# Exploratory\_Data\_Analysis\_Terrorism

In this task, we will be performing exploratory data analysis on the dataset "GlobalTerrorism" and try to find out the hot zone of terrorism. Also, we will derive the security issues and various insights.

```
In [1]: # Importing all the libraries needed in this notebook
   import math
   import warnings
   import numpy as np
   import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
   import seaborn as sns
   import plotly.offline as py
   import plotly.graph_objs as go
   import matplotlib.pyplot as plt
   warnings.filterwarnings('ignore')
```

In [2]: |global\_terror = pd.read\_csv('globalterrorismdb\_0718dist.csv',encoding='ISO-8859-1')

## **Loading and Reading The Data Set**

```
In [3]: global terror.head() # first five values of the dataset
Out[3]: approxdate extended resolution country country_txt region ... addnotes
                                                                        scite1 scite2 scite3 dbsource INT_LOG I
                                             Dominican
             NaN
                        0
                               NaN
                                        58
                                                          2 ...
                                                                   NaN
                                                                         NaN
                                                                               NaN
                                                                                      NaN
                                                                                              PGIS
                                                                                                         0
                                              Republic
             NaN
                        0
                                                                                              PGIS
                               NaN
                                       130
                                               Mexico
                                                                   NaN
                                                                         NaN
                                                                               NaN
                                                                                      NaN
                                                                                                         0
                        0
             NaN
                               NaN
                                       160
                                            Philippines
                                                          5
                                                                   NaN
                                                                         NaN
                                                                               NaN
                                                                                      NaN
                                                                                              PGIS
                                                                                                         -9
             NaN
                        0
                               NaN
                                        78
                                               Greece
                                                          8 ...
                                                                   NaN
                                                                         NaN
                                                                               NaN
                                                                                      NaN
                                                                                              PGIS
                                                                                                         -9
             NaN
                        0
                               NaN
                                       101
                                                          4 ...
                                                                   NaN
                                                                         NaN
                                                                               NaN
                                                                                      NaN
                                                                                              PGIS
                                                                                                         -9
                                                Japan
In [4]: global_terror.columns
Out[4]: Index(['eventid', 'iyear', 'imonth', 'iday', 'approxdate', 'extended',
                'resolution', 'country', 'country_txt', 'region',
                'addnotes', 'scite1', 'scite2', 'scite3', 'dbsource', 'INT_LOG',
                'INT_IDEO', 'INT_MISC', 'INT_ANY', 'related'],
              dtype='object', length=135)
'nwound':'Wounded','summary':'Summary','gname':'Group','targtype1_txt':'
'weaptype1_txt':'Weapon_type','motive':'Motive'},inplace=True)
```

## **Dropping out irrelevant columns**

	Year	Month	Day	Country	state	Region	city	latitude	longitude	AttackType	Killed	١
0	1970	7	2	Dominican Republic	NaN	Central America & Caribbean	Santo Domingo	18.456792	-69.951164	Assassination	1.0	
1	1970	0	0	Mexico	Federal	North America	Mexico city	19.371887	-99.086624	Hostage Taking (Kidnapping)	0.0	
2	1970	1	0	Philippines	Tarlac	Southeast Asia	Unknown	15.478598	120.599741	Assassination	1.0	
3	1970	1	0	Greece	Attica	Western Europe	Athens	37.997490	23.762728	Bombing/Explosion	NaN	
4	1970	1	0	Japan	Fukouka	East Asia	Fukouka	33.580412	130.396361	Facility/Infrastructure Attack	NaN	
glo				null val ll().sum(								
glo Ye	obal_t ar			11().sum( 0								
glo Yea Mon	obal_t ar nth			11().sum(								
Yea Mon Day	obal_t ar nth			0 0 0								
Yes Mor Day Cor	obal_t ar nth y		isnu	11().sum( 0 0 0								
Yes Mon Day Con sta	obal_t ar nth y untry		isnu	11().sum( 0 0 0 0								
Yes Mon Day Con sta Res ci	obal_t ar nth y untry ate gion ty	cerror.	isnu.	11().sum( 0 0 0 0 421 0 434								
Yes Mon Day Con sta Res cir	obal_t ar nth y untry ate gion ty titude	error.	isnu	11().sum( 0 0 0 0 421 0 434 556								
Yes Mon Day Con sta Res circles	ar nth y untry ate gion ty titude	error.	isnu	11().sum( 0 0 0 0 421 0 434 556 557								
Yes Mon Day Con sta Rep cir la lon At	obal_t ar nth y untry ate gion ty titude ngitud tackTy	error.	isnu 4 4	11().sum( 0 0 0 0 421 0 434 556 557 0								
Yes Mon Day Con sta Rep cir lan At- Ki	obal_t ar nth y untry ate gion ty titude ngitud tackTy	error.	isnu 4 4	11().sum( 0 0 0 0 421 0 434 556 557 0 313								
Year Moor Day Cook start Region Attributes Wood	ar nth y untry ate gion ty titude ngitud tackTy lled unded	error.	10 16	11().sum( 0 0 0 0 421 0 434 556 557 0 313 311								
Yes Mod Day Con stc in la lon At Ki Wood Ta	obal_t ar nth y untry ate gion ty titude ngitud tackTy lled unded rget	error.	19 10 16	11().sum( 0 0 0 0 421 0 434 556 557 0 313 311 636								
Year Mon Day Con sta Reg cir lan lon Atr Kii Won Tan Sun	obal_t ar nth y untry ate gion ty titude ngitud tackT) lled unded rget mmary	error.	19 10 16	11().sum( 0 0 0 0 421 0 434 556 557 0 313 311 636 129								
Year Mon Day Con star Rep cir la lon Atr Kii Wood Tan Sun Green	obal_t ar nth y untry ate gion ty titude ngitud tackTy lled unded rget mmary oup	error.	19 10 16	11().sum( 0 0 0 0 421 0 434 556 557 0 313 311 636 129 0								
Yes Mon Day Con sta Rep ci la lon At Ki Woo Tan Sun Gre Tan	obal_t ar nth y untry ate gion ty titude ngitud tackT) lled unded rget mmary	e de pe	19 10 16	11().sum( 0 0 0 0 421 0 434 556 557 0 313 311 636 129								

