

Plotly: Data Visualization

Presented by: Christina DePerro

What is Plotly?

- Data visualization library in Python founded in Canada 2012 and launched as open-source library in 2015
- Acts as a JavaScript-based library creating a JavaScript code but binds with other languages, where it can convert the code to other languages, like Python (commonly used)
- Optimal for:
 - Creating interactive charts, allowing users to engage with the visualizations
 - Supporting a wide range of charts and graphs (40+) such as line graphs, bar charts, stacked charts, box plots, scatter plots, heat maps, 3D models, etc.
 - Integrating with web applications and other tools including Pandas, NumPy, SciPy, and Matplotlib

Why would a data scientist use Plotly?

- Data Exploration & Dashboard Development
 - Visualize the relationships between the dataset variables, identify any outliers
 - View and analyze time series, geospatial and multidimensional models
 - Share visualizations with others using URLs or integrate them with web-based applications such as:
 - HTML, JavaScript, CSS, Jupyter Notebook
 - Use high-level interfaces for quick and easy visualizations or for more extensive plots:
 - 1. Plotly Express: High level API used to create simpler visuals, minimal code required
 - 2. Plotly Graph Objects: Low-level API used for customization purposes
 - 3. Dash: Python web framework, minimal code required
 - 4. Jupyter Notebook: Interactive environment to conduct Plotly data analysis
 - 5. JavaScript API: Web application for embedding Plotly visuals
 - 6. Plotly for R: Create interactive visuals in R
 - 7. Plotly for Pandas: Use Pandas DataFrames when creating plots

Getting started in Jupyter Notebook

Launch Jupyter Notebook and install Plotly

Ensure Jupyter is upgraded to latest version, then install plotly. (Access files from GitHub https://github.com/cdeperro1/Plotly and refer to "Plotly Installation Instructions" file for details)

Import data using Pandas

Today's Goal: Analyze national tornado trends from US Department of Commerce's NOAA dataset.

Dataset includes:

- 2020-2024 reporting of tornadoes in the US
- Tornado magnitude levels follow EF Scale, rating their intensities based on wind, speed and damage
 - **EFU:** Weakest, unknown speed
 - **EF0:** Weak: 65-85mph
 - **EF1:** Moderate: 86-110mph
 - **EF2:** Strong; 111-135mph
 - **EF3:** Severe; 136-165mph
 - EF4: Devastating; 166-200mph
 - **EF5:** Catastrophic; >200mph

pip install plotly

Requirement already satisfied: plotly in /opt/anaconda3/lib/python3.12/site-packages (6.0.1)
Requirement already satisfied: narwhals>=1.15.1 in /opt/anaconda3/lib/python3.12/site-packages (from plotly) (1.31.0)
Requirement already satisfied: packaging in /opt/anaconda3/lib/python3.12/site-packages (from plotly) (23.2)
Note: you may need to restart the kernel to use updated packages.

import pandas as pd df = pd.read csv('/Users/christinadeperro/Desktop/0IM7502/Data Sources/2020-2024 tornado.csv') print(df.head()) location county_zone state month year \ ANCHORAGE ANCHORAGE MUNI TO BIRD C... AK 4/19/24 April 2024.0 UNION CHAPEL PICKENS CO. 1/11/20 2020.0 January HOLLY POND CULLMAN CO. **JOPPA** CULLMAN CO. 1/11/20 2020.0 ARAB MARSHALL CO. AL 1/11/20

time time zone Type magnitude deaths injuries property damage 19:13 AKST-9 Tornado 0.00K EF2 11:11 CST-6 3.0 7.0 0.00K 13:03 Tornado EF0 0.0 0.00K 13:13 CST-6 Tornado EF1 0.0 0.0 0.00K 4 13:14 CST-6 Tornado 0.0 0.0 0.00K

