

Exploratory Data Analysis on Dataset -Terrorism

As a security/defense analyst,try to find out the hot zone of terrorism.reading and analyzing the dataset:

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Link for the dataset:<https://bit.ly/2TK5Xn5> (<https://bit.ly/2TK5Xn5>) 🔒

Importing the dependencies

In [1]:

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3 import seaborn as sns
4 import pandas as pd
5 import warnings
6 warnings.filterwarnings("ignore")
7 from sklearn.preprocessing import LabelEncoder
```

Loading the dataset

In [4]:

```
1 dataset=pd.read_csv("C:/Users/sakshi itnare/Downloads/Global Terrorism - START data.
```

In [6]:

```
1 dataset
```

Out[6]:

	eventid	iyear	imonth	iday	approxdate	extended	resolution	country	count
0	1970000000001	1970	7	2	NaN	0	NaN	58	Dorr Re
1	1970000000002	1970	0	0	NaN	0	NaN	130	M
2	1970010000001	1970	1	0	NaN	0	NaN	160	Philij
3	1970010000002	1970	1	0	NaN	0	NaN	78	C
4	1970010000003	1970	1	0	NaN	0	NaN	101	
...	
181686	201712310022	2017	12	31	NaN	0	NaN	182	Si
181687	201712310029	2017	12	31	NaN	0	NaN	200	
181688	201712310030	2017	12	31	NaN	0	NaN	160	Philij
181689	201712310031	2017	12	31	NaN	0	NaN	92	
181690	201712310032	2017	12	31	NaN	0	NaN	160	Philij
181691 rows × 135 columns									

Exploratory Data Analysis

In [7]:

```
1 dataset.head()
```

Out[7]:

	eventid	iyear	imonth	iday	approxdate	extended	resolution	country	country_txt
0	1970000000001	1970	7	2	NaN	0	NaN	58	Dominican Republic
1	1970000000002	1970	0	0	NaN	0	NaN	130	Mexico
2	1970010000001	1970	1	0	NaN	0	NaN	160	Philippines
3	1970010000002	1970	1	0	NaN	0	NaN	78	Greece
4	1970010000003	1970	1	0	NaN	0	NaN	101	Japan

5 rows × 135 columns



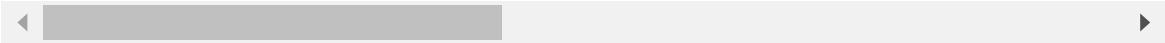
In [8]:

```
1 dataset.tail()
```

Out[8]:

	eventid	iyear	imonth	iday	approxdate	extended	resolution	country	count
181686	201712310022	2017	12	31	NaN	0	NaN	182	Si
181687	201712310029	2017	12	31	NaN	0	NaN	200	
181688	201712310030	2017	12	31	NaN	0	NaN	160	Philij
181689	201712310031	2017	12	31	NaN	0	NaN	92	
181690	201712310032	2017	12	31	NaN	0	NaN	160	Philij

5 rows × 135 columns



Printing the shape and features of dataset

In [9]:

```
1 dataset.shape
```

Out[9]:

(181691, 135)

In [10]:

```
1 dataset.columns
```

Out[10]:

```
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)
```

In [12]:

```
1 for columns in dataset.columns:  
2     print(columns)
```

```
eventid  
iyear  
imonth  
iday  
approxdate  
extended  
resolution  
country  
country_txt  
region  
region_txt  
provstate  
city  
latitude  
longitude  
specificity  
vicinity  
location  
summary  
...
```

Feature Selection

In [17]:

```
1 data=dataset[["eventid","iyear","imonth","iday","gname","country_txt","provstate","r  
2             "longitude","latitude","target1","nkill","nwound",  
3             "summary","gname","motive"]]
```

In [18]:

```
1 data.head()
```

Out[18]:

	eventid	iyear	imonth	iday	gname	country_txt	provstate	region_txt	longitu
0	1970000000001	1970	7	2	MANO-D	Dominican Republic	NaN	Central America & Caribbean	-69.951
1	1970000000002	1970	0	0	23rd of September Communist League	Mexico	Federal	North America	-99.086
2	1970010000001	1970	1	0	Unknown	Philippines	Tarlac	Southeast Asia	120.599
3	1970010000002	1970	1	0	Unknown	Greece	Attica	Western Europe	23.762
4	1970010000003	1970	1	0	Unknown	Japan	Fukouka	East Asia	130.396



In [19]:

```
1 data.shape
```

Out[19]:

(181691, 16)

In [20]:

