

Lets Grow more Data Science Internship Intermediate Level Task-2

Author : Sujata Gaikwad

```
In [1]: # import Libraries
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [2]: # Loaded dataset
df=pd.read_csv(r"C:\Users\admin\Desktop\globalterrorism.csv",encoding="latin-1")
df.sample(5)

C:\Users\admin\AppData\Local\Temp\ipykernel_12648\2061752522.py:1: 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.
df=pd.read_csv(r"C:\Users\admin\Desktop\globalterrorism.csv",encoding="latin-1")
```

```
Out[2]:
```

	eventid	iyear	imonth	iday	approxdate	extended	resolution	country	country_1
--	---------	-------	--------	------	------------	----------	------------	---------	-----------

115115	201303120011	2013	3	12	NaN	0	NaN	78	Gree
--------	--------------	------	---	----	-----	---	-----	----	------

82237	200701250004	2007	1	25	NaN	0	NaN	95	Ir
-------	--------------	------	---	----	-----	---	-----	----	----

27672	198603260054	1986	3	26	NaN	0	NaN	43	Ch
-------	--------------	------	---	----	-----	---	-----	----	----

124236	201312030041	2013	12	3	NaN	0	NaN	95	Ir
--------	--------------	------	----	---	-----	---	-----	----	----

88923	200809280013	2008	9	28	NaN	0	NaN	4	Afghanist
-------	--------------	------	---	----	-----	---	-----	---	-----------

5 rows × 135 columns

```
In [3]: df.rename(columns={'iyear':'Year','imonth':'Month','iday':'Day','country_txt':
                        'region_txt','attacktype1_txt':'AttackType','t
                        'nwound':'Wounded','summary':'Summary','gname':'Group',
                        'weaptype1_txt':'Weapon_type','motive':'Motive'},inplace=True)
```

```
In [4]: # I'm just taking important data from whole dataset, which I will use in further
df=df[['Year','Month','Day','Country','state','Region','city','latitude','longitude',
      'Wounded','Target','Summary','Group','Target_type','Weapon_type']]
```

```
In [5]: df.head()
```

```
Out[5]:
```

	Year	Month	Day	Country	state	Region	city	latitude	longitude	AttackType
0	1970	7	2	Dominican Republic	NaN	Central America & Caribbean	Santo Domingo	18.456792	-69.951164	Assassination
1	1970	0	0	Mexico	Federal	North America	Mexico city	19.371887	-99.086624	Hostage taking
2	1970	1	0	Philippines	Tarlac	Southeast Asia	Unknown	15.478598	120.599741	Assassination
3	1970	1	0	Greece	Attica	Western Europe	Athens	37.997490	23.762728	Bombing
4	1970	1	0	Japan	Fukouka	East Asia	Fukouka	33.580412	130.396361	Facility/Installation

```
In [6]: df.isnull().sum()
```

```
Out[6]: Year          0
Month          0
Day            0
Country        0
state         421
Region         0
city          434
latitude       4556
longitude      4557
AttackType     0
Killed        10313
Wounded       16311
Target         636
Summary       66129
Group          0
Target_type    0
Weapon_type    0
Motive       131130
dtype: int64
```

- there are null values in data

```
In [7]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 181691 entries, 0 to 181690
Data columns (total 18 columns):
#   Column          Non-Null Count  Dtype  
---  -
0   Year            181691 non-null int64  
1   Month           181691 non-null int64  
2   Day             181691 non-null int64  
3   Country         181691 non-null object 
4   state           181270 non-null object 
5   Region          181691 non-null object 
6   city            181257 non-null object 
7   latitude        177135 non-null float64 
8   longitude       177134 non-null float64 
9   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
```

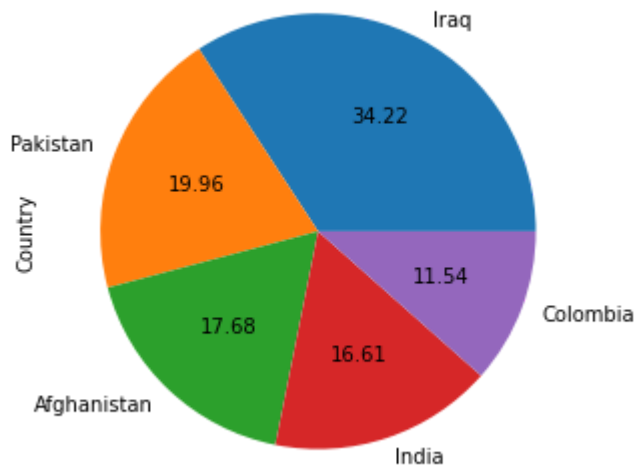
Top 5 Countries affected by terror attack

```
In [8]: print("Top 5 Countries with most attacks:",df['Country'].value_counts())
```

```
Top 5 Countries with most attacks: Iraq                24636
Pakistan                14368
Afghanistan             12731
India                   11960
Colombia                 8306
...
International              1
Wallis and Futuna          1
South Vietnam              1
Andorra                   1
Antigua and Barbuda        1
Name: Country, Length: 205, dtype: int64
```

```
In [9]: plt.figure(figsize=(10,5))
df['Country'].value_counts().head(5).plot(kind='pie',autopct='%.2f')
```

