THE SPARKS FOUNDATION INTERNSHIP

Task 4:Exploratory Data Analysis Global Terrorism

imports and loading of dataset:

```
In [58]: import pandas as pd
          import numpy as np # linear algebra
          import matplotlib.pvplot as plt
          import matplotlib.patches as mpatches
          import seaborn as sns
          import numpy as np
          import pandas as pd
          import numpy as np # linear algebra
          import matplotlib.pyplot as plt
          import matplotlib.patches as mpatches
          import seaborn as sns
          import numpy as np
         df = pd.read csv(r"D:\[CN] Data Structure\globalterrorismdb 0718dist.csv",encoding="ISO-8859-1")
          print(df)
          D:\anaconda 3\lib\site-packages\IPython\core\interactiveshell.py:3063: DtypeWarning: Columns (4,6,31,33,61,62,63,76,79,
          90,92,94,96,114,115,121) have mixed types. Specify dtype option on import or set low memory=False.
            interactivity=interactivity, compiler=compiler, result=result)
                       eventid iyear imonth iday approxdate extended resolution \
          0
                  197000000001
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          1
                  1970000000002
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                                        Mexico
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                                   Philippines
          2
                      160
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                       78
                                        Greece
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```

181688 181689 181690	160 92 160	Philippines India Philippines	5 6 5	NaN NaN NaN	
0 1 2 3 4 181686 181687 181688	"Putin's 'v	l-Shabaab Militants ictory' in Syria has o clashes trap tribe	turned into	a	\
181689 181690	"Trader esca	apes grenade attack ightened in Cotabato	in Imphal,"	Bus	
0 1 2 3 4				scite2 NaN NaN NaN NaN NaN	\
181686 181687 181688 181689 181690	"Two Russian	: Somalia Daily Medi n soldiers killed at ightened in Cotabato	: Hmeymim bas	e i NaN NaN	
0 1 2 3 4				scite3 NaN NaN NaN NaN NaN	\
181686 181687 181688 181689 181690		: Somalia Daily Medi n servicemen killed			

			dbsource	INT_LOG	INT_IDEO	<pre>INT_MISC</pre>	INT_ANY	related
0			PGIS	0	0	0	0	NaN
1			PGIS	0	1	1	1	NaN
2			PGIS	-9	-9	1	1	NaN
3			PGIS	-9	-9	1	1	NaN
4			PGIS	-9	-9	1	1	NaN
• • •			• • •	• • •				
181686	START	Primary	Collection	0	0	0	0	NaN
181687	START	Primary	Collection	-9	-9	1	1	NaN
181688	START	Primary	Collection	0	0	0	0	NaN
181689	START	Primary	Collection	-9	-9	0	-9	NaN
181690	START	Primary	Collection	-9	-9	0	-9	NaN

[181691 rows x 135 columns]

In [59]: df.head()

Out[59]:

	eventid	iyear	imonth	iday	approxdate	extended	resolution	country	country_txt	region	 addnotes	scite1	scite2	scite3	dbsource
0	197000000001	1970	7	2	NaN	0	NaN	58	Dominican Republic	2	 NaN	NaN	NaN	NaN	PGIS
1	197000000002	1970	0	0	NaN	0	NaN	130	Mexico	1	 NaN	NaN	NaN	NaN	PGIS
2	197001000001	1970	1	0	NaN	0	NaN	160	Philippines	5	 NaN	NaN	NaN	NaN	PGIS
3	197001000002	1970	1	0	NaN	0	NaN	78	Greece	8	 NaN	NaN	NaN	NaN	PGIS
4	197001000003	1970	1	0	NaN	0	NaN	101	Japan	4	 NaN	NaN	NaN	NaN	PGIS

5 rows × 135 columns

_

In [60]: df.rename(columns={'iyear':'Year','imonth':'Month','city':'City','iday':'Day','country_txt':'Country','region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Region_txt':'Regio

Out[60]:

	Year	Month	Day	Country	Region	City	latitude	longitude	AttackType	Killed	Wounded	Casualities	Target	G
0	1970	7	2	Dominican Republic	Central America & Caribbean	Santo Domingo	18.456792	-69.951164	Assassination	1.0	0.0	1.0	Julio Guzman	MAI
1	1970	0	0	Mexico	North America	Mexico city	19.371887	-99.086624	Hostage Taking (Kidnapping)	0.0	0.0	0.0	Nadine Chaval, daughter	2: Septe Comn Le
2	1970	1	0	Philippines	Southeast Asia	Unknown	15.478598	120.599741	Assassination	1.0	0.0	1.0	Employee	Unk
3	1970	1	0	Greece	Western Europe	Athens	37.997490	23.762728	Bombing/Explosion	NaN	NaN	NaN	U.S. Embassy	Unk
4	1970	1	0	Japan	East Asia	Fukouka	33.580412	130.396361	Facility/Infrastructure Attack	NaN	NaN	NaN	U.S. Consulate	Unk
4														>