## Checking the dataset's information

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 181691 entries, 0 to 181690

In [9]: global\_terror.info() # Returns the concise summary

Data columns (total 18 columns): # Column Non-Null Count Dtype Year 181691 non-null int64
Month 181691 non-null int64
Day 181691 non-null int64
Country 181691 non-null object
state 181270 non-null object
Region 181691 non-null object
city 181257 non-null object
latitude 177135 non-null float64
longitude 177134 non-null object
AttackType 181691 non-null object 1 2 3 4 5 6 7 8 AttackType 181691 non-null object 10 Killed 171378 non-null float64
11 Wounded 165380 non-null float64
12 Target 181055 non-null object
13 Summary 115562 non-null object
14 Group 181691 non-null object 15 Target\_type 181691 non-null object 16 Weapon\_type 181691 non-null object 17 Motive 50561 non-null object

dtypes: float64(4), int64(3), object(11)

memory usage: 25.0+ MB

### **Data Visualization**

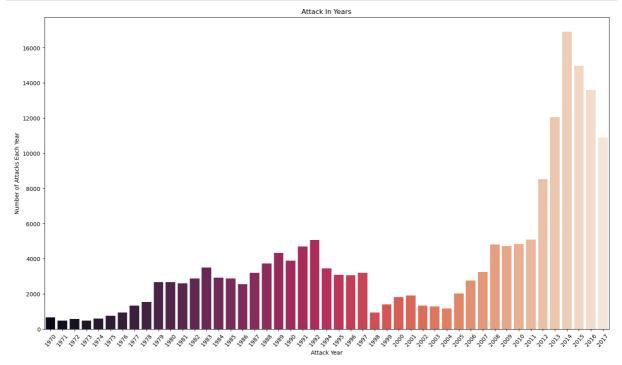
### **Destructive Features**

```
In [10]: print("Country with the most attacks:",global_terror['Country'].value_counts().idxmax())
          print("City with the most attacks:",global_terror['city'].value_counts().index[1]) #as first entry
          print("Region with the most attacks:",global_terror['Region'].value_counts().idxmax())
          print("Year with the most attacks:",global_terror['Year'].value_counts().idxmax())
         print("Month with the most attacks:",global_terror['Month'].value_counts().idxmax())
print("Group with the most attacks:",global_terror['Group'].value_counts().index[1])
          print("Most Attack Types:",global_terror['AttackType'].value_counts().idxmax())
          Country with the most attacks: Iraq
          City with the most attacks: Baghdad
          Region with the most attacks: Middle East & North Africa
          Year with the most attacks: 2014
          Month with the most attacks: 5
          Group with the most attacks: Taliban
          Most Attack Types: Bombing/Explosion
In [13]: | from wordcloud import WordCloud
          from scipy import signal
          cities = global_terror.state.dropna(False)
          plt.subplots(figsize=(20,10))
          wordcloud = WordCloud(background_color = 'black',
                                width = 500,
                                height = 400).generate(' '.join(cities))
          plt.axis('off')
          plt.imshow(wordcloud)
          plt.show()
```

