

Exploratory Data Analysis - Terrorism

-Vaishnavi Parvathy N

Importing all the libraries required for analysis

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
```

Importing and observing the data

```
In [3]: df=pd.read_csv("globalterrorismdb_0718dist.csv", encoding="latin1")
df=pd.DataFrame(df)
print("Data has been successfully imported")
df.head()
```

Data has been successfully imported

```
Out[3]:
```

	eventid	year	month	day	approxdate	extended	resolution	country	country_txt	region	...	addnotes	scite1	scite2	scite3	dbsource	IN
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 [4]: `df.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 181691 entries, 0 to 181690
Columns: 135 entries, eventid to related
dtypes: float64(55), int64(22), object(58)
memory usage: 187.1+ MB
```

In [5]: `df.shape`

Out[5]: (181691, 135)

In [6]: `df.columns`

Out[6]: 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 [7]: `for i in df.columns:
 print(i,end=" ", " ")`

```
eventid, iyear, imonth, iday, approxdate, extended, resolution, country, country_txt, region, region_txt, provstate, city, latitud
e, longitude, specificity, vicinity, location, summary, crit1, crit2, crit3, doubtterr, alternative, alternative_txt, multiple, su
ccess, suicide, attacktype1, attacktype1_txt, attacktype2, attacktype2_txt, attacktype3, attacktype3_txt, targtype1, targtype1_tx
t, targsubtype1, targsubtype1_txt, corp1, target1, natlty1, natlty1_txt, targtype2, targtype2_txt, targsubtype2, targsubtype2_txt,
corp2, target2, natlty2, natlty2_txt, targtype3, targtype3_txt, targsubtype3, targsubtype3_txt, corp3, target3, natlty3, natlty3_t
xt, gname, gsubname, gname2, gsubname2, gname3, gsubname3, motive, guncertain1, guncertain2, guncertain3, individual, nperps, nper
pcap, claimed, claimmode, claimmode_txt, claim2, claimmode2, claimmode2_txt, claim3, claimmode3, claimmode3_txt, compclaim, weapty
pe1, weaptype1_txt, weapsubtype1, weapsubtype1_txt, weaptype2, weaptype2_txt, weapsubtype2, weapsubtype2_txt, weaptype3, weaptype3
_txt, weapsubtype3, weapsubtype3_txt, weaptype4, weaptype4_txt, weapsubtype4, weapsubtype4_txt, weapdetail, nkill, nkillus, nkillt
er, nbound, nwoundus, nwoundte, property, propextent, propextent_txt, propvalue, propcomment, ishostkid, nhostkid, nhostkidus, nho
urs, ndays, divert, kidhijcountry, ransom, ransomamt, ransomamtus, ransompaid, ransompaidus, ransomnote, hostkidoutcome, hostkidou
tcome_txt, nreleased, addnotes, scite1, scite2, scite3, dbsource, INT_LOG, INT_IDEO, INT_MISC, INT_ANY, related,
```

Proccessing the data

```
In [8]: df=df[["iyear","imonth","iday","country_txt","region_txt","provstate","city",
            "latitude","longitude","location","summary","attacktype1_txt","targtype1_txt",
            "gname","motive","weaptype1_txt","nkill","nwound","addnotes"]]
df.head()
```

Out[8]:

	iyear	imonth	iday	country_txt	region_txt	provstate	city	latitude	longitude	location	summary	attacktype1_txt	targtype1_txt	gr
0	1970	7	2	Dominican Republic	Central America & Caribbean	NaN	Santo Domingo	18.456792	-69.951164	NaN	NaN	Assassination	Private Citizens & Property	MA
1	1970	0	0	Mexico	North America	Federal	Mexico city	19.371887	-99.086624	NaN	NaN	Hostage Taking (Kidnapping)	Government (Diplomatic)	Septe Comrr Le
2	1970	1	0	Philippines	Southeast Asia	Tarlac	Unknown	15.478598	120.599741	NaN	NaN	Assassination	Journalists & Media	Unki
3	1970	1	0	Greece	Western Europe	Attica	Athens	37.997490	23.762728	NaN	NaN	Bombing/Explosion	Government (Diplomatic)	Unki
4	1970	1	0	Japan	East Asia	Fukouka	Fukouka	33.580412	130.396361	NaN	NaN	Facility/Infrastructure Attack	Government (Diplomatic)	Unki

```
In [9]: df.rename(columns={"iyear":"Year","imonth":"Month","iday":"Day","country_txt":"Country",
                            "region_txt":"Region","provstate":"Province/State","city":"City",
                            "latitude":"Latitude","longitude":"Longitude","location":"Location",
                            "summary":"Summary","attacktype1_txt":"Attack Type","targtype1_txt":"Target Type",
                            "gname":"Group Name","motive":"Motive","weaptype1_txt":"Weapon Type",
                            "nkill":"Killed","nwound":"Wounded","addnotes":"Add Notes"},inplace=True)
df.head()
```

Out[9]:

	Year	Month	Day	Country	Region	Province/State	City	Latitude	Longitude	Location	Summary	Attack Type	Target Type	
0	1970	7	2	Dominican Republic	Central America & Caribbean	NaN	Santo Domingo	18.456792	-69.951164	NaN	NaN	Assassination	Private Citizens & Property	MA

	Year	Month	Day	Country	Region	Province/State	City	Latitude	Longitude	Location	Summary	Attack Type	Target Type	
1	1970	0	0	Mexico	North America	Federal	Mexico city	19.371887	-99.086624	NaN	NaN	Hostage Taking (Kidnapping)	Government (Diplomatic)	Sept 2
2	1970	1	0	Philippines	Southeast Asia	Tarlac	Unknown	15.478598	120.599741	NaN	NaN	Assassination	Journalists & Media	Unk
3	1970	1	0	Greece	Western Europe	Attica	Athens	37.997490	23.762728	NaN	NaN	Bombing/Explosion	Government (Diplomatic)	Unk
4	1970	1	0	Japan	East Asia	Fukouka	Fukouka	33.580412	130.396361	NaN	NaN	Facility/Infrastructure Attack	Government (Diplomatic)	Unk

In [10]:

df.info()

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 181691 entries, 0 to 181690
Data columns (total 19 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   Region                181691 non-null  object  
5   Province/State        181270 non-null  object  
6   City                  181257 non-null  object  
7   Latitude              177135 non-null  float64 
8   Longitude             177134 non-null  float64 
9   Location              55495 non-null   object  
10  Summary               115562 non-null  object  
11  Attack Type           181691 non-null  object  
12  Target Type           181691 non-null  object  
13  Group Name            181691 non-null  object  
14  Motive                 50561 non-null   object  
15  Weapon Type           181691 non-null  object  
16  Killed                171378 non-null  float64 
17  Wounded               165380 non-null  float64 
18  Add Notes             28289 non-null   object  

```

dtypes: float64(4), int64(3), object(12)
memory usage: 26.3+ MB

In [11]: `df.shape`

Out[11]: (181691, 19)

In [12]: `df.isnull().sum()`

Out[12]:

Year	0
Month	0
Day	0
Country	0
Region	0
Province/State	421
City	434
Latitude	4556
Longitude	4557
Location	126196
Summary	66129
Attack Type	0
Target Type	0
Group Name	0
Motive	131130
Weapon Type	0
Killed	10313
Wounded	16311
Add Notes	153402

dtype: int64

In [13]:

```
df["Killed"] = df["Killed"].fillna(0)
df["Wounded"] = df["Wounded"].fillna(0)
df["Casualty"] = df["Killed"] + df["Wounded"]
df.describe()
```

Out[13]:

	Year	Month	Day	Latitude	Longitude	Killed	Wounded	Casualty
count	181691.000000	181691.000000	181691.000000	177135.000000	1.771340e+05	181691.000000	181691.000000	181691.000000
mean	2002.638997	6.467277	15.505644	23.498343	-4.586957e+02	2.266860	2.883296	5.150156
std	13.259430	3.388303	8.814045	18.569242	2.047790e+05	11.227057	34.309747	40.555416

	Year	Month	Day	Latitude	Longitude	Killed	Wounded	Casualty
min	1970.000000	0.000000	0.000000	-53.154613	-8.618590e+07	0.000000	0.000000	0.000000
25%	1991.000000	4.000000	8.000000	11.510046	4.545640e+00	0.000000	0.000000	0.000000
50%	2009.000000	6.000000	15.000000	31.467463	4.324651e+01	0.000000	0.000000	1.000000
75%	2014.000000	9.000000	23.000000	34.685087	6.871033e+01	2.000000	2.000000	4.000000
max	2017.000000	12.000000	31.000000	74.633553	1.793667e+02	1570.000000	8191.000000	9574.000000

Observation: 1.The data consists of terrorist activities ranging from the year: 1970 to 2017 2.Maximum number of people killed in an event were: 1570 3.Maximum number of people wounded in an event were: 8191 4.Maximum number of total casualties in an event were: 9574

Visualizing the data

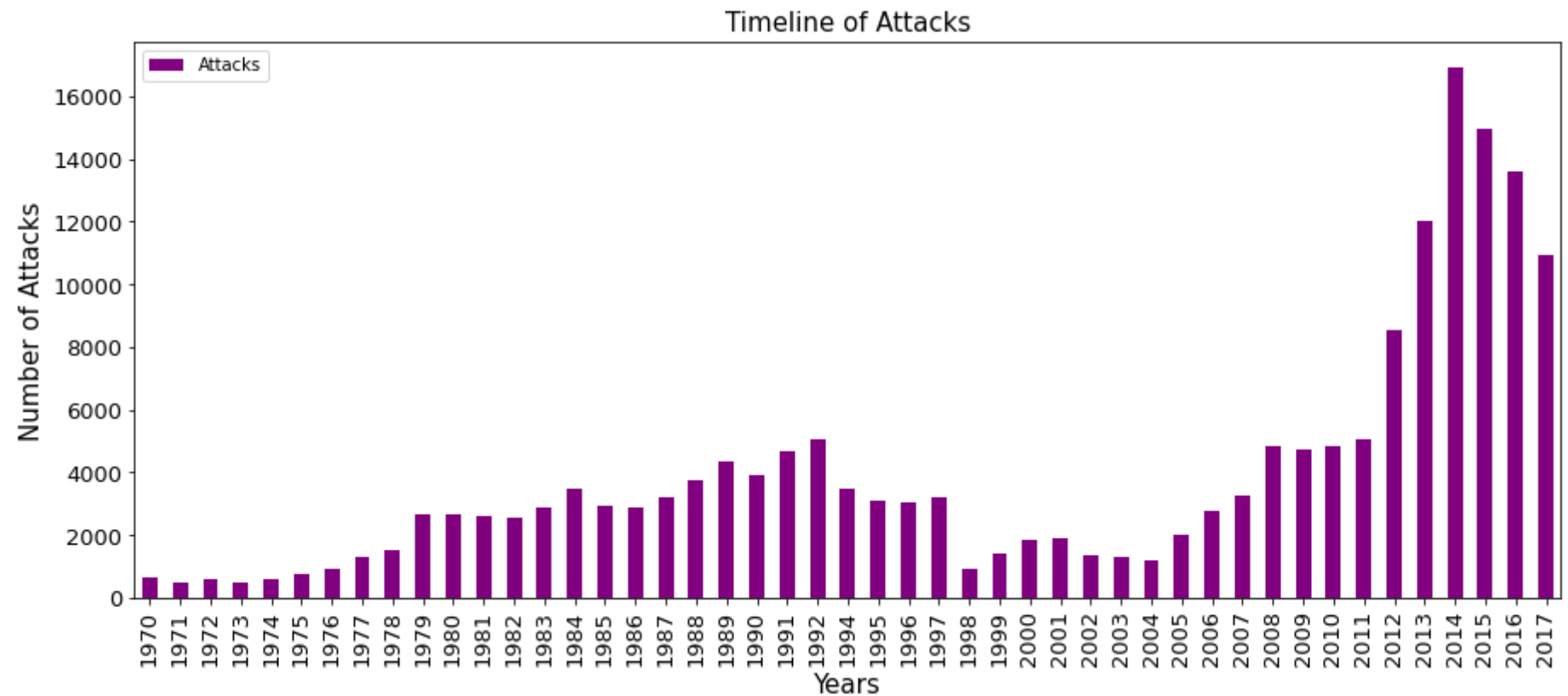
1. Year wise Attacks: Number of Attacks in each Year

```
In [14]: attacks=df["Year"].value_counts(dropna=False).sort_index().to_frame().reset_index().rename(columns={"index":"Year","Year":"Attacks"})
attacks.head()
```

Out[14]:

Attacks	
Year	
1970	651
1971	471
1972	568
1973	473
1974	581

```
In [15]: attacks.plot(kind="bar",color="purple",figsize=(15,6),fontsize=13)
plt.title("Timeline of Attacks",fontsize=15)
plt.xlabel("Years",fontsize=15)
plt.ylabel("Number of Attacks",fontsize=15)
plt.show()
```



(i). Most number of attacks(16903) in 2014 (ii). Least number of attacks(471) in 1971

Total Casualties (Killed + Wounded) in each Year

```
In [16]: yc=df[["Year","Casualty"]].groupby("Year").sum()
yc.head()
```

Out[16]: **Casualty**

Year	
1970	386.0
1971	255.0
1972	975.0

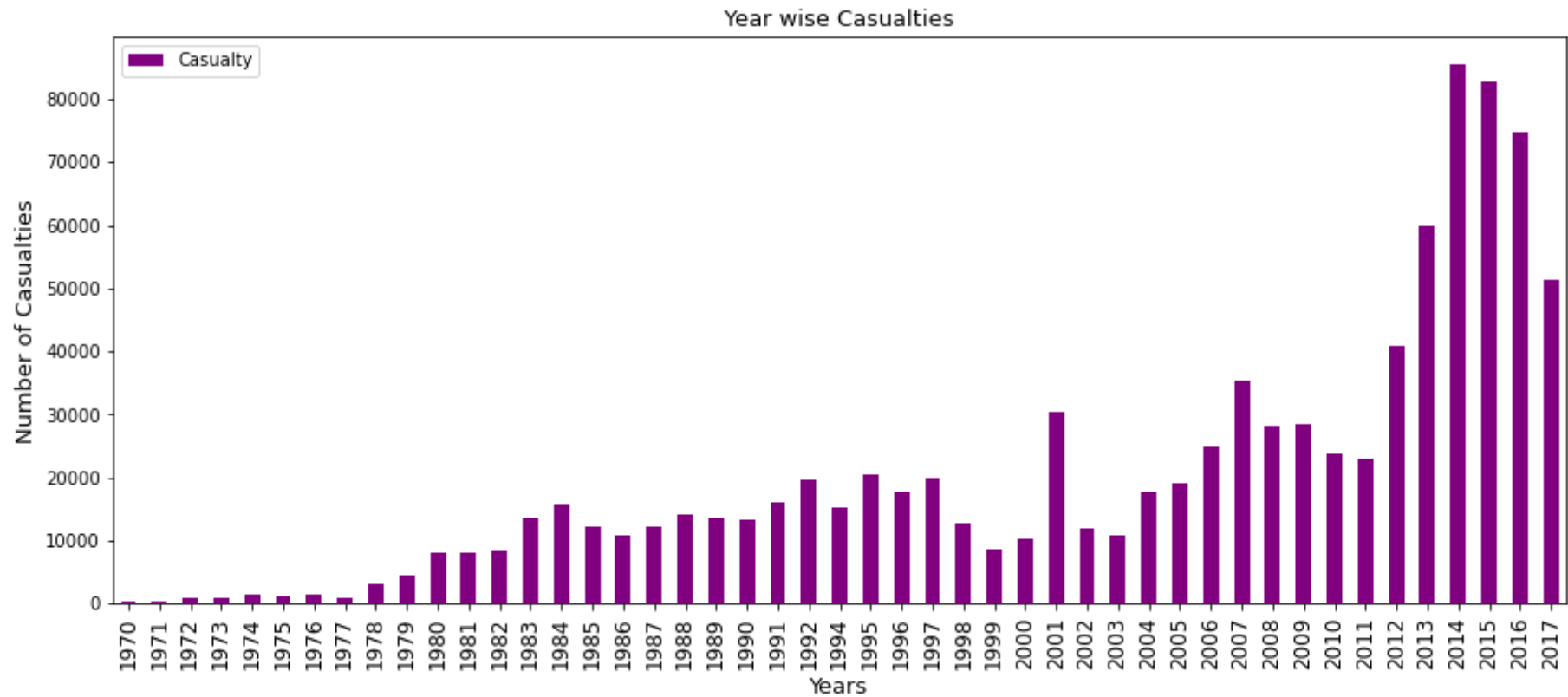
Casualty**Year****1973** 865.0**1974** 1404.0

In [17]:

```

yc.plot(kind="bar",color="purple",figsize=(15,6))
plt.title("Year wise Casualties",fontsize=13)
plt.xlabel("Years",fontsize=13)
plt.xticks(fontsize=12)
plt.ylabel("Number of Casualties",fontsize=13)
plt.show()

```



Killed in each Year


```
In [18]: yk=df[["Year","Killed"]].groupby("Year").sum()  
yk.head()
```

```
Out[18]:
```

	Killed
Year	
1970	174.0
1971	173.0
1972	566.0
1973	370.0
1974	539.0

Wounded in each Region

