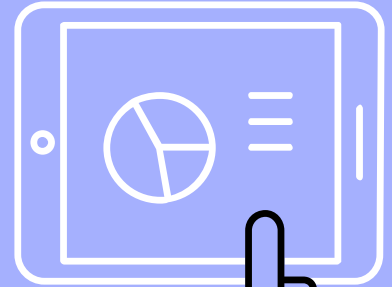
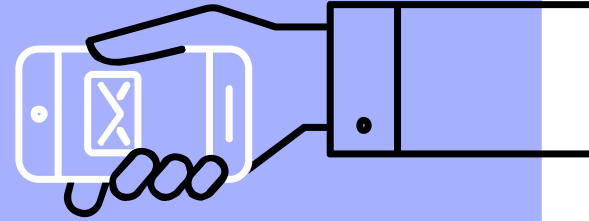
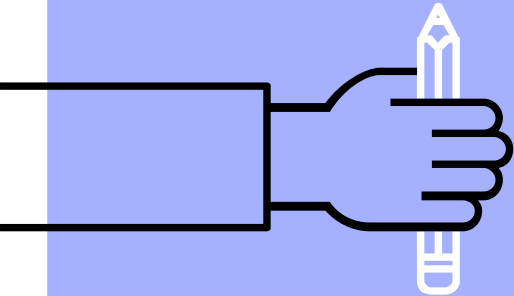
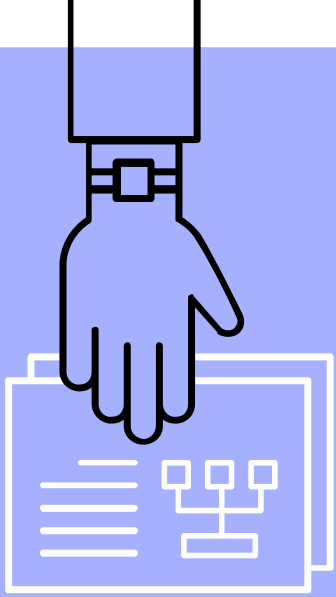