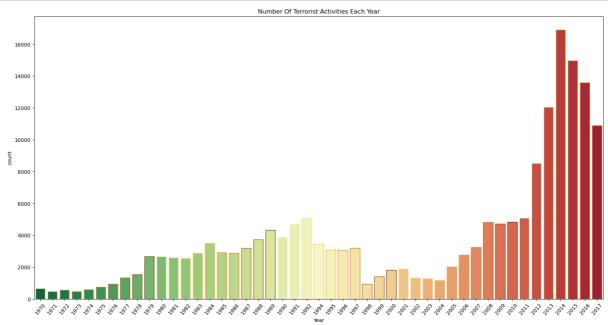


```
In [14]: global_terror['Year'].value_counts(dropna = False).sort_index()
Out[14]: 1970
                    651
         1971
                    471
         1972
                    568
         1973
                    473
         1974
                    581
                    740
         1975
         1976
                    923
         1977
                   1319
         1978
                   1526
         1979
                   2662
         1980
                   2662
         1981
                   2586
         1982
                   2544
         1983
                   2870
         1984
                   3495
         1985
                   2915
         1986
                   2860
         1987
                   3183
         1988
                   3721
         1989
                   4324
         1990
                   3887
                   4683
         1991
         1992
                   5071
         1994
                   3456
                   3081
         1995
         1996
                   3058
         1997
                   3197
         1998
                   934
         1999
                   1395
         2000
                   1814
                   1906
         2001
         2002
                   1333
         2003
                   1278
         2004
                   1166
         2005
                   2017
         2006
                   2758
         2007
                   3242
         2008
                   4805
         2009
                   4721
         2010
                   4826
         2011
                   5076
         2012
                   8522
         2013
                  12036
         2014
                  16903
         2015
                  14965
         2016
                  13587
         2017
                  10900
         Name: Year, dtype: int64
```

## **Terrorist Activities Each Year**

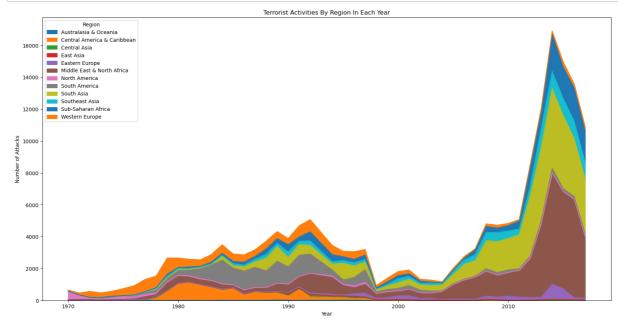


```
In [18]: plt.subplots(figsize=(20,10))
    sns.countplot(x='Year', data=global_terror, palette='RdYlGn_r', edgecolor=sns.color_palette("YlOri
    plt.xticks(rotation=50)
    plt.title('Number Of Terrorist Activities Each Year')
    plt.show()
```



## **Terrorist Activities By Region In Each Year**

```
In [19]: pd.crosstab(global_terror.Year, global_terror.Region).plot(kind='area',figsize=(20,10))
    plt.title('Terrorist Activities By Region In Each Year')
    plt.ylabel('Number of Attacks')
    plt.show()
```



```
In [20]: global_terror['Wounded'] = global_terror['Wounded'].fillna(0).astype(int)
    global_terror['Killed'] = global_terror['Killed'].fillna(0).astype(int)
    global_terror['Casualities'] = global_terror['Killed'] + global_terror['Wounded']
```

```
In [21]: # Top 50 worst terrorist attacks
global_terror1 = global_terror.sort_values(by='Casualities',ascending=False)[:50]
```

In [22]: heat=global\_terror1.pivot\_table(index='Country',columns='Year',values='Casualities')
heat.fillna(0,inplace=True)

In [23]: heat.head()

Out[23]:	Year	1982	1984	1987	1992	1994	1995	1996	1997	1998	2001	 2005	2006	2007	2008	2009	2013	2
	Country																	
	Afghanistan	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.0	0.0	
	Algeria	0.0	0.0	0.0	0.0	0.0	0.0	0.0	450.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.0	0.0	
	Chad	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	1161.0	0.0	0.0	
	Egypt	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.0	0.0	
	Ethiopia	0.0	0.0	0.0	500.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.0	0.0	

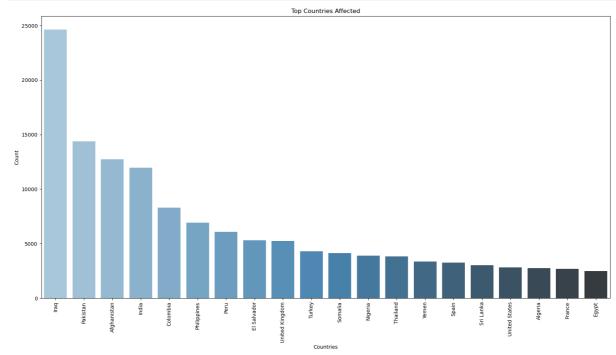
5 rows × 21 columns

```
In [24]: import plotly.offline as py
py.init_notebook_mode(connected=True)
import plotly.graph_objs as go
colorscale = [[0, '#edf8fb'], [.3, '#00BFFF'], [.6, '#8856a7'], [1, '#810f7c']]
heatmap = go.Heatmap(z=heat.values, x=heat.columns, y=heat.index, colorscale=colorscale)
data = [heatmap]
layout = go.Layout(
    title='Top 50 Worst Terror Attacks in History from 1982 to 2017',
    xaxis = dict(ticks='', nticks=20),
    yaxis = dict(ticks='')
)
fig = go.Figure(data=data, layout=layout)
py.iplot(fig, filename='heatmap',show_link=False)
```

```
In [25]: global_terror.Country.value_counts()[:21]
Out[25]: Iraq
                          24636
                         14368
         Pakistan
         Afghanistan
                          12731
                         11960
         India
         Colombia
                           8306
         Philippines
                          6908
                           6096
         El Salvador 5320
United Kingdom 5235
         Turkey
                           4292
         Somalia
                           4142
         Nigeria
                          3907
         Thailand
                          3849
                          3347
         Yemen
         Spain
                           3249
         Sri Lanka
                           3022
         United States
                           2836
                           2743
         Algeria
         France
                           2693
         Egypt
                           2479
                           2478
         Lebanon
         Name: Country, dtype: int64
```