crop_damage 0.00K 0.00K 0.00K

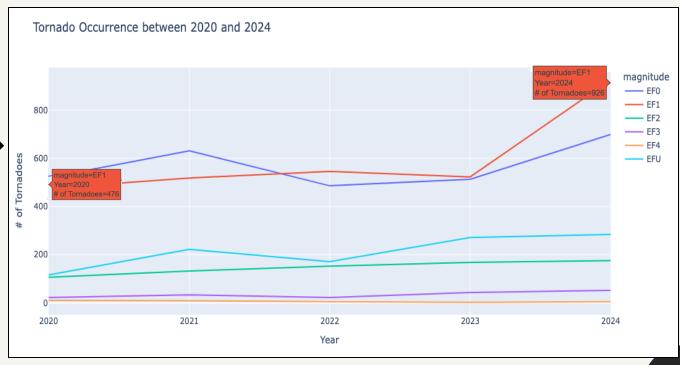
3 0.00K 4 0.00K

Creating a Basic Line Graph

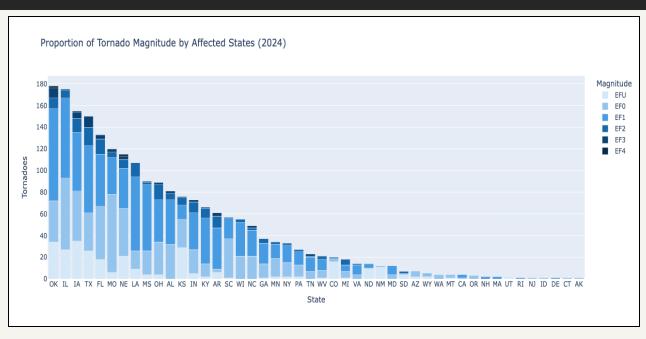
```
Provides flexibility to manipulate our DataFrame: "2020-2024 Tornado" csv
import plotly.express as px
import pandas as pd
                                                                                                        Group data by 'year' and 'magnitude' (df.groupby) and count rows in each
                                                                                                        group (.size). Add new column called 'count' (.reset index). Outputs are stored
magnitude counts = df.groupby(['year', 'magnitude']).size().reset index(name='count')
                                                                                                        in magnitude counts.
magnitude_counts['year'] = pd.to_datetime(magnitude_counts['year'], format='%Y')
                                                                                                        Convert values in 'year' column to datetime object using format='%Y'
fig = px.line(
    magnitude_counts,
    x="vear",
                                                                                                              Create line graph using plotly express (px.line). Pull in data from new
    y="count",
                                                                                                              DataFrame magnitude counts. Identify x-value and y-values. The line
    color="magnitude",
                                                                                                              graph will color code the tornados by their respective magnitude levels.
    title='Tornado Occurrence Across Years',
    labels={'count': '# of Tornadoes', 'year': 'Year'}
                                                                                                              Provide a title and customize the axis labels.
fig.update layout(
    xaxis=dict(
        tickformat='%Y'.
                                                                                                              Update the layout for the x-axis using (dict) to modify the ticks. '%Y'
        tickvals=magnitude_counts['year'].dt.year.unique()
                                                                                                             must display the year in a 4-digital format. Unique values from
                                                                                                             magnitude counts will be extracted to appear as ticks on x-axis.
fig.show()
```

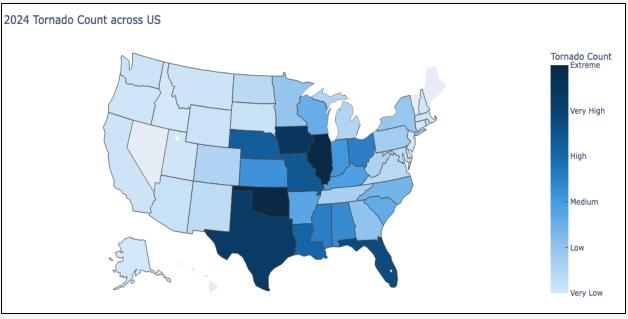
Visualizing the Line Graph

```
import plotly.express as px
import pandas as pd
magnitude counts = df.groupby(['year', 'magnitude']).size().reset index(name='count')
magnitude counts['year'] = pd.to datetime(magnitude counts['year'], format='%Y')
fig = px.line(
    magnitude_counts,
    x="year",
    y="count",
    color="magnitude",
    title='Tornado Occurrence Across Years',
    labels={'count': '# of Tornadoes', 'year': 'Year'}
fig.update_layout(
    xaxis=dict(
        tickformat='%Y'.
        tickvals=magnitude_counts['year'].dt.year.unique()
fig.show()
```



Analyzing Aggregated Data Results





Stacked bar chart displaying the distribution of tornado magnitude levels for each affected state in 2024.

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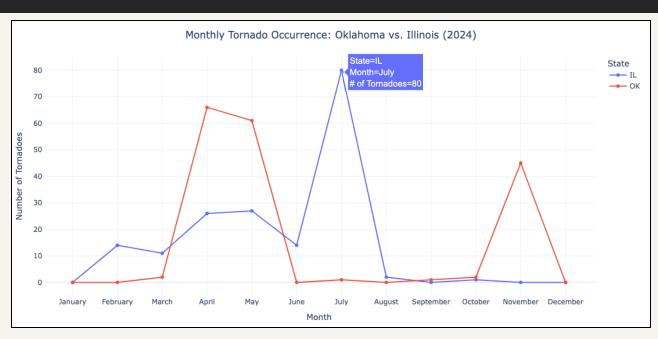
For dashboard purposes, a data scientist might add a filter by region to aggregate the data and provide a cleaner visualization.

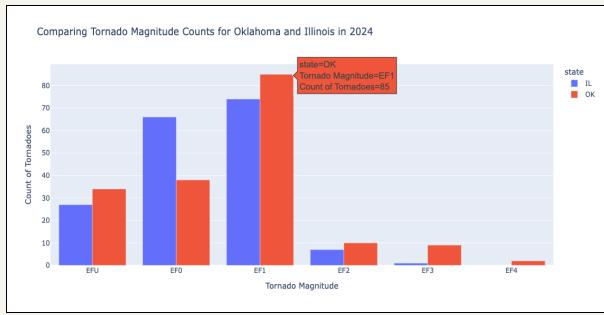
Choropleth map showing the frequency levels of tornado occurrence across each state in 2024.



Increase granularity and use coordinate data to create a heat map on the county or zip code level.

Analyzing Granular Data Results





Time series line graph comparing the top two states with the highest tornado occurrence in 2024.

Grouped bar chart comparing tornado magnitude levels against the top two tornado-prone states in 2024.

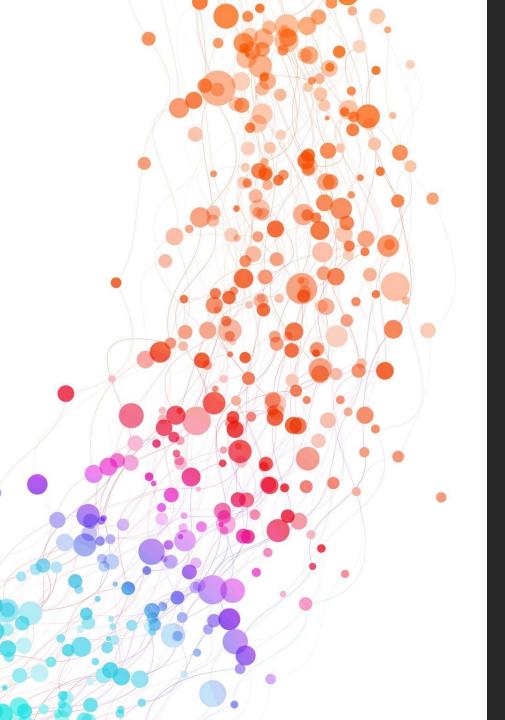


Import weather data to compare and analyze storm conditions attributing to tornado frequency.

Class Tutorial

GitHub File Name Reference: "Tornado_Pie_Shell"

- 1 Create a pie chart
- 2 Filter data to 2024
- 3 Group the data to view the proportion of tornado magnitudes across all states



Thank You