```
In [19]: yw=df[["Year","Wounded"]].groupby("Year").sum()  
yw.head()
```

```
Out[19]:
```

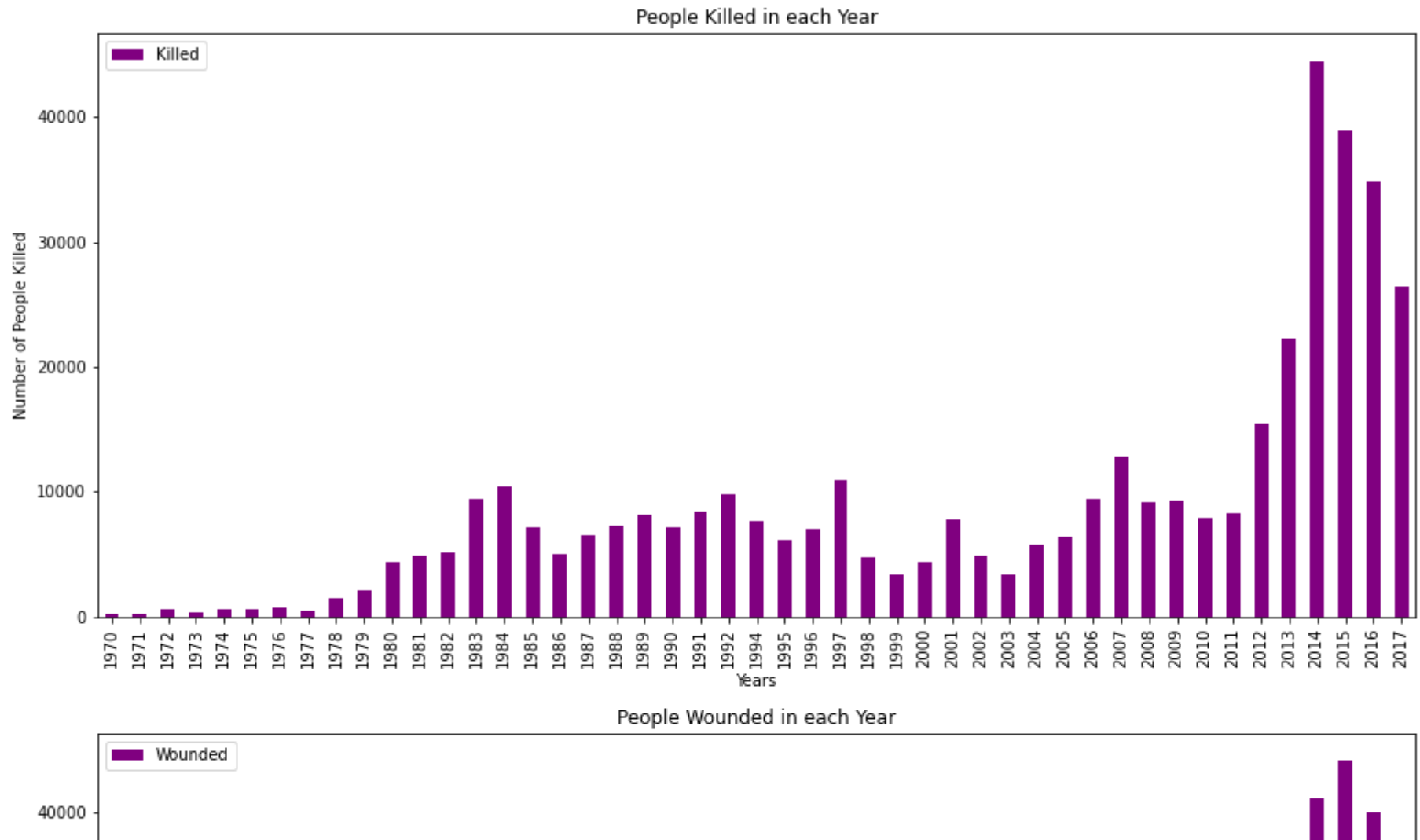
	Wounded
Year	
1970	212.0
1971	82.0
1972	409.0
1973	495.0
1974	865.0

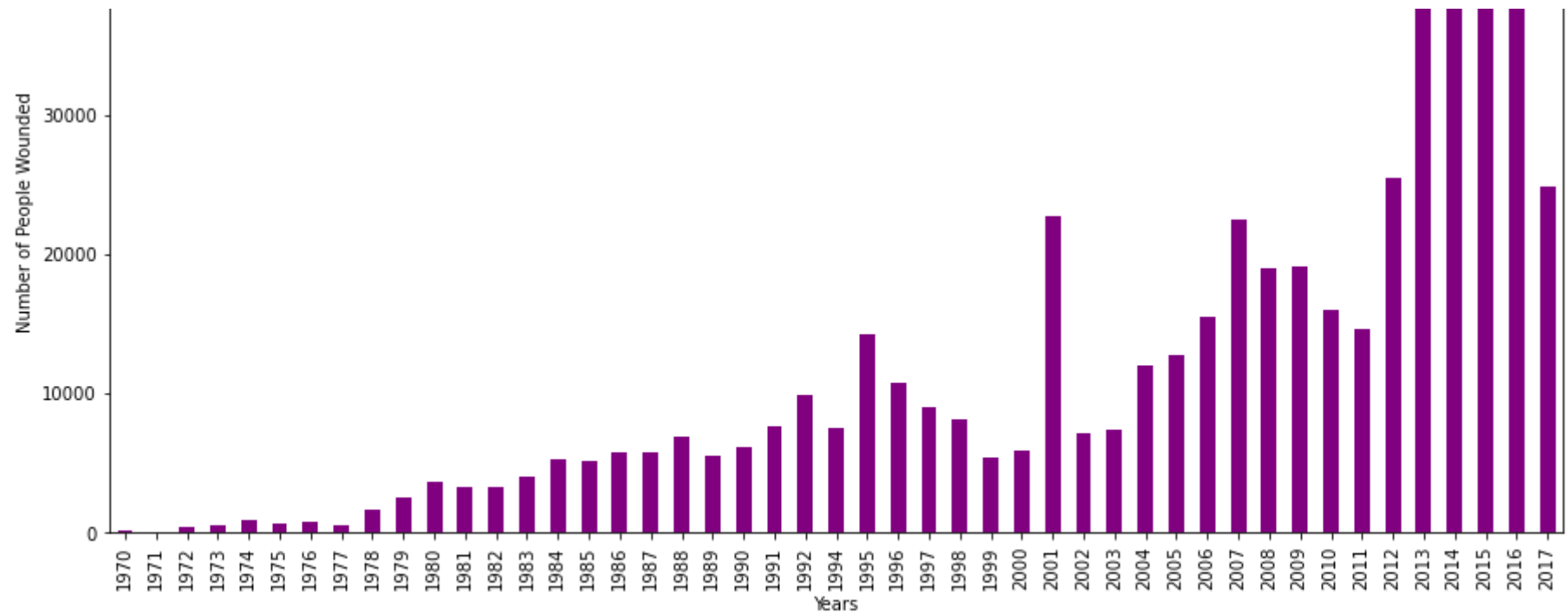
```
In [20]: fig=plt.figure()  
ax0=fig.add_subplot(2,1,1)  
ax1=fig.add_subplot(2,1,2)  
  
#Killed  
yk.plot(kind="bar",color="purple",figsize=(15,15),ax=ax0)  
ax0.set_title("People Killed in each Year")
```

```
ax0.set_xlabel("Years")
ax0.set_ylabel("Number of People Killed")

#Wounded
yw.plot(kind="bar", color="purple", figsize=(15,15), ax=ax1)
ax1.set_title("People Wounded in each Year")
ax1.set_xlabel("Years")
ax1.set_ylabel("Number of People Wounded")

plt.show()
```





1. Region wise Attacks Distribution of Terrorist Attacks over Regions from 1970-2017

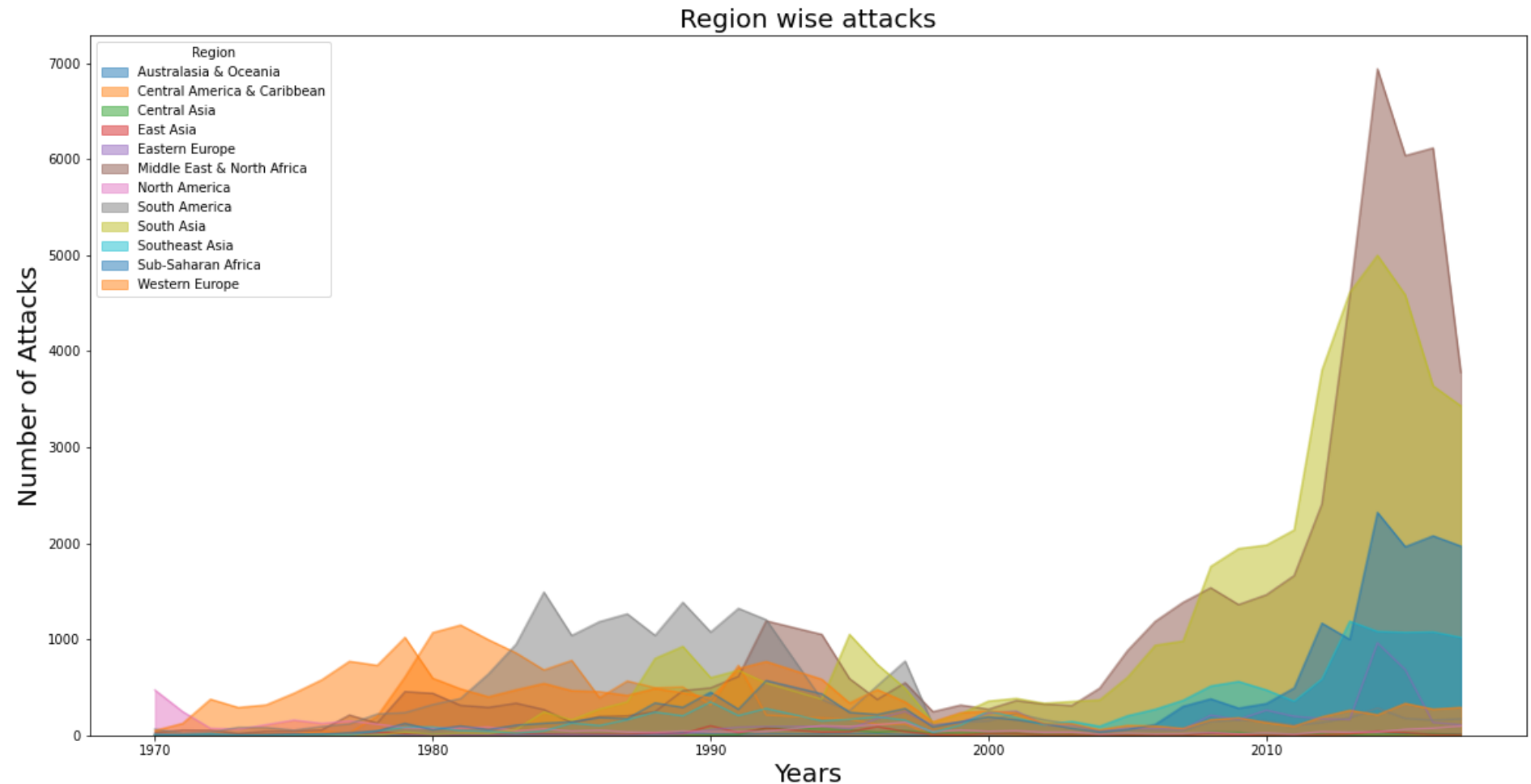
In [21]:

```
reg=pd.crosstab(df.Year,df.Region)
reg.head()
```

Out[21]:

Region	Australasia & Oceania	Central America & Caribbean	Central Asia	East Asia	Eastern Europe	Middle East & North Africa	North America	South America	South Asia	Southeast Asia	Sub-Saharan Africa	Western Europe
Year												
1970	1	7	0	2	12	28	472	65	1	10	3	50
1971	1	5	0	1	5	55	247	24	0	6	2	125
1972	8	3	0	0	1	53	73	33	1	16	4	376
1973	1	6	0	2	1	19	64	83	1	2	4	290
1974	1	11	0	4	2	42	111	81	2	3	7	317

```
In [22]: reg.plot(kind="area", stacked=False, alpha=0.5, figsize=(20,10))
plt.title("Region wise attacks", fontsize=20)
plt.xlabel("Years", fontsize=20)
plt.ylabel("Number of Attacks", fontsize=20)
plt.show()
```



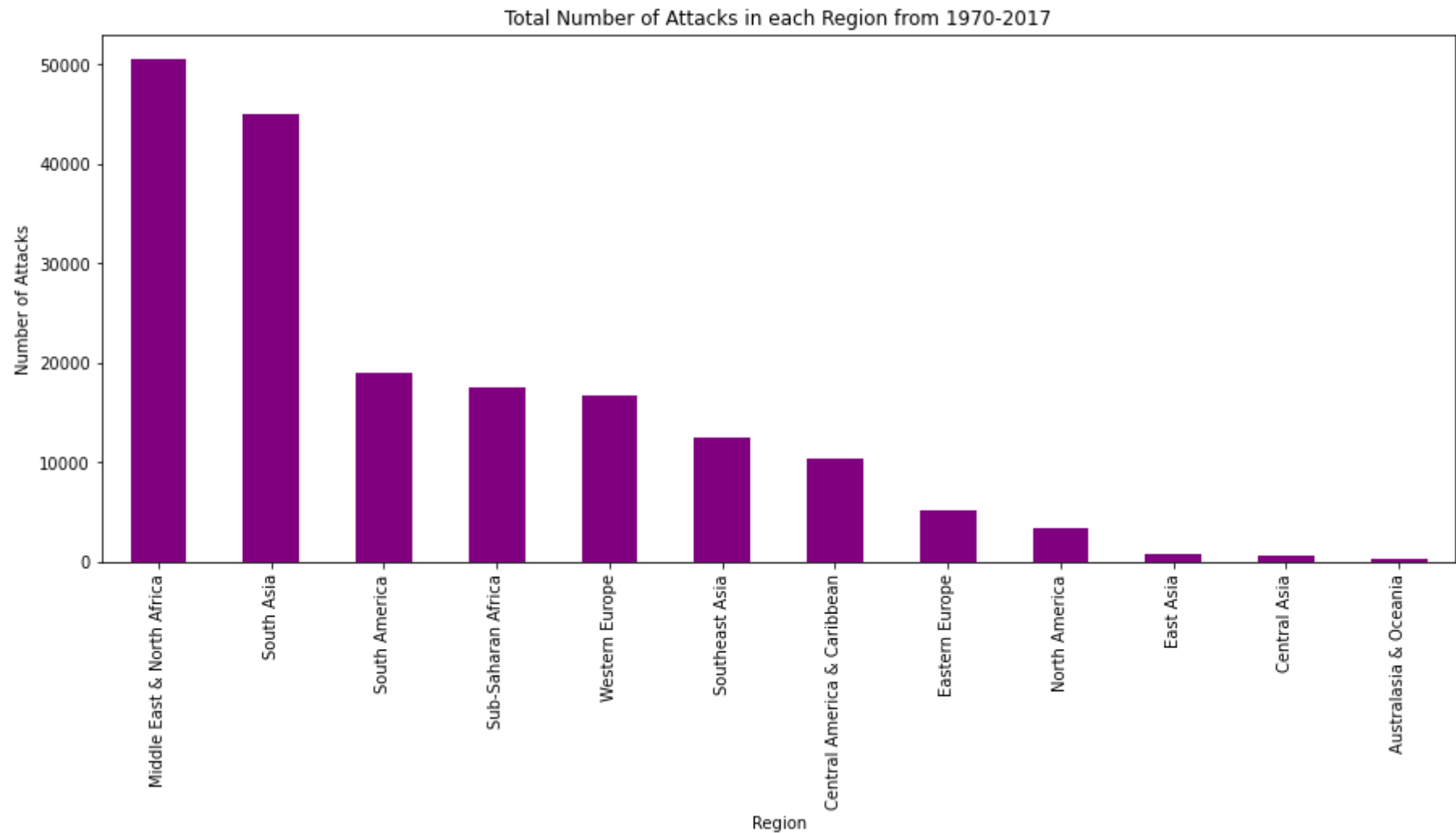
Total Terrorist Attacks in each Region from 1970-2017

```
In [23]: regt=reg.transpose()
regt["Total"]=regt.sum(axis=1)
```

```
ra=regt["Total"].sort_values(ascending=False)
ra
```

```
Out[23]: Region
Middle East & North Africa    50474
South Asia                    44974
South America                 18978
Sub-Saharan Africa            17550
Western Europe                16639
Southeast Asia                12485
Central America & Caribbean   10344
Eastern Europe                5144
North America                 3456
East Asia                     802
Central Asia                   563
Australasia & Oceania         282
Name: Total, dtype: int64
```

```
In [24]: ra.plot(kind="bar",color="purple",figsize=(15,6))
plt.title("Total Number of Attacks in each Region from 1970-2017")
plt.xlabel("Region")
plt.ylabel("Number of Attacks")
plt.show()
```



```
In [25]: rc=df[["Region","Casualty"]].groupby("Region").sum().sort_values(by="Casualty",ascending=False)
rc
```

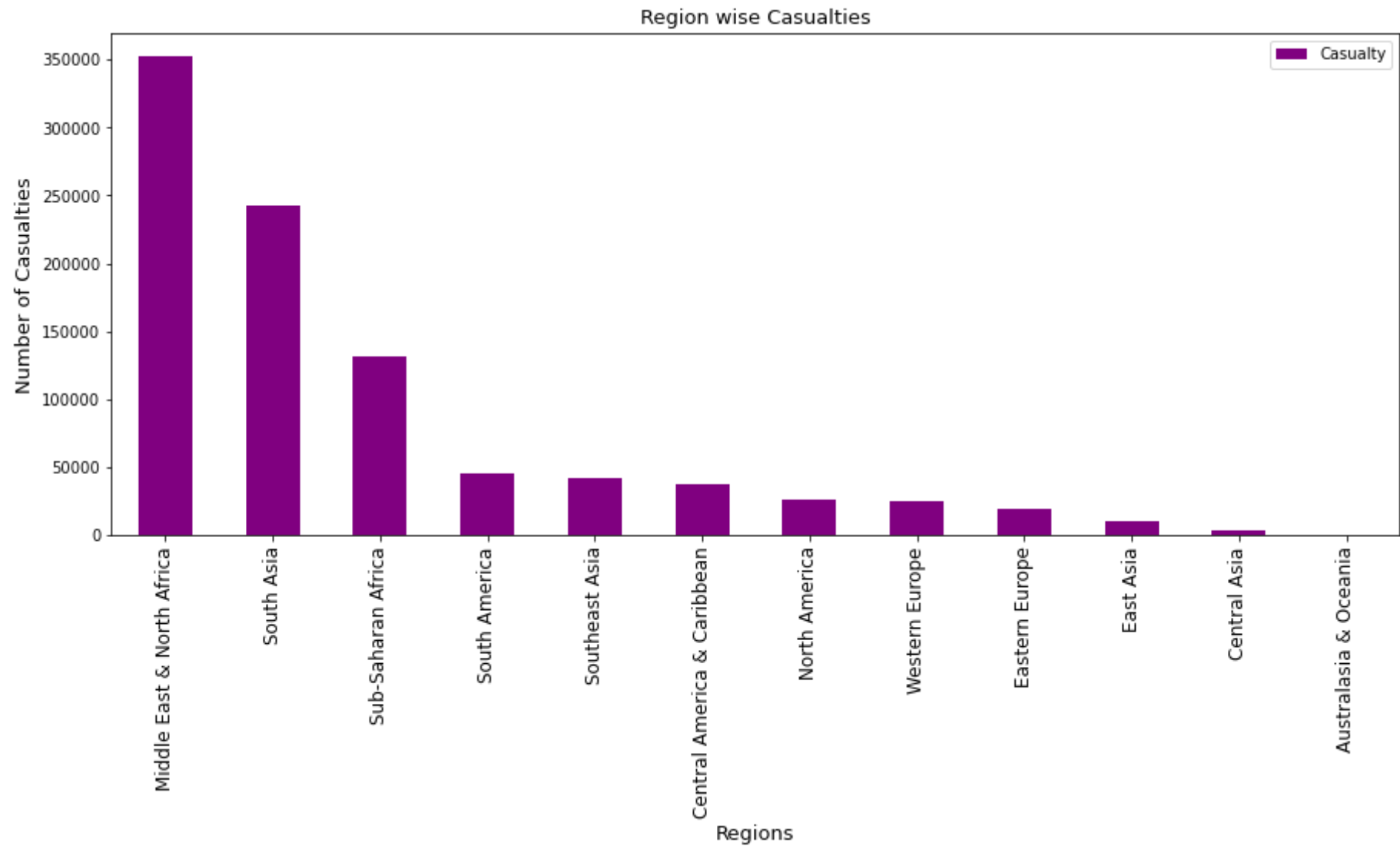
```
Out[25]:
```

	Casualty
Region	
Middle East & North Africa	351950.0
South Asia	242679.0

	Casualty
Region	
Sub-Saharan Africa	131243.0
South America	45553.0
Southeast Asia	41896.0
Central America & Caribbean	37699.0
North America	26447.0
Western Europe	25026.0
Eastern Europe	19460.0
East Asia	10365.0
Central Asia	3009.0
Australasia & Oceania	410.0

In [26]:

```
rc.plot(kind="bar", color="purple", figsize=(15,6))
plt.title("Region wise Casualties", fontsize=13)
plt.xlabel("Regions", fontsize=13)
plt.xticks(fontsize=12)
plt.ylabel("Number of Casualties", fontsize=13)
plt.show()
```



Killed in each Region

```
In [27]: rk=df[["Region","Killed"]].groupby("Region").sum().sort_values(by="Killed",ascending=False)
rk
```

```
Out[27]: Killed
```


Region	Killed
Region	
Middle East & North Africa	137642.0
South Asia	101319.0
Sub-Saharan Africa	78386.0
South America	28849.0
Central America & Caribbean	28708.0
Southeast Asia	15637.0
Eastern Europe	7415.0
Western Europe	6694.0
North America	4916.0
East Asia	1152.0
Central Asia	1000.0
Australasia & Oceania	150.0

Wounded in each Region

In [28]:

```
rw=df[["Region","Wounded"]].groupby("Region").sum().sort_values(by="Wounded",ascending=False)
rw
```

Out[28]:

Region	Wounded
Middle East & North Africa	214308.0
South Asia	141360.0
Sub-Saharan Africa	52857.0
Southeast Asia	26259.0

Wounded	
Region	
North America	21531.0
Western Europe	18332.0
South America	16704.0
Eastern Europe	12045.0
East Asia	9213.0
Central America & Caribbean	8991.0
Central Asia	2009.0
Australasia & Oceania	260.0

In [29]:

```

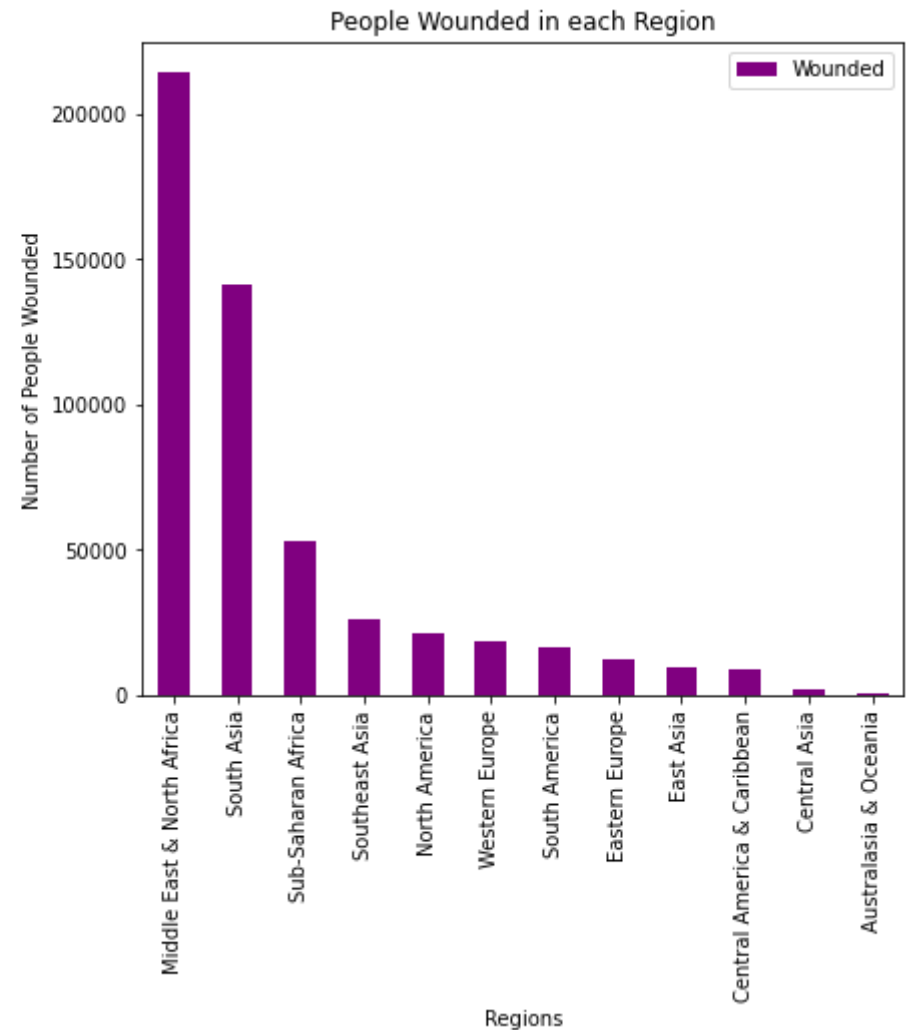
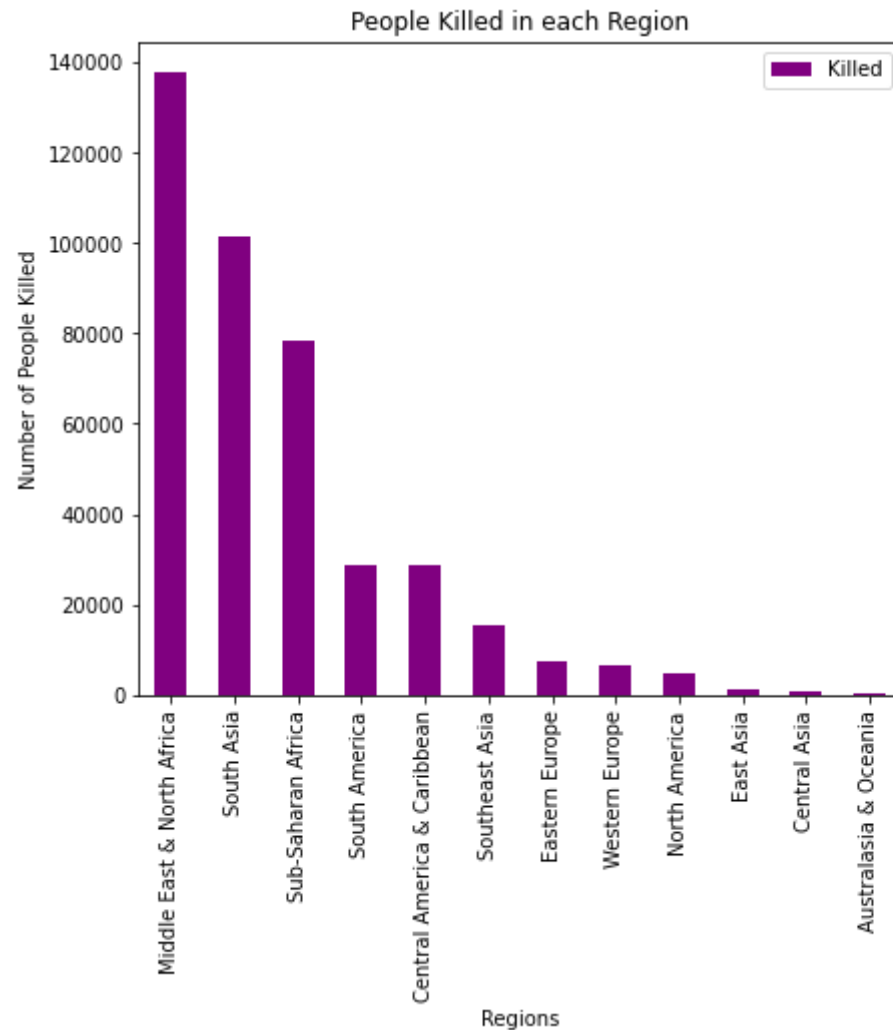
fig=plt.figure()
ax0=fig.add_subplot(1,2,1)
ax1=fig.add_subplot(1,2,2)

#Killed
rk.plot(kind="bar",color="purple",figsize=(15,6),ax=ax0)
ax0.set_title("People Killed in each Region")
ax0.set_xlabel("Regions")
ax0.set_ylabel("Number of People Killed")

#Wounded
rw.plot(kind="bar",color="purple",figsize=(15,6),ax=ax1)
ax1.set_title("People Wounded in each Region")
ax1.set_xlabel("Regions")
ax1.set_ylabel("Number of People Wounded")

plt.show()

```



3. Country wise Attacks - Top 10

Number of Attacks in each Country

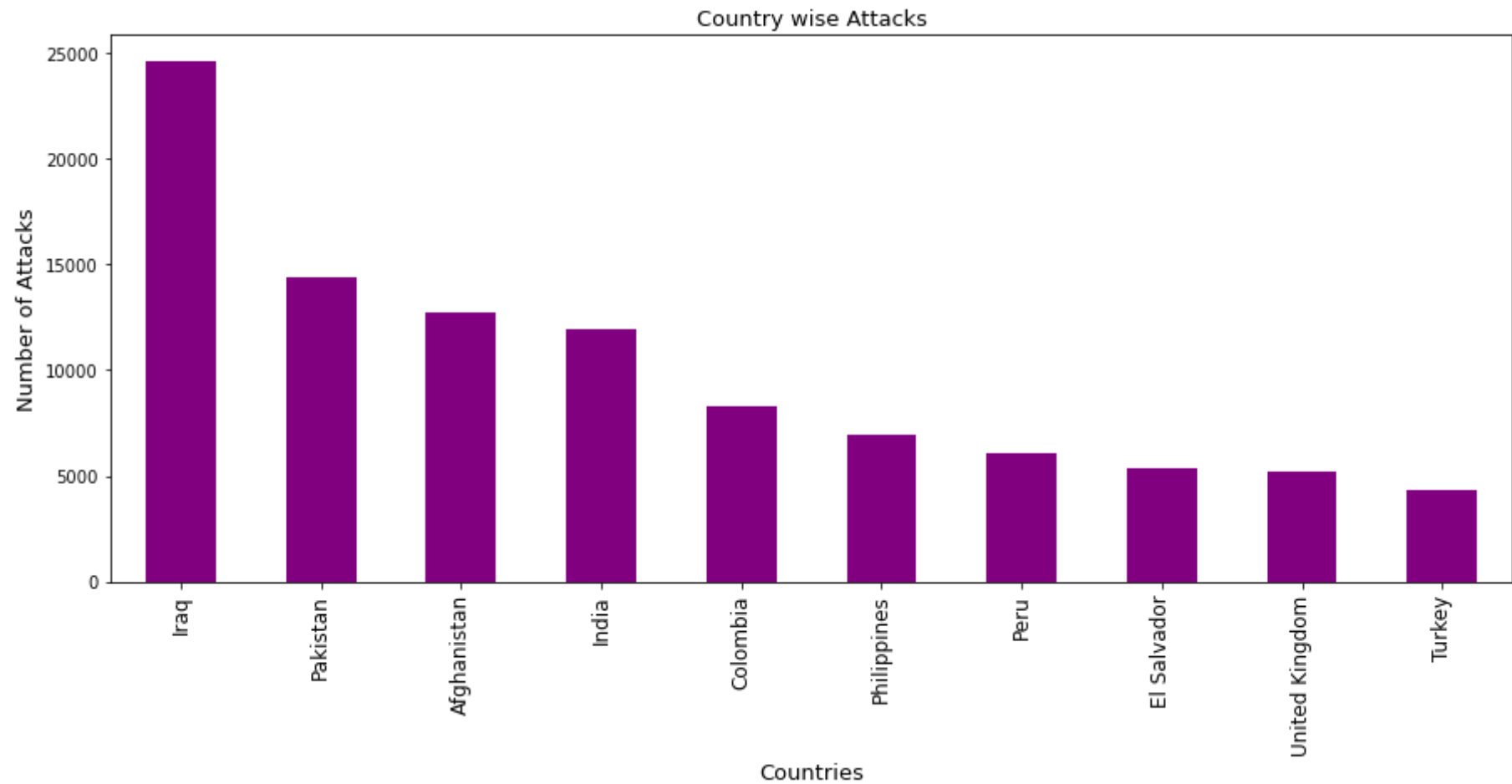
```
In [30]: ct=df["Country"].value_counts().head(10)
ct
```

```
Out[30]: Iraq                24636
```

Pakistan	14368
Afghanistan	12731
India	11960
Colombia	8306
Philippines	6908
Peru	6096
El Salvador	5320
United Kingdom	5235
Turkey	4292

Name: Country, dtype: int64

```
In [31]: ct.plot(kind="bar", color="purple", figsize=(15,6))
plt.title("Country wise Attacks", fontsize=13)
plt.xlabel("Countries", fontsize=13)
plt.xticks(fontsize=12)
plt.ylabel("Number of Attacks", fontsize=13)
plt.show()
```



Total Casualties (Killed + Wounded) in each Country

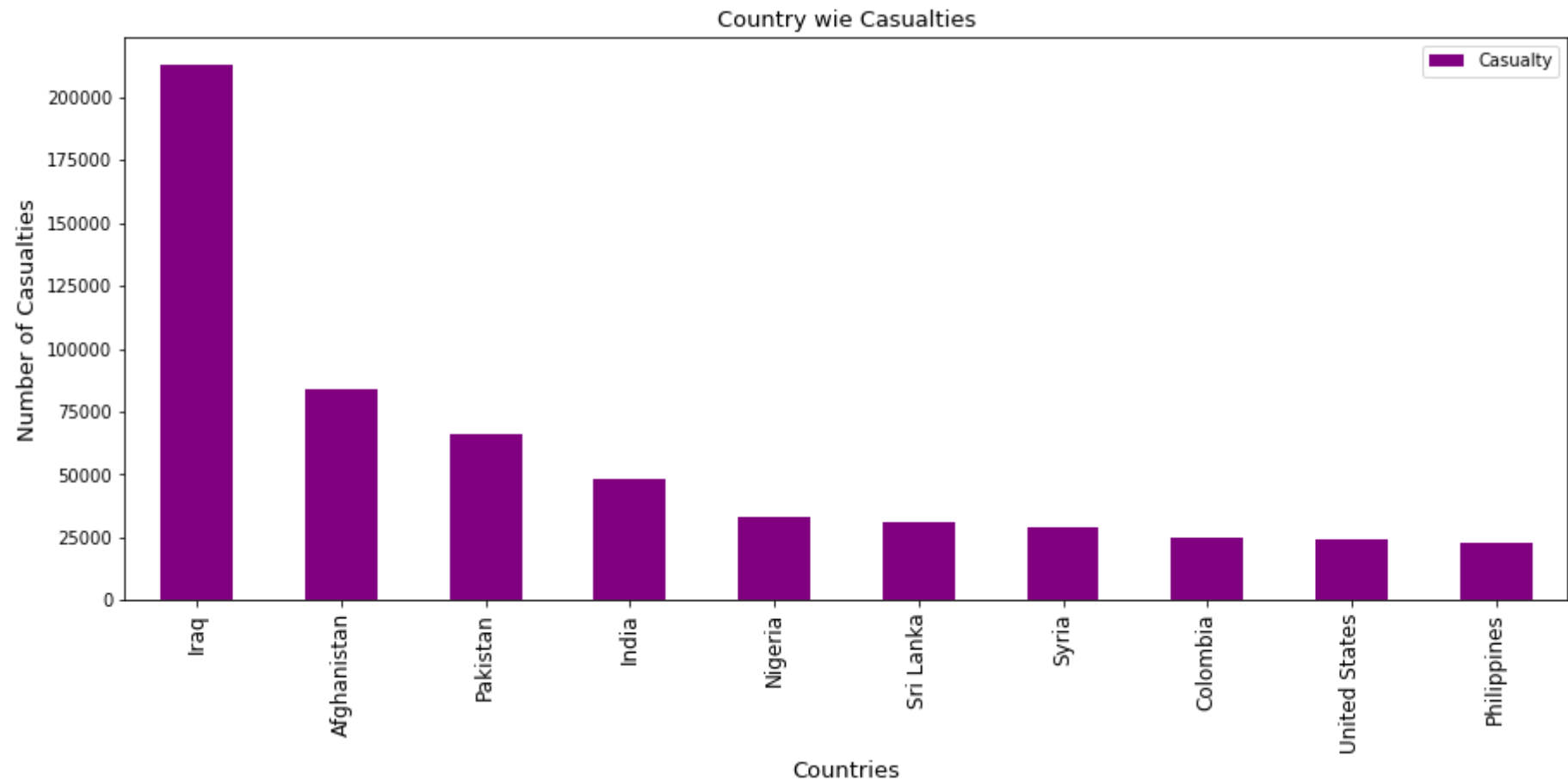
```
In [32]: cnc=df[["Country","Casualty"]].groupby("Country").sum().sort_values(by="Casualty",ascending=False)
cnc.head(10)
```

```
Out[32]:
```

	Casualty
Country	
Iraq	213279.0
Afghanistan	83661.0

Casualty	
Country	
Pakistan	65860.0
India	48321.0
Nigeria	32921.0
Sri Lanka	31091.0
Syria	29338.0
Colombia	25026.0
United States	24473.0
Philippines	22926.0

```
In [41]: cnc[:10].plot(kind="bar",color="purple",figsize=(15,6))
plt.title("Country wie Casualties",fontsize=13)
plt.xlabel("Countries",fontsize=13)
plt.xticks(fontsize=12)
plt.ylabel("Number of Casualties",fontsize=13)
plt.show()
```



Killed in each Country

```
In [42]: cnk=df[["Country","Killed"]].groupby("Country").sum().sort_values(by="Killed",ascending=False)
cnk.head(10)
```

Out[42]:

Killed	
Country	
Iraq	78589.0
Afghanistan	39384.0
Pakistan	23822.0

Killed	
Country	
Nigeria	22682.0
India	19341.0
Sri Lanka	15530.0
Syria	15229.0
Colombia	14698.0
Peru	12771.0
El Salvador	12053.0

Wounded in each Country

```
In [35]: cnw=df[["Country","Wounded"]].groupby("Country").sum().sort_values(by="Wounded",ascending=False)
cnw.head(10)
```

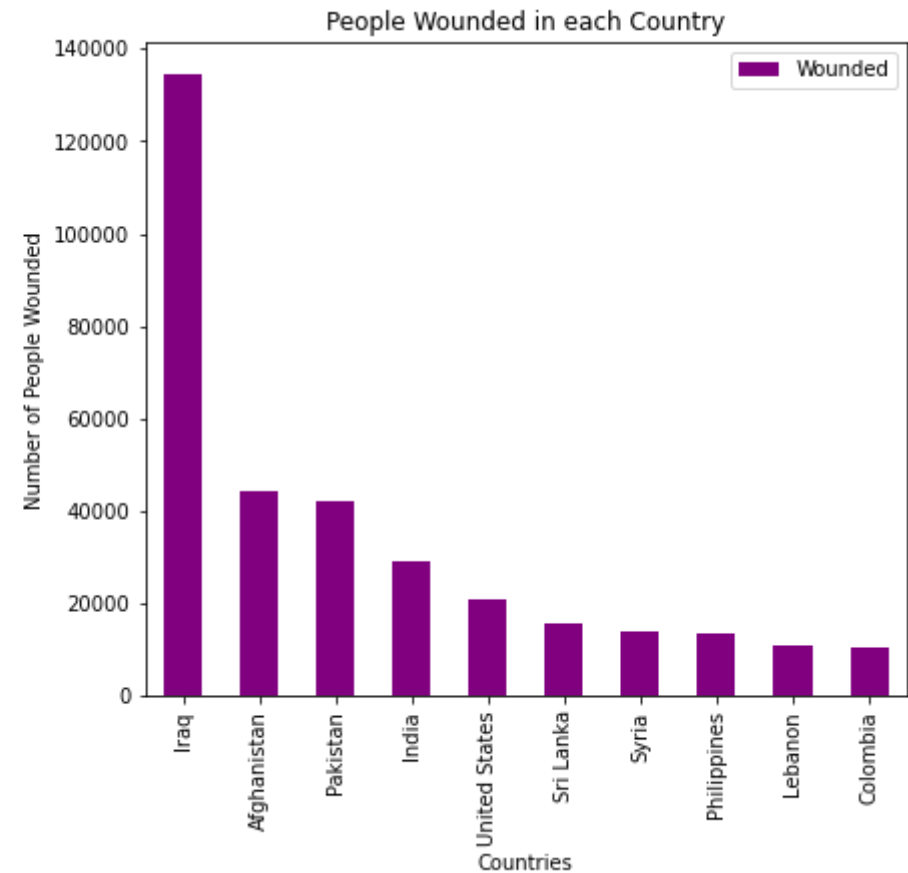
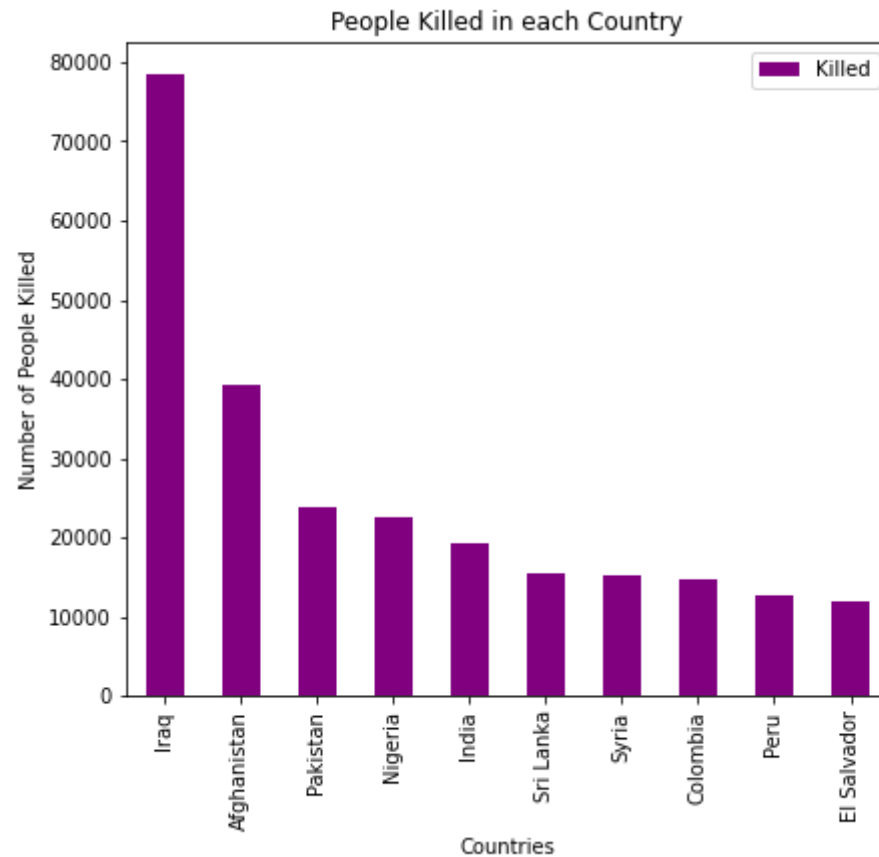
Wounded	
Country	
Iraq	134690.0
Afghanistan	44277.0
Pakistan	42038.0
India	28980.0
United States	20702.0
Sri Lanka	15561.0
Syria	14109.0
Philippines	13367.0
Lebanon	10904.0
Colombia	10328.0