## **Top Countries Affected By Terrorist Attacks**

```
In [27]: plt.subplots(figsize=(20,10))
    sns.barplot(x=global_terror['Country'].value_counts()[:20].index, y=global_terror['Country'].value
    plt.title('Top Countries Affected')
    plt.xlabel('Countries')
    plt.ylabel('Count')
    plt.xticks(rotation=90)
    plt.show()
```



## **ANALYSIS ON CUSTOMIZED DATA**

### Terrorist Attacks of a Particular year and their Locations

Let's look at the terrorist acts in the world over a certain year.

Out[33]: Make this Notebook Trusted to load map: File -> Trust Notebook

From the above map, we can depict that the maximum attacks carried out in the year 2001 was on the African Continent, almost 1325 attacks. Then, the continent South America faced the highest number of attacks, i.e. 258.

## **Terrorist's Origanizations Operations In Each Country**

```
In [35]: global_terror.Group.value_counts()[1:20]
Out[35]: Taliban
                                                              7478
         Islamic State of Iraq and the Levant (ISIL)
                                                              5613
         Shining Path (SL)
                                                              4555
         Farabundo Marti National Liberation Front (FMLN)
                                                              3351
         Al-Shabaab
                                                              3288
         New People's Army (NPA)
                                                              2772
         Irish Republican Army (IRA)
                                                              2671
         Revolutionary Armed Forces of Colombia (FARC)
                                                              2487
                                                              2418
         Kurdistan Workers' Party (PKK)
                                                              2310
         Basque Fatherland and Freedom (ETA)
                                                              2024
         Communist Party of India - Maoist (CPI-Maoist)
                                                              1878
                                                              1630
         Maoists
         Liberation Tigers of Tamil Eelam (LTTE)
                                                              1606
         National Liberation Army of Colombia (ELN)
                                                              1561
         Tehrik-i-Taliban Pakistan (TTP)
                                                              1351
         Palestinians
                                                              1125
         Houthi extremists (Ansar Allah)
                                                              1062
         Al-Qaida in the Arabian Peninsula (AQAP)
                                                              1020
         Name: Group, dtype: int64
In [36]: test = global_terror[global_terror.Group.isin(['Shining Path (SL)','Taliban','Islamic State of Ir
```

```
In [37]: test.Country.unique()
Out[37]: array(['Peru', 'Bolivia', 'Colombia', 'Argentina', 'Brazil', 'Mexico', 'Afghanistan', 'Pakistan', 'Syria', 'Iraq', 'Turkey', 'Tunisia', 'Lebanon', 'Turkmenistan', 'Israel', 'Belgium', 'Egypt', 'Libya',
                     'Saudi Arabia', 'West Bank and Gaza Strip', 'France', 'Bahrain', 'Jordan', 'Somalia', 'Germany', 'Yemen', 'Philippines', 'Malaysia', 'Indonesia', 'Russia', 'Georgia', 'United Kingdom', 'Iran',
                     'Australia'], dtype=object)
In [38]: |global_terror_df_group = global_terror.dropna(subset=['latitude','longitude'])
In [39]: global_terror_df_group = global_terror_df_group.drop_duplicates(subset=['Country','Group'])
In [40]: terrorist_groups = global_terror.Group.value_counts()[1:8].index.tolist()
            global_terror_df_group = global_terror_df_group.loc[global_terror_df_group.Group.isin(terrorist_g
            print(global_terror_df_group.Group.unique())
            ["New People's Army (NPA)" 'Irish Republican Army (IRA)'
              'Shining Path (SL)' 'Farabundo Marti National Liberation Front (FMLN)'
             'Taliban' 'Al-Shabaab' 'Islamic State of Iraq and the Levant (ISIL)']
In [41]: map = folium.Map(location=[50, 0], tiles="CartoDB positron", zoom start=2)
            markerCluster = folium.plugins.MarkerCluster().add_to(map)
            for i in range(0,len(global_terror_df_group)):
                 folium.Marker([global_terror_df_group.iloc[i]['latitude'],global_terror_df_group.iloc[i]['long
                                  popup='Group:{}<br/>Country:{}'.format(global_terror_df_group.iloc[i]['Group'],
global_terror_df_group.iloc[i]['Country'])).add_to(map)
            map
```

