

DATA VISULATION PROJECT - NATURAL DISASTERS

Junnubala Gowtham(20181CSE0295)

Kalangi Sai Nikhil(20181CSE0307)

Karlapudi Praveen(20181CSE0317)

K Sai Rama Kaushik(20181CSE0318)

K Venkata Sai Abhishek(20181CSE0320)

TYPES OF NATURAL DISASTERS



```
#To mount the Colab and google drive
from google.colab import drive
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mour

```
◀ ▶
```

```
#Importing the required libraries
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns
sns.set(color_codes=True)
```

```
#Read CSV
df = pd.read_csv("/content/drive/MyDrive/DV/NATURAL DISASTERS.csv")
print(df.info())

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8824 entries, 0 to 8823
Data columns (total 19 columns):
```

#	Column	Non-Null Count	Dtype	
0	Year	8824	non-null	int64
1	Disaster Subgroup	8824	non-null	object
2	Disaster Type	8824	non-null	object
3	Disaster Subtype	8824	non-null	object
4	Country	8824	non-null	object
5	ISO	8824	non-null	object
6	Region	8824	non-null	object
7	Location	8535	non-null	object
8	Start Year	8823	non-null	float64
9	Start Month	8769	non-null	float64
10	Start Day	7503	non-null	float64
11	End Year	8824	non-null	int64
12	End Month	8669	non-null	float64
13	End Day	7544	non-null	float64
14	Total Deaths	6350	non-null	float64
15	No Injured	2339	non-null	float64
16	No Homeless	1172	non-null	float64
17	Total Affected	6997	non-null	float64
18	Total Damages	2735	non-null	float64

dtypes: float64(10), int64(2), object(7)
memory usage: 1.3+ MB
None

```
#To Remove Unnamed Columns
df = df.loc[:, ~df.columns.str.contains('^\u0333nna\u0333d')]

print(df.info())
```

#	Column	Non-Null Count	Dtype	
0	Year	8824	non-null	int64
1	Disaster Subgroup	8824	non-null	object
2	Disaster Type	8824	non-null	object
3	Disaster Subtype	8824	non-null	object
4	Country	8824	non-null	object
5	ISO	8824	non-null	object
6	Region	8824	non-null	object
7	Location	8535	non-null	object
8	Start Year	8823	non-null	float64
9	Start Month	8769	non-null	float64
10	Start Day	7503	non-null	float64
11	End Year	8824	non-null	int64
12	End Month	8669	non-null	float64
13	End Day	7544	non-null	float64
14	Total Deaths	6350	non-null	float64
15	No Injured	2339	non-null	float64
16	No Homeless	1172	non-null	float64
17	Total Affected	6997	non-null	float64
18	Total Damages	2735	non-null	float64

dtypes: float64(10), int64(2), object(7)

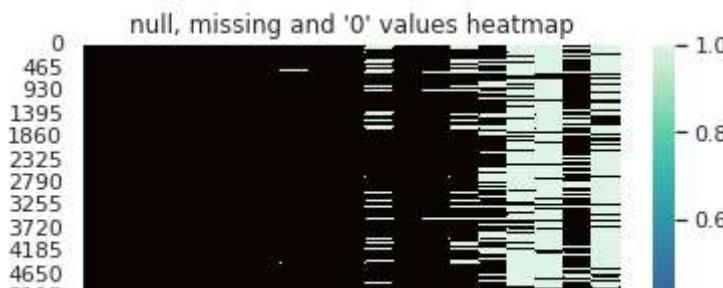
```
memory usage: 1.3+ MB
None
```

```
#Replace Null Values with 0
df.fillna(value = 0,inplace = True)
df.isnull().sum()
df.head()
```

	Year	Disaster Subgroup	Disaster Type	Disaster Subtype	Country	ISO	Region	Location	Start Year	Start Month
0	2000	Hydrological	Landslide	Landslide	Angola	AGO	Middle Africa	Chassuala village (Saurimo district, Lunda-Sul...)	2000.0	10.0
1	2000	Biological	Epidemic	Epidemic	Angola	AGO	Middle Africa	Bie, Lunda Sul, Benguela	2000.0	8.0
2	2000	Hydrological	Flood	Flood	Angola	AGO	Middle Africa	Caxito village (Dande district)	2000.0	3.0

HEAT MAP

```
plt.title('null, missing and \'0\' values heatmap')
sns.heatmap((df.isnull()) | (df == 0), cmap = 'mako')
plt.show()
print('\nDUPLICATE VALUE COUNT: ', df.duplicated().sum())
```



SWARM PLOT



```
# groupby()- To group df based on Year
y=df.groupby('Year')
y
# Iterate through he Year column and their corresponding df
for Year,Year_df in y:
    print(Year) # Prints Key Value - Year
    print(Year_df) # Prints df corresponding to the key as value
```

1900

	Year	Disaster	Subgroup	...	Total	Affected	Total	Damages
8016	1900		Geophysical	...		0.0		0.0

[1 rows x 19 columns]

1902

	Year	Disaster	Subgroup	...	Total	Affected	Total	Damages
6732	1902		Geophysical	...		17540.0		0.0
6733	1902		Geophysical	...		0.0		0.0

[2 rows x 19 columns]

1904

	Year	Disaster	Subgroup	...	Total	Affected	Total	Damages
7347	1904		Geophysical	...		1620.0		0.0
8019	1904		Geophysical	...		208.0		0.0
8020	1904		Geophysical	...		2141.0		0.0

[3 rows x 19 columns]

1905

	Year	Disaster	Subgroup	...	Total	Affected	Total	Damages
6734	1905		Geophysical	...		0.0		23800.0
6735	1905		Geophysical	...		0.0		0.0
7348	1905		Geophysical	...		177.0		0.0

[3 rows x 19 columns]

1906

	Year	Disaster	Subgroup	...	Total	Affected	Total	Damages
6736	1906		Geophysical	...		10000.0		0.0
8021	1906		Geophysical	...		5466.0		0.0

[2 rows x 19 columns]

1909