```
In [61]: df.isnull().sum()
Out[61]: Year
                             0
          Month
                             0
          Day
                             0
          Country
          Region
                             0
         Citv
                           434
         latitude
                          4556
         longitude
                          4557
         AttackType
                             0
         Killed
                         10313
          Wounded
                         16311
         Casualities
                         16874
         Target
                           636
         Group
                             0
         Target type
         Weapon type
          dtype: int64
```

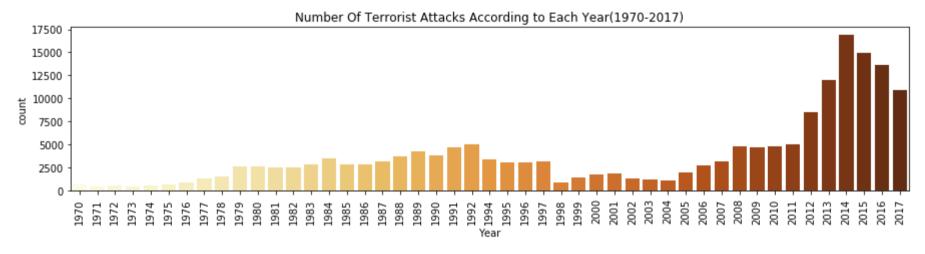
Most Attack-prone regions of the world:

The highest terrorist attacks were committed in Iraq with 24636 attacks Top 4 countries with highest terrorist attacks following iraq are:

- 2. Pakistan with 14368 attacks
- 3. Afghanistan with 12731 attacks
- 4. India with 11960 attacks
- 5. Colombia with 8306 attacks

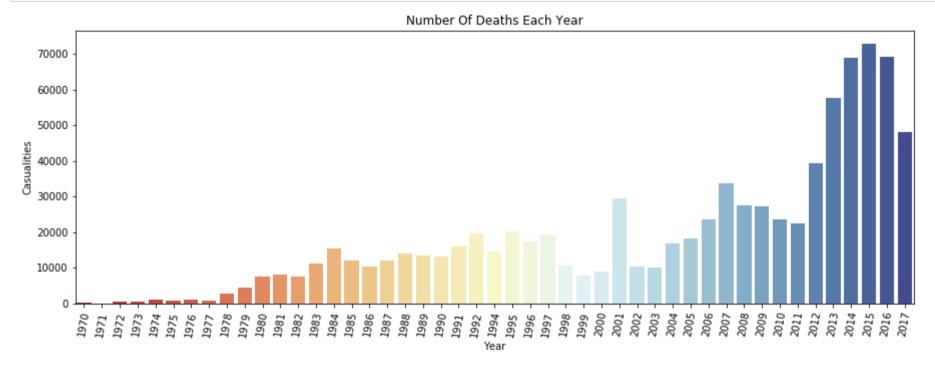
Timeline of Terror attacks, over the past 45 years:

```
In [63]: plt.subplots(figsize=(15,3))
    sns.countplot('Year',data=df,palette='YlOrBr')
    plt.xticks(rotation=90)
    plt.title('Number Of Terrorist Attacks According to Each Year(1970-2017)')
    plt.show()
```



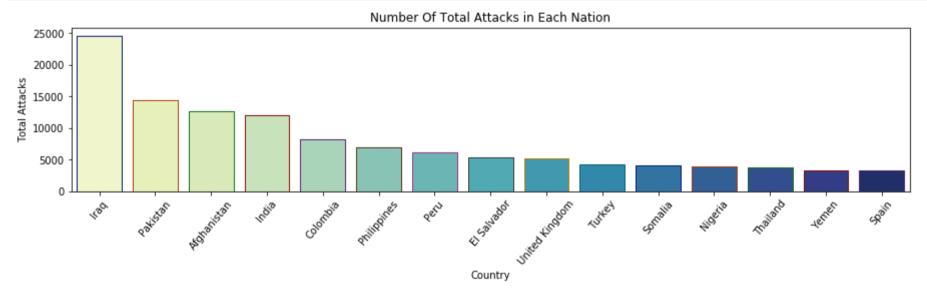
Timeline of Death:

```
In [64]: plt.subplots(figsize=(15,5))
    year_cas = df.groupby('Year').Casualities.sum().to_frame().reset_index()
    year_cas.columns = ['Year','Casualities']
    sns.barplot(x=year_cas.Year, y=year_cas.Casualities, palette='RdY1Bu')
    plt.xticks(rotation=80)
    plt.title('Number Of Deaths Each Year')
    plt.show()
```



Nation-wise total attacks:

```
In [65]: plt.subplots(figsize=(15,3))
    country_attacks = df.Country.value_counts()[:15].reset_index()
    country_attacks.columns = ['Country', 'Total Attacks']
    sns.barplot(x=country_attacks.Country, y=country_attacks['Total Attacks'], palette= 'YlGnBu',edgecolor=sns.color_palette
    plt.xticks(rotation=50)
    plt.title('Number Of Total Attacks in Each Nation')
    plt.show()
```



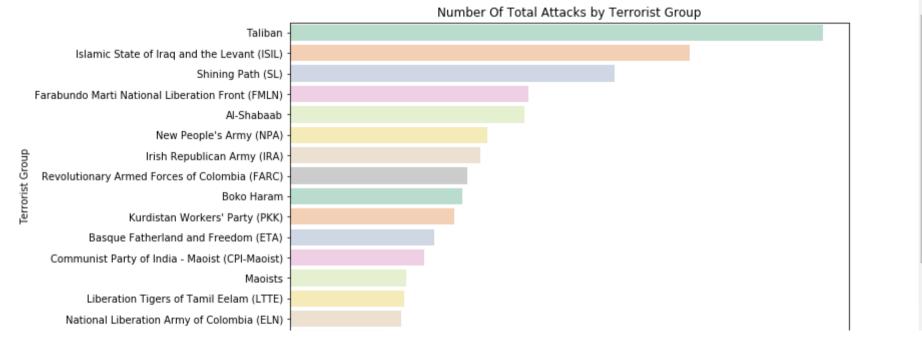
Terrain wise attack:

```
In [66]: region_attacks = df.Region.value_counts().to_frame().reset_index()
    region_attacks.columns = ['Region', 'Total Attacks']
    plt.subplots(figsize=(15,8))
    sns.barplot(x=region_attacks.Region, y=region_attacks['Total Attacks'], palette='Accent', edgecolor=sns.color_palette('daplt.xticks(rotation=46))
    plt.title('Number Of Total Attacks in Each Terrrain')
    plt.show()
```

```
In [67]: group_attacks = df.Group.value_counts().to_frame().drop('Unknown').reset_index()[:16]
    group_attacks.columns = ['Terrorist Organizations', 'Total Attacks']
    group_attacks
```

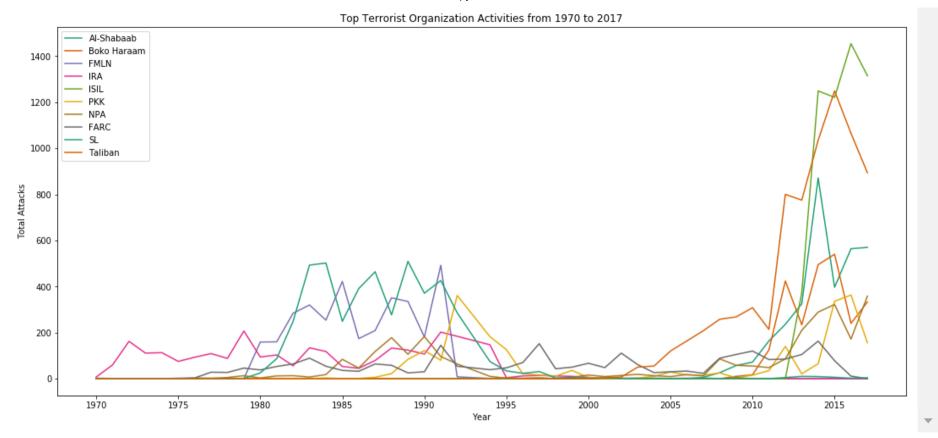
Out[67]:

	Terrorist Organizations	Total Attacks
0	Taliban	7478
1	Islamic State of Iraq and the Levant (ISIL)	5613
2	Shining Path (SL)	4555
3	Farabundo Marti National Liberation Front (FMLN)	3351
4	Al-Shabaab	3288
5	New People's Army (NPA)	2772
6	Irish Republican Army (IRA)	2671
7	Revolutionary Armed Forces of Colombia (FARC)	2487
8	Boko Haram	2418
9	Kurdistan Workers' Party (PKK)	2310
10	Basque Fatherland and Freedom (ETA)	2024
11	Communist Party of India - Maoist (CPI-Maoist)	1878
12	Maoists	1630
13	Liberation Tigers of Tamil Eelam (LTTE)	1606
14	National Liberation Army of Colombia (ELN)	1561
15	Tehrik-i-Taliban Pakistan (TTP)	1351



Terrorist Organizations Attack Trend

```
In [69]: groups_10 = df[df.Group.isin(df.Group.value_counts()[1:11].index)]
         pd.crosstab(groups_10.Year, groups_10.Group).plot(color=sns.color_palette('Dark2', 10))
         fig=plt.gcf()
         fig.set_size_inches(18,8)
         plt.xticks(range(1970, 2017, 5))
         plt.ylabel('Total Attacks')
         plt.title('Top Terrorist Organization Activities from 1970 to 2017')
         plt.legend(labels=['Al-Shabaab',
                             'Boko Haraam',
                             'FMLN',
                             'IRA',
                             'ISIL',
                             'PKK',
                             'NPA',
                             'FARC',
                             'SL',
                             'Taliban'], loc='upper left')
         plt.show()
```



In []: