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## Why Tennessee Tornadoes

The members of this team live or have lived in the state of Tennessee. We have noticed a change in the frequency of tornadoes in the state during our collective time living in the state. The team was curious to see if data supports our concerning hypotheses. We believe this information can be used for homebuyers and insurance companies.

The data was obtained from data.world and represents tornado tracks from the United States, Puerto Rico, and the US Virgin Islands. For this project we filtered the data for tornadoes in the state of Tennessee.

### What will the data show?

Our Research questions for the data to answer are:

- 1. Have tornadoes increased in intensity in the last 50 years in the state of Tennessee?
- 2. What counties are most likely to have more tornadoes?
- 3. Has the frequency of tornadoes in Tennessee increased since 1950?

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## Data Exploration

After deciding which data to use, we narrowed our data to just the state of Tennessee using Excel. Then using Python and Pandas to search for missing values, removing columns that were providing information that wasn't needed for our research questions, and to adjust values that switched in 1996 with reporting protocols to match the previous years. Each tornadoes starting and ending counties were calculated in <a href="GetCounties.ipynb">GetCounties.ipynb</a> with the geopy library and exported to <a href="GetCounties.csv">Counties.csv</a>

## Analysis

Machine learning models will be created to predict the following:

- 1. Magnitude of tornadoes
- 2. Location of tornadoes
- 3. Amount of property damage

## **Tools and Technology**

#### **Data Cleaning:**

- Python 3.7.13 (pandas and geopy libraries)
- Jupyter Notebook 6.4.8

#### Database:

- PostgreSQL 11.16
- pgAdmin 4 v6.8
- AWS

#### **Connecting Database:**

- Psycopg2

#### **Machine Learning:**

- Python (pandas, imbalance-learn, scikit-Learn, numpy libraries)
- Jupyter Notebook

#### Dashboard:

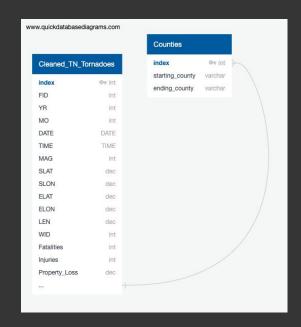
- Tableau
- Javascript
- Bootstrap
- Leaflet
- D3
- HTML
- CSS

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## **Connecting the Database**

Overview of ERD and screenshot of connection stream

```
# declare a new PostgreSQL connection object
    conn = connect(
        dbname = "",
        user = "postgres",
        host = "tennesseetornadoes.cdilutmdgtwo.us-east-1.rds.amazonaws.com",
        port = "5432",
        password = password
    # print the connection if successful
    print ("psycopg2 connection:", conn)
except Exception as err:
    print ("psycopg2 connect() ERROR:", err)
psycopg2 connection:
cr = conn.cursor()
cr.execute('SELECT * FROM total tn tornadoes;')
tmp = cr.fetchall()
# Extract the column names
col names = []
for elt in cr.description:
    col names.append(elt[0])
# Create the dataframe, passing in the list of col names extracted from the description
df = pd.DataFrame(tmp, columns=col names)
df.head()
```

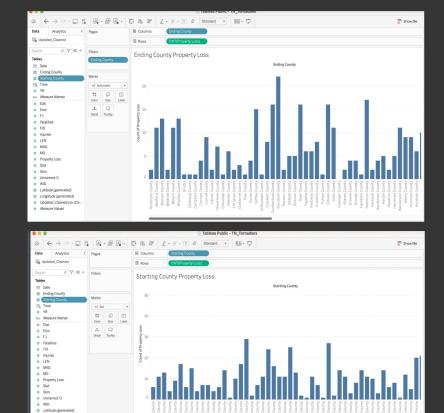


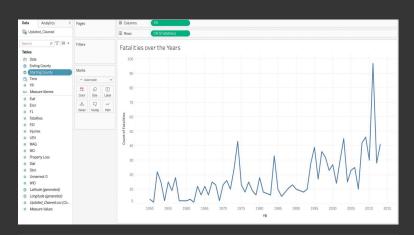
### **Database**

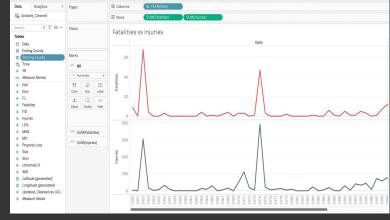
⊕ Longitude (generated)

# Updated\_Cleaned.csv (Co.

#### Tables were created in Tableau







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## **Machine Learning**

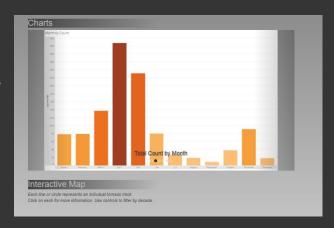
```
# prepare the dataframe
   df 3 = df.drop(['Unnamed: 0','Unnamed: 0.1','TIME',"WID","starting county",
                  'SLAT', 'SLON', 'ELAT', 'ELON', 'LEN', 'ending county', 'FID'], axis=1)
   df.columns
   target = 'Property Loss'
   X = pd.get dummies(df 3.drop([target],axis = 1))
10
11
   # Create our target
   y = df[target]
14
15
   X train, X test, y train, y test = train test split(X,
17
18
                                                         random state=1)
19
   from sklearn.pipeline import make pipeline
   from sklearn.linear model import SGDClassifier
   from sklearn.preprocessing import StandardScaler
23
   model = make pipeline(StandardScaler(),LogisticRegression(solver='newton-cg'))
   model.fit(X train, y train)
26
   # model = LogisticRegression(solver='newton-cg', random state=1)
   #'liblinear', 'sag', 'saga
```

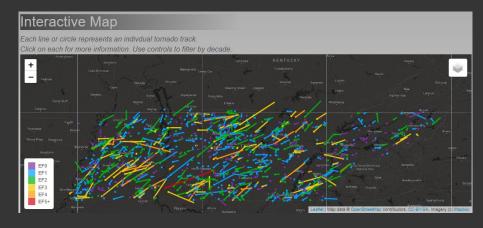
# **Machine Learning:**

### Dashboard

The Dashboard has been created using Javascript to be displayed as an interactive webpage. Tableau, CSS, D3, HTML, and Bootstrap components have been used to enhance the displays. The map was created using Leaflet







### Result of Analysis:

Based on data, the count of tornadoes is, in fact, *increasing* over time. However, the average magnitude appears to decrease over time. This could be a result of a higher frequency of lower magnitude tornadoes. So, tornadoes may not necessarily be getting stronger, but they are becoming more numerous.

Combined magnitudes based on time of day show that tornadoes are more frequent and strongest between 4PM and 7PM. Property damage losses are measured by categorical range of dollar values instead of actual dollar amounts. Therefore, property loss statistics cannot give an actual dollar amount but can show the trend in each county.

With 80% accuracy we are able to predict the months in which tornadoes are likely to occur.

### Recommendations for Future Analysis: