# Project Name: Storm is Coming.. Are You Safe!?

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### **Background and Motivation:**

Predicting the intensity of a storm by different weather forecasts is a boon which helps us predict what kind of storm is going to hit and by what intensity, so that we can take preventive measures for such storms beforehand.

However, predicting the intensities of storms on the basis of damage and the casualties caused by them can help in taking better preventive measures for different types of storms.

In 2019 alone, the Atlantic storms caused damage of approximately \$22 billion in America, according to a report by AccuWeather.

On an average:

- 80 deaths and 1,500 injuries each year are directly attributed to thousands of tornadoes reported. (source: www.nssl.noaa.gov)
- 31 deaths per year are caused by thunderstorm winds; Number of thunderstorms occurring in the United States a year: 100,000.

## **Project objective:**

Our goal is to Analyze different types of storms in different parts of America based on historical data gathered from the National Centers for Environmental Information datasets of the past 20 years. Also, analyzing the deaths caused by different storms in a geospatial way.

The most important questions that arise when we hear about Storms are:

- What was the number of casualties due to that storm?
- What was the damage caused because of the storm?

With this project, we are trying to convert these "was" questions into the prediction of the damage that "will be" caused by the storm, i.e. We are trying to predict the damage caused by the Storm in any location based on the multiple factors like Magnitude of the Storms.

```
In []: # imports and setup
import pandas as pd
import scipy as sc
import numpy as np
import seaborn as sns

#%matplotlib notebook
import matplotlib.pyplot as plt
plt.style.use('ggplot')
%matplotlib inline
plt.rcParams['figure.figsize'] = (10, 6)
```

#### Data:

We have gathered the data from the following website:

 $\underline{\text{https://www1.ncdc.noaa.gov/pub/data/swdi/stormevents/csvfiles/?C=M;O=D} \ (\underline{\text{https://www1.ncdc.noaa.gov/pub/data/swdi/stormevents/csvfiles/?C=M;O=D}) \ (\underline{\text{https://www1.ncdc.noaa.gov/pub/data/swdi/stormevents/csvfiles/?C=M;O=D})} \ (\underline{\text{https://www1.ncdc.noaa.gov/pub/data/swdi/stormevents/csvfiles/?C=M;O=D})} \ (\underline{\text{https://www1.ncdc.noaa.gov/pub/data/swd$ 

It has a dataset of storms that occurred in the USA every year (for 2019, 2018, 2017...).

Our goal is to merge all of the different datasets of the different year into one single dataset and perform descriptive analysis, and predictive analysis of the storm data generated.

The Metadata of the columns is described in the following link:

http://www1.ncdc.noaa.gov/pub/data/swdi/stormevents/csvfiles/Storm-Data-Export-Format.docx (http://www1.ncdc.noaa.gov/pub/data/swdi/stormevents/csvfiles/Storm-Data-Export-Format.docx)

```
In [288]: | # Data Retrieval
          storm = pd.read_csv("StormEvents_2019details.csv")
          storm.head()
          storm.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 64439 entries, 0 to 64438
          Data columns (total 51 columns):
          BEGIN_YEARMONTH
                                64439 non-null int64
          BEGIN_DAY
                                64439 non-null int64
          BEGIN_TIME
                                64439 non-null int64
          END_YEARMONTH
                                64439 non-null int64
          END_DAY
                                64439 non-null int64
          END_TIME
                                64439 non-null int64
          EPISODE_ID
                                64439 non-null int64
          EVENT ID
                                64439 non-null int64
          STATE
                                64439 non-null object
                                64439 non-null int64
          STATE_FIPS
          YEAR
                                64439 non-null int64
                                64439 non-null object
          MONTH_NAME
          EVENT_TYPE
                                64439 non-null object
          CZ_TYPE
                                64439 non-null object
                                64439 non-null int64
          CZ_FIPS
          CZ_NAME
                                64439 non-null object
                                64439 non-null object
          WFO
                                64439 non-null object
          BEGIN_DATE_TIME
          CZ_TIMEZONE
                                64439 non-null object
          END_DATE_TIME
                                64439 non-null object
          INJURIES_DIRECT
                                 64439 non-null int64
          INJURIES_INDIRECT
                                64439 non-null int64
                                64439 non-null int64
          DEATHS_DIRECT
                                64439 non-null int64
          DEATHS_INDIRECT
                                52112 non-null object
          DAMAGE_PROPERTY
          DAMAGE_CROPS
                                52558 non-null object
          SOURCE
                                64439 non-null object
          MAGNITUDE
                                 35031 non-null float64
          MAGNITUDE_TYPE
                                 26025 non-null object
          FLOOD_CAUSE
                                8946 non-null object
          CATEGORY
                                10 non-null float64
          TOR_F_SCALE
                                1649 non-null object
          TOR_LENGTH
                                1649 non-null float64
          TOR WIDTH
                                1649 non-null float64
          TOR_OTHER_WFO
                                194 non-null object
                                194 non-null object
          TOR_OTHER_CZ_STATE
                                194 non-null float64
          TOR_OTHER_CZ_FIPS
                                194 non-null object
          TOR_OTHER_CZ_NAME
          BEGIN_RANGE
                                42858 non-null float64
          BEGIN_AZIMUTH
                                42858 non-null object
          BEGIN_LOCATION
                                42858 non-null object
          END_RANGE
                                42858 non-null float64
          END_AZIMUTH
                                42858 non-null object
          END_LOCATION
                                42858 non-null object
          BEGIN_LAT
                                42858 non-null float64
          BEGIN_LON
                                42858 non-null float64
          END_LAT
                                42858 non-null float64
                                42858 non-null float64
          END_LON
          EPISODE_NARRATIVE
                                64439 non-null object
          EVENT_NARRATIVE
                                50629 non-null object
                                64439 non-null object
          DATA_SOURCE
          dtypes: float64(11), int64(15), object(25)
          memory usage: 25.1+ MB
```

#### **Ethical Consideration:**

Data collected or recorded are accurate. We are performing this analysis only for America, as the number of storm cases in America contributes to 40% of the entire world.

The data is openly available at <a href="https://www.ncdc.noaa.gov/stormevents/ftp.jsp">https://www.ncdc.noaa.gov/stormevents/ftp.jsp</a> by the government of America for public interest and we have verified that anyone is allowed to access and perform ethical data analysis.

Since we are working on a dataset of different types of Storms, which are Natural Disasters and is not in control of humankind, the risk of violating ethical consideration reduces to a great extent. The only bias our models might have is of showing more storms in the areas which are prone to that kind of storms and that is a valid bias as well.

We also made sure that our data doesn't have any sort of bias that might impair any individual or community. Therefore, we strongly believe that our data is ethical and our analysis will help mankind.

## **Data Processing:**

For this proposal, we have considered only the 2019 dataset and as stated above, for the submission of our final project, we will try to get data for at least 20 years.

We have created new fields like Tot\_Deaths & Tot\_Injuries by adding columns of direct and indirect deaths and injuries to get a better initial understanding of the effects of the storms.

```
In [292]: # Addition of Deaths & Injuries into one new Column
storm['Tot_Deaths'] = storm['DEATHS_DIRECT'] + storm['DEATHS_INDIRECT']
storm['Tot_Injures'] = storm['INJURIES_DIRECT'] + storm['INJURIES_INDIRECT']
```

### **Exploratory Data Analysis:**

Name: TOR\_LENGTH, dtype: float64

After performing a final EDA, we might find a few columns that have outliers and needs to be removed before further analysis.

We have also converted the "Damage Property" column from string datatype to int (1k to 1000).

We found a few columns with null values and further research is required whether we should remove those values or whether we are supposed to perform null value imputation on them.

Also, we can remove a few columns because they might not be significant in performing analysis or they might are highly correlated with other columns in the dataset.

```
In [310]:
            storm.describe()
Out[310]:
                    BEGIN YEARMONTH
                                                                                                                  EPISODE ID
                                         BEGIN_DAY
                                                     BEGIN_TIME END_YEARMONTH
                                                                                        END_DAY
                                                                                                     END_TIME
                                                                                                                                  EVENT_ID
                                                                                                                                              STATE_FI
                           52112.000000
                                        52112.000000
                                                                                                                 52112.000000
                                                                                                                                52112.000000
             count
                                                     52112.000000
                                                                        52112.000000 52112.000000
                                                                                                  52112.000000
                                                                                                                                             52112.0000
                         201905.644170
                                                                       201905.644170
                                                                                                                              830239.033006
                                           16.240732
                                                      1291.740290
                                                                                        17.309276
                                                                                                    1426.107921
                                                                                                                138112.086640
                                                                                                                                                32.4976
             mean
                                            8.821335
                               2.804604
                                                       656.440183
                                                                            2.804604
                                                                                         8.756046
                                                                                                     620.178898
                                                                                                                  3222.463571
                                                                                                                               20202.124489
                                                                                                                                                18.7755
               std
                         201901.000000
                                                                       201901.000000
                                            1.000000
                                                         0.000000
                                                                                         1.000000
                                                                                                      0.000000
                                                                                                                132253.000000
                                                                                                                              791304.000000
                                                                                                                                                 1.0000
              min
              25%
                         201903.000000
                                            9.000000
                                                       830.000000
                                                                       201903.000000
                                                                                        10.000000
                                                                                                    1016.000000
                                                                                                                135439.000000
                                                                                                                              813159.750000
                                                                                                                                                19.0000
              50%
                         201906.000000
                                                      1450.000000
                                                                       201906.000000
                                           17.000000
                                                                                        18.000000
                                                                                                    1545.000000
                                                                                                                137824.000000
                                                                                                                              830208.500000
                                                                                                                                                31.0000
              75%
                         201908.000000
                                           23.000000
                                                      1800.000000
                                                                       201908.000000
                                                                                        25.000000
                                                                                                    1900.000000
                                                                                                                140807.000000
                                                                                                                              847617.250000
                                                                                                                                                46.0000
                          201911.000000
                                           31.000000
                                                      2359.000000
                                                                       201911.000000
                                                                                        31.000000
                                                                                                   2359.000000 144628.000000 869176.000000
                                                                                                                                                99.0000
              max
            8 rows × 29 columns
In [311]:
            # Outlier Suspection Method
            storm.TOR_LENGTH.describe()
Out[311]: count
                      1430.000000
            mean
                          2.929126
                          3.821778
            std
            min
                          0.010000
            25%
                          0.500000
            50%
                          1.445000
            75%
                          3.987500
                         31.400000
            max
```

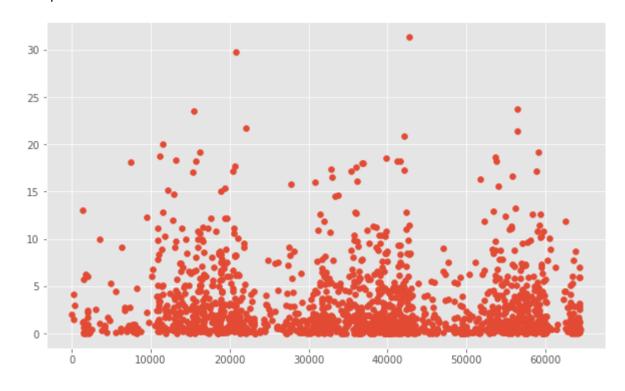
```
In [301]: | # Coverting Column Datatype (String to Integer)
          for i in range(0, len(storm.DAMAGE PROPERTY)):
              if(type(storm.DAMAGE_PROPERTY[i]) == str):
                  if ('K' in storm.DAMAGE_PROPERTY[i]):
                      storm.DAMAGE_PROPERTY[i] = storm.DAMAGE_PROPERTY[i].replace("K","")
                      storm.DAMAGE_PROPERTY[i] =int(float(storm.DAMAGE_PROPERTY[i])*1000)
                  elif('M' in storm.DAMAGE_PROPERTY[i]):
                      storm.DAMAGE_PROPERTY[i] = storm.DAMAGE_PROPERTY[i].replace("M","")
                      storm.DAMAGE_PROPERTY[i] =int(float(storm.DAMAGE_PROPERTY[i])*1000000)
                  elif('B' in storm.DAMAGE_PROPERTY[i]):
                      storm.DAMAGE_PROPERTY[i] = storm.DAMAGE_PROPERTY[i].replace("B","")
                      storm.DAMAGE_PROPERTY[i] =int(float(storm.DAMAGE_PROPERTY[i])*1000000000)
          C:\Users\madas\Anaconda3\lib\site-packages\ipykernel_launcher.py:4: SettingWithCopyWarning:
          A value is trying to be set on a copy of a slice from a DataFrame
          See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-
          a-view-versus-a-copy
            after removing the cwd from sys.path.
          C:\Users\madas\Anaconda3\lib\site-packages\ipykernel_launcher.py:5: SettingWithCopyWarning:
          A value is trying to be set on a copy of a slice from a DataFrame
          See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-
          a-view-versus-a-copy
          C:\Users\madas\Anaconda3\lib\site-packages\ipykernel_launcher.py:7: SettingWithCopyWarning:
          A value is trying to be set on a copy of a slice from a DataFrame
          See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-
          a-view-versus-a-copy
            import sys
          C:\Users\madas\Anaconda3\lib\site-packages\ipykernel launcher.py:8: SettingWithCopyWarning:
          A value is trying to be set on a copy of a slice from a DataFrame
          See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-
          a-view-versus-a-copy
          C:\Users\madas\Anaconda3\lib\site-packages\ipykernel_launcher.py:10: SettingWithCopyWarning:
          A value is trying to be set on a copy of a slice from a DataFrame
          See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-
          a-view-versus-a-copy
            # Remove the CWD from sys.path while we load stuff.
          C:\Users\madas\Anaconda3\lib\site-packages\ipykernel_launcher.py:11: SettingWithCopyWarning:
          A value is trying to be set on a copy of a slice from a DataFrame
          See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-
          a-view-versus-a-copy
            # This is added back by InteractiveShellApp.init_path()
In [313]: | # Removing NAs
          storm = storm[storm['DAMAGE_PROPERTY'].notna()]
```

#### **Data Visualization**

```
In [291]: # Outlier Detection Method
x = storm.TOR_LENGTH.index.values
plt.scatter(x, storm.TOR_LENGTH)
```

Out[291]: <matplotlib.collections.PathCollection at 0x1c68e7fbb48>

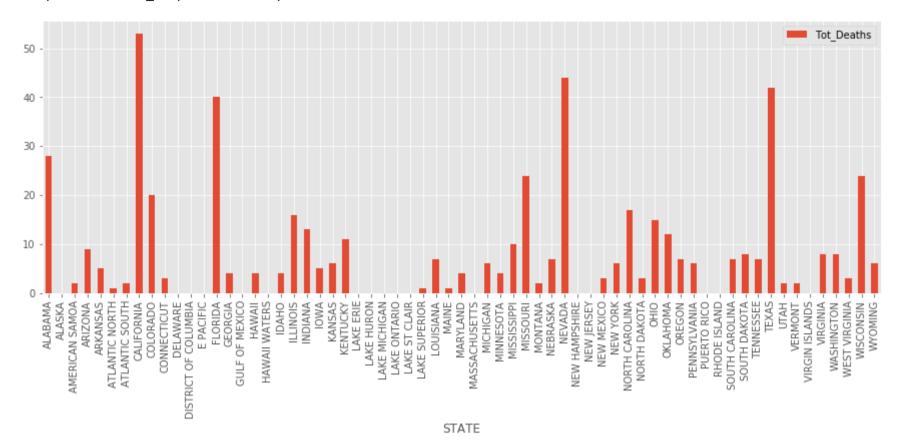
storm.DAMAGE\_PROPERTY = storm.DAMAGE\_PROPERTY.astype(int)



```
In [293]: # Group by states
pd.set_option('display.max_rows',65)
storm_dea = storm.groupby('STATE').agg({'Tot_Deaths':'sum'})
#storm.plot(x='STATE', y='Tot_Deaths', kind="bar")
```

In [294]: storm\_dea.plot(kind="bar",figsize=(15,5))

Out[294]: <matplotlib.axes.\_subplots.AxesSubplot at 0x1c68cd44188>



```
In [296]: #Group by states
storm_type_dea = storm.groupby('EVENT_TYPE').agg({'Tot_Deaths':'sum'})
storm_type_dea
```

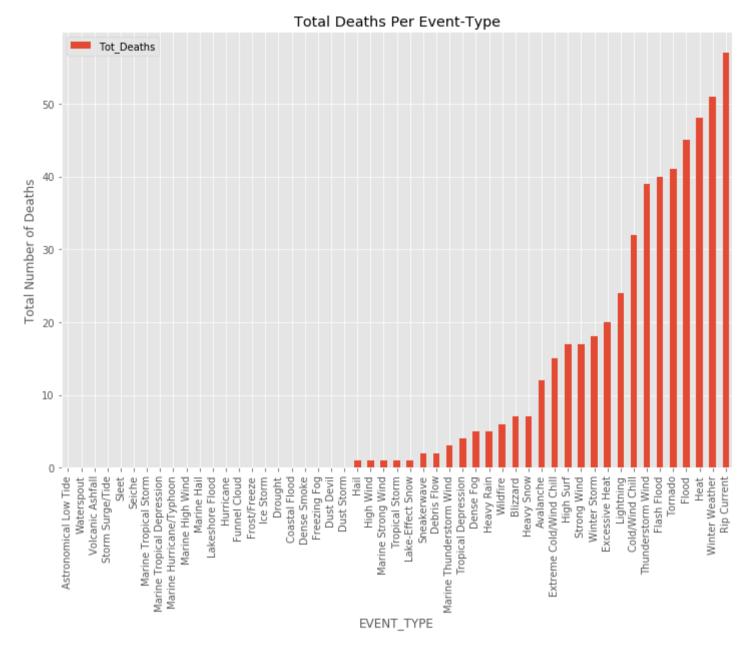
Out[296]:

#### Tot\_Deaths

	Tot_Deaths
EVENT_TYPE	
Astronomical Low Tide	0
Avalanche	12
Blizzard	7
Coastal Flood	. 0
Cold/Wind Chill	32
Debris Flow	2
Dense Fog	5
Dense Smoke	0
Drought	0
Dust Devil	0
Dust Storm	0
Excessive Heat	20
Extreme Cold/Wind Chill	15
Flash Flood	40
Flood	45
Freezing Fog	0
Frost/Freeze	0
Funnel Cloud	0
Hail	1
Heat	48
Heavy Rain	5
Heavy Snow	7
High Surf	17
High Wind	1
Hurricane	0
Ice Storm	0
Lake-Effect Snow	1
Lakeshore Flood	0
Lightning	24
Marine Hail	0
Marine High Wind	0
Marine Hurricane/Typhoon	0
Marine Strong Wind	1
Marine Thunderstorm Wind	3
Marine Tropical Depression	0
Marine Tropical Storm	0
Rip Current	57
Seiche	0
Sleet	0
Sneakerwave	2
Storm Surge/Tide	0
Strong Wind	17
Thunderstorm Wind	39
Tornado	41
Tropical Depression	4
Tropical Storm	1
	-
Volcanic Ashfall	0
Waterspout	0
Wildfire	6
Winter Storm	18
Winter Weather	51

```
In [317]: storm_type_dea.sort_values(by='Tot_Deaths',ascending=True).plot(kind="bar",figsize=(12,8))
    plt.title("Total Deaths Per Event-Type")
    plt.ylabel("Total Number of Deaths")
```

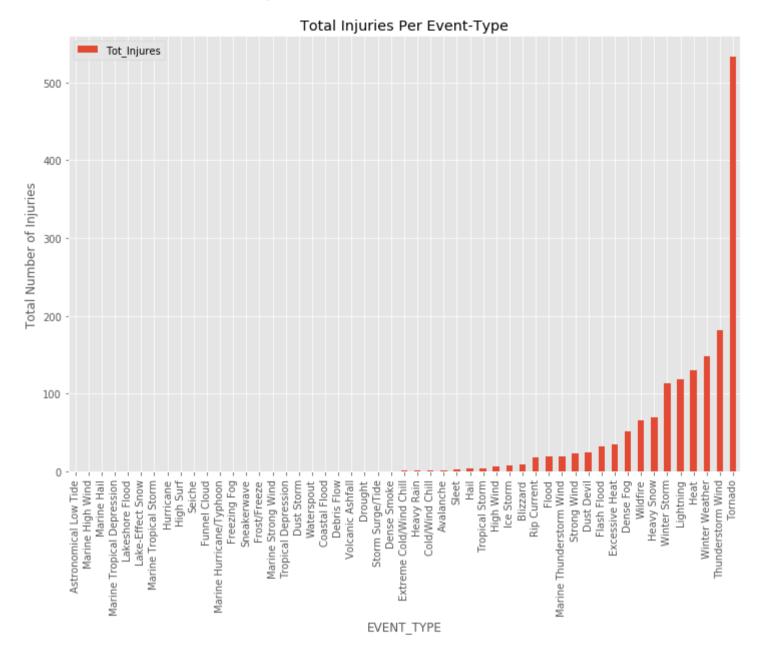
Out[317]: Text(0, 0.5, 'Total Number of Deaths')



In [298]: storm\_type\_inj = storm.groupby('EVENT\_TYPE').agg({'Tot\_Injures':'sum'})

```
In [318]: storm_type_inj.sort_values(by='Tot_Injures',ascending=True).plot(kind="bar",figsize=(12,8))
    plt.title("Total Injuries Per Event-Type")
    plt.ylabel("Total Number of Injuries")
```

Out[318]: Text(0, 0.5, 'Total Number of Injuries')

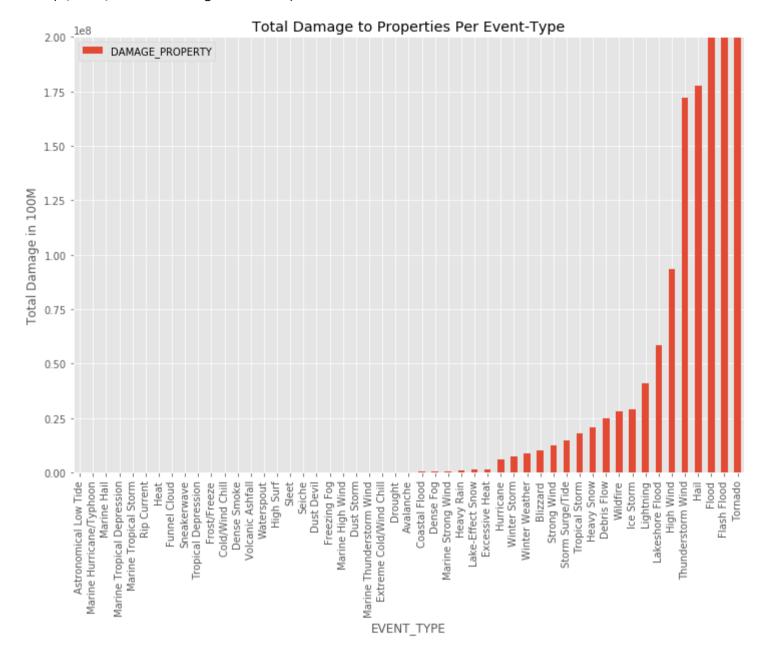


In [314]: storm\_type\_dam\_prop = storm.groupby('EVENT\_TYPE').agg({'DAMAGE\_PROPERTY':'sum'})

```
In [320]: fig_dp = storm_type_dam_prop.sort_values(by='DAMAGE_PROPERTY',ascending=True).plot(kind="bar",figsize=(12,8))
fig_dp.set_ylim(0, 200000000)
plt.title("Total Damage to Properties Per Event-Type")
plt.ylabel("Total Damage in 100M")

#NOTE: Floods, Flash Floods, and Tornados have damage of more than $200M.
```

Out[320]: Text(0, 0.5, 'Total Damage in 100M')



# **Analysis Methodology:**

Our major considerations are some important attributes from the dataset, like the number of deaths or injuries, and locations that were affected by the storms also the total damage caused.

We are interested in finding the relationships, if any, on the above-stated questions and understanding what kind of magnitudes of the storm are influencing how many deaths, injuries in America.

Based on this we are planning to model our data and predict the number of deaths in the future years due to a particular storm of some specific magnitude to take preventive measures in a much better way.

# **Project Schedule:**

Proposal Date: February 28
Meeting with the staff: TBD
Milestone: March 29
Meeting with the staff: TBD
Final Submission: April 19
Project Presentation: April 21

#### References:

http://www-das.uwyo.edu/~geerts/cwx/notes/chap03/nat\_hazard.html (http://www-das.uwyo.edu/~geerts/cwx/notes/chap03/nat\_hazard.html)
https://www.depts.ttu.edu/nwi/research/DebrisImpact/Reports/DDS.pdf (https://www.depts.ttu.edu/nwi/research/DebrisImpact/Reports/DDS.pdf)
https://www.nbcnews.com/news/weather/atlantic-hurricane-seasons-2019-2010-graphics-data-n1091986 (https://www.nbcnews.com/news/weather/atlantic-hurricane-seasons-2019-2010-graphics-data-n1091986)
https://www.c2es.org/content/hurricanes-and-climate-change/ (https://www.c2es.org/content/hurricanes-and-climate-change/)