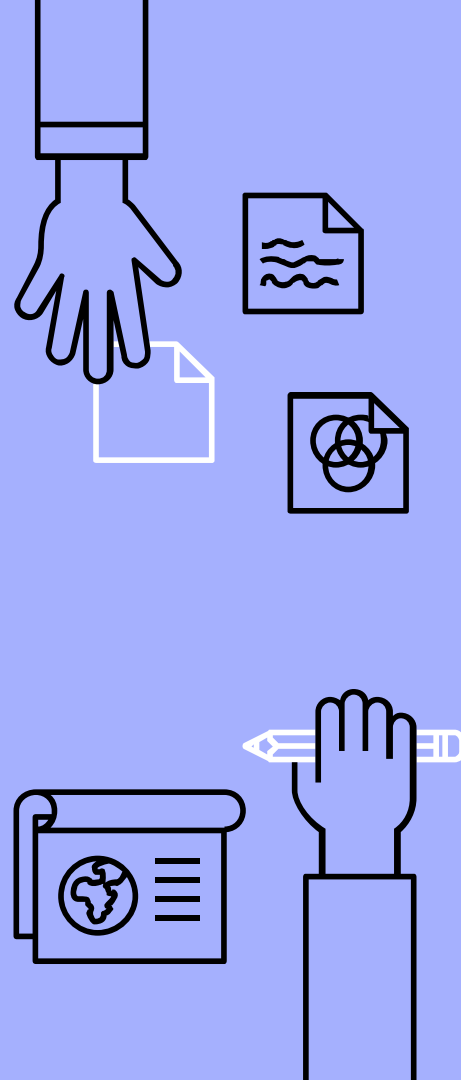
Storms in Florida

Presentation by:
Karen S, Winnie D,
and Jessica A



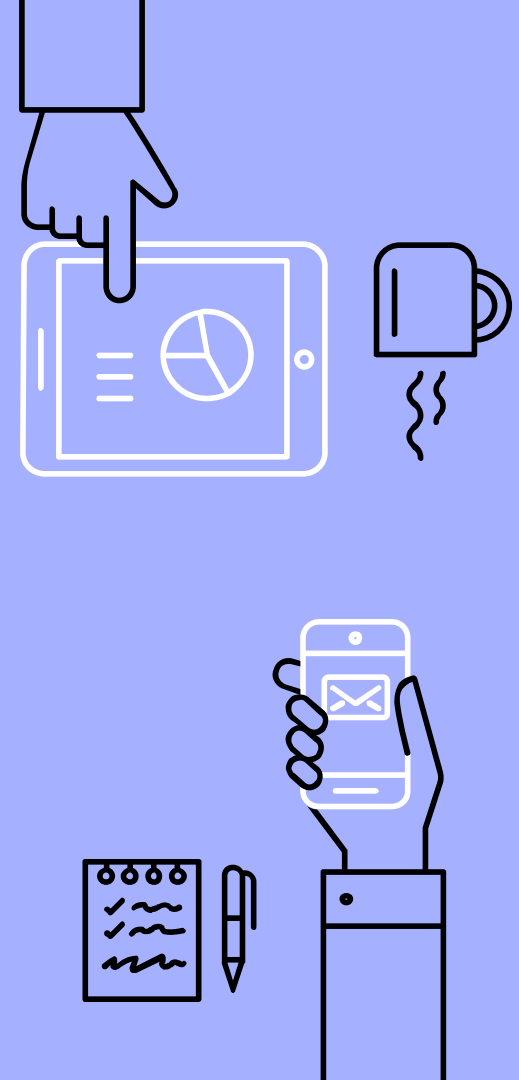
Basic Information/Objective

In this project, we look into a variety of storm events that occur in the state of Florida, from thunderstorms to flooding. We then retrieved data from NOAA that determines how long each storm event occurs for, and how the aftermath of the storm affected officials and residents.



Description of Dataset

The dataset describes the weather events that occurred in Florida 2013. It gives the 67 counties of Florida, the event types, the dates of the event types. In the dataset, it shows every storm event across all states, however, Florida is the main focus. The dataset also includes the month, day, and year of these storm events.



Data Preparation

Data Loading:

The data was loaded as a CSV file

code:

```
df1 = pd.read_csv("stormdata_2013.csv")
```

df1

Data Cleaning:

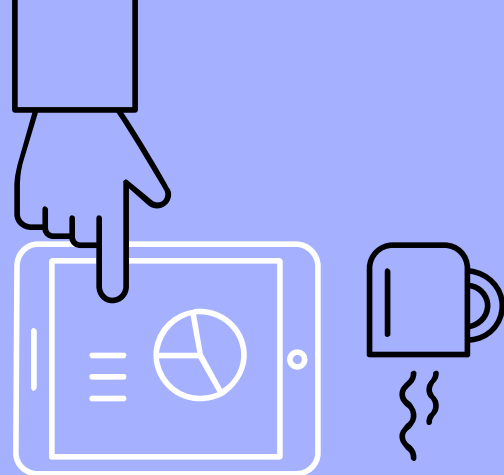
We scaled the data down to Florida and took out data we did not understand or felt was unnecessary to keep. We wanted to focus on Florida out of all the states and we wanted to see the event type, the start and end of the event, the damages, and magnitude.

To select only Florida, we made a new dataframe and called it df2. We equaled df2 to our first dataframe, df1, and selected the column "STATE". Afterwards we equaled it to "FLORIDA", so the data would only be retrieved from the state of Florida. We lastly called df2, so it could print the result.

code:

```
df2 = df1[df1['STATE'] == 'FLORIDA']
```

df2

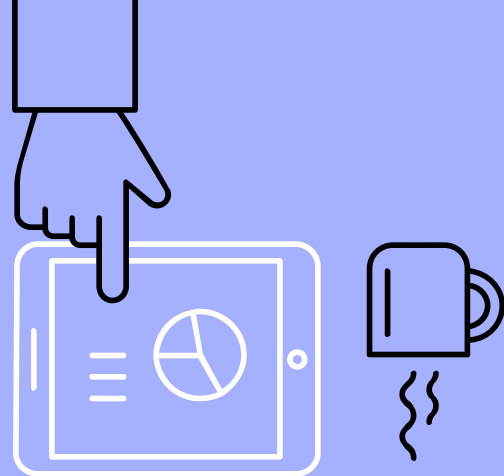


Data Preparation (cont.)