Out[41]: Make this Notebook Trusted to load map: File -> Trust Notebook

```
In [42]: |m1 = folium.Map(location=[50, 0], tiles="CartoDB positron", zoom_start=2)
         marker_cluster = MarkerCluster(
              name='clustered icons',
              overlay=True,
              control=False,
              icon_create_function=None
         for i in range(0,len(global_terror_df_group)):
              marker=folium.Marker([global_terror_df_group.iloc[i]['latitude'],global_terror_df_group.iloc[
              popup='Group:{}<br>Country:{}'.format(global_terror_df_group.iloc[i]['Group'],
                                                      global_terror_df_group.iloc[i]['Country'])
              folium.Popup(popup).add_to(marker)
              marker_cluster.add_child(marker)
         marker_cluster.add_to(m1)
         folium.TileLayer('openstreetmap').add_to(m1)
#folium.TileLayer('Mapbox Bright').add_to(m1)
         folium.TileLayer('cartodbdark_matter').add_to(m1)
         folium.TileLayer('stamentoner').add_to(m1)
         folium.LayerControl().add_to(m1)
         m1
```

Out[42]: Make this Notebook Trusted to load map: File -> Trust Notebook

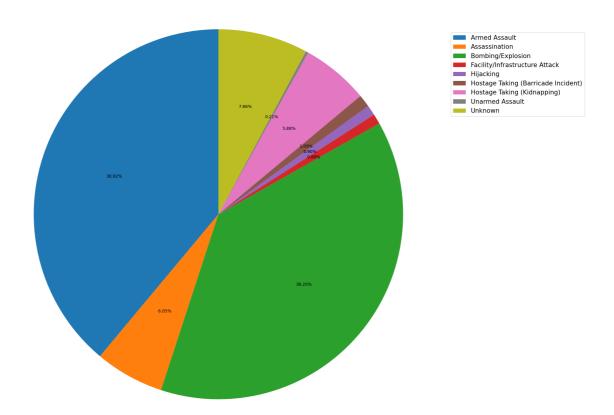
]: g	10	bal_t	error.	head	()								
		Year	Month	Day	Country	state	Region	city	latitude	longitude	AttackType	Killed	Wound
•	0	1970	7	2	Dominican Republic	NaN	Central America & Caribbean	Santo Domingo	18.456792	-69.951164	Assassination	1	
,	1	1970	0	0	Mexico	Federal	North America	Mexico city	19.371887	-99.086624	Hostage Taking (Kidnapping)	0	
:	2	1970	1	0	Philippines	Tarlac	Southeast Asia	Unknown	15.478598	120.599741	Assassination	1	
;	3	1970	1	0	Greece	Attica	Western Europe	Athens	37.997490	23.762728	Bombing/Explosion	0	
•	4	1970	1	0	Japan	Fukouka	East Asia	Fukouka	33.580412	130.396361	Facility/Infrastructure Attack	0	
4													<b>&gt;</b>

```
In [44]: # Total Number of people killed in terror attack
          killData = global_terror.loc[:,'Killed']
         print('Number of people killed by terror attack:', int(sum(killData.dropna())))# drop the NaN value
          Number of people killed by terror attack: 411868
In [45]: # Let's look at what types of attacks these deaths were made of.
          attackData = global_terror.loc[:,'AttackType']
         typeKillData = pd.concat([attackData, killData], axis=1)
In [46]: typeKillData.head()
Out[46]:
                         AttackType Killed
          0
                        Assassination
          1 Hostage Taking (Kidnapping)
                                       0
          2
                        Assassination
                                        1
                    Bombing/Explosion
          3
                                       0
              Facility/Infrastructure Attack
                                       0
In [47]: typeKillFormatData = typeKillData.pivot_table(columns='AttackType', values='Killed', aggfunc='sum
          typeKillFormatData
Out[47]:
                                                                                        Hostage
                                                                                                    Hostage
                                                           Facility/Infrastructure
                                                                                         Taking
                                                                                                            Unarm
                      Armed
          AttackType
                            Assassination Bombing/Explosion
                                                                             Hijacking
                                                                                                     Taking
                                                                                       (Barricade
                     Assault
                                                                       Attack
                                                                                                             Assa
                                                                                                (Kidnapping)
                                                                                        Incident)
               Killed 160297
                                   24920
                                                    157321
                                                                        3642
                                                                                 3718
                                                                                           4478
                                                                                                      24231
                                                                                                                 8
In [48]: typeKillFormatData.info()
          <class 'pandas.core.frame.DataFrame'>
          Index: 1 entries, Killed to Killed
         Data columns (total 9 columns):
          # Column
                                                      Non-Null Count Dtype
          0
              Armed Assault
                                                      1 non-null
                                                                       int32
          1
              Assassination
                                                      1 non-null
                                                                       int32
              Bombing/Explosion
                                                      1 non-null
                                                                       int32
          3
              Facility/Infrastructure Attack
                                                      1 non-null
                                                                       int32
                                                      1 non-null
          4
              Hijacking
                                                                       int32
              Hostage Taking (Barricade Incident)
                                                      1 non-null
                                                                       int32
          6
              Hostage Taking (Kidnapping)
                                                      1 non-null
                                                                       int32
          7
              Unarmed Assault
                                                      1 non-null
                                                                       int32
              Unknown
                                                      1 non-null
                                                                       int32
          dtypes: int32(9)
         memory usage: 152.0+ bytes
```

```
In [50]: labels = typeKillFormatData.columns.tolist()
    transposed = typeKillFormatData.T
    values = transposed.values.flatten().tolist() # FLatten the 2D array

fig, ax = plt.subplots(figsize=(20, 20), subplot_kw=dict(aspect="equal"))
    plt.pie(values, startangle=90, autopct='%.2f%')
    plt.title('Types Of Terrorist Attacks That Cause Deaths', fontsize=30)
    plt.legend(labels, loc='upper right', bbox_to_anchor=(1.3, 0.9), fontsize=15) # Location of the plt.show()
```