Out[9]: <AxesSubplot:ylabel='Country'>



- Most of the times Iraq has been attacked by terrorists

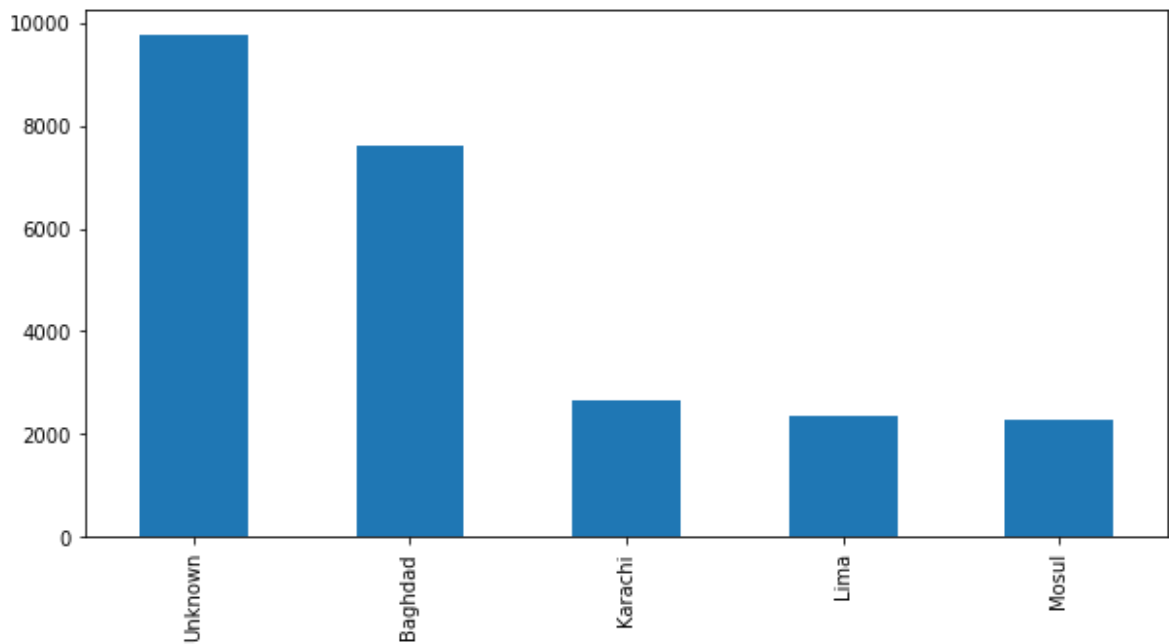
Top 5 Cities with most attack

```
In [10]: print("Top 5 Cities with most attacks:",df['city'].value_counts().head(5))
```

```
Top 5 Cities with most attacks: Unknown      9775
Baghdad      7589
Karachi      2652
Lima         2359
Mosul        2265
Name: city, dtype: int64
```

```
In [11]: plt.figure(figsize=(10,5))
df['city'].value_counts().head(5).plot(kind='bar')
```

Out[11]: <AxesSubplot:>



- As here 1st city is unknown, so, city with most attacks is Baghdad

State With the Most attacks

```
In [12]: df['state'].value_counts()
```

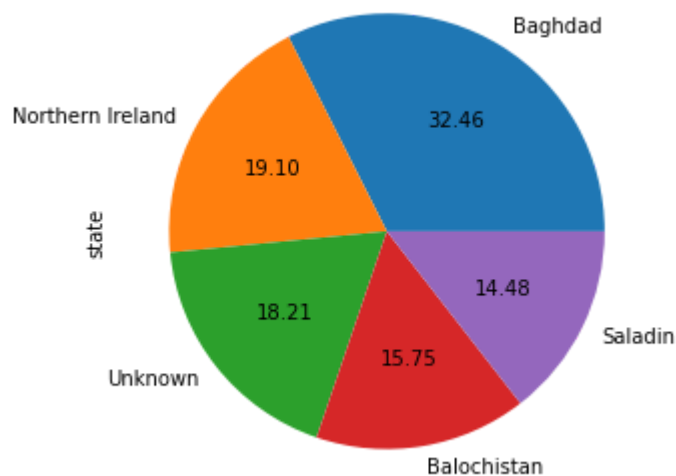
Out[12]:

Baghdad	7645
Northern Ireland	4498
Unknown	4290
Balochistan	3710
Saladin	3411
...	
Federal Territory	1
Topaz (Province)	1
Obilic (Municipality)	1
East Timor (Region)	1
Vidzeme	1

Name: state, Length: 2855, dtype: int64

```
In [13]: plt.figure(figsize=(10,5))
df['state'].value_counts().head(5).plot(kind='pie', autopct='%.2f')
```

```
Out[13]: <AxesSubplot:ylabel='state'>
```



- Baghdad State has been Mostly Attacked by Terrorists.

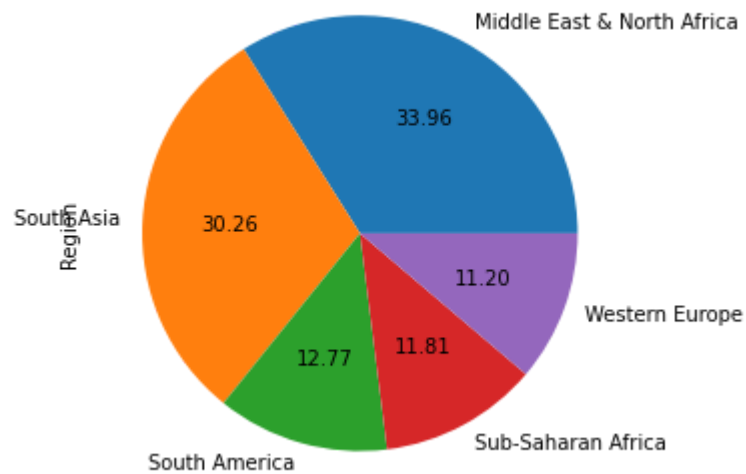
Top 5 Regions with most attack

```
In [14]: print("Top 5 Regions with most attacks:",df['Region'].value_counts().head(5))
```