```
Year Disaster Subgroup ... Total Affected Total Damages
6737 1909 Geophysical ... 0.0 0.0
7349 1909 Geophysical ... 2928.0 0.0
7350 1909 Geophysical ... 0.0 0.0
7351 1909 Geophysical ... 0.0 0.0
8022 1909 Geophysical ... 417.0 0.0
```

[5 rows x 19 columns]

1910

```
Year Disaster Subgroup ... Total Affected Total Damages
6738 1910 Geophysical ... 0.0 0.0
8023 1910 Geophysical ... 0.0 0.0
```

[2 rows x 19 columns]

1911

```
Year Disaster Subgroup ... Total Affected Total Damages
7352 1911 Geophysical ... 0.0 0.0
7353 1911 Geophysical ... 1266.0 0.0
7354 1911 Geophysical ... 0.0 0.0
7355 1911 Geophysical ... 22.0 0.0
7356 1911 Geophysical ... 0.0 0.0
```

[5 rows x 19 columns]

1913

```
Year Disaster Subgroup ... Total Affected Total Damages
```

```
# To access a specific df with key value 2021
```

```
gg=y.get_group(2021)
```

```
gg
```

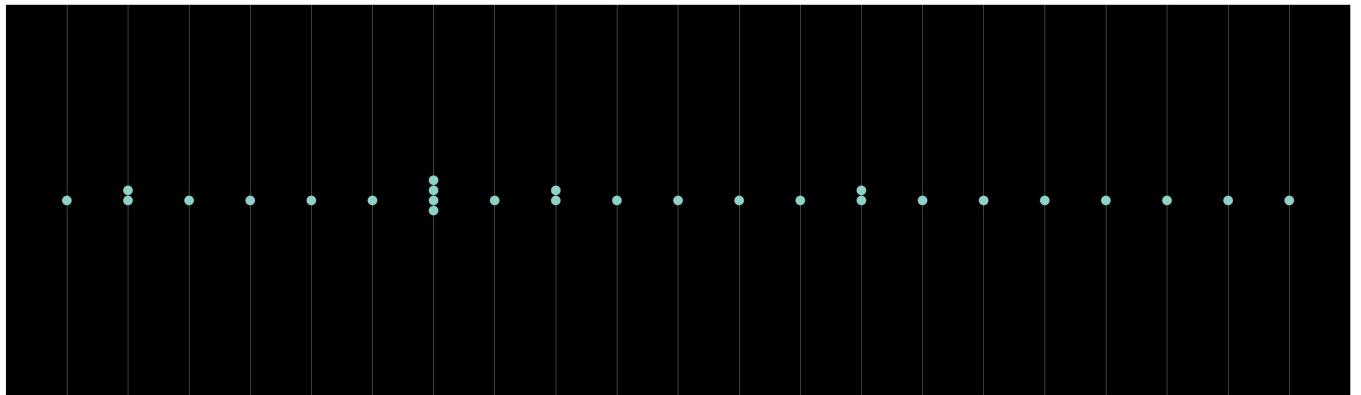
	Year	Disaster Subgroup	Disaster Type	Disaster Subtype	Country	ISO	Region	Loc
8384	2021	Meteorological	Storm	Storm	Chile	CHL	South America	Santiago, Metro
8401	2021	Meteorological	Storm	Storm	Spain	ESP	Southern Europe	Madrid, Madrid
8419	2021	Geophysical	Earthquake	Earthquake	Indonesia	IDN	South-Eastern Asia	Mamuju, Mamuju districts
8428	2021	Hydrological	Flood	Flood	Indonesia	IDN	South-Eastern Asia	Kalin Sulawesi Maluku
8449	2021	Hydrological	Flood	Flood	Paraguay	PRY	South America	Distrito Central Paraguayan
8464	2021	Hydrological	Flood	Flood	Bolivia (Plurinational State of)	BOL	South America	Sucumbíos (Chucuri Depa)
8472	2021	Meteorological	Storm	Storm	Japan	JPN	Eastern Asia	Hokuriku, Tohoku
8474	2021	Meteorological	Storm	Storm	Taiwan (Province of China)	TWN	Eastern Asia	Taipei, New Taipei, Taoyuan, Keelung

```
# Display countries wrt year 2021
yc=gg[['Year','Country']]
yc
```

	Year	Country
8384	2021	Chile
8401	2021	Spain
8419	2021	Indonesia
8428	2021	Indonesia
8449	2021	Paraguay
8464	2021	Bolivia (Plurinational State of)
8472	2021	Japan
8474	2021	Taiwan (Province of China)
8483	2021	India
8489	2021	Mozambique
8508	2021	Malaysia
8520	2021	Australia
8532	2021	Bolivia (Plurinational State of)
8540	2021	China
8566	2021	Fiji
8601	2021	Indonesia
8612	2021	Indonesia
8621	2021	Iran (Islamic Republic of)

```
#Width=100 and Height=30
plt.figure(figsize=(100,30))
plt.style.use('dark_background')
sns.swarmplot(x = 'Country', data = yc,s=40).set_title('Seaborn Plot')
```

```
Text(0.5, 1.0, 'Seaborn Plot')
```



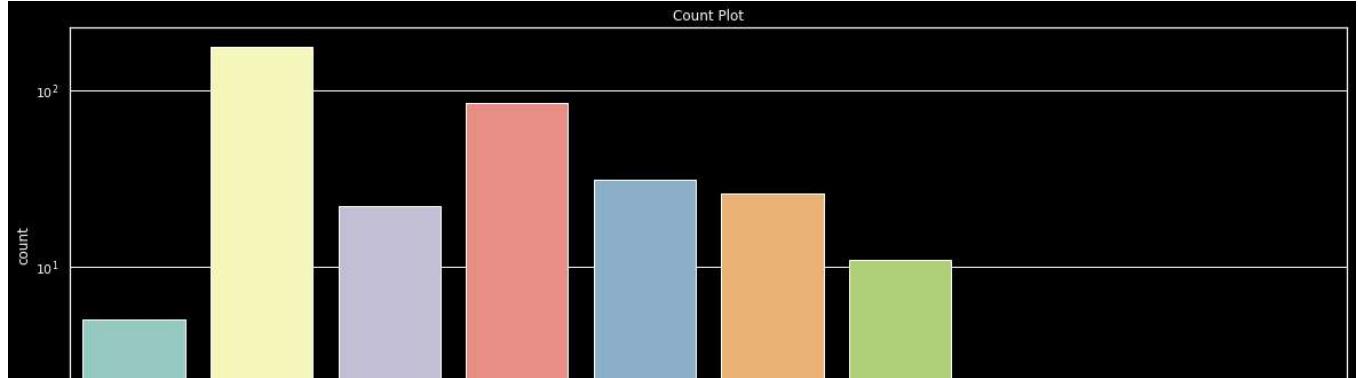
COUNT PLOT

```
plt.figure(figsize=(20,9))
plt.yscale('log') # Set Y axis scale to log
plt.style.use('dark_background')
# x,hue - inputs
sns.countplot(x = 'Disaster Subgroup', hue='Disaster Type' ,data = df).set_title('Count Plot')
plt.legend(loc="upper right")
# In df,extract all row values of India
i=df[df['Country'] == 'India']
i
```

	Year	Disaster Subgroup	Disaster Type	Disaster Subtype	Country	ISO	Region	Location	St Y
118	2000	Climatological	Drought	Drought	India	IND	Southern Asia	New Delhi city (Delhi province), Gujarat, Raja...	200
132	2000	Hydrological	Flood	Flood	India	IND	Southern Asia	Yingkiang,Tuting villages (Upper Siang (70051)...	200
215	2000	Hydrological	Flood	Flood	India	IND	Southern Asia	East Siang, 70041, West Siang, 70054, Upper Si...	200
216	2000	Hydrological	Flood	Flood	India	IND	Southern Asia	Adilabad, Cuddapah, East Godavari, Guntur, Hyd...	200
217	2000	Hydrological	Flood	Flood	India	IND	Southern Asia	Katihar, Sitamarhi districts (Bihar province),...	200
...
8615	2020	Meteorological	Storm	Storm	India	IND	Southern Asia	Raigad, Pune	0 2020

```
# Extract Disaster Type and Total Affected of India
DT = i[['Disaster Type','Total Affected']]
DT
plt.figure(figsize=(20,7))
plt.style.use('dark_background')
plt.yscale('log')
sns.countplot(x='Disaster Type',data=DT).set_title('Count Plot')
```

```
Text(0.5, 1.0, 'Count Plot')
```



STRIP PLOT

```
# Sort Total Affected column in Descending order and print first 10 values
TA=df.sort_values(by='Total Affected',ascending=False).head(10)
# Prints first 10 Countries with Highest Total Affected values
print(TA['Country'])
# Prints first 10 Highest Total Affected values
14
print(TA['Total Affected'])
plt.figure(figsize=(20,6))
# Sort Total Affected column in Descending order and print first 40 values
TA=df.sort_values(by='Total Affected',ascending=False).head(40)
t=TA['Total Affected']
c=TA['Country']
plt.style.use('dark_background')
# x axis - Countries , y axis - Total Affected Count
sns.stripplot(x=c,y=t,hue='Country',data=TA,jitter=True,s=9)
```

```
6492           India
1154           India
1567           China
4297           China
3484           China
746            China
7005  United States of America (the)
664            China
3725           China
4448           China
```

Name: Country, dtype: object

```
6492    330000000.0
1154    300000000.0
1567    150146000.0
4297    134000000.0
3484    105004000.0
746     100000000.0
7005    85000012.0
664     80035257.0
3725    77000000.0
4448    67900000.0
```

Name: Total Affected, dtype: float64

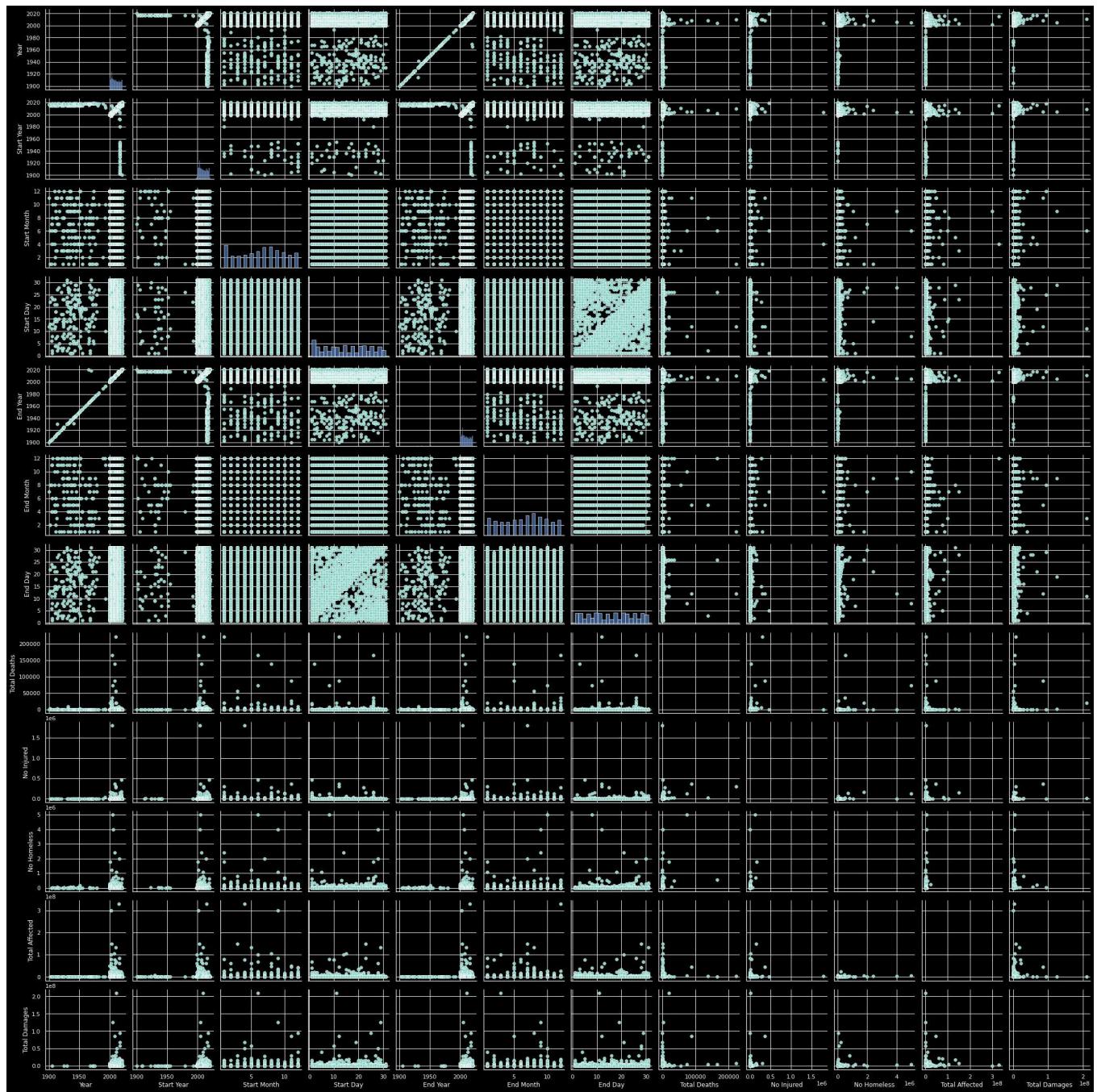
```
<matplotlib.axes._subplots.AxesSubplot at 0x7fd84ee6e050>
```



PAIR PLOT

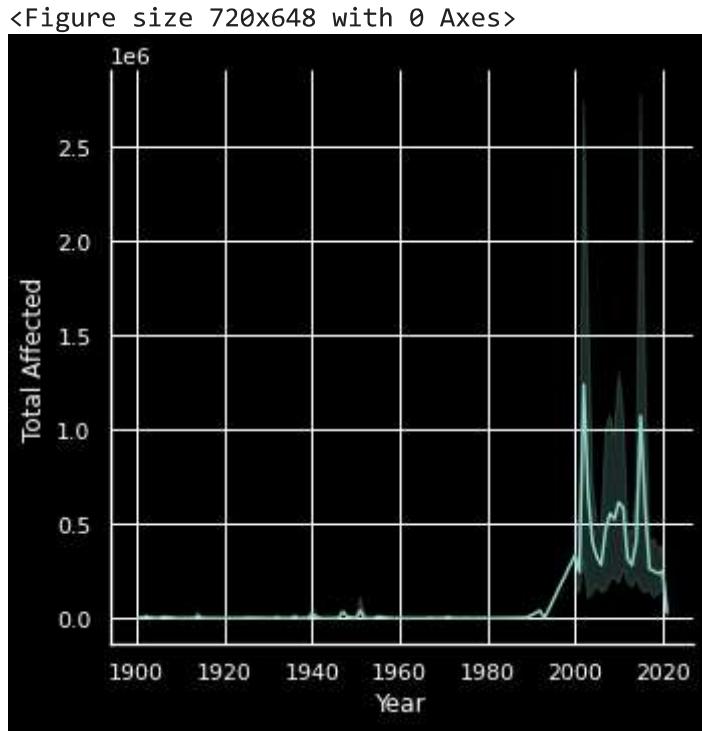


```
sns.pairplot(data=df)
plt.show()
```



RED PLOT

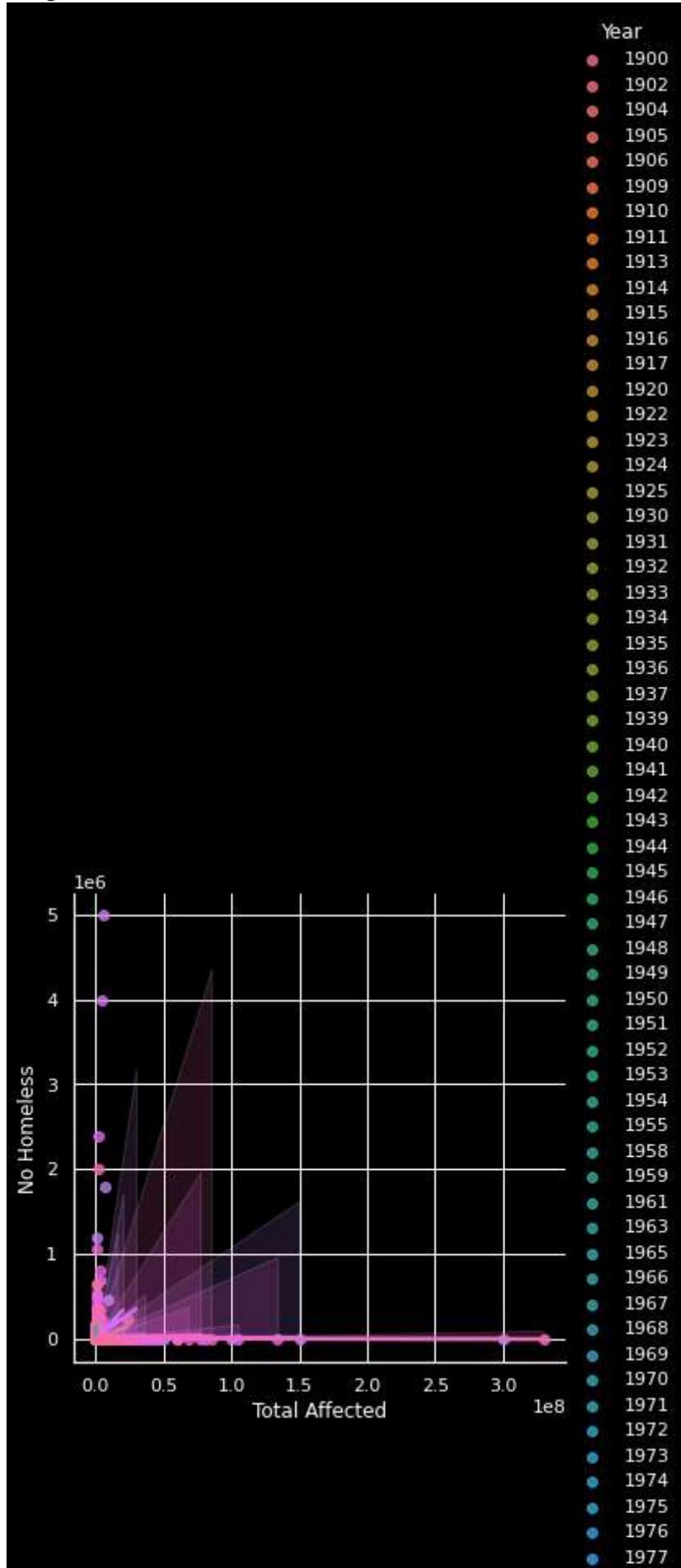
```
plt.figure(figsize=(10,9))
sns.relplot(x=df['Year'], y=df['Total Affected'], kind='line');
```

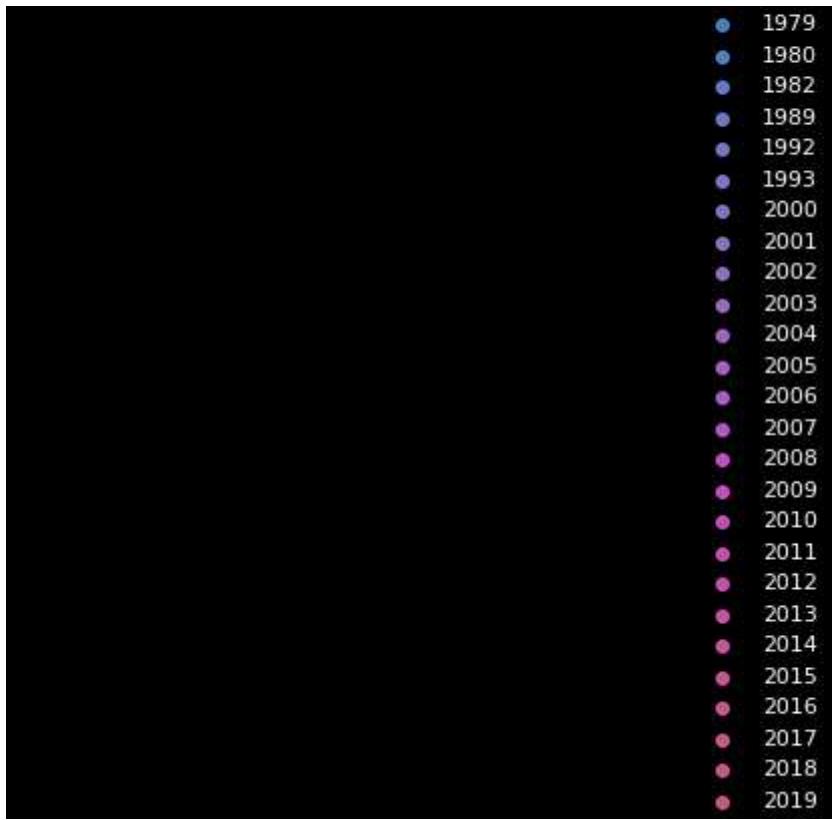


LM PLOT

```
plt.figure(figsize=(10,9))
sns.lmplot(x='Total Affected', y='No Homeless', hue='Year', data= df, fit_reg= True)
plt.show()
```

<Figure size 720x648 with 0 Axes>





VIOLIN PLOT

```
plt.figure(figsize=(15,8))
sns.violinplot(data=df, x='Year', y='No Homeless', linewidth=1,)
plt.show()
```

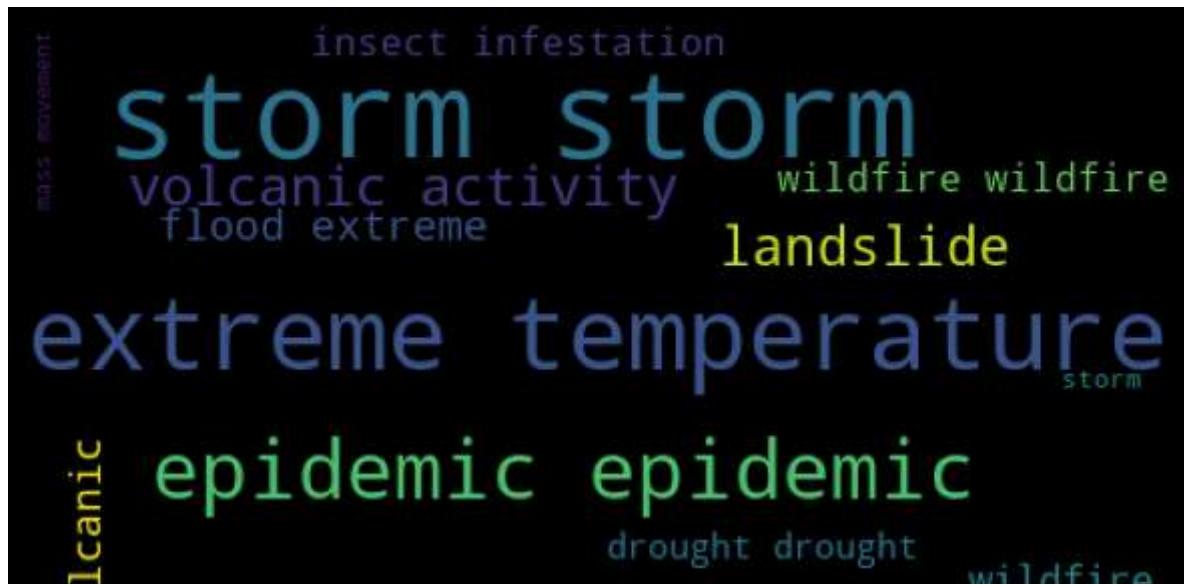


WORD CLOUD



```
from PIL import Image
import matplotlib.pyplot as plt
from wordcloud import WordCloud, STOPWORDS
import pandas as pd
df=pd.read_csv("/content/drive/MyDrive/DV/NATURAL DISASTERS.csv")
comment_words = " "
stopwords = set(STOPWORDS)
#Iterating disaster subtype values
for val in df['Disaster Subtype'][0:3000]:
    val = str(val)
    tokens = val.split()
    #converting to lowercase
    for i in range(len(tokens)):
        tokens[i] = tokens[i].lower()
    comment_words += " ".join(tokens)+" "
#generate wordcloud
wc = WordCloud(width = 500, height = 500, background_color ='black', stopwords = stopwords, m
plt.figure(figsize = (8, 8), facecolor = None)

#display wordcloud
plt.imshow(wc)
plt.axis('off')
plt.tight_layout(pad = 0)
plt.show()
```



WAFFLE CHART



```
!pip install pywaffle
from pywaffle import Waffle
data=pd.read_csv("/content/drive/MyDrive/DV/NATURAL DISASTERS.csv")
data['Disaster Type'].value_counts()
dis_data={'Disaster_Type':[ 'Insect infestation','Mass movement','Glacial lake outburst','Impacts','Volcanic activity','Flood','Extreme weather','Drought','Landslide','Wildfire']}
#converting data to dataframe
dataframe = pd.DataFrame(dis_data)
plt.style.use('dark_background')
#plotting waffle chart with five rows of cells
fig = plt.figure(FigureClass = Waffle,values = dataframe.Count,rows=5,labels = list(dataframe['Disaster Type']))
```



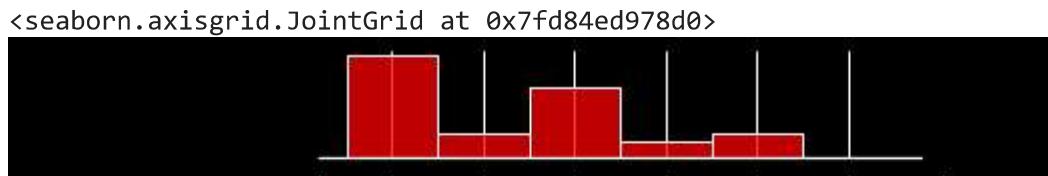
```
Collecting pywaffle
  Downloading pywaffle-0.6.4-py2.py3-none-any.whl (565 kB)
    |██████████| 565 kB 7.5 MB/s
Requirement already satisfied: matplotlib in /usr/local/lib/python3.7/dist-packages (from pywaffle==0.6.4)
Requirement already satisfied: numpy>=1.11 in /usr/local/lib/python3.7/dist-packages (from pywaffle==0.6.4)
Requirement already satisfied: python-dateutil>=2.1 in /usr/local/lib/python3.7/dist-packages (from pywaffle==0.6.4)
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.7/dist-packages (from pywaffle==0.6.4)
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.7/dist-packages (from pywaffle==0.6.4)
Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in /usr/local/lib/python3.7/dist-packages (from pywaffle==0.6.4)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.7/dist-packages (from pywaffle==0.6.4)
Installing collected packages: pywaffle
Successfully installed pywaffle-0.6.4
```



JOINT PLOT



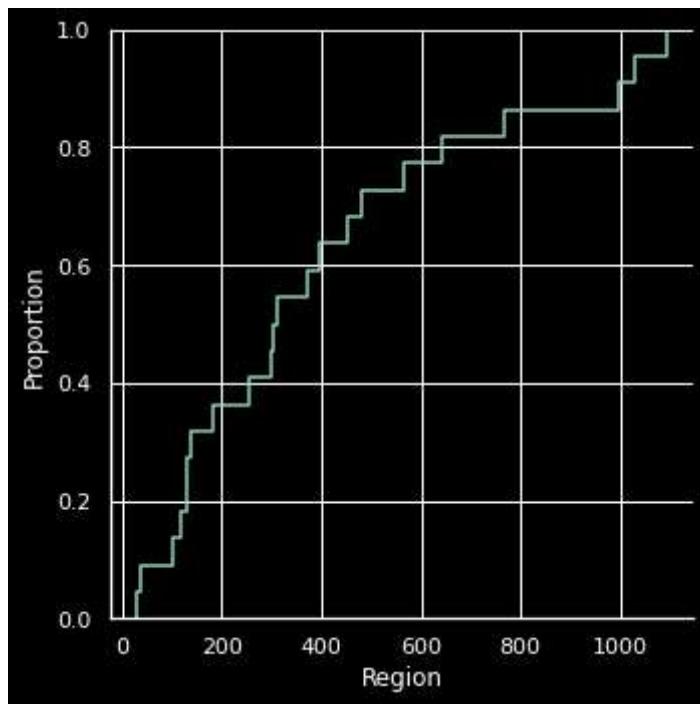
```
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd
data=pd.read_csv("/content/drive/MyDrive/DV/NATURAL DISASTERS.csv")
#renaming column names
data=data.replace(['Hydrological'],'Hydro')
data.replace(['Biological'],'Bio')
data.replace(['Meteorological'],'Meteoro')
data.replace(['Climatological'],'Climat')
data.replace(['Geophysical'],'Geo')
data.replace(['Extra-terrestrial'],'Extra')
#plotting jointplot
sns.jointplot(x = 'Disaster Subgroup',y = 'Disaster Type',color='red',data = data)
```



DIS PLOT