```
In [36]: fig=plt.figure()
ax0=fig.add_subplot(1,2,1)
ax1=fig.add_subplot(1,2,2)

#Killed
cnk[:10].plot(kind="bar",color="purple",figsize=(15,6),ax=ax0)
ax0.set_title("People Killed in each Country")
ax0.set_xlabel("Countries")
ax0.set_ylabel("Number of People Killed")

#Wounded
cnw[:10].plot(kind="bar",color="purple",figsize=(15,6),ax=ax1)
ax1.set_title("People Wounded in each Country")
ax1.set_xlabel("Countries")
ax1.set_ylabel("Number of People Wounded")

plt.show()
```



4. City wise Attacks - Top 10

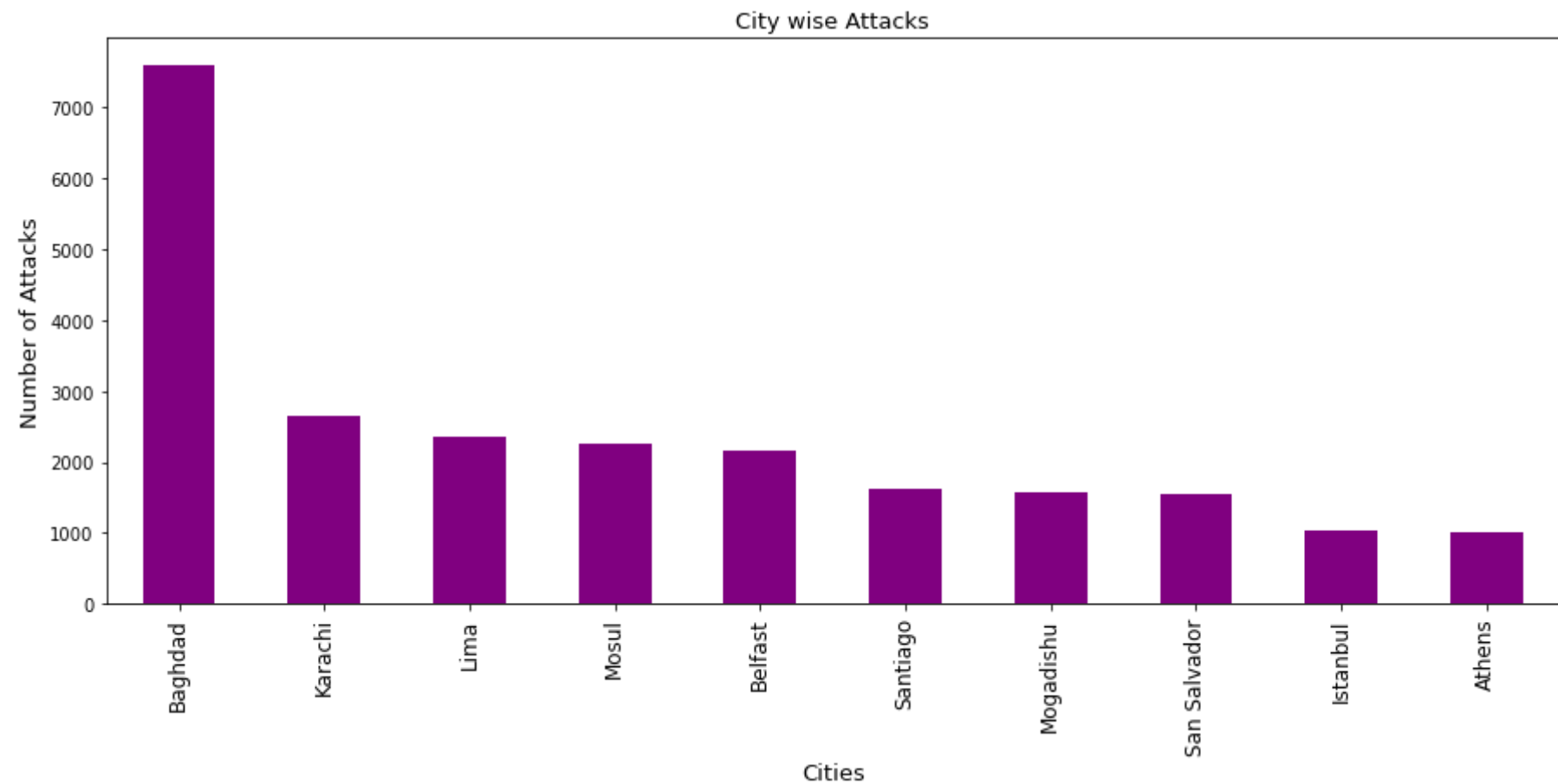
Number of Attacks in each city

```
In [37]: city=df["City"].value_counts()[1:11]
city
```

```
Out[37]: Baghdad      7589
Karachi      2652
Lima      2359
Mosul      2265
Belfast      2171
Santiago      1621
```

```
Mogadishu      1581
San Salvador    1558
Istanbul        1048
Athens          1019
Name: City, dtype: int64
```

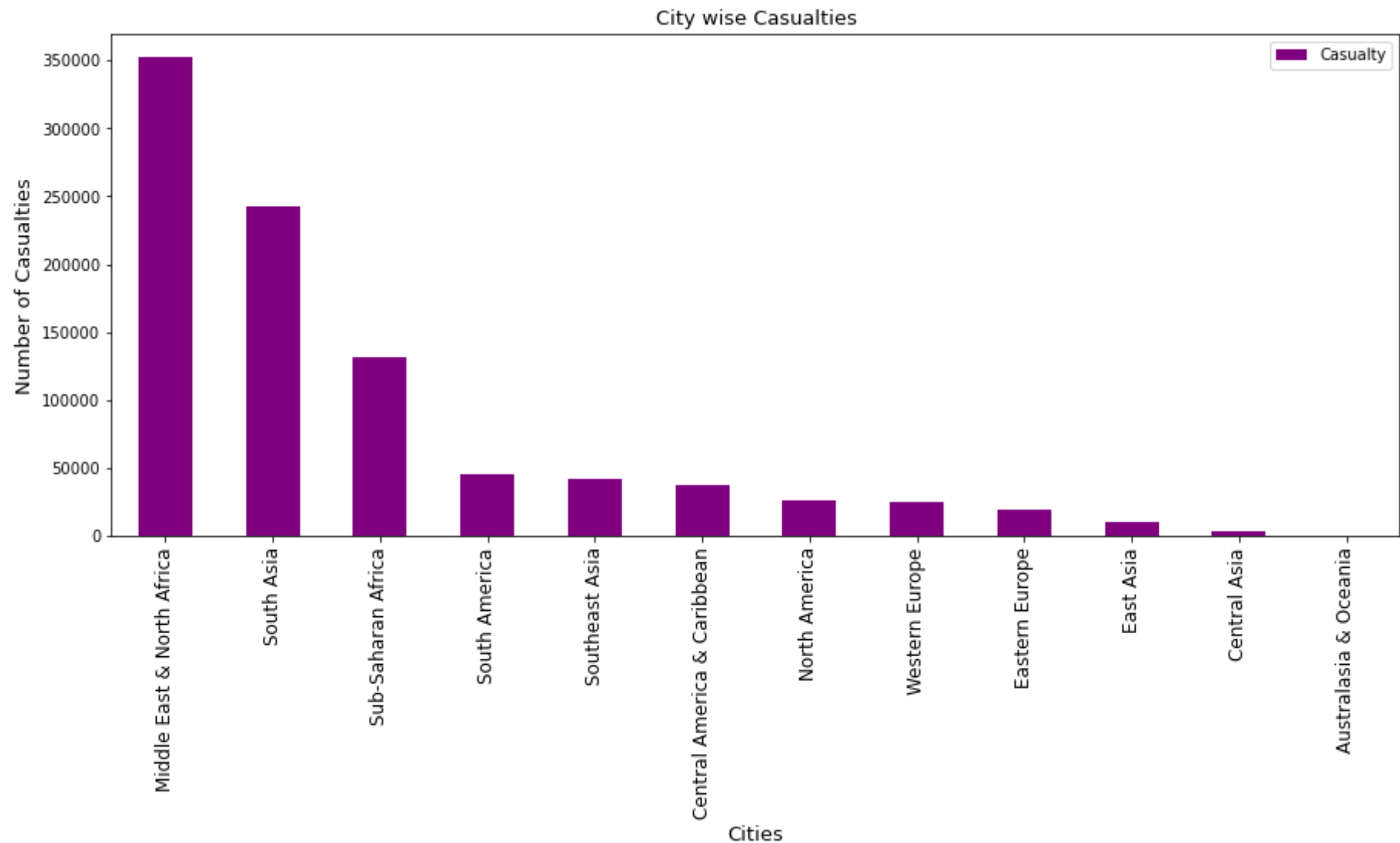
```
In [38]: city.plot(kind="bar",color="purple",figsize=(15,6))
plt.title("City wise Attacks",fontsize=13)
plt.xlabel("Cities",fontsize=13)
plt.xticks(fontsize=12)
plt.ylabel("Number of Attacks",fontsize=13)
plt.show()
```



Total Casualties (Killed + Wounded) in each City

In [43]:

```
rc.plot(kind="bar",color="purple",figsize=(15,6))  
plt.title("City wise Casualties",fontsize=13)  
plt.xlabel("Cities",fontsize=13)  
plt.xticks(fontsize=12)  
plt.ylabel("Number of Casualties",fontsize=13)  
plt.show()
```



```
In [44]: cc=df[["City","Casualty"]].groupby("City").sum().sort_values(by="Casualty",ascending=False).drop("Unknown")
cc.head(10)
```

Out[44]:

Casualty	
City	
Baghdad	77876.0
New York City	19619.0
Mosul	12927.0
Karachi	9376.0
Mogadishu	8868.0
Kabul	8466.0
Beirut	7257.0
Kirkuk	6636.0
Colombo	5906.0
Aleppo	5748.0

Killed in each City

```
In [45]: ck=df[["City","Killed"]].groupby("City").sum().sort_values(by="Killed",ascending=False).drop("Unknown")
ck.head(10)
```

Out[45]:

Killed	
City	
Baghdad	21151.0
Mosul	7140.0
Mogadishu	3913.0
Karachi	3688.0
New York City	2838.0

	Killed
City	
Tikrit	2679.0
Kabul	2493.0
Ramadi	2313.0
Maiduguri	2235.0
Aleppo	2125.0

Wounded in each City

```
In [46]: cw=df[["City","Wounded"]].groupby("City").sum().sort_values(by="Wounded",ascending=False).drop("Unknown")
cw.head(10)
```

```
Out[46]:
```

	Wounded
City	
Baghdad	56725.0
New York City	16781.0
Kabul	5973.0
Mosul	5787.0
Karachi	5688.0
Tokyo	5542.0
Beirut	5341.0
Nairobi	5024.0
Kirkuk	5008.0
Mogadishu	4955.0

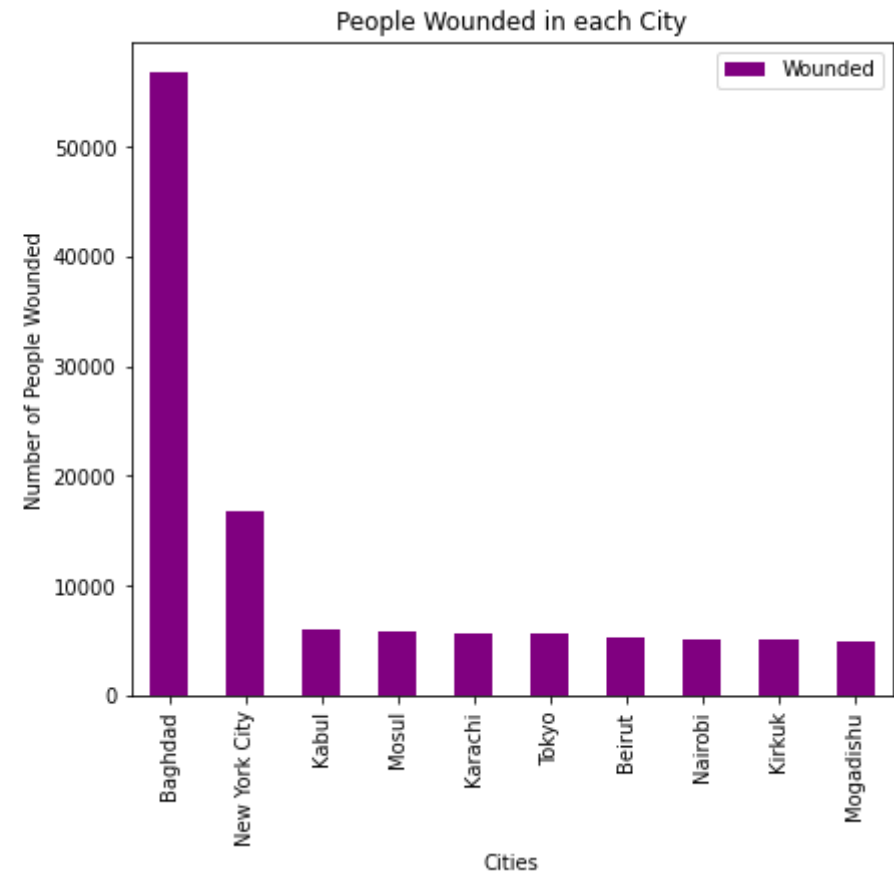
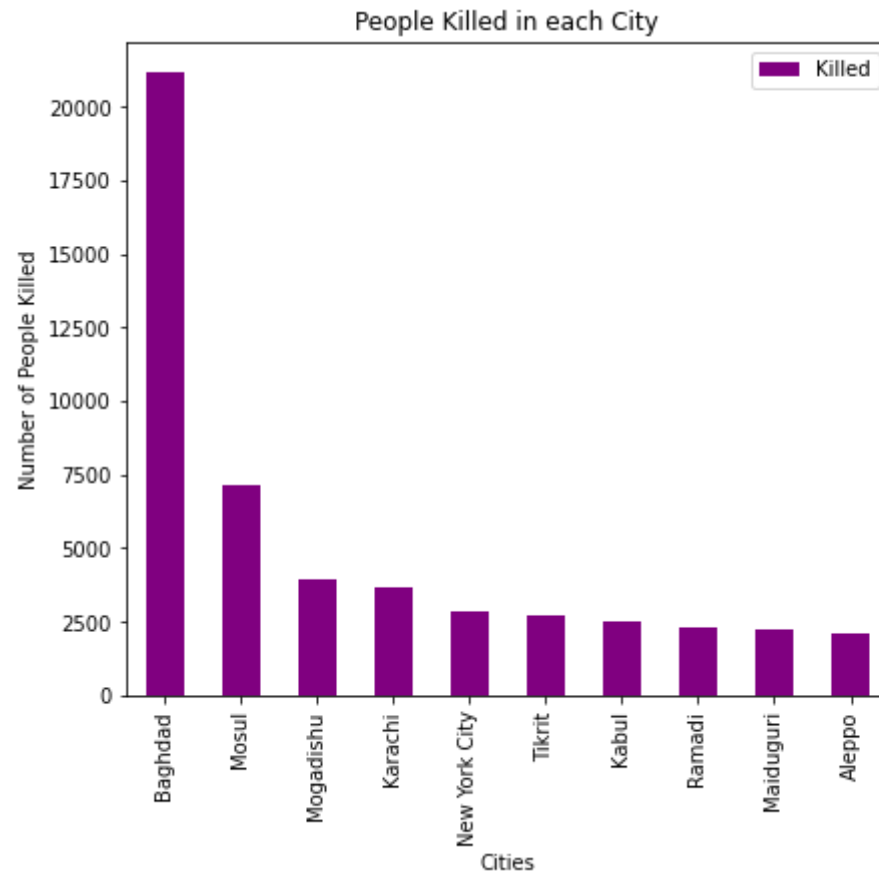
```
In [47]: fig=plt.figure()
```

```
ax0=fig.add_subplot(1,2,1)
ax1=fig.add_subplot(1,2,2)

#Killed
ck[:10].plot(kind="bar",color="purple",figsize=(15,6),ax=ax0)
ax0.set_title("People Killed in each City")
ax0.set_xlabel("Cities")
ax0.set_ylabel("Number of People Killed")

#Wounded
cw[:10].plot(kind="bar",color="purple",figsize=(15,6),ax=ax1)
ax1.set_title("People Wounded in each City")
ax1.set_xlabel("Cities")
ax1.set_ylabel("Number of People Wounded")

plt.show()
```



5. Terrorist Group wise Attacks - Top 10

Number of Attacks by each Group