In addition to prepping the dataset, we dropped many columns that were not relevant to what we were looking for and didn't have much information to use. These columns were dropped using the drop function.

```
In [4]: florida2 = florida.drop(['EVENT_ID', 'STATE_FIPS', 'EPISODE_NARRATIVE', 'EVENT_NARRATIVE', 'LAST_MOD_DATE', 'LAST_MOD_TIME',  
                                'LAST_CERT_DATE', 'LAST_CERT_TIME', 'ADDCORR_FLG', 'ADDCORR_DATE', 'BEGIN_LOCATION', 'YEAR',  
                                'CZ_FIPS', 'BEGIN_AZIMUTH', 'END_RANGE', 'END_AZIMUTH', 'BEGIN_RANGE', 'END_LOCATION',  
                                'BEGIN_LAT', 'BEGIN_LON', 'END_LAT', 'END_LON', 'CATEGORY', 'FLOOD_CAUSE', 'MAGNITUDE_TYPE',  
                                'TOR_F_SCALE', 'TOR_LENGTH', 'TOR_WIDTH', 'TOR_OTHER_WFO', 'SOURCE', 'TOR_OTHER_CZ_STATE',  
                                'TOR_OTHER_CZ_FIPS', 'TOR_OTHER_CZ_NAME', 'INJURIES_DIRECT', 'INJURIES_INDIRECT', 'DEATHS_DIRECT',  
                                'DEATHS_INDIRECT', 'DAMAGE_CROPS', 'WFO', 'CZ_TIMEZONE', 'STATE', 'EPISODE_ID'], axis='columns')
```

florida2



Data Preparation Images/Code

```
In [22]: import pandas as pd
import numpy as np
import csv
import matplotlib.pyplot as plt
import seaborn as sns
from datetime import datetime
```

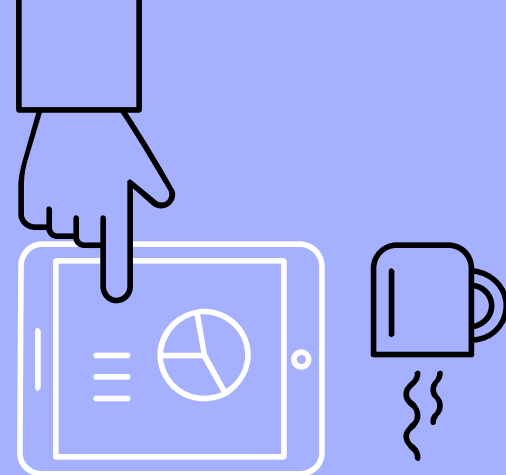
```
In [23]: df = pd.read_csv('stormdata_2013.csv')
df
```

```
Out[23]:
```

	BEGIN_YEAR	BEGIN_MONTH	BEGIN_DAY	BEGIN_TIME	END_YEAR	END_MONTH	END_DAY	END_TIME	EPISODE_ID	EVENT_ID	STATE	STATE_FIPS	...	
0	2013	1	1	0	2013	1	31	2359	70300	422180	ALABAMA	1	...	
1	2013	1	1	0	2013	1	31	2359	70580	423795	COLORADO	8	...	
2	2013	1	1	1230	2013	1	1230	71728	431896	CALIFORNIA	6	...		
3	2013	1	1	1800	2013	1	1400	70755	424932	WYOMING	56	...		
4	2013	1	1	1800	2013	1	1400	70753	424935	SOUTH DAKOTA	46	...		

```
In [4]: df2 = df1[df1['STATE'] == 'FLORIDA']
df2
```

YEAR	MONTH	BEGIN_DAY	BEGIN_TIME	END_YEAR	END_MONTH	END_DAY	END_TIME	EPISODE_ID	EVENT_ID	STATE	MONTH_NAME	EVENT_TYPE	CZ_NAME	BEGIN_DATE
2013	1	4	1330	2013	1	4	1342	69951	419577	FLORIDA	January	Funnel Cloud	MIAMI-DADE	1/4/20
2013	1	4	1540	2013	1	4	1545	69951	419578	FLORIDA	January	Funnel Cloud	MIAMI-DADE	1/4/20
2013	1	29	0	2013	1	31	2359	70302	422188	FLORIDA	January	Drought	MADISON	1/29/2
2013	1	29	0	2013	1	31	2359	70302	422187	FLORIDA	January	Drought	INLAND JEFFERSON	1/29/2
2013	1	29	0	2013	1	31	2359	70302	422186	FLORIDA	January	Drought	LEON	1/29/2
...
2013	10	15	1754	2013	10	15	1754	79203	475171	FLORIDA	October	Coastal Flood	Duval	10/15/20
2013	10	15	600	2013	10	15	600	79203	475170	FLORIDA	October	Coastal Flood	Duval	10/15/2
2013	10	15	536	2013	10	15	536	79203	475169	FLORIDA	October	Coastal Flood	Duval	10/15/2

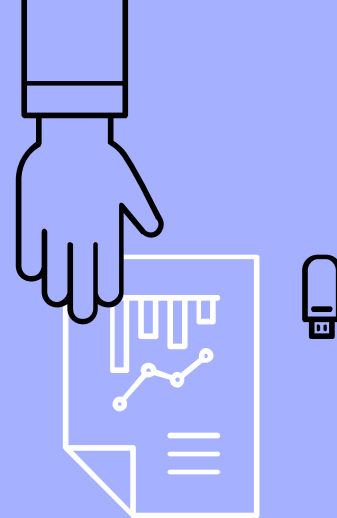


Data Preparation Images/Code (cont)

Out[4]:

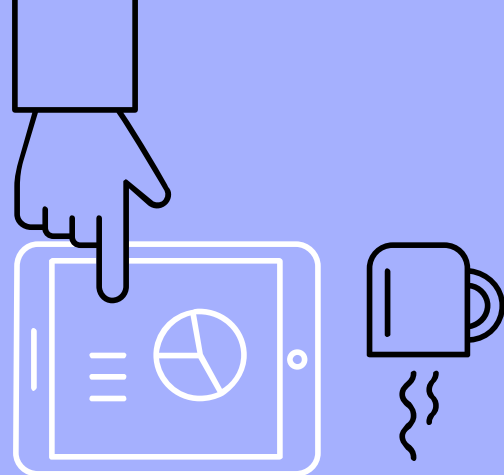
	BEGIN_YEAR	MONTH	BEGIN_DAY	BEGIN_TIME	END_YEAR	MONTH	END_DAY	END_TIME	MONTH_NAME	EVENT_TYPE	CZ_TYPE	CZ_NAME	B
798	2013	01	4	1330	2013	01	4	1342	January	Funnel Cloud	C	MIAMI-DADE	
799	2013	01	4	1540	2013	01	4	1545	January	Funnel Cloud	C	MIAMI-DADE	
4039	2013	01	29	0	2013	01	31	2359	January	Drought	Z	MADISON	
4040	2013	01	29	0	2013	01	31	2359	January	Drought	Z	INLAND JEFFERSON	
4041	2013	01	29	0	2013	01	31	2359	January	Drought	Z	LEON	
...
51161	2013	10	15	1754	2013	10	15	1754	October	Coastal Flood	Z	Duval	
51162	2013	10	15	600	2013	10	15	600	October	Coastal Flood	Z	Duval	
51163	2013	10	15	536	2013	10	15	536	October	Coastal Flood	Z	Duval	
51189	2013	10	17	1648	2013	10	17	1648	October	Funnel Cloud	C	COLLIER	
51231	2013	10	19	1611	2013	10	19	1631	October	Heavy Rain	C	ST. JOHNS	