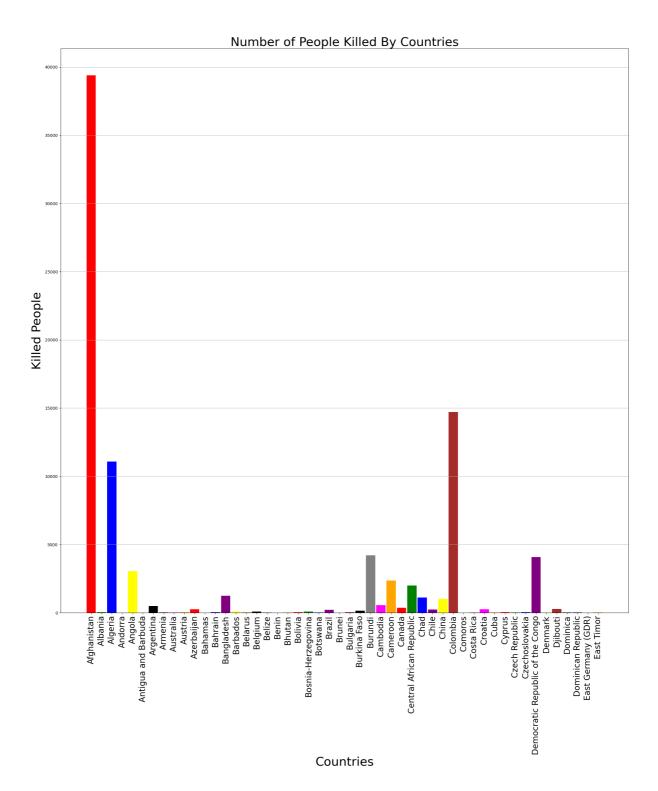
Types Of Terrorist Attacks That Cause Deaths



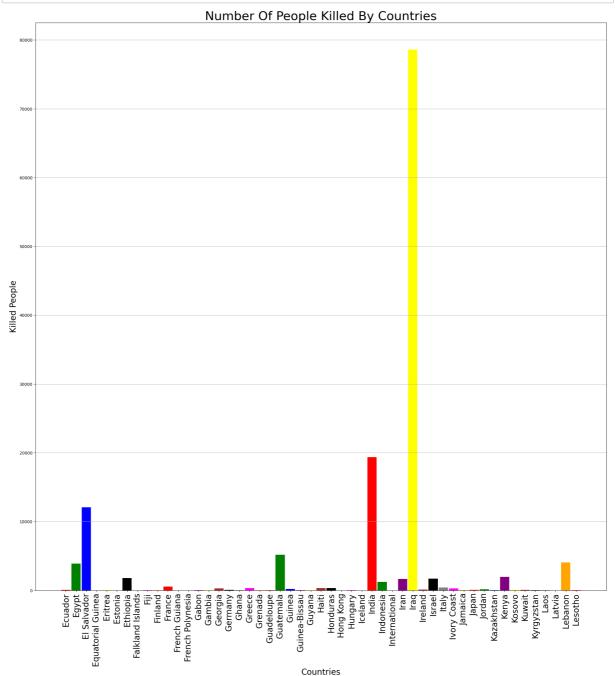
```
In [51]: global_terror.head(2)
Out[51]:
              Year Month Day
                                 Country
                                            state
                                                    Region
                                                                city
                                                                       latitude
                                                                                longitude
                                                                                           AttackType Killed Wounded
                                                                                                                         T
                                                    Central
                                Dominican
                                                              Santo
           0 1970
                        7
                                                                     18.456792 -69.951164 Assassination
                                            NaN
                                                  America &
                                                                                                           1
                                                                                                                    0
                                 Republic
                                                            Domingo
                                                                                                                       Gu
                                                  Caribbean
                                                                                              Hostage
                                                      North
                                                             Mexico
                             0
                                                                                                                    0
           1 1970
                        0
                                  Mexico Federal
                                                                     19.371887 -99.086624
                                                                                                Taking
                                                                                                          0
                                                                                                                        CI
                                                   America
                                                                citv
                                                                                           (Kidnapping)
                                                                                                                       daι
In [52]: # Number of Killed in Terrorist Attacks by Countries
          countryData = global_terror.loc[:,'Country']
          # countyData
          countryKillData = pd.concat([countryData, killData], axis=1)
```

```
In [53]: countryKillFormatData = countryKillData.pivot_table(columns='Country', values='Killed', aggfunc='s
          countryKillFormatData
Out[53]:
                                                             Antigua and
           Country Afghanistan Albania Algeria Andorra Angola
                                                                     Argentina Armenia Australia Austria ... Vietnam
                                                             Barbuda
                                                                                                    30 ...
            Killed
                        39384
                                       11066
                                                   0
                                                       3043
                                                                   0
                                                                          490
                                                                                    37
                                                                                            23
                                                                                                                1
                                  42
          1 rows × 205 columns
In [54]: fig_size = plt.rcParams["figure.figsize"]
          fig_size[0]=25
          fig_size[1]=25
          plt.rcParams["figure.figsize"] = fig_size
```

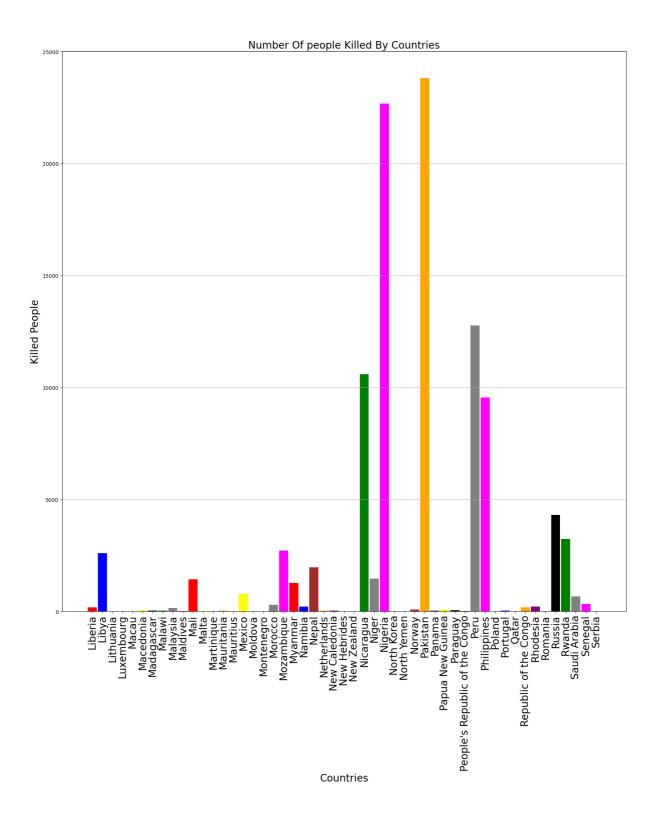
```
In [55]: labels = countryKillFormatData.columns.tolist()
         labels = labels[:50] #50 bar provides nice view
         index = np.arange(len(labels))
         transpoze = countryKillFormatData.T
         values = transpoze.values.tolist()
         values = values[:50]
         values = [int(i[0]) for i in values] # convert float to int
         colors = ['red', 'green', 'blue', 'purple', 'yellow', 'brown', 'black', 'gray', 'magenta', 'orange
fig, ax = plt.subplots(1, 1)
         ax.yaxis.grid(True)
         fig_size = plt.rcParams["figure.figsize"]
         fig_size[0]=25
         fig_size[1]=25
         plt.rcParams["figure.figsize"] = fig_size
         plt.bar(index, values, color = colors, width = 0.9)
         plt.ylabel('Killed People', fontsize=30)
         plt.xlabel('Countries', fontsize = 30)
         plt.xticks(index, labels, fontsize=20, rotation=90)
         plt.title('Number of People Killed By Countries', fontsize = 30)
         # print(fig_size)
         plt.show()
```



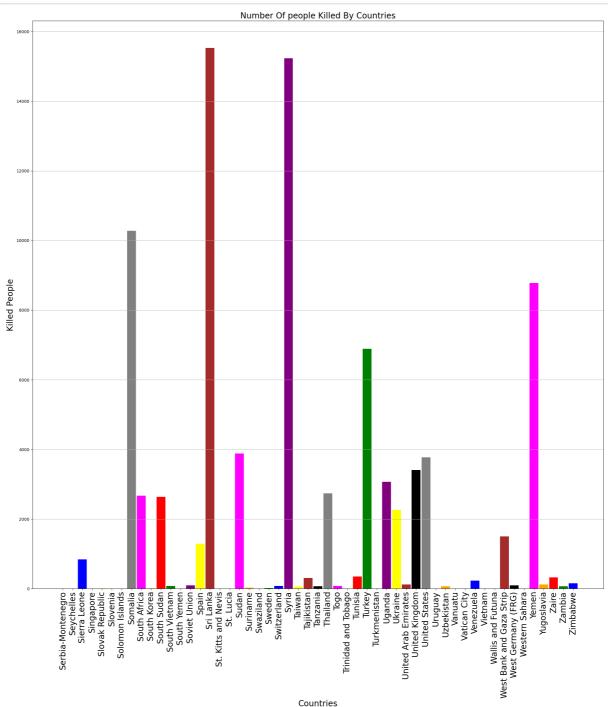
```
In [56]: labels = countryKillFormatData.columns.tolist()
         labels = labels[50:101]
         index = np.arange(len(labels))
         transpoze = countryKillFormatData.T
         values = transpoze.values.tolist()
         values = values[50:101]
         values = [int(i[0]) for i in values]
         colors = ['red', 'green', 'blue', 'purple', 'yellow', 'brown', 'black', 'gray', 'magenta', 'orange
fig, ax = plt.subplots(1, 1)
         ax.yaxis.grid(True)
         fig_size = plt.rcParams["figure.figsize"]
         fig_size[0]=20
         fig_size[1]=20
         plt.rcParams["figure.figsize"] = fig_size
         plt.bar(index, values, color = colors, width = 0.9)
         plt.ylabel('Killed People', fontsize=20)
         plt.xlabel('Countries', fontsize = 20)
         plt.xticks(index, labels, fontsize=20, rotation=90)
         plt.title('Number Of People Killed By Countries', fontsize = 30)
         plt.show()
```



```
In [57]: labels = countryKillFormatData.columns.tolist()
         labels = labels[101:152]
         index = np.arange(len(labels))
         transpoze = countryKillFormatData.T
         values = transpoze.values.tolist()
         values = values[101:152]
         values = [int(i[0]) for i in values]
         colors = ['red', 'blue', 'brown', 'orange', 'purple', 'yellow', 'black', 'green', 'gray', 'magenta'
         fig, ax = plt.subplots(1, 1)
         ax.yaxis.grid(True)
         fig_size = plt.rcParams["figure.figsize"]
         fig_size[0]=25
         fig_size[1]=25
         plt.rcParams["figure.figsize"] = fig_size
         plt.bar(index, values, color = colors, width = 0.9)
         plt.ylabel('Killed People', fontsize=20)
         plt.xlabel('Countries', fontsize = 20)
         plt.xticks(index, labels, fontsize=20, rotation=90)
         plt.title('Number Of people Killed By Countries', fontsize = 20)
         plt.show()
```



```
In [58]: labels = countryKillFormatData.columns.tolist()
         labels = labels[152:206]
         index = np.arange(len(labels))
         transpoze = countryKillFormatData.T
         values = transpoze.values.tolist()
         values = values[152:206]
         values = [int(i[0]) for i in values]
         colors = ['red', 'green', 'blue', 'purple', 'yellow', 'brown', 'black', 'gray', 'magenta', 'orange
fig, ax = plt.subplots(1, 1)
         ax.yaxis.grid(True)
         fig_size = plt.rcParams["figure.figsize"]
         fig_size[0]=25
         fig_size[1]=25
         plt.rcParams["figure.figsize"] = fig_size
         plt.bar(index, values, color = colors, width = 0.9)
         plt.ylabel('Killed People', fontsize=20)
         plt.xlabel('Countries', fontsize = 20)
         plt.xticks(index, labels, fontsize=20, rotation=90)
         plt.title('Number Of people Killed By Countries', fontsize = 20)
         plt.show()
```



Countries

### **CONCLUSIONS:**

From the above graphs, we can see that the countries where most people are killed are: Afghanisthan, Columbia, Iran, Sri lanka, Syria, Somalia, Yemen naming a few. Even though there is a perception that Muslims are supporters of terrorism, but Muslims are the people who are most damaged by terrorist attacks.

So after different type of analyzation ,Overall terrorism is suddenly increased from 2010 and I have ranked the Hot zone in terms of terrorist activities in all regions -

#### **#Middle East & North Africa**

#### Iraq

Overall having highest Number of terrorist rate and sudden increase in terrorist activity after year 2010

#### Libya

Overall having low terrorism activity, Second Highest in terms of successful terrorist activities which takes place in this country and an increased in terrorism rate after 2010.

#### Yemen

Overall having low terrorism activity, Third Highest in terms of successful terrorist activities which takes place in this country and an increased in terrorism rate after 2010.

#### **#South Asia**

#### **Afghanistan**

Overall having 1st Highest terrorist rates and 1st highest in terms of extended terrorist rates and a very high increase in terrorism rate after 2010.

#### **Pakistan**

Overall having 2nd highest terrorist rates but there is a good decrease in terrorist rates if we consider the 20s century and an increase in terrorism rate after 2010 but not as that of increase like Afghanistan and India.

#### India

Overall having 3rd highest terrorist rates but if we consider extended terrorist rate so this country is 2nd highest and a very high increase in terrorism rate after 2010.

#### **#Sub-Saharan Africa**

#### Nigeria

Overall having low terrorism activity but 1st Highest in terms of extended terrorist activities which takes place in this country and 2nd in terms of increased of terrorism rate after 2010.

### Somalia

Overall having low terrorism activity but 2nd Highest in terms of extended terrorist activities which takes place in this country and 1st in terms of increased of terrorism rate after 2010.

#### Sudan

Overall having low terrorism activity but 3rd Highest in terms of extended terrorist activities which takes place in this country and a bit increase in terrorism rate after 2010.

### #Solution

- More security surveillance required at Iraq.
- Noticing the trends of terrorism activities , hugely populated regions suffer major kill ratios. This must be controlled with strict border rules

# Thank You!

GitHub: <a href="https://github.com/anujtiwari21?tab=repositories">https://github.com/anujtiwari21?tab=repositories</a> <a href="https://github.com/anujtiwari21?tab=repositories">(https://github.com/anujtiwari21?tab=repositories</a>)