```
In [48]: grp=df["Group Name"].value_counts()[1:10]
grp
```

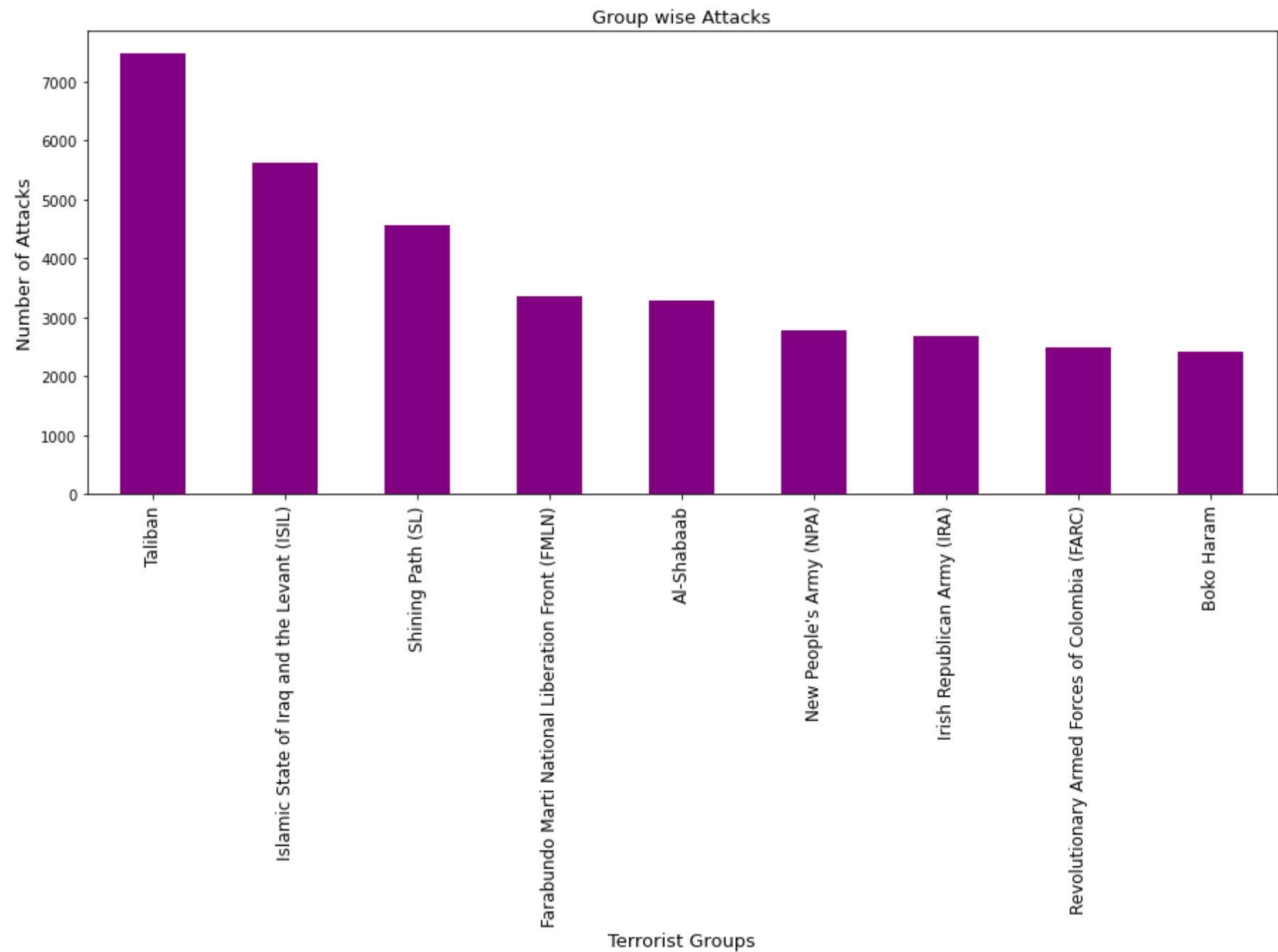
```
Out[48]: Taliban 7478
Islamic State of Iraq and the Levant (ISIL) 5613
Shining Path (SL) 4555
Farabundo Marti National Liberation Front (FMLN) 3351
Al-Shabaab 3288
New People's Army (NPA) 2772
```


Irish Republican Army (IRA)	2671
Revolutionary Armed Forces of Colombia (FARC)	2487
Boko Haram	2418

Name: Group Name, dtype: int64

In [49]:

```
grp.plot(kind="bar",color="purple",figsize=(15,6))  
plt.title("Group wise Attacks",fontsize=13)  
plt.xlabel("Terrorist Groups",fontsize=13)  
plt.xticks(fontsize=12)  
plt.ylabel("Number of Attacks",fontsize=13)  
plt.show()
```



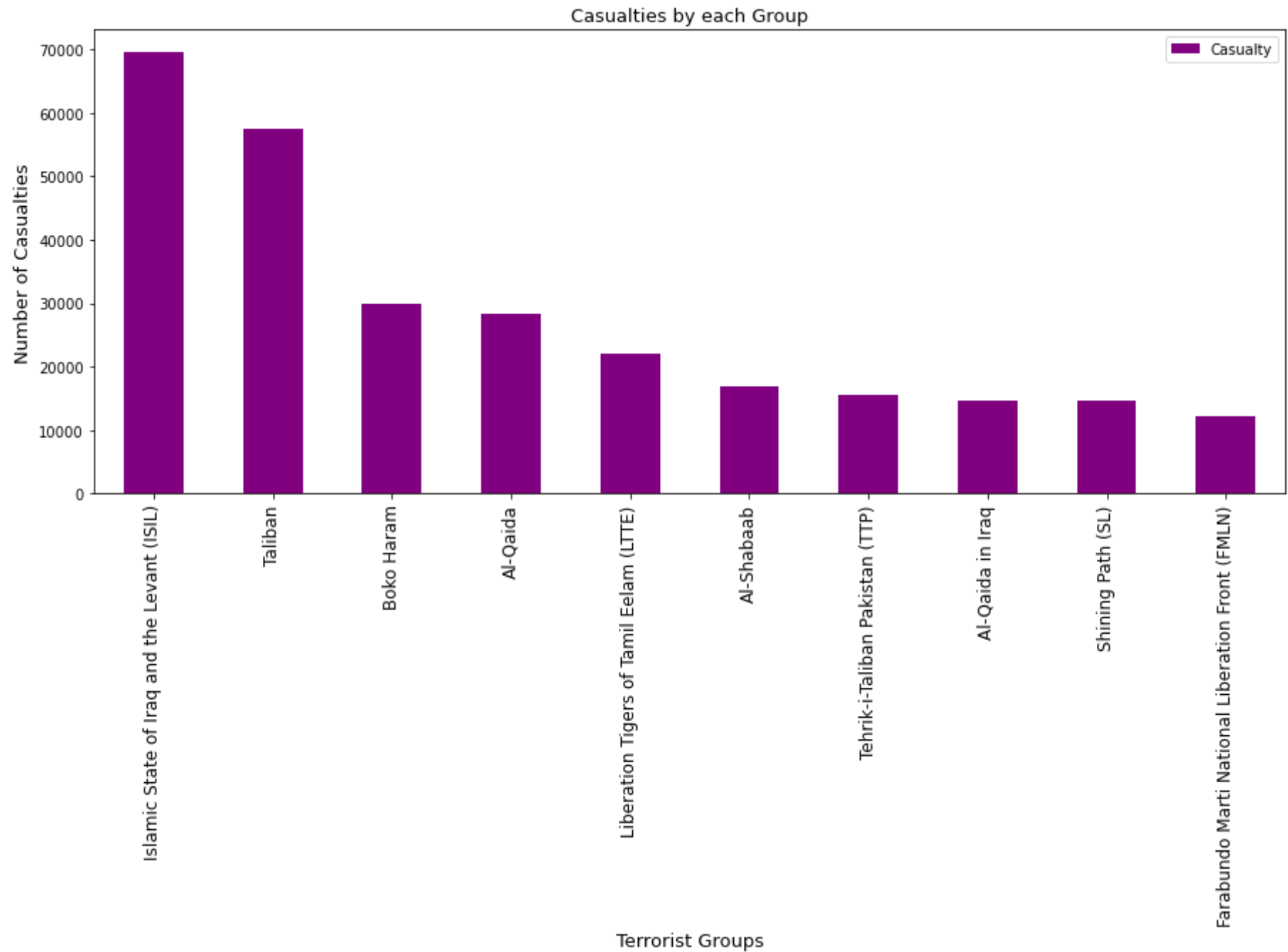
Total Casualties(Killed + Wounded) by each Group

```
In [50]: gc=df[["Group Name", "Casualty"]].groupby("Group Name").sum().sort_values(by="Casualty",ascending=False).drop("Unknown")
gc.head(10)
```

```
Out[50]:
```

	Casualty
Group Name	
Islamic State of Iraq and the Levant (ISIL)	69595.0
Taliban	57342.0
Boko Haram	29801.0
Al-Qaida	28372.0
Liberation Tigers of Tamil Eelam (LTTE)	22020.0
Al-Shabaab	16954.0
Tehrik-i-Taliban Pakistan (TTP)	15574.0
Al-Qaida in Iraq	14724.0
Shining Path (SL)	14632.0
Farabundo Marti National Liberation Front (FMLN)	12130.0

```
In [51]: gc.head(10).plot(kind="bar",color="purple",figsize=(15,6))
plt.title("Casualties by each Group",fontsize=13)
plt.xlabel("Terrorist Groups",fontsize=13)
plt.xticks(fontsize=12)
plt.ylabel("Number of Casualties",fontsize=13)
plt.show()
```



Killed by each Group

```
In [52]: gk=df[["Group Name", "Killed"]].groupby("Group Name").sum().sort_values(by="Killed", ascending=False).drop("Unknown")
gk.head(10)
```

```
Out[52]:
```

	Killed
Group Name	
Islamic State of Iraq and the Levant (ISIL)	38923.0
Taliban	29410.0
Boko Haram	20328.0
Shining Path (SL)	11601.0
Liberation Tigers of Tamil Eelam (LTTE)	10989.0
Al-Shabaab	9330.0
Farabundo Marti National Liberation Front (FMLN)	8065.0
Nicaraguan Democratic Force (FDN)	6662.0
Tehrik-i-Taliban Pakistan (TTP)	6042.0
Revolutionary Armed Forces of Colombia (FARC)	5661.0

Wounded by each Group

```
In [53]: gw=df[["Group Name", "Wounded"]].groupby("Group Name").sum().sort_values(by="Wounded", ascending=False).drop("Unknown")
gw.head(10)
```

```
Out[53]:
```

	Wounded
Group Name	
Islamic State of Iraq and the Levant (ISIL)	30672.0
Taliban	27932.0
Al-Qaida	24512.0
Liberation Tigers of Tamil Eelam (LTTE)	11031.0

Wounded	
Group Name	
Al-Qaida in Iraq	10343.0
Tehrik-i-Taliban Pakistan (TTP)	9532.0
Boko Haram	9473.0
Al-Shabaab	7624.0
Aum Shinri Kyo	6003.0
Kurdistan Workers' Party (PKK)	4908.0

In [54]:

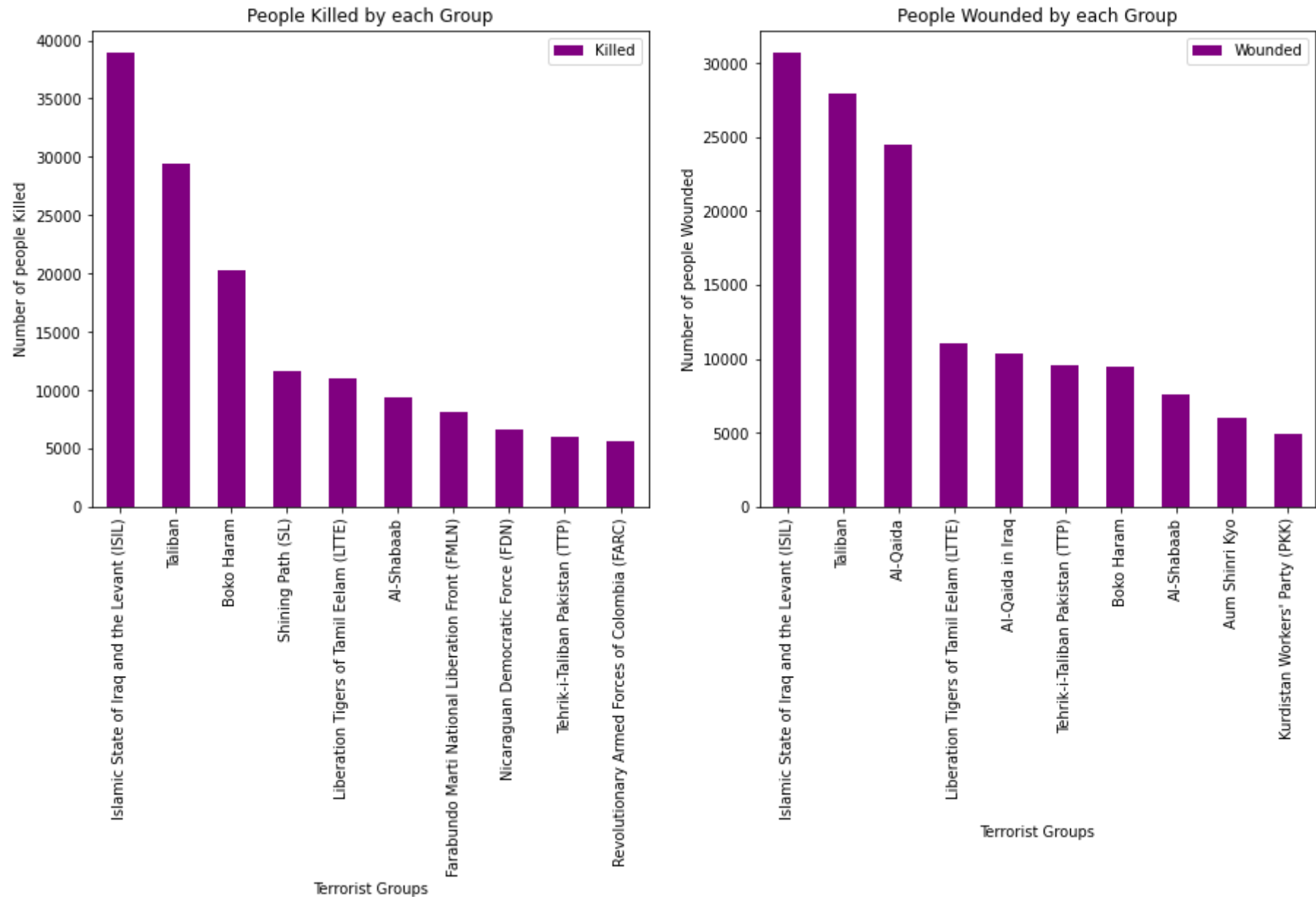
```

fig=plt.figure()
ax0=fig.add_subplot(1,2,1)
ax1=fig.add_subplot(1,2,2)

#Killed
gk[:10].plot(kind="bar",color="purple",figsize=(15,6),ax=ax0)
ax0.set_title("People Killed by each Group")
ax0.set_xlabel("Terrorist Groups")
ax0.set_ylabel("Number of people Killed")

#Wounded
gw[:10].plot(kind="bar",color="purple",figsize=(15,6),ax=ax1)
ax1.set_title("People Wounded by each Group")
ax1.set_xlabel("Terrorist Groups")
ax1.set_ylabel("Number of people Wounded")
plt.show()

```



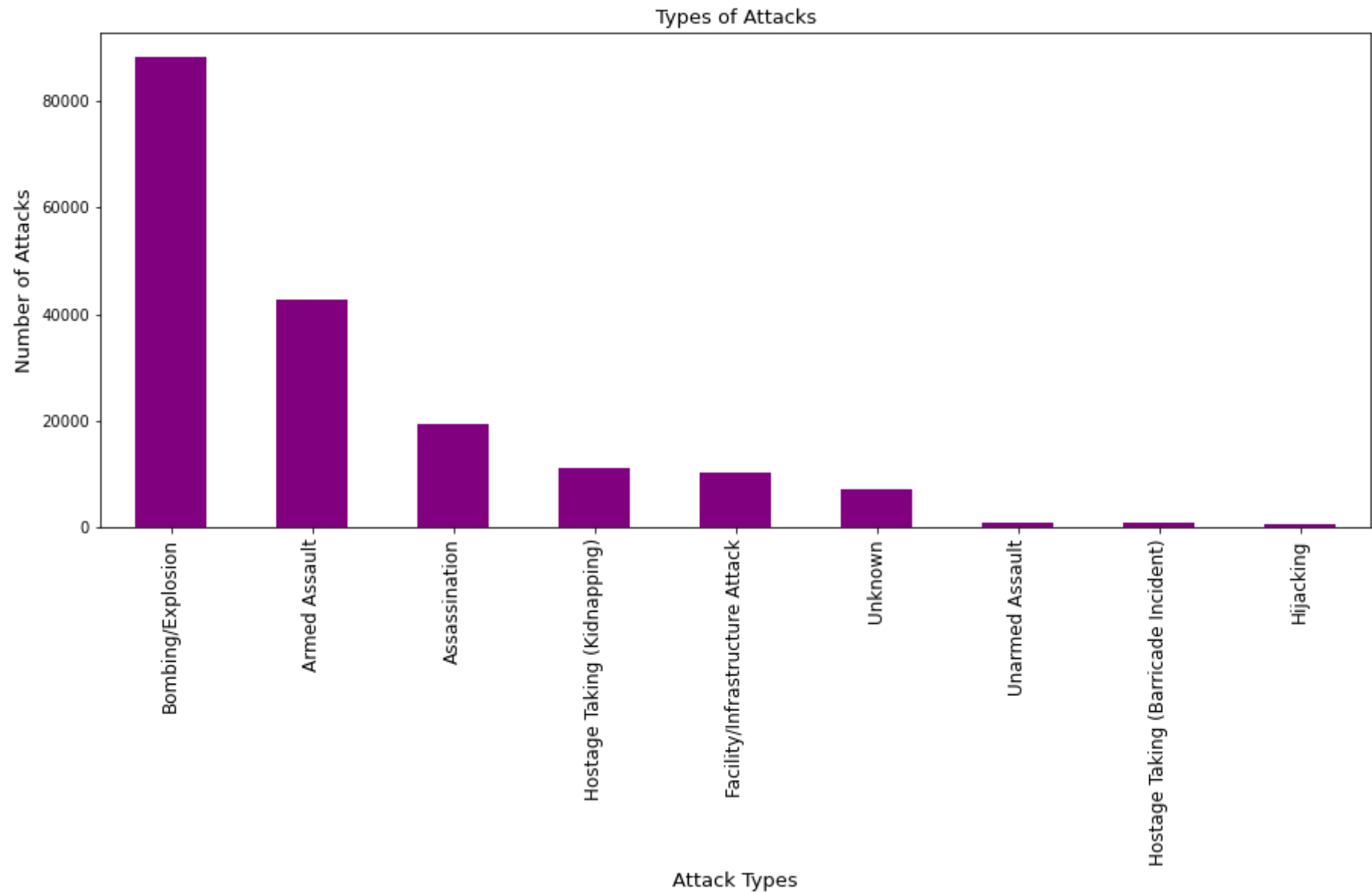
6. Attack Type wise Attacks

Number of Attacks by each Attack Type

```
In [55]: at=df["Attack Type"].value_counts()  
at
```

```
Out[55]: 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: Attack Type, dtype: int64
```