[illegible]

```
In [15]: plt.figure(figsize=(10,5))  
df['Region'].value_counts().head(5).plot(kind='pie',autopct='%.2f')
```

```
Out[15]: <AxesSubplot:ylabel='Region'>
```



- Region with most attacks is Middle East & North Africa

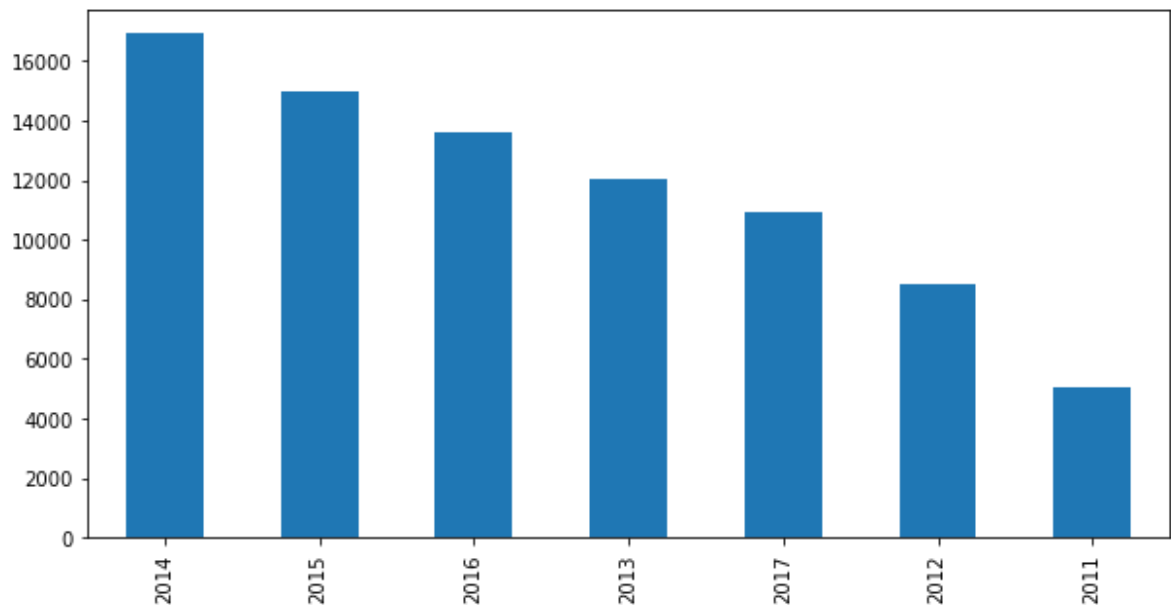
Top Year with most attacks


```
In [16]: print("Year with most attacks:",df['Year'].value_counts())
```

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

```
In [17]: plt.figure(figsize=(10,5))
df['Year'].value_counts().head(7).plot(kind='bar')
```

Out[17]: <AxesSubplot:>



- 2014 is the year in which most of the terror attacks happened.

Month with most attack

```
In [18]: print("Month with most attacks:",df['Month'].value_counts())
```

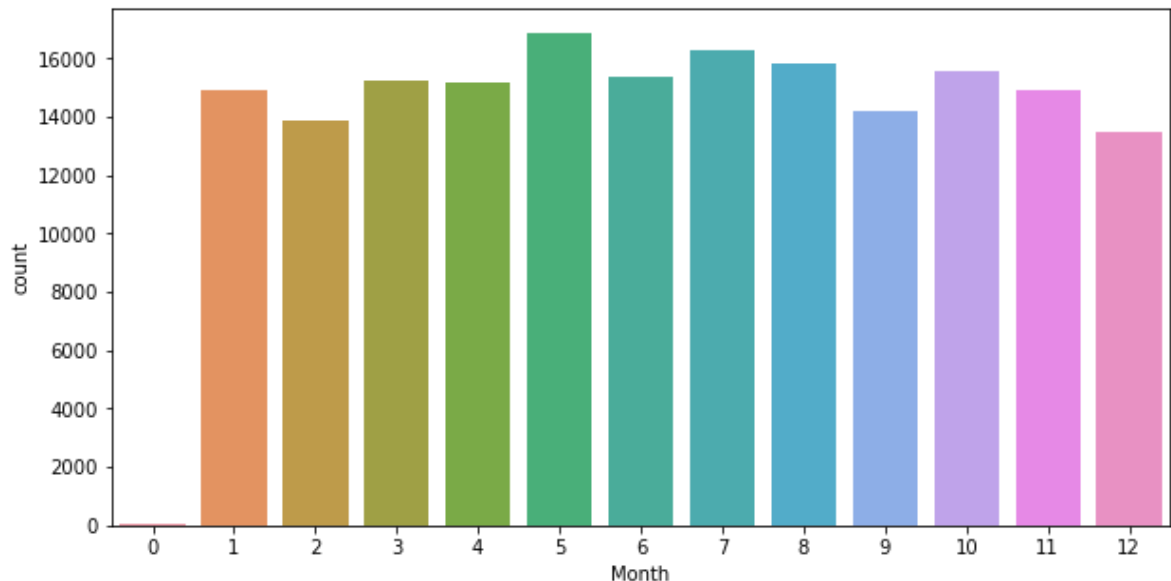
```
Month with most attacks: 5      16875
7       16268
8       15800
10      15563
6       15359
3       15257
4       15152
1       14936
11      14906
9       14180
2       13879
12      13496
0         20
Name: Month, dtype: int64
```

```
In [19]: plt.figure(figsize=(10,5))
sns.countplot(df['Month'])
```

C:\ProgramData\Anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

```
Out[19]: <AxesSubplot:xlabel='Month', ylabel='count'>
```



- In Month of 5 or May Most of the terror attack has been done.

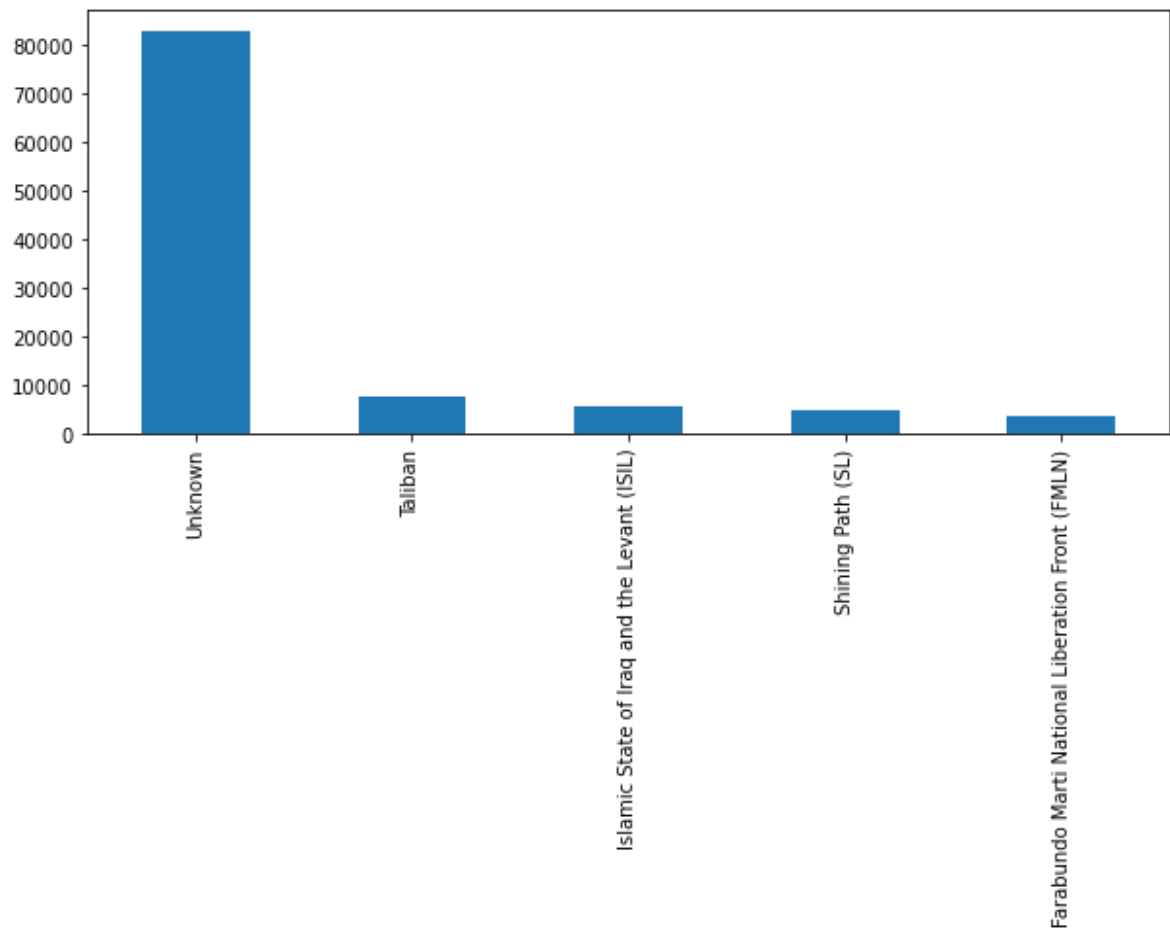
Groups which attacks most

```
In [20]: print("Top 5 Groups which attacks most:",df['Group'].value_counts())
```

```
Top 5 Groups which attacks most: Unknown
82782
Taliban                                7478
Islamic State of Iraq and the Levant (ISIL)  5613
Shining Path (SL)                        4555
Farabundo Marti National Liberation Front (FMLN)  3351
...
Ansar Sarallah                          1
Sword of Islam                          1
Support of Ocalan-The Hawks of Thrace   1
Arab Revolutionary Front                 1
MANO-D                                  1
Name: Group, Length: 3537, dtype: int64
```

```
In [21]: plt.figure(figsize=(10,4))
df['Group'].value_counts().head(5).plot(kind='bar')
```

Out[21]: <AxesSubplot:>



- as 1st group is unknown so, Taliban attacks or do more terror.