```
import seaborn as sns
#empirical cumulative distribution function plot
#plotting displot
sns.displot(x =data['Region'].value_counts(),kind='ecdf')
plt.show()
```

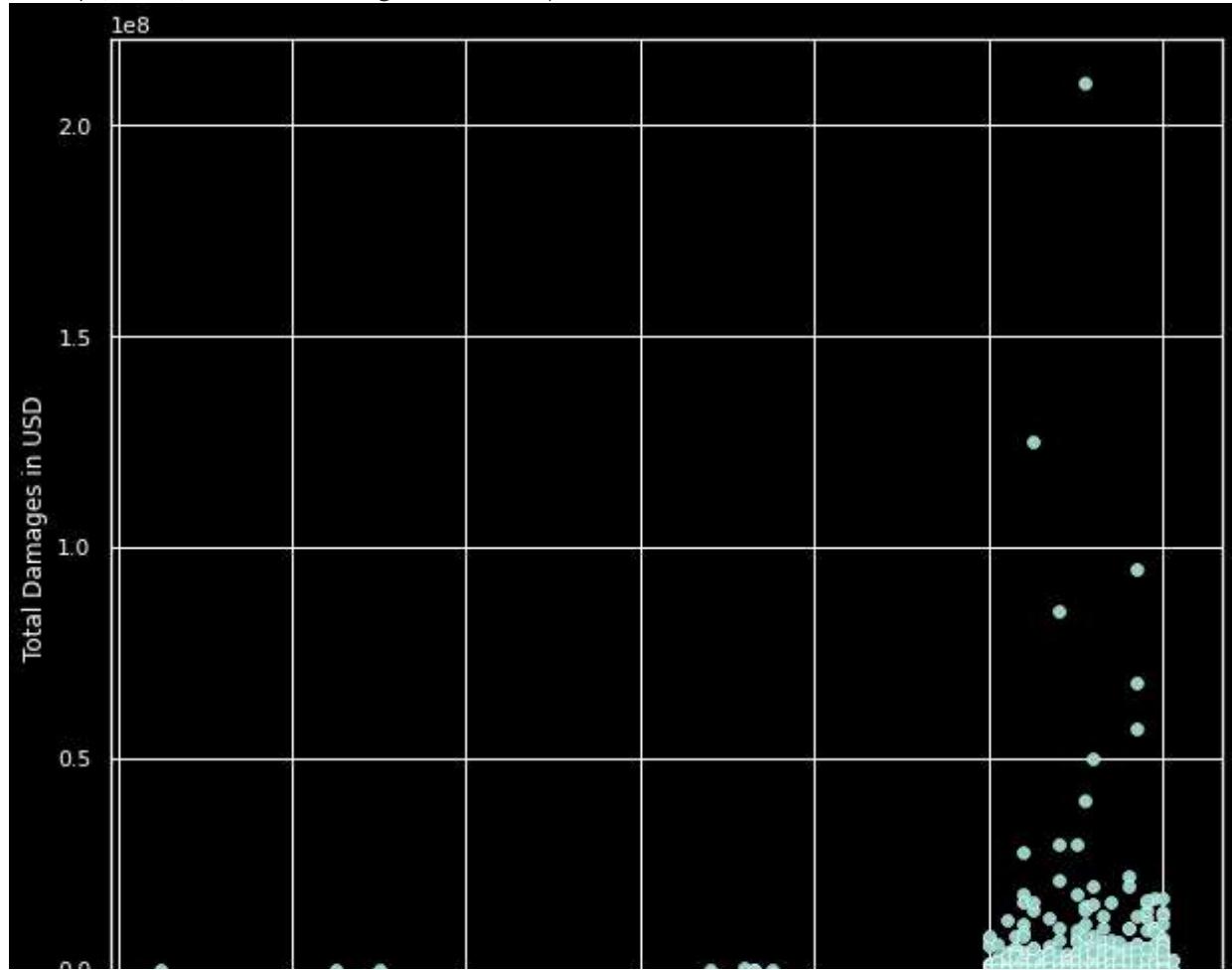


SCATTER PLOT

```
#selecting data to make dataframe sn
plt.figure(figsize=(10,9))
df=pd.read_csv("/content/drive/MyDrive/DV/NATURAL DISASTERS.csv")