```
In [56]: at.plot(kind="bar",color="purple",figsize=(15,6))  
plt.title("Types of Attacks",fontsize=13)  
plt.xlabel("Attack Types",fontsize=13)  
plt.xticks(fontsize=12)  
plt.ylabel("Number of Attacks",fontsize=13)  
plt.show()
```

Total Casualties (Killed + Wounded) by each Attack Type

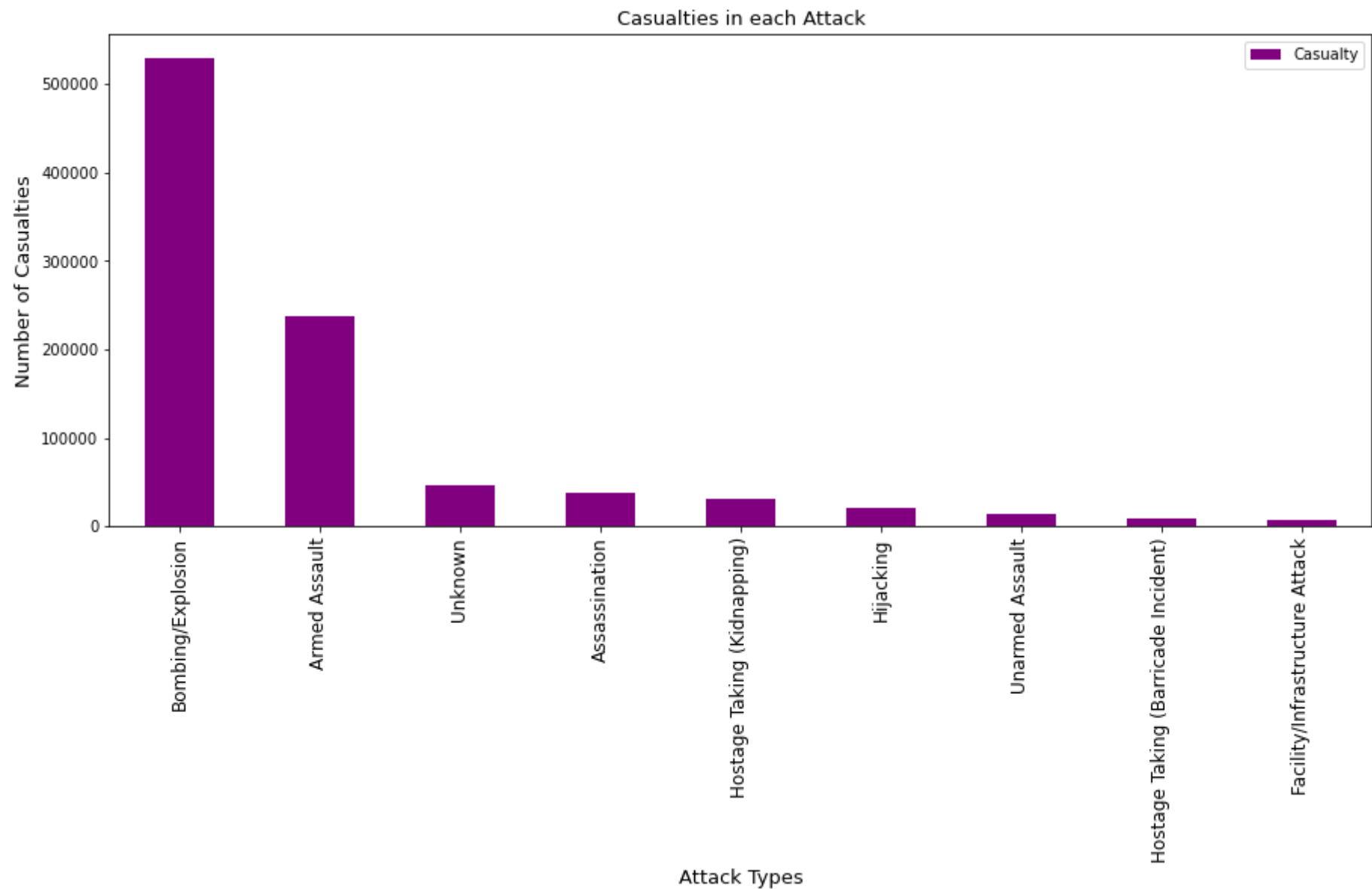
```
In [57]: ac=df[["Attack Type","Casualty"]].groupby("Attack Type").sum().sort_values(by="Casualty",ascending=False)
ac
```

Out[57]:

	Casualty
Attack Type	
Bombing/Explosion	530007.0
Armed Assault	237663.0
Unknown	47106.0
Assassination	38807.0
Hostage Taking (Kidnapping)	30677.0
Hijacking	20719.0
Unarmed Assault	14907.0
Hostage Taking (Barricade Incident)	8444.0
Facility/Infrastructure Attack	7407.0

In [58]:

```
ac.plot(kind="bar",color="purple",figsize=(15,6))
plt.title("Casualties in each Attack",fontsize=13)
plt.xlabel("Attack Types",fontsize=13)
plt.xticks(fontsize=12)
plt.ylabel("Number of Casualties",fontsize=13)
plt.show()
```



Killed by each Attack Type

```
In [59]: ak=df[["Attack Type","Killed"]].groupby("Attack Type").sum().sort_values(by="Killed",ascending=False)
ak
```

Out[59]:

	Killed
Attack Type	
Armed Assault	160297.0
Bombing/Explosion	157321.0
Unknown	32381.0
Assassination	24920.0
Hostage Taking (Kidnapping)	24231.0
Hostage Taking (Barricade Incident)	4478.0
Hijacking	3718.0
Facility/Infrastructure Attack	3642.0
Unarmed Assault	880.0

Wounded by each Attack Type

In [60]:

```
aw=df[["Attack Type","Wounded"]].groupby("Attack Type").sum().sort_values(by="Wounded",ascending=False)
aw
```

Out[60]:

	Wounded
Attack Type	
Bombing/Explosion	372686.0
Armed Assault	77366.0
Hijacking	17001.0
Unknown	14725.0
Unarmed Assault	14027.0
Assassination	13887.0
Hostage Taking (Kidnapping)	6446.0
Hostage Taking (Barricade Incident)	3966.0

Wounded**Attack Type**

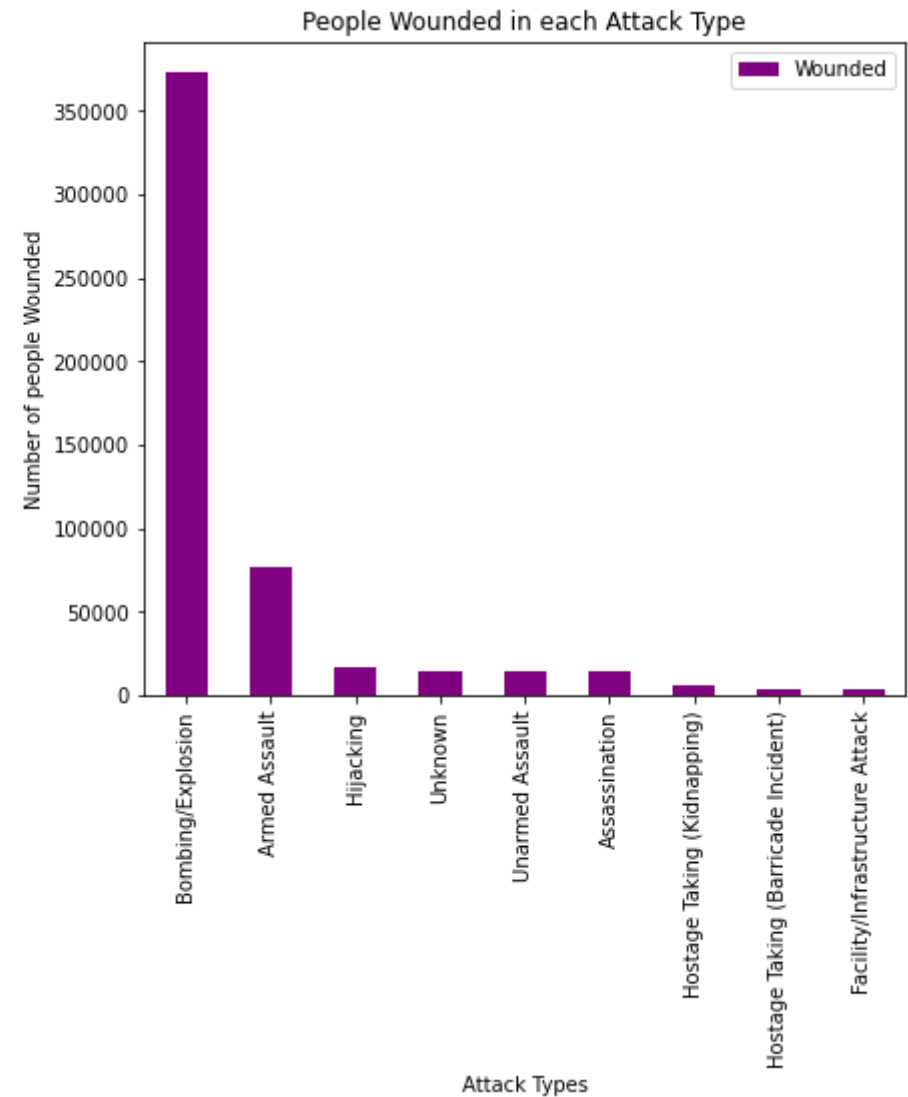
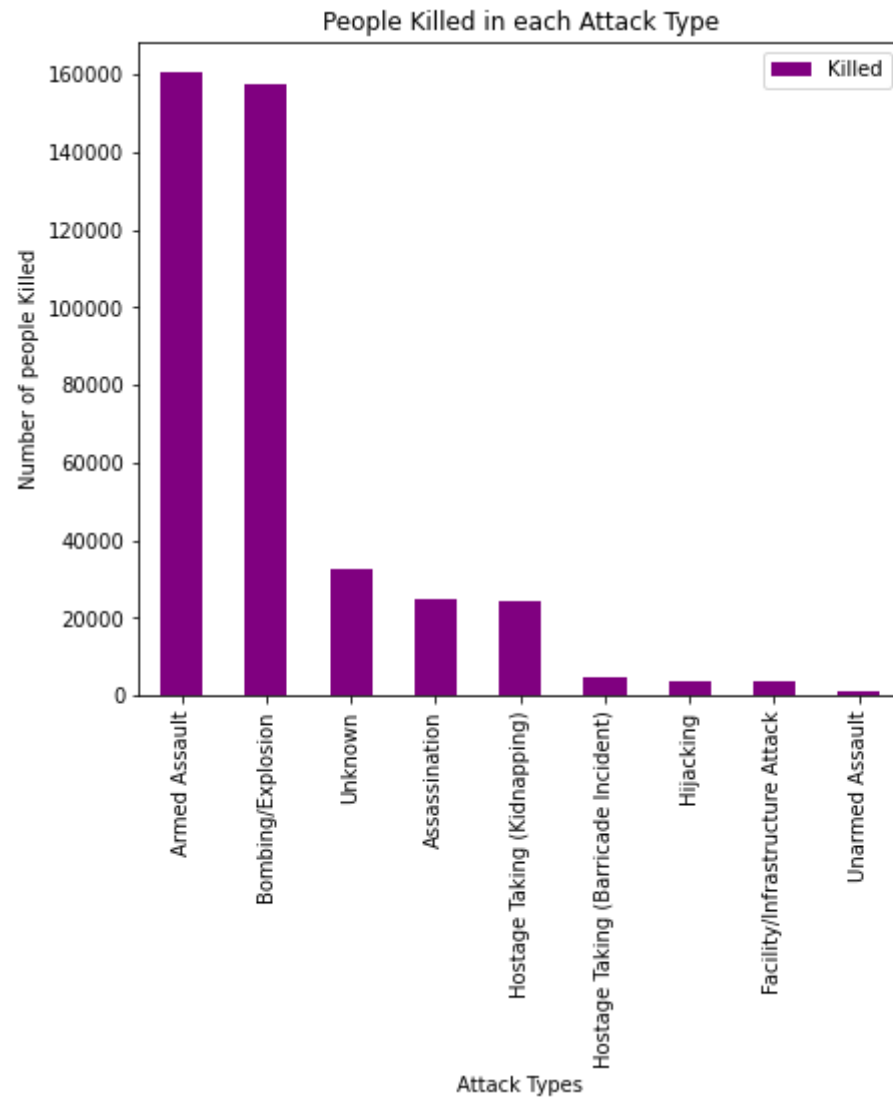
Facility/Infrastructure Attack	3765.0
--------------------------------	--------

In [61]:

```
fig=plt.figure()
ax0=fig.add_subplot(1,2,1)
ax1=fig.add_subplot(1,2,2)

#Killed
ak.plot(kind="bar",color="purple",figsize=(15,6),ax=ax0)
ax0.set_title("People Killed in each Attack Type")
ax0.set_xlabel("Attack Types")
ax0.set_ylabel("Number of people Killed")

#Wounded
aw.plot(kind="bar",color="purple",figsize=(15,6),ax=ax1)
ax1.set_title("People Wounded in each Attack Type")
ax1.set_xlabel("Attack Types")
ax1.set_ylabel("Number of people Wounded")
plt.show()
```



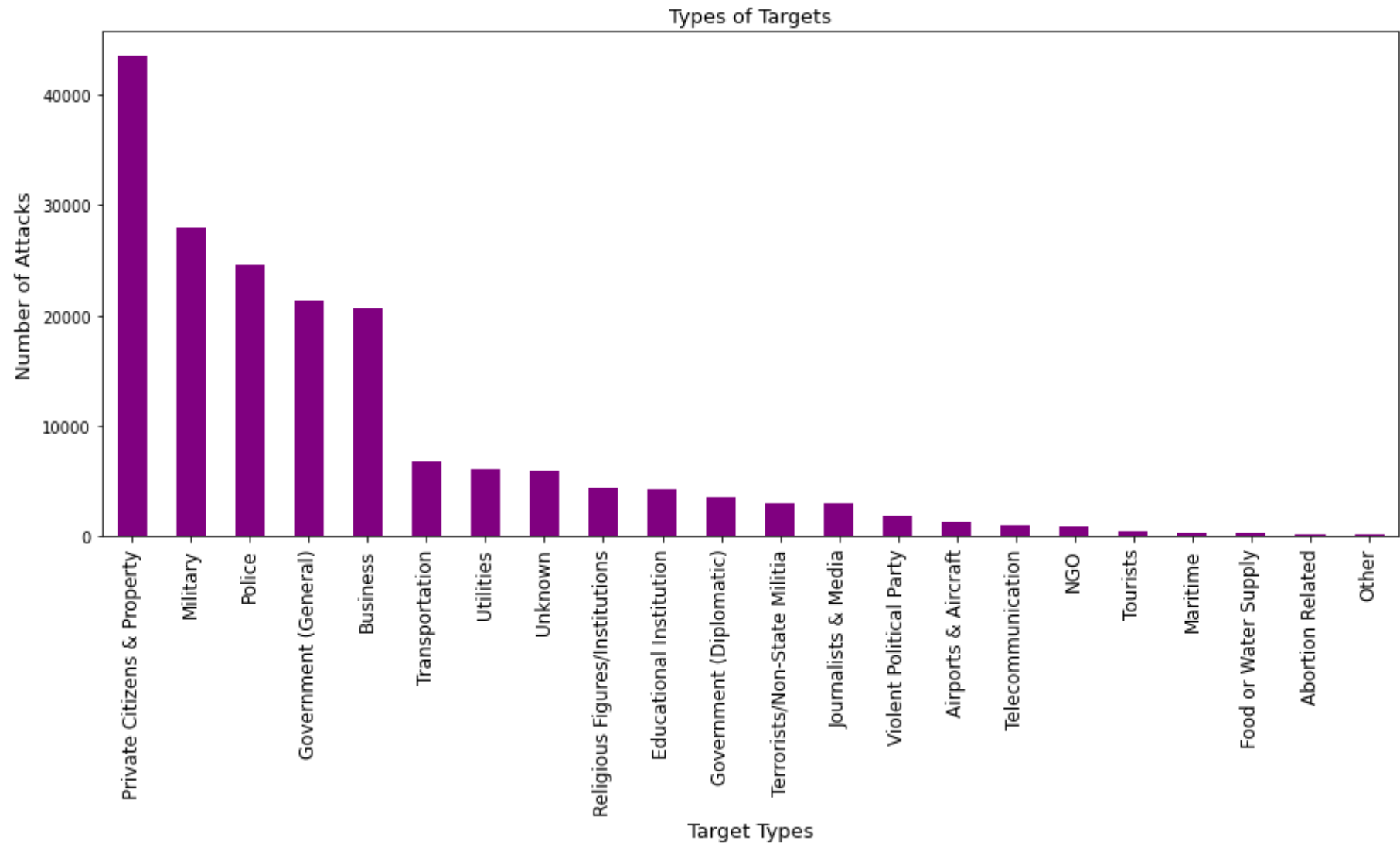
7.Target Type wise Attacks

Number of Attacks over each Target Type

```
In [62]: ta=df["Target Type"].value_counts()  
ta
```

```
Out[62]: Private Citizens & Property    43511
Military                                27984
Police                                 24506
Government (General)                   21283
Business                               20669
Transportation                         6799
Utilities                              6023
Unknown                               5898
Religious Figures/Institutions         4440
Educational Institution                4322
Government (Diplomatic)                 3573
Terrorists/Non-State Militia           3039
Journalists & Media                     2948
Violent Political Party                 1866
Airports & Aircraft                     1343
Telecommunication                      1009
NGO                                     970
Tourists                               440
Maritime                               351
Food or Water Supply                   317
Abortion Related                       263
Other                                  137
Name: Target Type, dtype: int64
```

```
In [63]: ta.plot(kind="bar", color="purple", figsize=(15,6))
plt.title("Types of Targets", fontsize=13)
plt.xlabel("Target Types", fontsize=13)
plt.xticks(fontsize=12)
plt.ylabel("Number of Attacks", fontsize=13)
plt.show()
```



```
In [64]: tc=df[["Target Type","Casualty"]].groupby("Target Type").sum().sort_values(by="Casualty",ascending=False)
tc
```

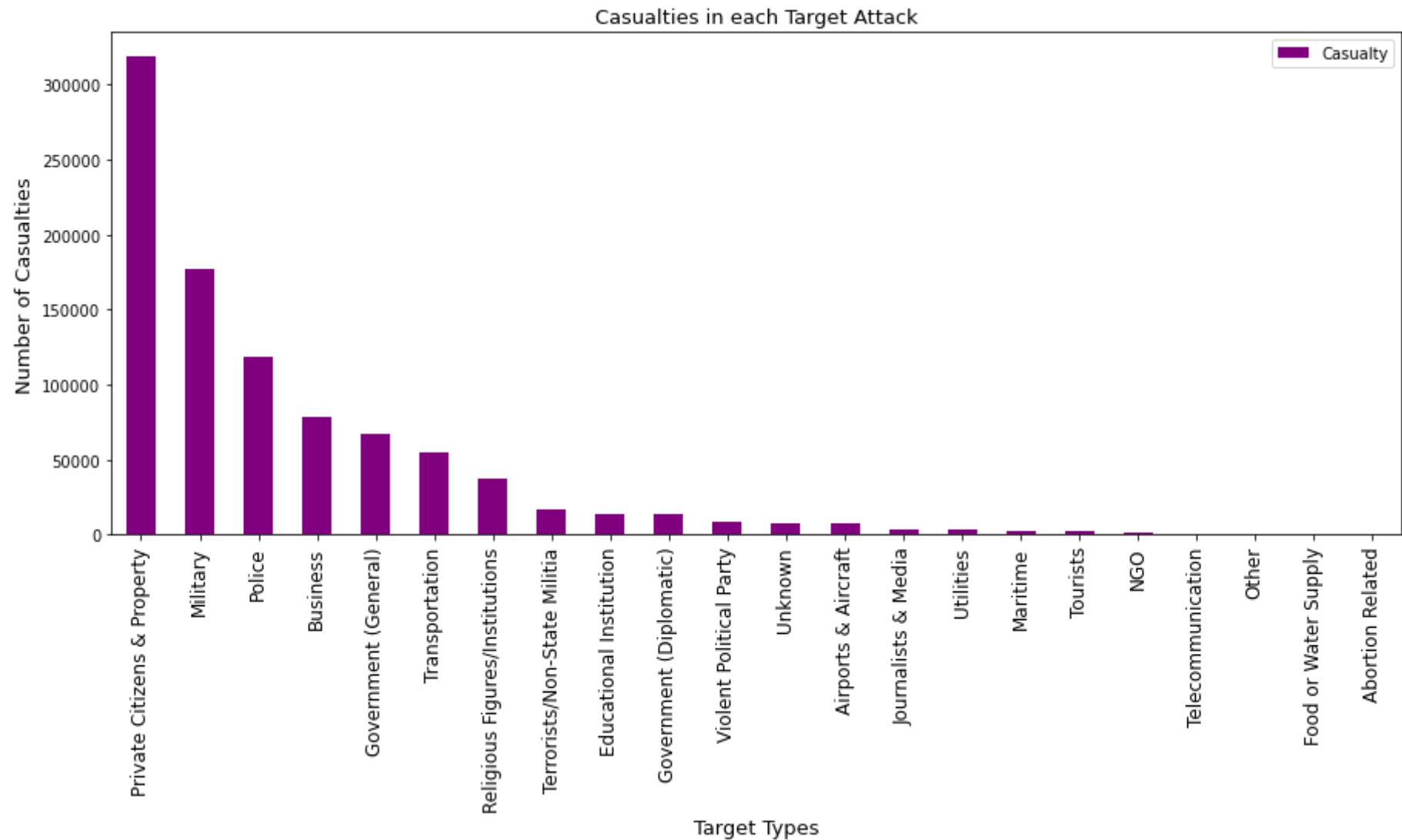
```
Out[64]:
```

Target Type	Casualty
Private Citizens & Property	319176.0

Casualty	
Target Type	
Military	177085.0
Police	118407.0
Business	78018.0
Government (General)	67255.0
Transportation	54595.0
Religious Figures/Institutions	37890.0
Terrorists/Non-State Militia	17311.0
Educational Institution	13972.0
Government (Diplomatic)	13398.0
Violent Political Party	8920.0
Unknown	7888.0
Airports & Aircraft	7245.0
Journalists & Media	3297.0
Utilities	3227.0
Maritime	2099.0
Tourists	2048.0
NGO	1950.0
Telecommunication	679.0
Other	674.0
Food or Water Supply	547.0
Abortion Related	56.0