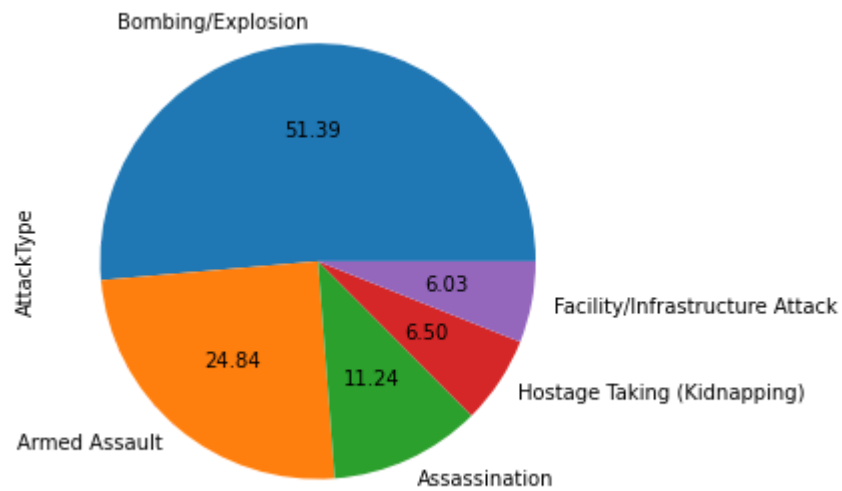
Most of terrorism done by using which AttackType

```
In [22]: print("Top 5 attack types:",df['AttackType'].value_counts())
```

Top 5 attack types: Bombing/Explosion	88255
Armed Assault	42669
Assassination	19312
Hostage Taking (Kidnapping)	11158
Facility/Infrastructure Attack	10356
Unknown	7276
Unarmed Assault	1015
Hostage Taking (Barricade Incident)	991
Hijacking	659
Name: AttackType, dtype: int64	

```
In [23]: plt.figure(figsize=(10,5))
df['AttackType'].value_counts().head(5).plot(kind='pie',autopct='%.2f')
```

Out[23]: <AxesSubplot:ylabel='AttackType'>



- Most of the time attack has been done with Bombing/Explosion

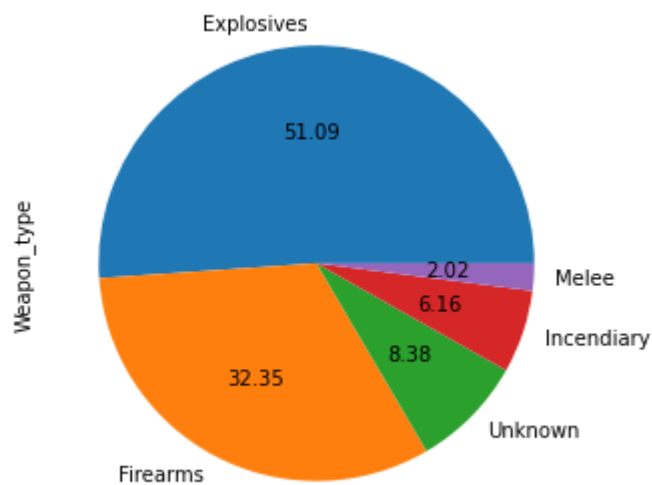
Weapons used for attack

```
In [24]: print("Top 5 attack types:",df['Weapon_type'].value_counts())
```

```
Top 5 attack types: Explosives
92426
Firearms
58524
Unknown
15157
Incendiary
11135
Melee
3655
Chemical
321
Sabotage Equipment
141
Vehicle (not to include vehicle-borne explosives, i.e., car or truck bombs)
136
Other
114
Biological
35
Fake Weapons
33
Radiological
14
Name: Weapon_type, dtype: int64
```

```
In [25]: plt.figure(figsize=(10,5))
df['Weapon_type'].value_counts().head(5).plot(kind='pie',autopct='%.2f')
```

```
Out[25]: <AxesSubplot:ylabel='Weapon_type'>
```



- Explosives are mostly used by terrorist to carryout attacks

```
In [26]: df['Wounded'] = df['Wounded'].fillna(0).astype(int)
df['Killed'] = df['Killed'].fillna(0).astype(int)
df['casualties'] = df['Killed'] + df['Wounded']
```

```
In [27]: df1 = df.sort_values(by='casualties',ascending=False)[:40]
```

```
In [28]: df1.head()
```

Out[28]:

	Year	Month	Day	Country	state	Region	city	latitude	longitude	Att:
73126	2001	9	11	United States	New York	North America	New York City	40.697132	-73.931351	+
73127	2001	9	11	United States	New York	North America	New York City	40.697132	-73.931351	+
58841	1995	3	20	Japan	Tokyo	East Asia	Tokyo	35.689125	139.747742	Unarmec
68071	1998	8	7	Kenya	Nairobi	Sub-Saharan Africa	Nairobi	-1.285180	36.821107	Bombing/E
133518	2014	6	12	Iraq	Saladin	Middle East & North Africa	Tikrit	34.621521	43.668377	Hostag (Kidn

- as we can see in above dataframe most casualties done in US

```
In [29]: heat=df1.pivot_table(index='Country',columns='Year',values='casualties')
heat.fillna(0,inplace=True)
```