```
1 data.isnull()
```

Out[20]:

	eventid	iyear	imonth	iday	gname	country_txt	provstate	region_txt	longitude
0	False	False	False	False	False	False	True	False	False
1	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False
...
181686	False	False	False	False	False	False	False	False	False
181687	False	False	False	False	False	False	False	False	False
181688	False	False	False	False	False	False	False	False	False
181689	False	False	False	False	False	False	False	False	False
181690	False	False	False	False	False	False	False	False	False

181691 rows × 16 columns



In [21]:

```
1 data.isnull().sum()
```

Out[21]:

```
eventid      0
iyear        0
imonth       0
iday         0
gname        0
country_txt   0
provstate    421
region_txt   0
longitude    4557
latitude     4556
target1      636
nkill        10313
nwound       16311
summary      66129
gname        0
motive       131130
dtype: int64
```

In [25]:

```
1 data.isnull().sum()
```

Out[25]:

```
eventid      0
iyear        0
imonth       0
iday         0
gname        0
country_txt   0
provstate    421
region_txt   0
longitude    4557
latitude     4556
target1      636
nkill        10313
nwound       16311
summary      66129
gname        0
motive       131130
dtype: int64
```

Displaying the shape and info of the data

In [26]:

```
1 data.shape
```

Out[26]:

```
(181691, 16)
```

In [28]:

```
1 data.describe()
```

Out[28]:

	eventid	iyear	imonth	iday	longitude	lati
count	1.816910e+05	181691.000000	181691.000000	181691.000000	1.771340e+05	177135.00
mean	2.002705e+11	2002.638997	6.467277	15.505644	-4.586957e+02	23.49
std	1.325957e+09	13.259430	3.388303	8.814045	2.047790e+05	18.56
min	1.970000e+11	1970.000000	0.000000	0.000000	-8.618590e+07	-53.15
25%	1.991021e+11	1991.000000	4.000000	8.000000	4.545640e+00	11.51
50%	2.009022e+11	2009.000000	6.000000	15.000000	4.324651e+01	31.46
75%	2.014081e+11	2014.000000	9.000000	23.000000	6.871033e+01	34.68
max	2.017123e+11	2017.000000	12.000000	31.000000	1.793667e+02	74.63

In [29]:

```
1 data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 181691 entries, 0 to 181690
Data columns (total 16 columns):
#   Column          Non-Null Count  Dtype
---  -
0   eventid         181691 non-null int64
1   iyear           181691 non-null int64
2   imonth          181691 non-null int64
3   iday            181691 non-null int64
4   gname           181691 non-null object
5   country_txt     181691 non-null object
6   provstate       181270 non-null object
7   region_txt      181691 non-null object
8   longitude       177134 non-null float64
9   latitude        177135 non-null float64
10  target1         181055 non-null object
11  nkill           171378 non-null float64
12  nwound          165380 non-null float64
13  summary         115562 non-null object
14  gname           181691 non-null object
15  motive          50561 non-null  object
dtypes: float64(4), int64(4), object(8)
memory usage: 22.2+ MB
```


Correlation and Heatmap

In [30]:

```
1 data.corr()
```

Out[30]:

	eventid	year	imonth	iday	longitude	latitude	nkill	nwound
eventid	1.000000	0.999996	0.002706	0.018336	0.003907	0.166886	0.015351	0.015282
year	0.999996	1.000000	0.000139	0.018254	0.003917	0.166933	0.015341	0.015273
imonth	0.002706	0.000139	1.000000	0.005497	-0.003880	-0.015978	0.003463	0.002938
iday	0.018336	0.018254	0.005497	1.000000	-0.002285	0.003423	-0.003693	-0.001268
longitude	0.003907	0.003917	-0.003880	-0.002285	1.000000	0.001463	-0.000562	0.000223
latitude	0.166886	0.166933	-0.015978	0.003423	0.001463	1.000000	-0.018124	0.015988
nkill	0.015351	0.015341	0.003463	-0.003693	-0.000562	-0.018124	1.000000	0.534375
nwound	0.015282	0.015273	0.002938	-0.001268	0.000223	0.015988	0.534375	1.000000

In [31]:

```
1 plt.figure(figsize=(10,5))
2 sns.heatmap(data.corr(),annot=True)
3 plt.show()
```



Number of unique values

In [32]:

```
1 data.nunique()
```

Out[32]:

eventid	181691
iyear	47
imonth	13
iday	32
gname	3537
country_txt	205
provstate	2855
region_txt	12
longitude	48039
latitude	48322
target1	86006
nkill	205
nwound	238
summary	112492
gname	3537
motive	14490

dtype: int64