```
In [65]: tc.plot(kind="bar",color="purple",figsize=(15,6))
plt.title("Casualties in each Target Attack",fontsize=13)
```

```
plt.xlabel("Target Types",fontsize=13)
plt.xticks(fontsize=12)
plt.ylabel("Number of Casualties",fontsize=13)
plt.show()
```



```
In [66]: tk=df[["Target Type","Killed"]].groupby("Target Type").sum().sort_values(by="Killed",ascending=False)
tk
```

Out[66]:

Killed	
Target Type	
Private Citizens & Property	140504.0
Military	106047.0
Police	53704.0
Government (General)	26071.0
Business	23487.0
Transportation	13916.0
Religious Figures/Institutions	13413.0
Terrorists/Non-State Militia	9088.0
Unknown	4329.0
Airports & Aircraft	3767.0
Educational Institution	3745.0
Violent Political Party	3617.0
Government (Diplomatic)	3039.0
Utilities	1874.0
Journalists & Media	1501.0
Maritime	1191.0
NGO	1057.0
Tourists	758.0
Food or Water Supply	313.0
Other	255.0
Telecommunication	182.0
Abortion Related	10.0

```
In [67]: tw=df[["Target Type", "Wounded"]].groupby("Target Type").sum().sort_values(by="Wounded", ascending=False)
tw
```

Out[67]:

Target Type	Wounded
Private Citizens & Property	178672.0
Military	71038.0
Police	64703.0
Business	54531.0
Government (General)	41184.0
Transportation	40679.0
Religious Figures/Institutions	24477.0
Government (Diplomatic)	10359.0
Educational Institution	10227.0
Terrorists/Non-State Militia	8223.0
Violent Political Party	5303.0
Unknown	3559.0
Airports & Aircraft	3478.0
Journalists & Media	1796.0
Utilities	1353.0
Tourists	1290.0
Maritime	908.0
NGO	893.0
Telecommunication	497.0
Other	419.0
Food or Water Supply	234.0

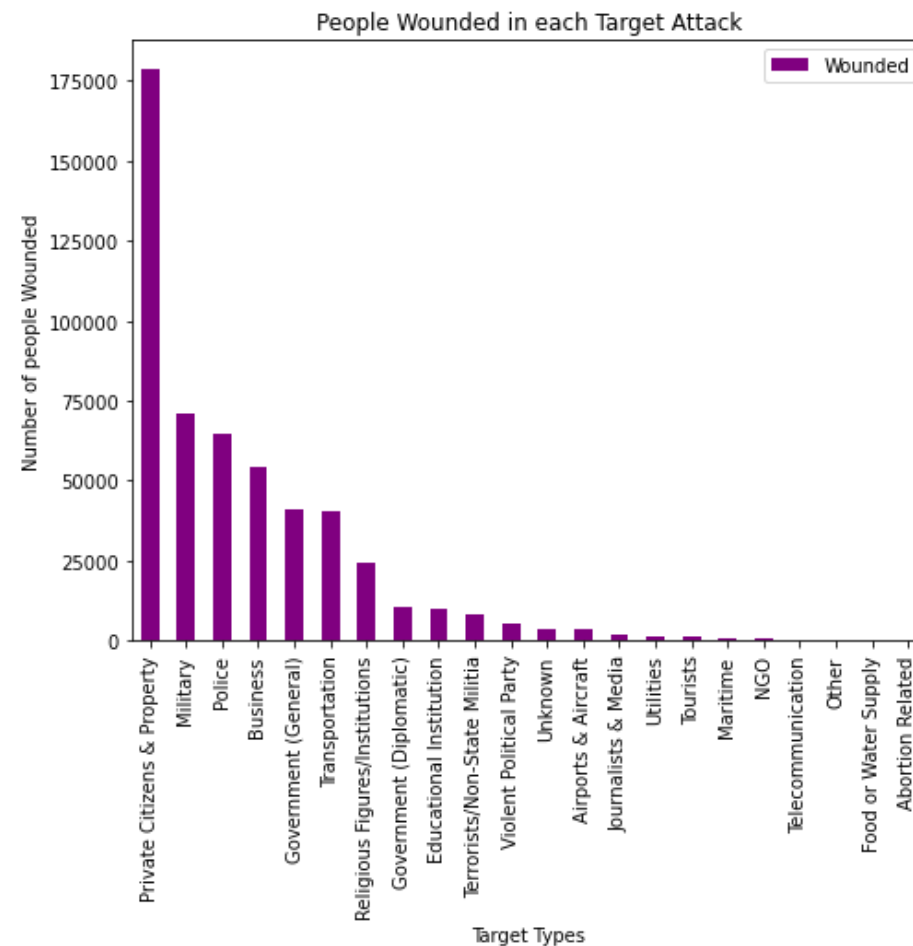
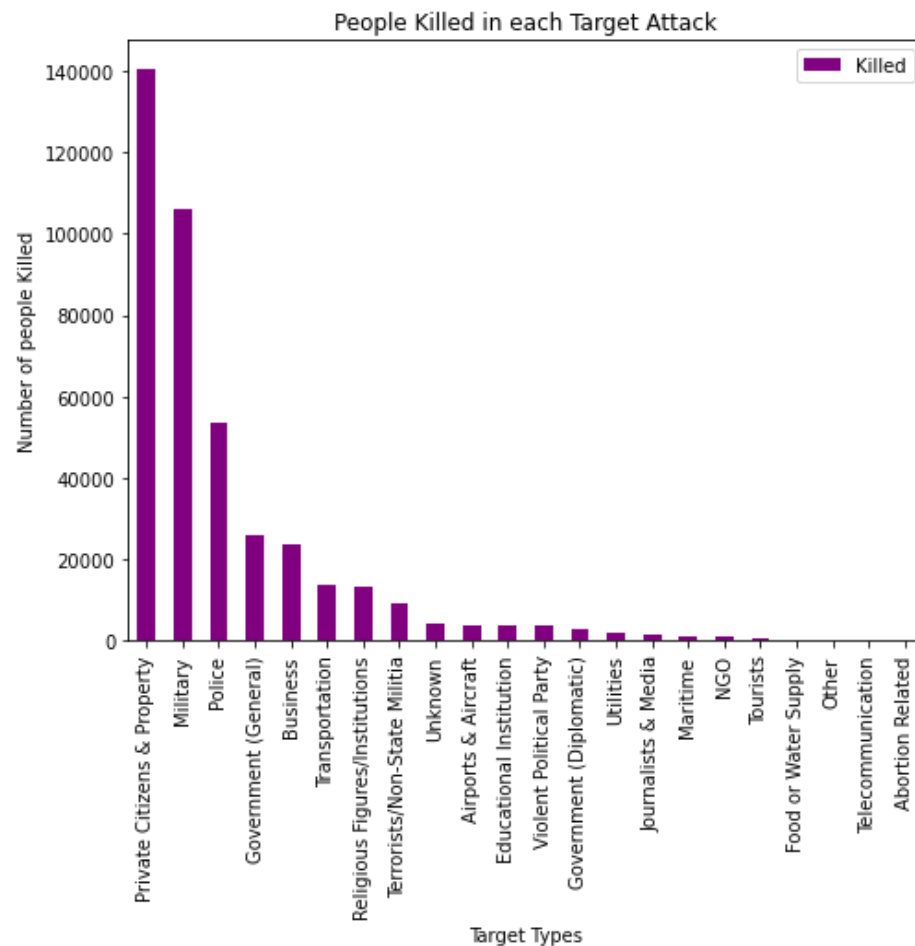
Wounded**Target Type****Abortion Related** 46.0

In [68]:

```
fig=plt.figure()
ax0=fig.add_subplot(1,2,1)
ax1=fig.add_subplot(1,2,2)

#Killed
tk.plot(kind="bar",color="purple",figsize=(17,6),ax=ax0)
ax0.set_title("People Killed in each Target Attack")
ax0.set_xlabel("Target Types")
ax0.set_ylabel("Number of people Killed")

#Wounded
tw.plot(kind="bar",color="purple",figsize=(17,6),ax=ax1)
ax1.set_title("People Wounded in each Target Attack")
ax1.set_xlabel("Target Types")
ax1.set_ylabel("Number of people Wounded")
plt.show()
```



8. Group + Country wise - Top10

Sorting by number of Attacks

```
In [69]: gca=df[["Group Name","Country"]].value_counts().drop("Unknown")
          gca.head(10)
```

C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\generic.py:4153: PerformanceWarning: dropping on a non-lexsorted multi-index without a level parameter may impact performance.

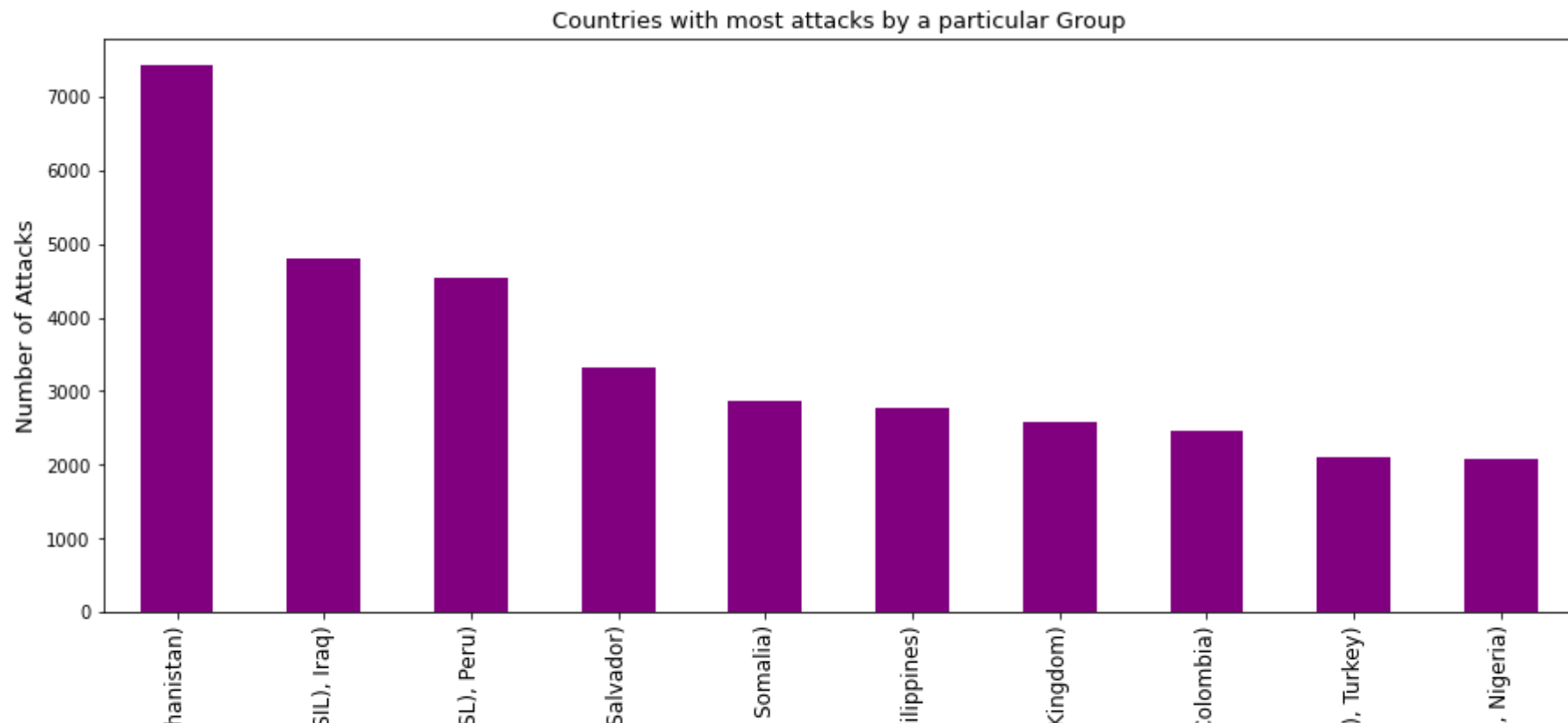
```
obj = obj._drop_axis(labels, axis, level=level, errors=errors)
```

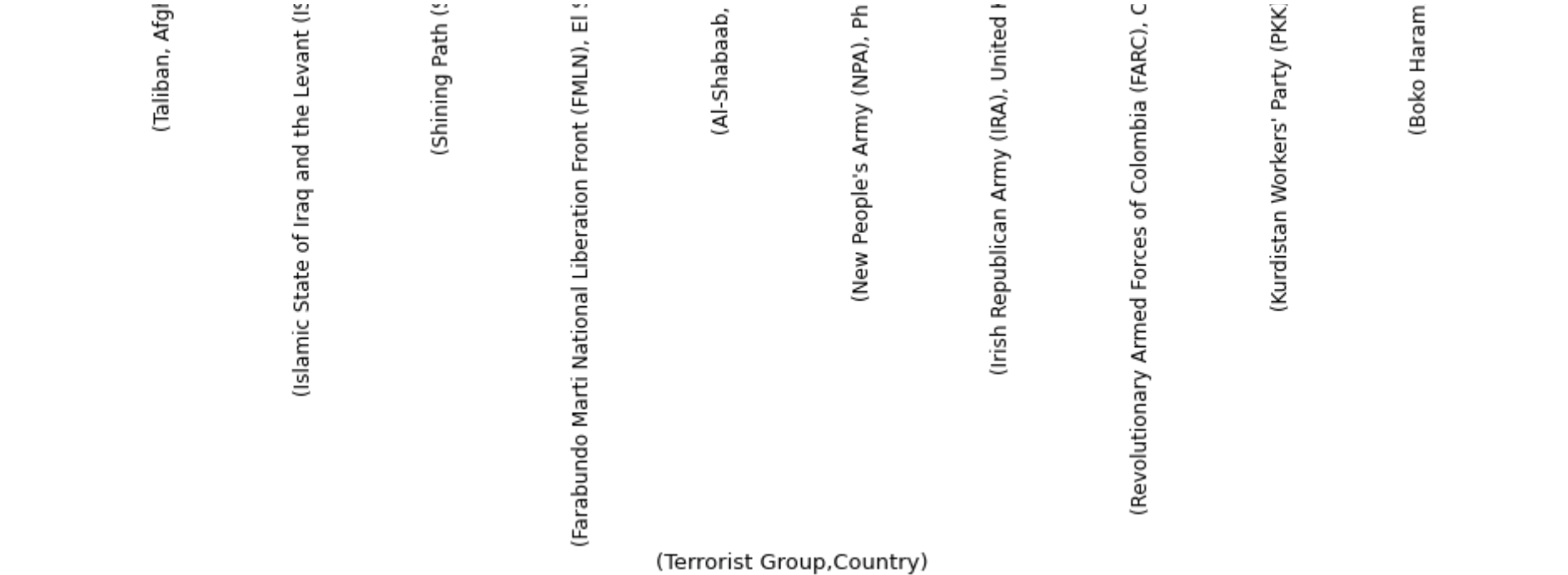
Group Name

Country

```
Out[69]: Taliban Afghanistan 7423
Islamic State of Iraq and the Levant (ISIL) Iraq 4797
Shining Path (SL) Peru 4541
Farabundo Marti National Liberation Front (FMLN) El Salvador 3330
Al-Shabaab Somalia 2867
New People's Army (NPA) Philippines 2770
Irish Republican Army (IRA) United Kingdom 2575
Revolutionary Armed Forces of Colombia (FARC) Colombia 2468
Kurdistan Workers' Party (PKK) Turkey 2109
Boko Haram Nigeria 2087
dtype: int64
```

```
In [70]: gca.head(10).plot(kind="bar",color="purple",figsize=(15,6))
plt.title("Countries with most attacks by a particular Group",fontsize=13)
plt.xlabel("(Terrorist Group,Country)",fontsize=13)
plt.xticks(fontsize=12)
plt.ylabel("Number of Attacks",fontsize=13)
plt.show()
```





Sorting by Number of Casualties