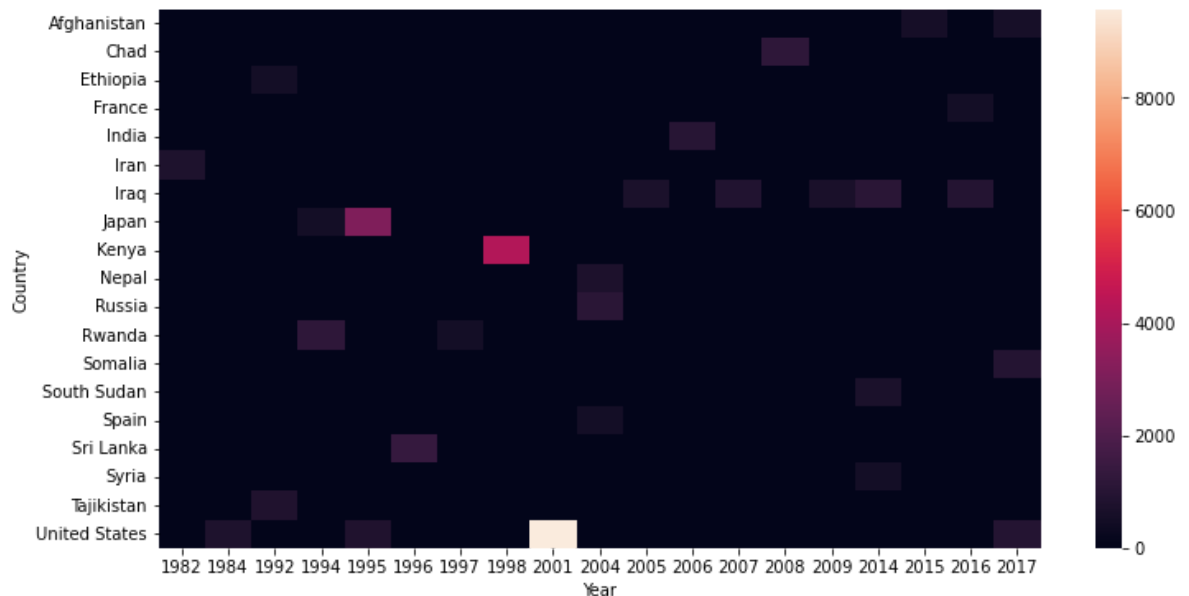
```
In [30]: heat.head()
```

```
Out[30]:
```

	Year	1982	1984	1992	1994	1995	1996	1997	1998	2001	2004	2005	2006	2007
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
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
Ethiopia		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
France		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
India		0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1005.0	0.0

```
In [39]: plt.figure(figsize=(12,6))
sns.heatmap(heat)
```

```
Out[39]: <AxesSubplot:xlabel='Year', ylabel='Country'>
```



- Most Casualties happened in US in 2001

```
In [31]: #Number of Killed in Terrorist Attacks by Countries
countryData = df.loc[:, 'Country']
killData = df.loc[:, 'Killed']
# countyData
countryKillData = pd.concat([countryData, killData], axis=1)
```



```
In [32]: countryKillData.head()
```

```
Out[32]:
```

	Country	Killed
0	Dominican Republic	1
1	Mexico	0
2	Philippines	1
3	Greece	0
4	Japan	0

```
In [33]: countryKill = countryKillData.pivot_table(columns='Country', values='Killed',  
countryKill
```

```
Out[33]:
```

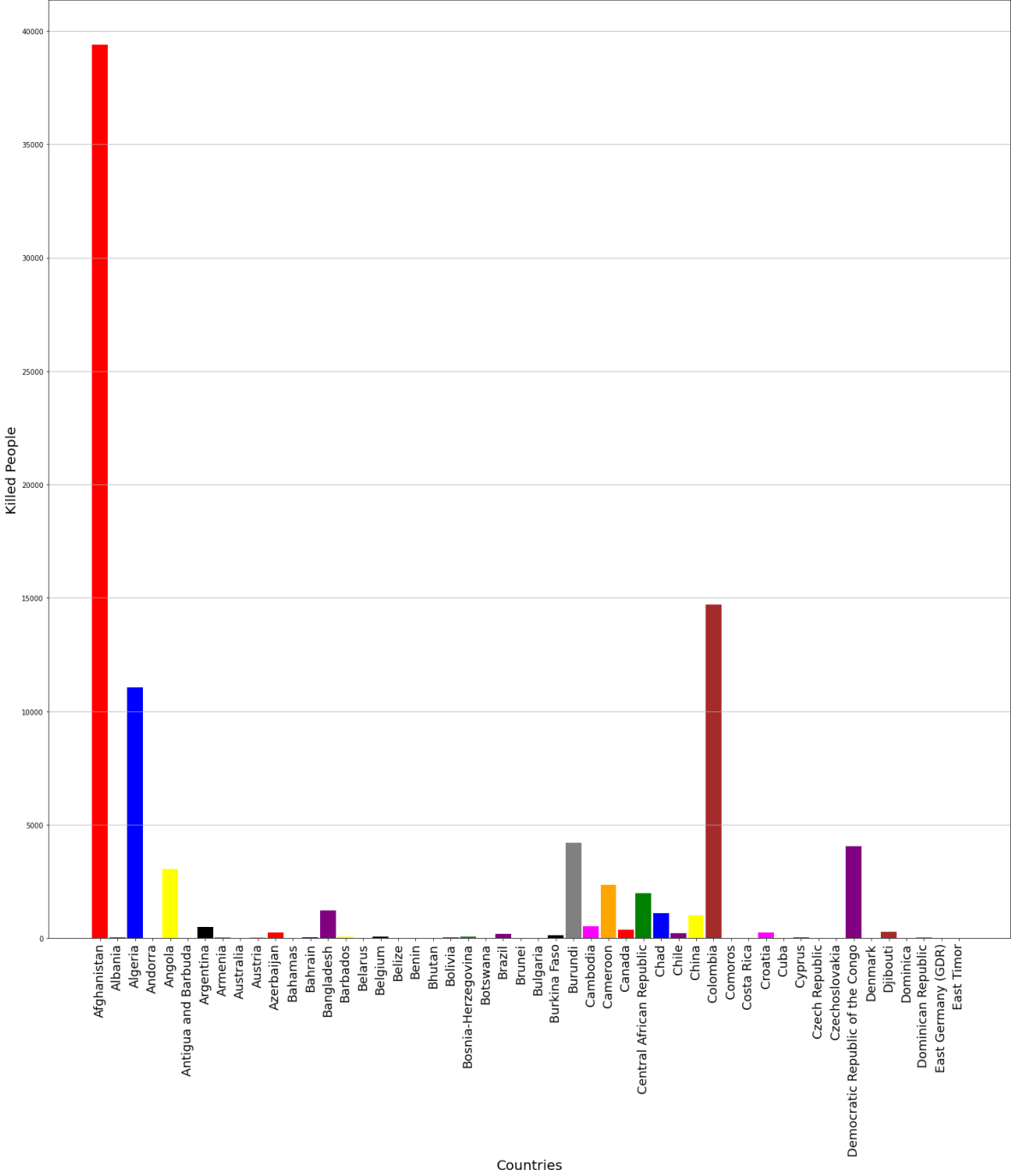
Country	Afghanistan	Albania	Algeria	Andorra	Angola	Antigua and Barbuda	Argentina	Armenia	Australi
Killed	39384	42	11066	0	3043	0	490	37	2