842 rows × 16 columns



Data Analysis using Descriptive Statistics Images/Code

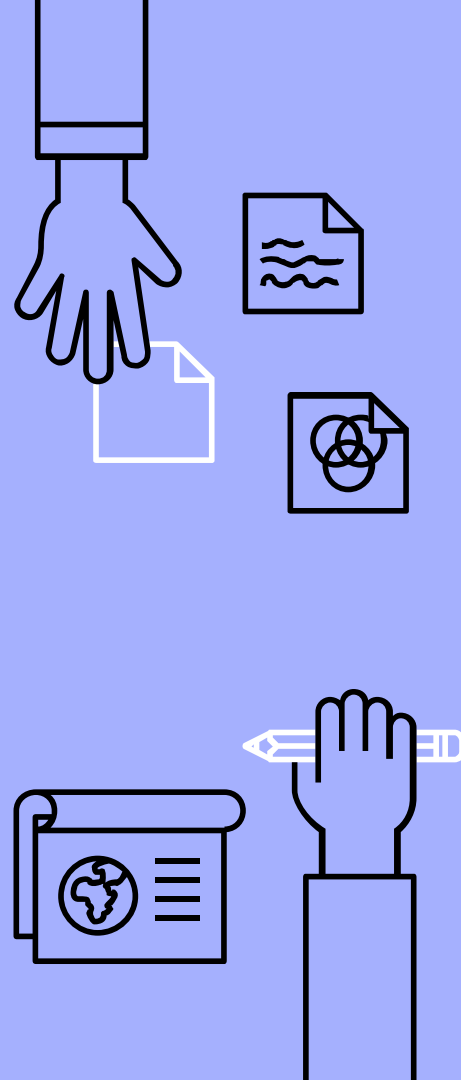
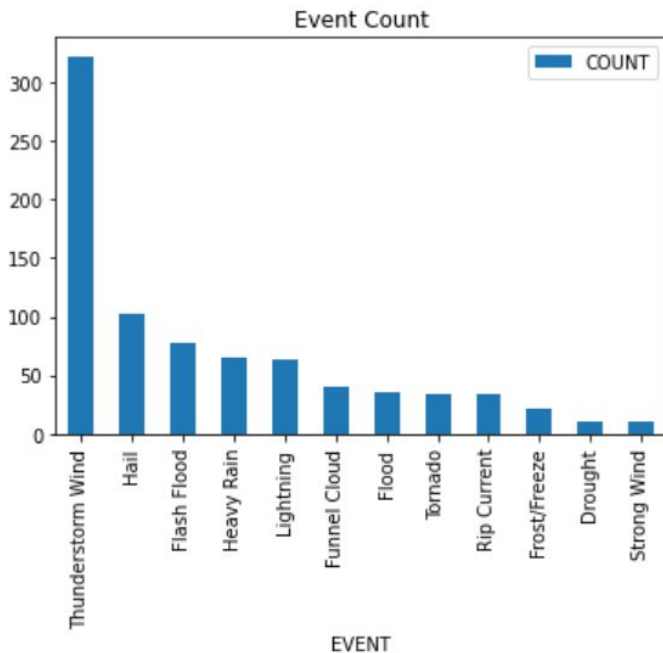
```
In [66]: ▶ df['MAGNITUDE'].mean()  
df['MAGNITUDE'].describe()  
  
Out[66]: count      51932.000000  
mean          17.144878  
std           24.671429  
min            0.000000  
25%            0.000000  
50%            0.750000  
75%           50.000000  
max          105.000000  
Name: MAGNITUDE, dtype: float64
```



Data Visualization

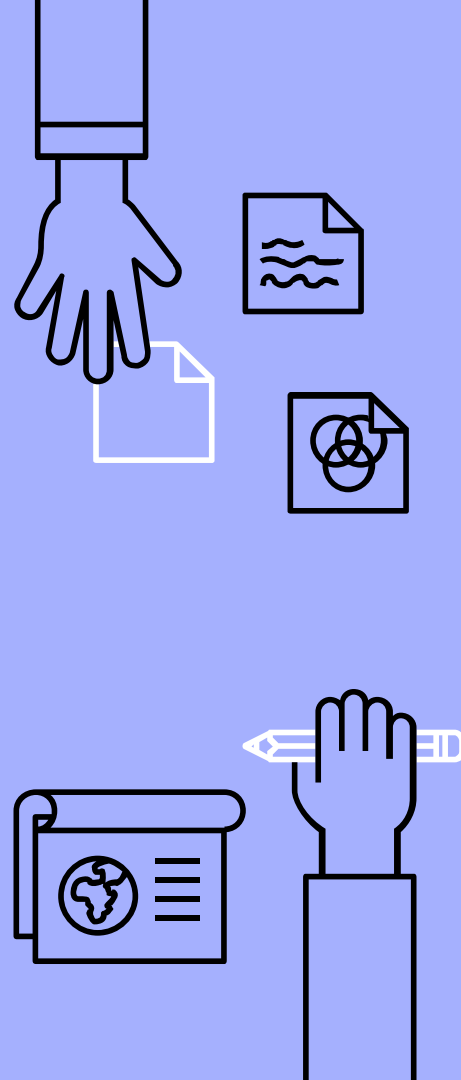
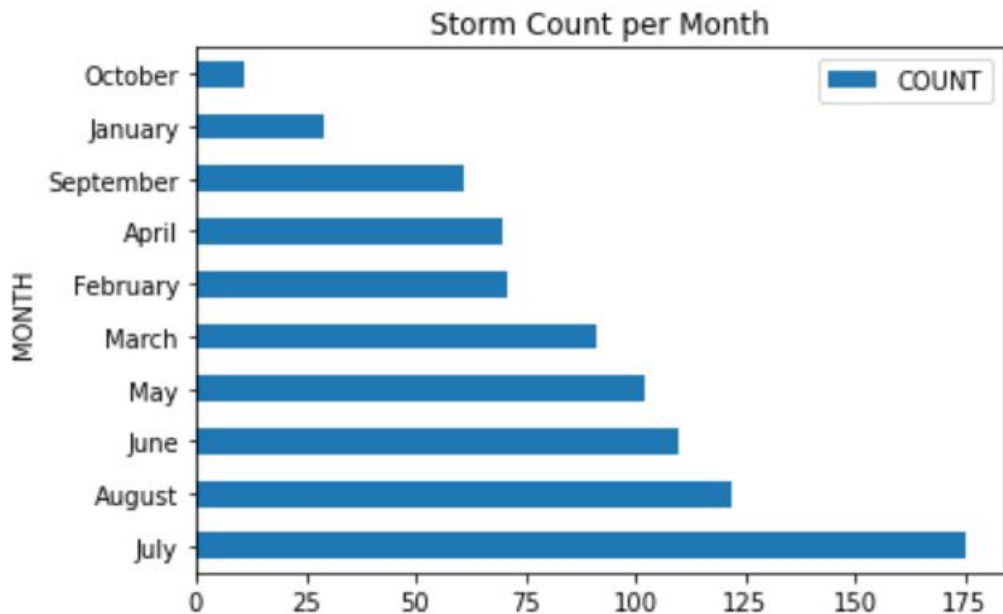
```
In [78]: florida6 = florida5.sort_values(by='COUNT', ascending=False)
florida6 = florida6[:10]
florida6.plot.barh(title="Storm Count per Month",rot=0)
```

```
Out[78]: <AxesSubplot:title={'center':'Storm Count per Month'}, ylabel='MONTH'>
```



