# Which cities in North Carolina have the least crime?

### Why are we looking into this topic?

- North Carolina is a growing state with Charlotte and Raleigh leading the way
- Industries such as science and finance are becoming a hub in the state
- Majority of those in class live in NC and may move around the state to pursue new opportunities.



## **Every city offers something different**

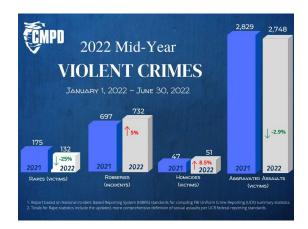
- Cost of living differs from city to city
- One may want a more affordable house/cheaper rent but work in a higher cost of living area
- One may want a suburb feel or a city feel
- A common example we have seen is up north. Many way work in NYC however live in New Jersey where the cost is cheaper. (New Jersey is on average about 30% cheaper to live in than NYC)





## Why does this deal with city safety?

- For those wanting to move anywhere in the state or to the state, it is important to have an understanding of where you're going.
- Safety should be a top concern as the environment is new
- Seeing trends in city safety overs years can give direction to the future



## What we are looking to explore?

- 1. Which cities have the least crime?
- 2. What cities are safe in North Carolina?
- 3. What is the difference in crime from 2016 and 2019?

#### **Data Source**

- The data sets we used to explore were from the FBI and we choose to use the years 2019 and 2016.
- These data sets contained violent and non-violent crimes.



### **Data Exploration**

What we were looking for in our data

- Recent crime reports
- Credible website we can download information from (in our case it was an excel file)

| Table 6  NORTH CAROLINA Offenses Known to Law Enforcement by City, 2016  Data Declaration |        |     |    |  |    |     |     |       |     |       |     |    |          |            |               |                                      |                                               |                                              |         |                    |                |          |                   |                           |       |
|-------------------------------------------------------------------------------------------|--------|-----|----|--|----|-----|-----|-------|-----|-------|-----|----|----------|------------|---------------|--------------------------------------|-----------------------------------------------|----------------------------------------------|---------|--------------------|----------------|----------|-------------------|---------------------------|-------|
|                                                                                           |        |     |    |  |    |     |     |       |     |       |     |    | City     | Population | Violent crime | Murder and nonnegligent manslaughter | Rape<br>(revised<br>definition <sup>1</sup> ) | Rape<br>(legacy<br>definition <sup>2</sup> ) | Robbery | Aggravated assault | Property crime | Burglary | Larceny-<br>theft | Motor<br>vehicle<br>theft | Arson |
|                                                                                           |        |     |    |  |    |     |     |       |     |       |     |    | Aberdeen | 7,549      | 25            | 0                                    |                                               | 3                                            | 10      | 12                 | 289            | 52       | 226               | 11                        | 1     |
| Ahoskie                                                                                   | 4,883  | 43  | 2  |  | 3  | 8   | 30  | 291   | 81  | 201   | 9   | 0  |          |            |               |                                      |                                               |                                              |         |                    |                |          |                   |                           |       |
| Albemarle                                                                                 | 16,024 | 107 | 1  |  | 6  | 24  | 76  | 771   | 201 | 538   | 32  | 4  |          |            |               |                                      |                                               |                                              |         |                    |                |          |                   |                           |       |
| Angier                                                                                    | 5,097  | 14  | 0  |  | 2  | 2   | 10  | 129   | 62  | 59    | 8   | 3  |          |            |               |                                      |                                               |                                              |         |                    |                |          |                   |                           |       |
| Apex                                                                                      | 47,324 | 40  | 1  |  | 2  | 13  | 24  | 601   | 83  | 501   | 17  | 1  |          |            |               |                                      |                                               |                                              |         |                    |                |          |                   |                           |       |
| Asheboro                                                                                  | 26,230 | 104 | 1  |  | 11 | 39  | 53  | 1,501 | 300 | 1,161 | 40  | 10 |          |            |               |                                      |                                               |                                              |         |                    |                |          |                   |                           |       |
| Asheville                                                                                 | 89,546 | 537 | 10 |  | 48 | 144 | 335 | 4,357 | 752 | 3,345 | 260 | 7  |          |            |               |                                      |                                               |                                              |         |                    |                |          |                   |                           |       |
| Atlantic Beach                                                                            | 1,503  | 14  | 0  |  | 4  | 1   | 9   | 130   | 42  | 84    | 4   | 0  |          |            |               |                                      |                                               |                                              |         |                    |                |          |                   |                           |       |

#### **Data Analysis Phase**

- The first cleaned our data to drop columns/rows that were titles and double counted values
- We determined a threshold that the model would learn to detect a safe and unsafe city in NC using excel by calculating a crime score.
- Total crimes = Property Crimes + Violent crimes

Total Number of crimes/Population of the city\*100= Crime Index Score

• We needed to create definition of "safe" which we determined by using the third quartile.

#### **Data Analysis Phase**

- In SQL, we created three tables and a database that our machine learning model would pull from.
- We used a Logistic Regression Model because it predicts binary outcomes (ie. safe or unsafe)
- This model analyzes the available data, and when presented a new sample, mathematically determines its probability of belonging to a class. If the probability is above a certain cutoff point, the sample is assigned to that class. If the probability is less than the cutoff point, the sample is assigned to the other class.

#### **Machine Learning Model**

#### **Logistic Regression Model**

#### • Limitations:

 It is difficult to capture complex relationships using logistic regression. More powerful and complex algorithms such as Neural Networks can easily outperform this algorithm.

#### Benefits:

- Good accuracy for many simple data sets and it performs well when the dataset is linearly separable.
- It is simple to implement and doesn't require high computation power.