```
In [71]: gcc=df[["Group Name","Country","Casualty"]].groupby(["Group Name","Country"],axis=0).sum().sort_values(by="Casualty",ascending=False)
gcc
```

Out[71]:

		Casualty
Group Name	Country	
Taliban	Afghanistan	57140.0
Islamic State of Iraq and the Levant (ISIL)	Iraq	54755.0
Boko Haram	Nigeria	24588.0
Liberation Tigers of Tamil Eelam (LTTE)	Sri Lanka	21919.0
Al-Qaida	United States	19494.0
Tehrik-i-Taliban Pakistan (TTP)	Pakistan	15532.0
Shining Path (SL)	Peru	14625.0

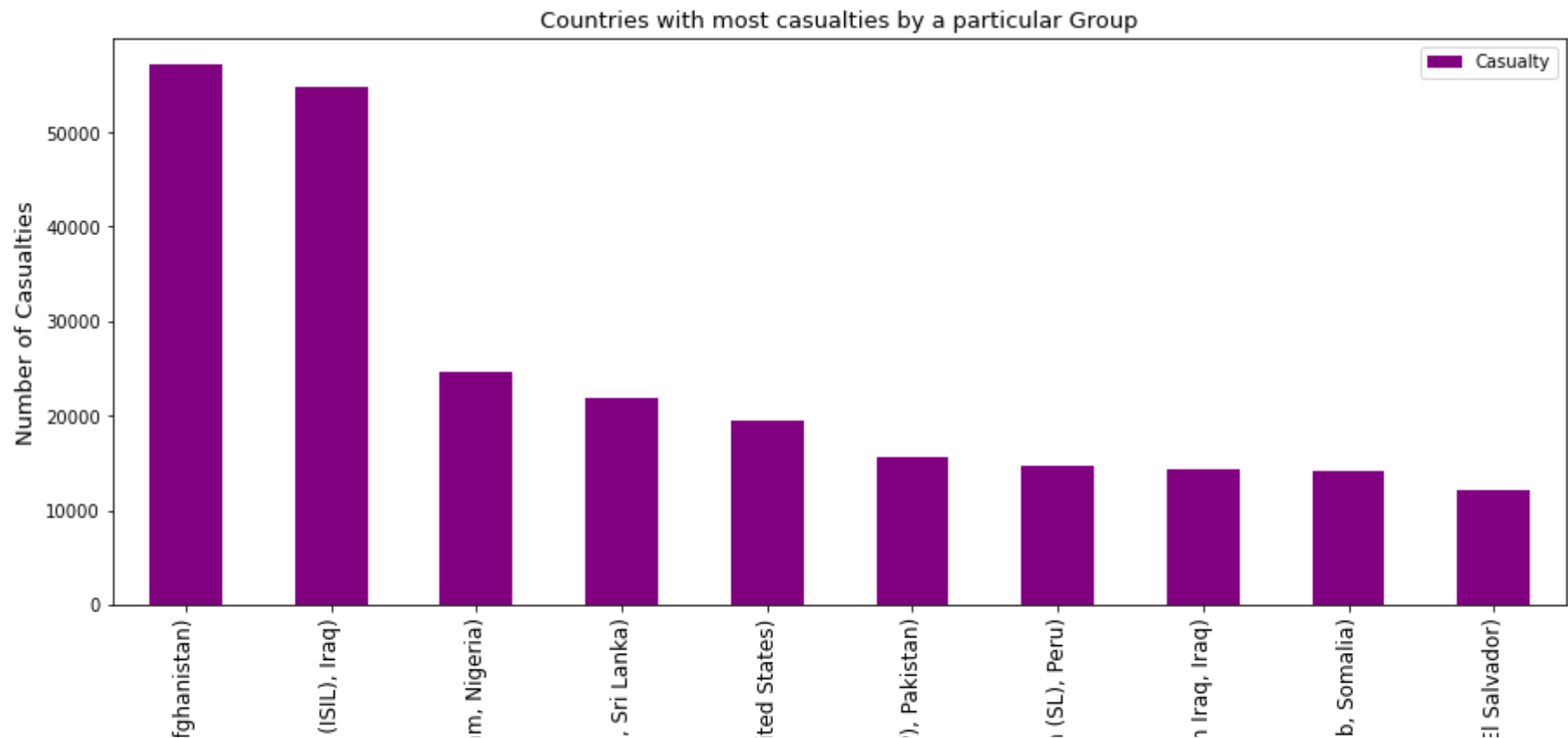
Casualty		
Group Name	Country	
Al-Qaida in Iraq	Iraq	14348.0
Al-Shabaab	Somalia	14201.0
Farabundo Marti National Liberation Front (FMLN)	El Salvador	12068.0

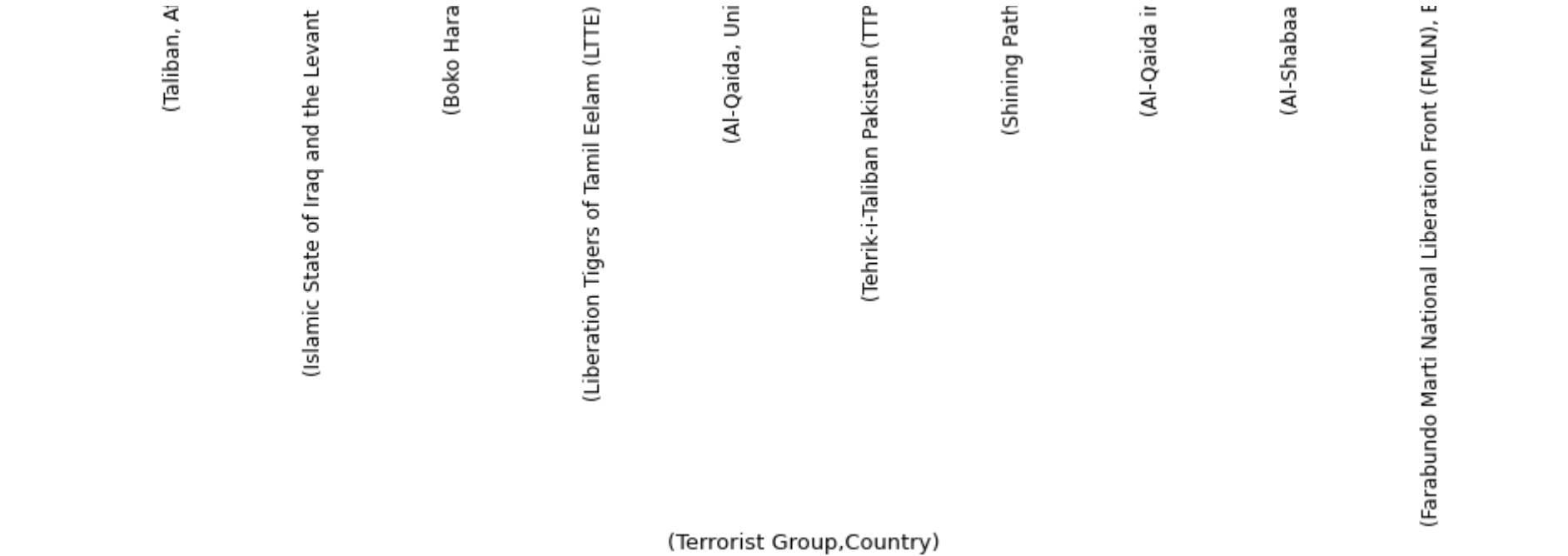
In [72]:

```

gcc.plot(kind="bar", color="purple", figsize=(15,6))
plt.title("Countries with most casualties by a particular Group", fontsize=13)
plt.xlabel("(Terrorist Group,Country)", fontsize=13)
plt.xticks(fontsize=12)
plt.ylabel("Number of Casualties", fontsize=13)
plt.show()

```





Sorting by Number of People Killed

```
In [73]: gck=df[["Group Name","Country","Killed"]].groupby(["Group Name","Country"],axis=0).sum().sort_values(by="Killed",ascending=False).gck
```

C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\generic.py:4153: PerformanceWarning: dropping on a non-lexsorted multi-index without a level parameter may impact performance.
obj = obj._drop_axis(labels, axis, level=level, errors=errors)

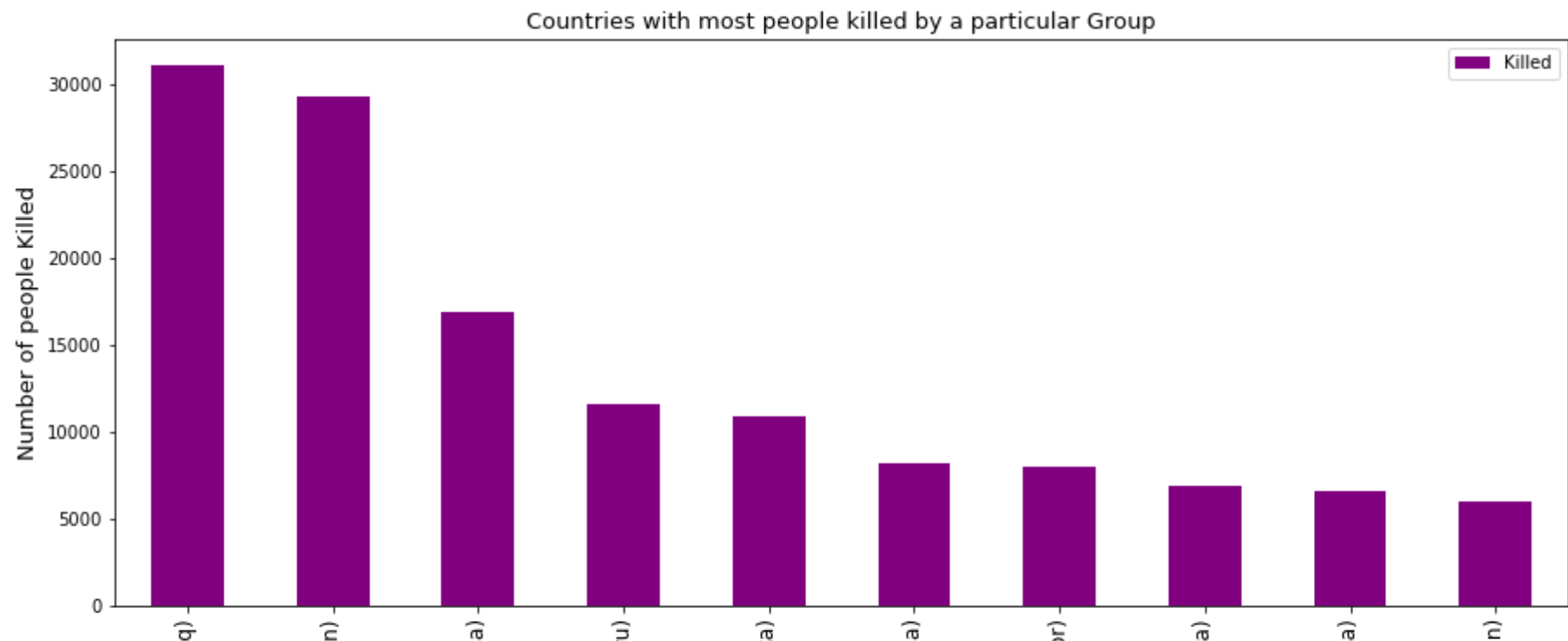
Out[73]:

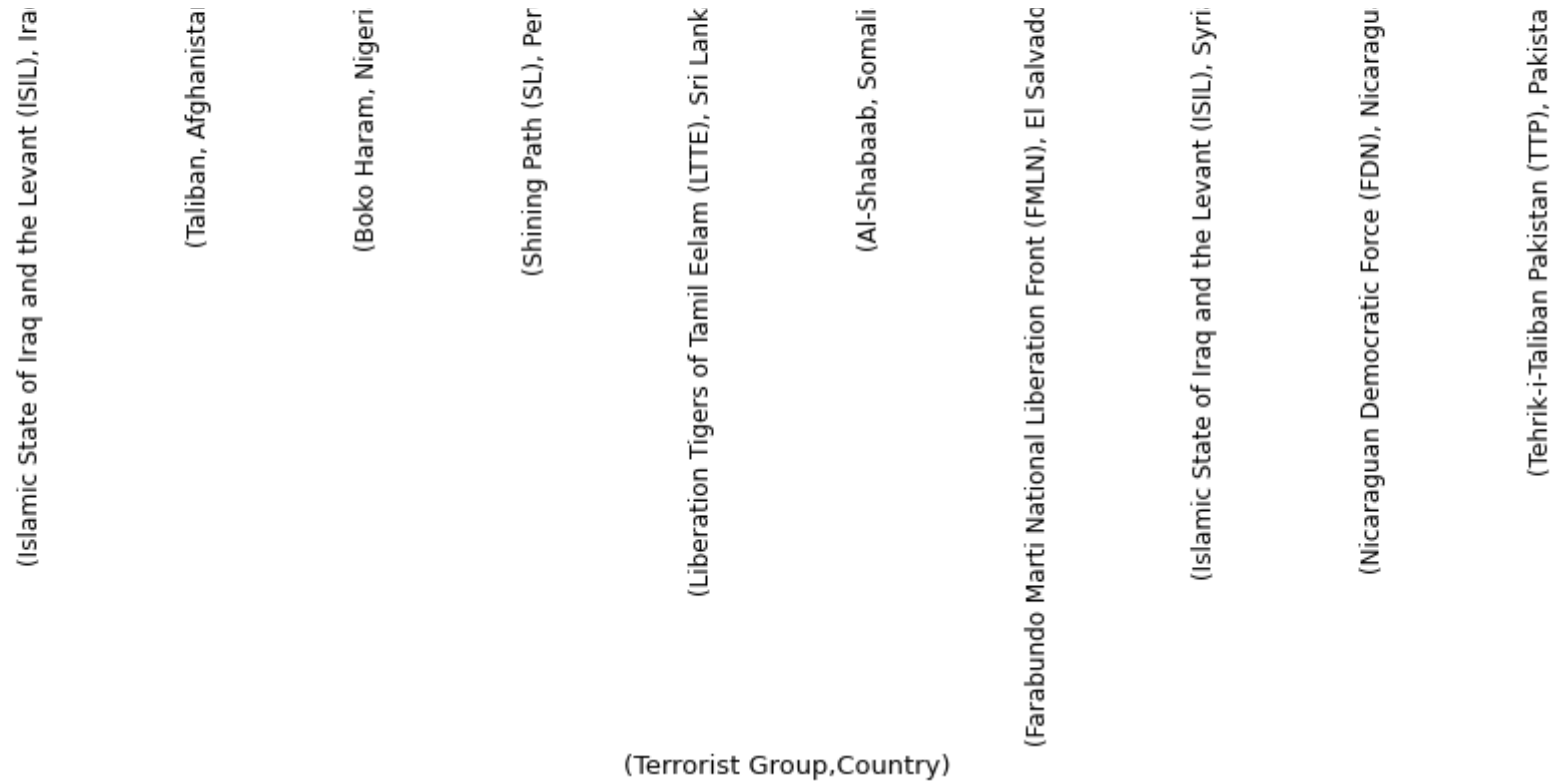
		Killed
Group Name	Country	
Islamic State of Iraq and the Levant (ISIL)	Iraq	31058.0
Taliban	Afghanistan	29269.0
Boko Haram	Nigeria	16917.0
Shining Path (SL)	Peru	11595.0
Liberation Tigers of Tamil Eelam (LTTE)	Sri Lanka	10928.0

		Killed
Group Name	Country	
Al-Shabaab	Somalia	8176.0
Farabundo Marti National Liberation Front (FMLN)	El Salvador	8019.0
Islamic State of Iraq and the Levant (ISIL)	Syria	6883.0
Nicaraguan Democratic Force (FDN)	Nicaragua	6630.0
Tehrik-i-Taliban Pakistan (TTP)	Pakistan	6014.0

In [74]:

```
gck.plot(kind="bar", color="purple", figsize=(15,6))
plt.title("Countries with most people killed by a particular Group", fontsize=13)
plt.xlabel("(Terrorist Group,Country)", fontsize=13)
plt.xticks(fontsize=12)
plt.ylabel("Number of people Killed", fontsize=13)
plt.show()
```





Sorting by Number of People Wounded

In [75]:

```
gcw=df[["Group Name", "Country", "Wounded"]].groupby(["Group Name", "Country"],axis=0).sum().sort_values(by="Wounded",ascending=False)
gcw
```

C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\generic.py:4153: PerformanceWarning: dropping on a non-lexsorted multi-index without a level parameter may impact performance.

```
obj = obj._drop_axis(labels, axis, level=level, errors=errors)
```

Out[75]:

		Wounded
Group Name	Country	
Taliban	Afghanistan	27871.0
Islamic State of Iraq and the Levant (ISIL)	Iraq	23697.0
Al-Qaida	United States	16493.0

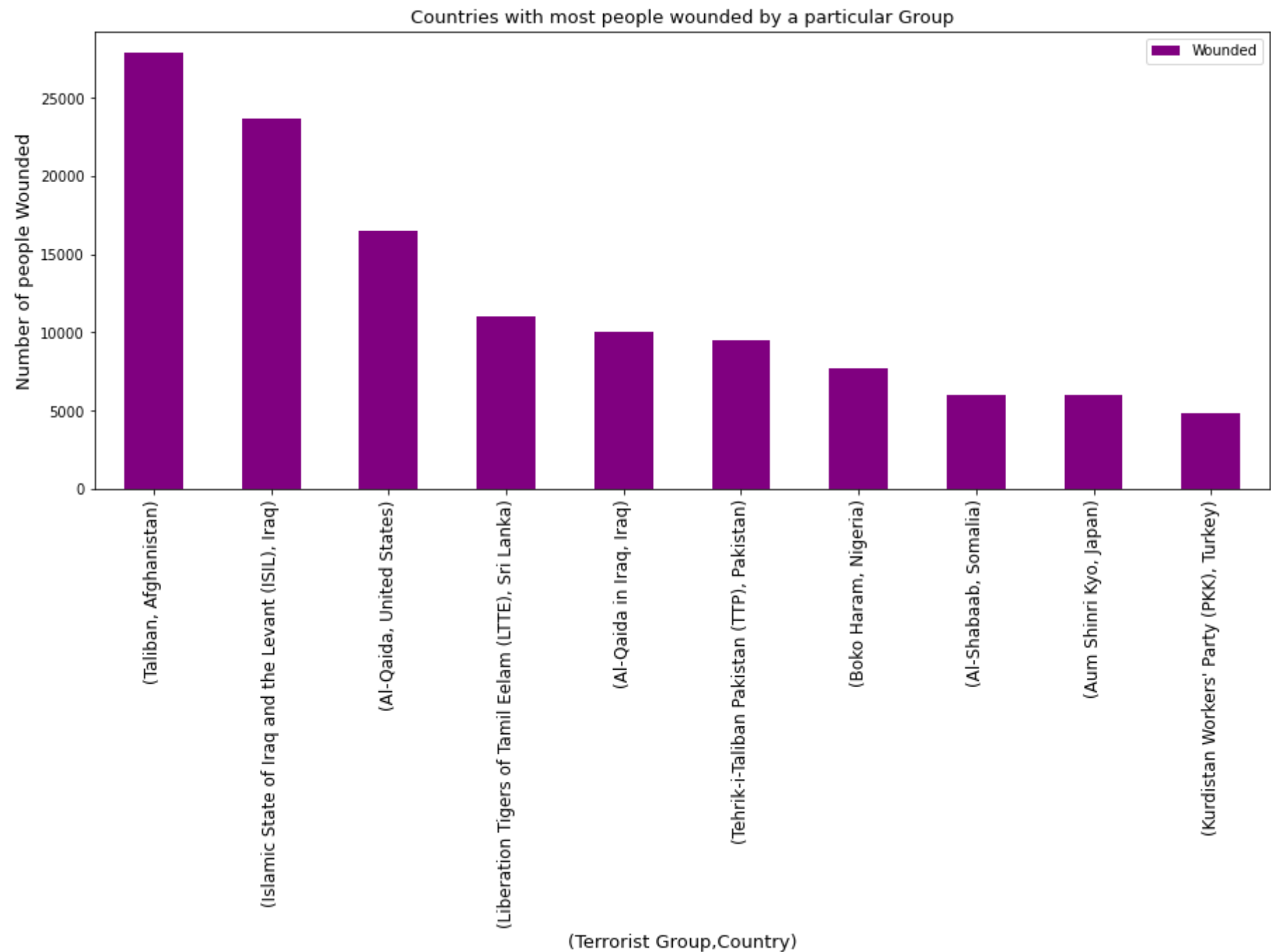
		Wounded
Group Name	Country	
Liberation Tigers of Tamil Eelam (LTTE)	Sri Lanka	10991.0
Al-Qaida in Iraq	Iraq	10075.0
Tehrik-i-Taliban Pakistan (TTP)	Pakistan	9518.0
Boko Haram	Nigeria	7671.0
Al-Shabaab	Somalia	6025.0
Aum Shinri Kyo	Japan	6003.0
Kurdistan Workers' Party (PKK)	Turkey	4795.0

In [76]:

```

gcw.plot(kind="bar",color="purple",figsize=(15,6))
plt.title("Countries with most people wounded by a particular Group",fontsize=13)
plt.xlabel("(Terrorist Group,Country)",fontsize=13)
plt.xticks(fontsize=12)
plt.ylabel("Number of people Wounded",fontsize=13)
plt.show()

```



9. Humanity Affected (World-wide) by Terrorist Attacks from 1970 to 2017

Total Casualties (Killed + Wounded) due to Terrorist Attacks

```
In [77]: casualty=df.loc[:, "Casualty"].sum()  
print("Total number of Casualties due to Terrorist Attacks from 1970 to 2017 across the world :\n",casualty)
```

```
Total number of Casualties due to Terrorist Attacks from 1970 to 2017 across the world :  
935737.0
```

Killed due to Terrorist Attacks

```
In [78]: kill=df.loc[:, "Killed"].sum()  
print("Total number of people killed due to Terrorist Attacks from 1970 to 2017 across the world :\n",kill)
```

```
Total number of people killed due to Terrorist Attacks from 1970 to 2017 across the world :  
411868.0
```

Wounded due to Terrorist Attacks

```
In [79]: wound=df.loc[:, "Wounded"].sum()  
print("Total number of people killed due to Terrorist Attacks from 1970 to 2017 across the world :\n",wound)
```

```
Total number of people killed due to Terrorist Attacks from 1970 to 2017 across the world :  
523869.0
```

```
In [ ]:
```