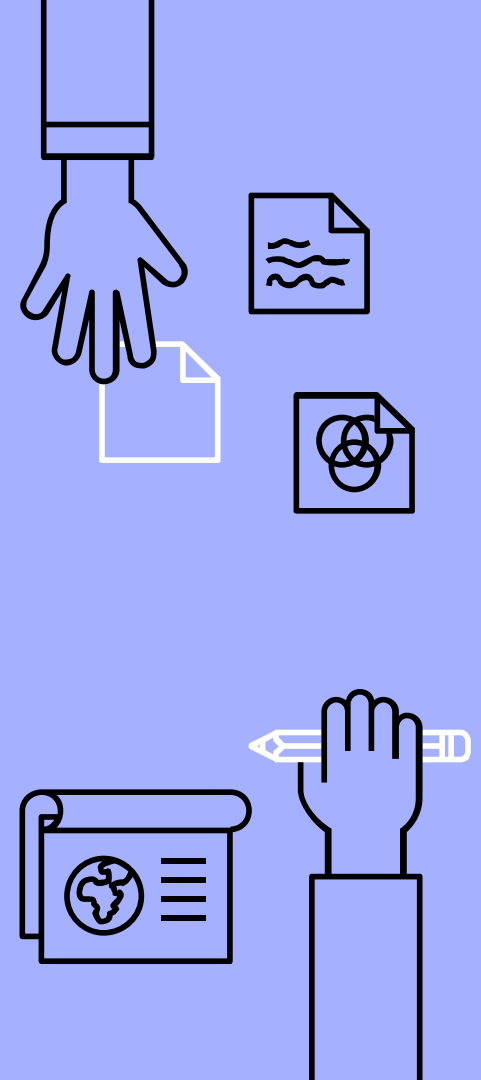
```
In [77]: florida8 = florida7.sort_values(by='COUNT', ascending=False)
florida8 = florida8[:12]
florida8.plot.bar(title="Event Count",rot=90)
```

```
Out[77]: <AxesSubplot:title={'center':'Event Count'}, xlabel='EVENT'>
```



Conclusion

Based on the graphs, we can conclude that most of the storm events occurred in the month of July and the most common event that occurred was thunderstorm winds, with hail being a close second that occurred in 2013. In the beginning of 2013, the longest duration of a drought was for 25 days. The property damage after a lightning storm was the most costly, being \$95,000. Magnitude was used to measure wind speeds and hail size and in 2013, the average magnitude was 17.144878 and the maximum was 105. Overall, these kinds of events are seen more frequently due to climate change and it will only increase in the future.



Source

https://www.ncdc.noaa.gov/stormevents/listevents.jsp?eventType=%28Z%29+Wildfire&beginDate_mm=01&beginDate_dd=01&beginDate_yyyy=2021&endDate_mm=01&endDate_dd=01&endDate_yyyy=2022&county=ALL&hailfilter=0.00&tornadofilter=0&windfilter=000&sort=DT&submitbutton=Search&statefips=6%2CCALIFORNIA