1 rows × 205 columns

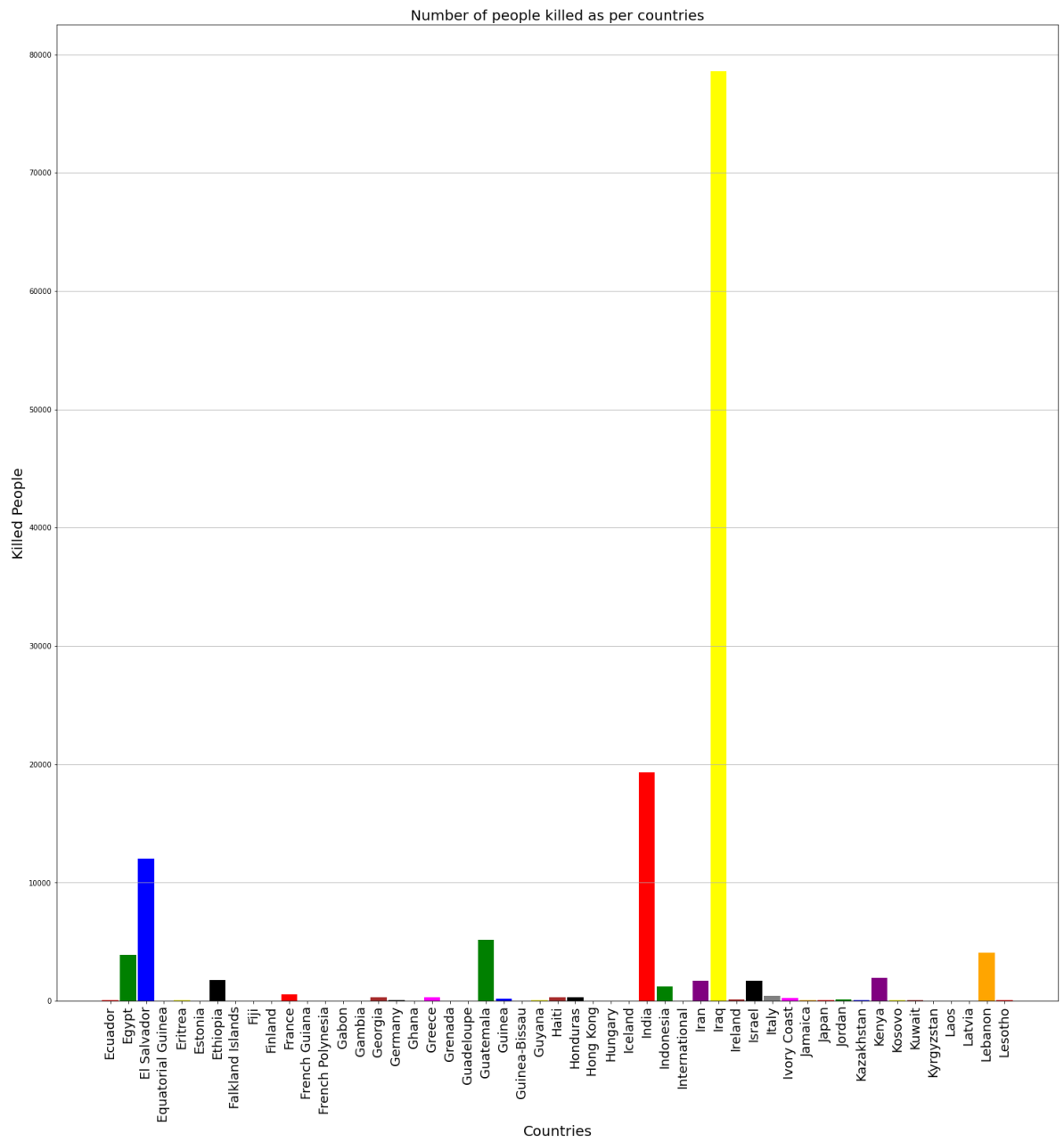
```
In [34]: fig_size = plt.rcParams["figure.figsize"]  
fig_size[0]=25  
fig_size[1]=25  
plt.rcParams["figure.figsize"] = fig_size
```

```
In [35]: labels = countryKill.columns.tolist()
labels = labels[:50] #50 bar provides nice view
index = np.arange(len(labels))
transpoze = countryKill.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']
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=18, rotation=90)
plt.title('Number of people killed as per countries', fontsize = 20)
# print(fig_size)
plt.show()
```

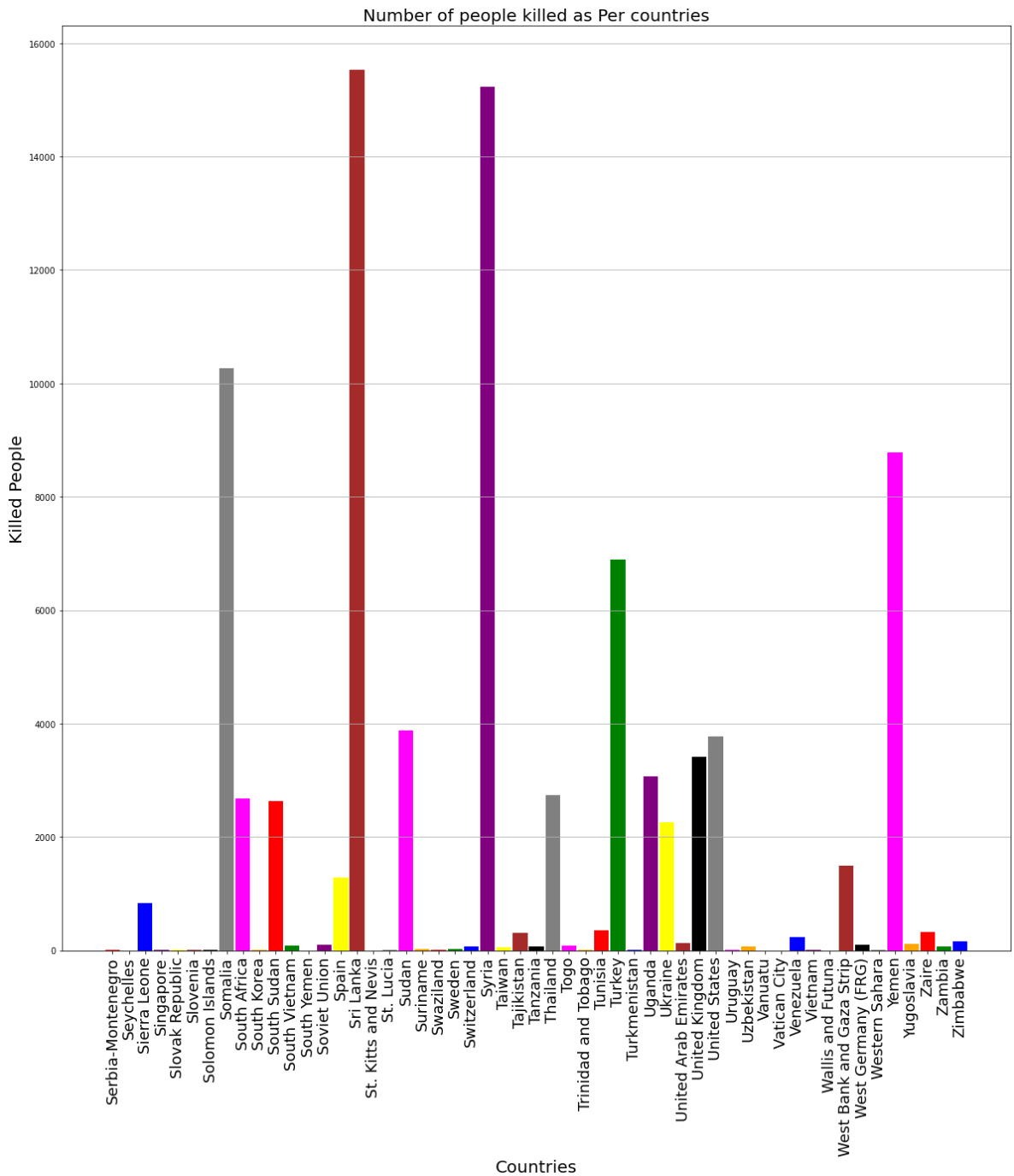
Number of people killed as per countries



```
In [36]: labels = countryKill.columns.tolist()
labels = labels[50:101]
index = np.arange(len(labels))
transpose = countryKill.T
values = transpose.values.tolist()
values = values[50:101]
values = [int(i[0]) for i in values]
colors = ['red', 'green', 'blue', 'purple', 'yellow', 'brown', 'black', 'gray']
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=18, rotation=90)
plt.title('Number of people killed as per countries', fontsize = 20)
plt.show()
```



```
In [37]: labels = countryKill.columns.tolist()
labels = labels[152:206]
index = np.arange(len(labels))
transpoze = countryKill.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']
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=18, rotation=90)
plt.title('Number of people killed as Per countries', fontsize = 20)
plt.show()
```



- Terrorist acts in the Middle East and northern Africa have been seen to have fatal consequences. The Middle East and North Africa are seen to be the places of serious terrorist attacks. In addition, even though there is a perception that Muslims are supporters of terrorism, Muslims are the people who are most damaged by terrorist attacks. If you look at the graphics, it appears that Iraq, Afghanistan and Pakistan are the most damaged countries. All of these countries are Muslim countries.

