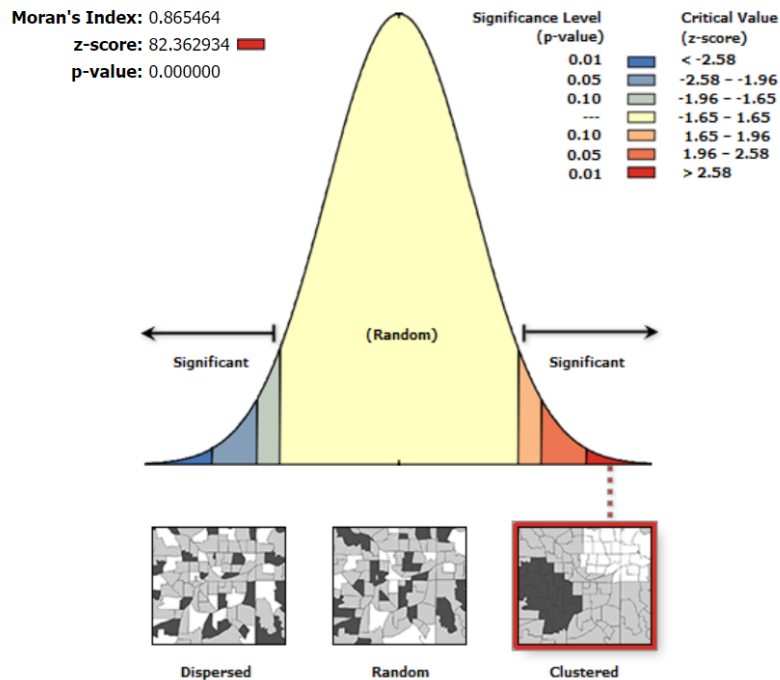
STATS 101: The Moran's Index is a measure of spatial autocorrelation. In the example below: the white and black squares are perfectly dispersed so Moran's I would be -1. If the white squares were stacked to one half of the board and the black squares to the other, Moran's I would be close to +1. A random arrangement of square colors would give Moran's I a value that is close to 0.



The expected index is simply a measure of spatial autocorrelation in a hypothetical random distribution indicating a spatial random pattern. A variance gives us a value indicating how far away, or the difference of the actual index, from the expected index. The z-scores and p-values tell you if you can reject the null hypothesis

13. Open the **Report File** and examine the results.

Spatial Autocorrelation Report



Given the z-score of 82.362934, there is a less than 1% likelihood that this clustered pattern could be the result of random chance.

INTERPRETATION: the report states very clearly that: Given the z-score of 82.36 there is less than 1% likelihood that this clustered pattern could be the result of random chance. We are 99% confident that the clustering in this dataset is "NOT" random



Our data is clustered indicating a correlation between BigFoot sightings and forest damage by bears. When the p-value returned by this tool is small (statistically significant), the null hypothesis can be rejected. The p-value is statistically significant and the z-score value is positive, indicating high values and low values for the attributes are clustered in the study area. With these results, we can confirm our theory that there is a correlation between BigFoot Sightings and areas known to have bears.

14. **Save** your project.



Congratulations! You have completed Exercise 2.