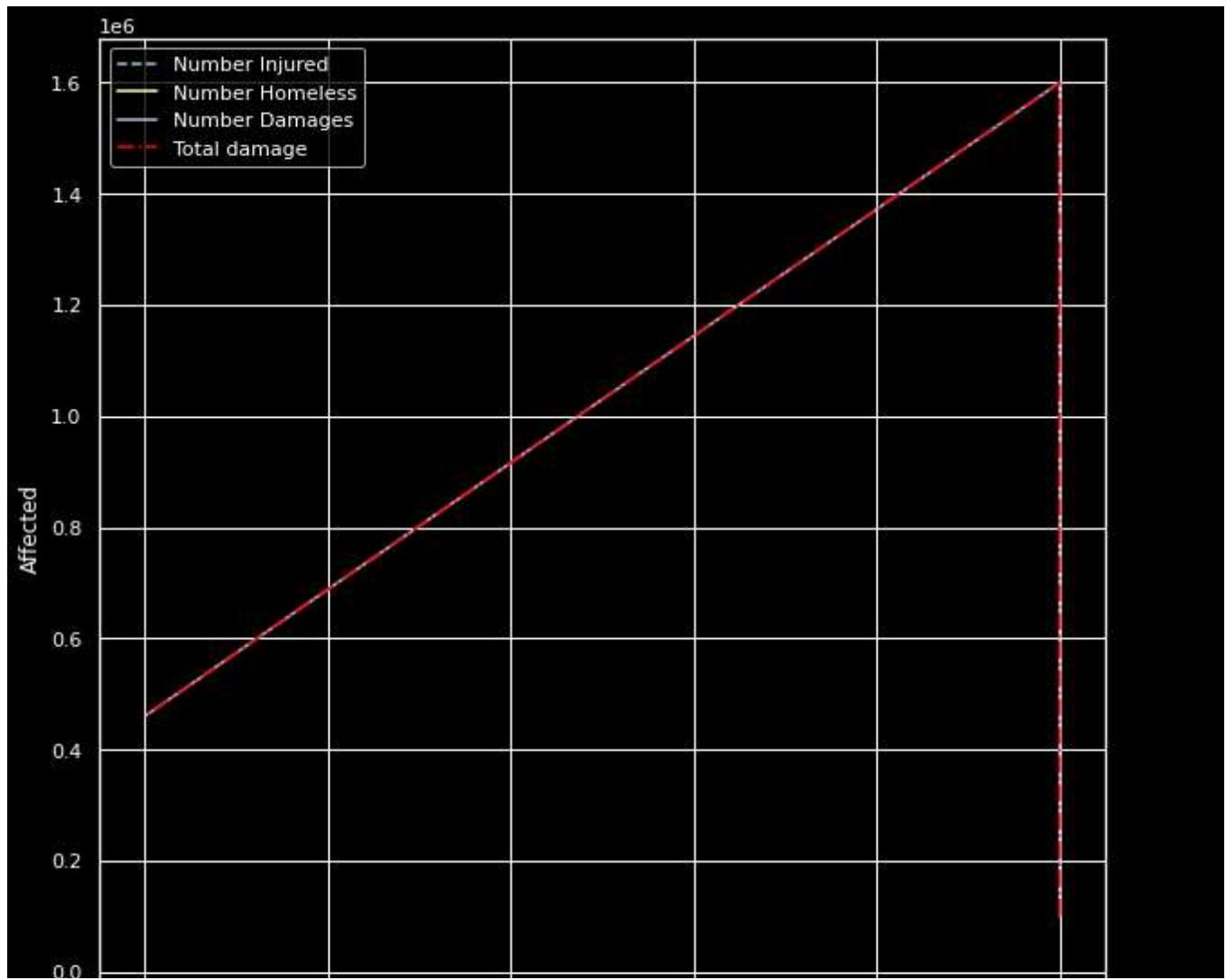
sn=df['Total Damages'].iloc[100:9000:500]
sns.scatterplot(x=df['Year'],y=df['Total Damages'],data=sn)
plt.ylabel('Total Damages in USD')
```

Text(0, 0.5, 'Total Damages in USD')



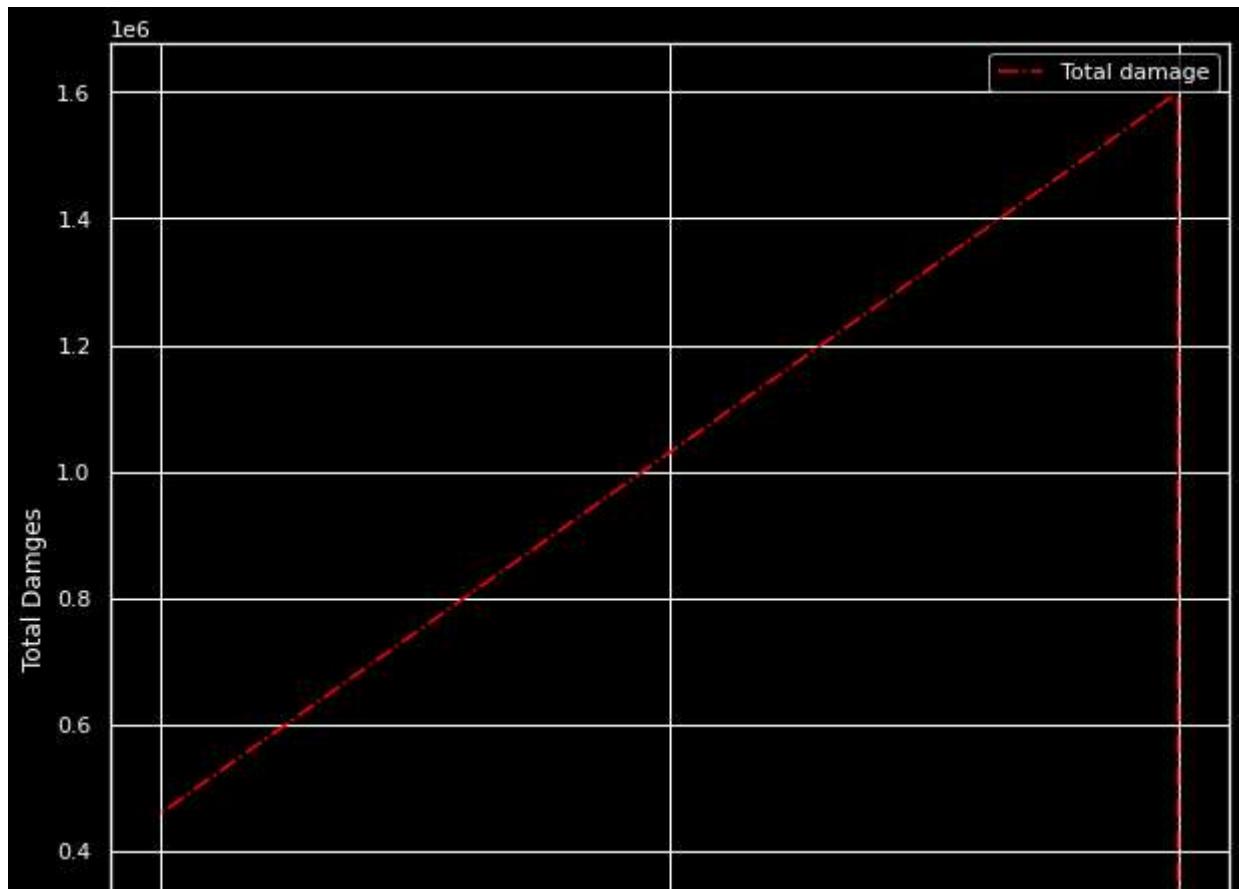
LINE PLOT

```
#setting up size of the figure
plt.figure(figsize=[10,10])
state=df['Country'].iloc[1000:9000:1000] #selecting data to make dataframe state
ar=df['No Injured'].iloc[1000:9000:1000] #selecting data to make dataframe ar
disr=df['No Homeless'].iloc[1000:9000:1000] #selecting data to make dataframe disr
dr=df['Total Damages'].iloc[1000:9000:1000] #selecting data to make dataframe dr
plt.plot(state,ar,linestyle = "--",label='Number Injured') #plotting for the data in datafram
plt.plot(state,disr,label='Number Homeless') #plotting for the data in dataframe disr
plt.plot(state,dr,label='Number Damages') #plotting for the data in dataframe dr
plt.plot(state,dr,linestyle = "-.",label='Total damage',color='red')
plt.xlabel('Disaster')
plt.ylabel('Affected')
plt.legend()
plt.show()
```



```
#setting up size of the figure
plt.figure(figsize=[10,10])

#selecting data to make from dataframe state
state=df['Disaster Subtype'].iloc[1000:9000:1000]
#selecting data to make from dataframe dr
dr=df['Total Damages'].iloc[1000:9000:1000]
plt.plot(state,dr,linestyle = "-.",label='Total damage',color='red')
plt.xlabel('Disaster')
plt.ylabel('Total Damages')
plt.legend()
plt.show()
```



HISTOGRAM



```
#setting up size of the figure
plt.figure(figsize=[10,5])
max_occurred=df.sort_values(by='Total Damages',ascending=False)
md=max_occurred.head(100)
#selecting data from dataframe md
a=md['Disaster Type']
plt.hist(a,bins=10,edgecolor='Black',color='Cyan',histtype='stepfilled')
plt.title('Histogram')
plt.ylabel('Most frequently occurred calamity')
plt.show()
```