

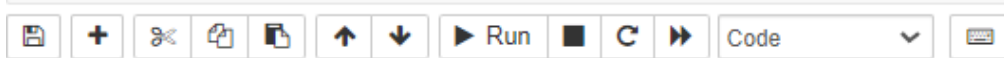
```
In [184]: 1 import numpy as np
          2 import pandas as pd
```

```
In [185]: 1 data=pd.read_csv('fatalities_isr_pse_conflict_2000_to_2023.csv')
```

```
In [186]: 1 data.head()
```

```
Out[186]:
```

	name	date_of_event	age	citizenship	event_location	event_location_district	event_location_region	date_of_death	gender	took_part_in_the_hostilities
0	'Abd a-Rahman Suleiman Muhammad Abu Daghash	2023-09-24	32.0	Palestinian	Nur Shams R.C.	Tulkarm	West Bank	2023-09-24	M	NaN
1	Usayed Farhan Muhammad 'Ali Abu 'Ali	2023-09-24	21.0	Palestinian	Nur Shams R.C.	Tulkarm	West Bank	2023-09-24	M	NaN
2	'Abdallah 'Imad Sa'ed Abu Hassan	2023-09-22	16.0	Palestinian	Kfar Dan	Jenin	West Bank	2023-09-22	M	NaN
3	Durgham Muhammad Yihya al-Akhras	2023-09-20	19.0	Palestinian	'Aqbat Jaber R.C.	Jericho	West Bank	2023-09-20	M	NaN
4	Raafat 'Omar Ahmad Khamaisah	2023-09-19	15.0	Palestinian	Jenin R.C.	Jenin	West Bank	2023-09-19	M	NaN



# Data Cleaning and Statistics

In [187]:

1 data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 11124 entries, 0 to 11123
Data columns (total 16 columns):
#   Column                Non-Null Count  Dtype
---  ---
0   name                   11124 non-null  object
1   date_of_event          11124 non-null  object
2   age                    10995 non-null  float64
3   citizenship            11124 non-null  object
4   event_location         11124 non-null  object
5   event_location_district 11124 non-null  object
6   event_location_region  11124 non-null  object
7   date_of_death          11124 non-null  object
8   gender                 11104 non-null  object
9   took_part_in_the_hostilities 9694 non-null  object
10  place_of_residence      11056 non-null  object
11  place_of_residence_district 11056 non-null  object
12  type_of_injury          10833 non-null  object
13  ammunition              5871 non-null  object
14  killed_by              11124 non-null  object
15  notes                  10844 non-null  object
dtypes: float64(1), object(15)
memory usage: 1.4+ MB
```

In [188]:

1 data.describe()

Out[188]:

	age
count	10995.000000
mean	26.745703
std	13.780548
min	1.000000
25%	19.000000
50%	23.000000
75%	31.000000
max	112.000000



```
In [189]: 1 data.isna().sum() #CHECKING NULL VALUES
```

```
Out[189]: name          0
date_of_event         0
age          129
citizenship          0
event_location        0
event_location_district  0
event_location_region  0
date_of_death         0
gender              20
took_part_in_the_hostilities 1430
place_of_residence     68
place_of_residence_district 68
type_of_injury         291
ammunition            5253
killed_by             0
notes                280
dtype: int64
```

```
In [190]: 1 data.drop('notes', axis=1, inplace=True)
```

```
In [191]: 1 #REPLACING NULL VALUES
2
3 data['age']=data['age'].fillna(data['age'].mean())
4 data['gender']=data['gender'].fillna(data['gender'].mode()[0])
5 data['type_of_injury']=data['type_of_injury'].fillna(data['type_of_injury'].mode()[0])
6 data['took_part_in_the_hostilities']=data['took_part_in_the_hostilities'].fillna(data['took_part_in_the_hostilities'].mode()[0])
7 data['place_of_residence']=data['place_of_residence'].fillna(data['place_of_residence'].mode()[0])
8 data['place_of_residence_district']=data['place_of_residence_district'].fillna(data['place_of_residence_district'].mode()[0])
```



In [192]: 1 data.isna().sum()

```
Out[192]: name          0
date_of_event          0
age                    0
citizenship            0
event_location          0
event_location_district 0
event_location_region   0
date_of_death           0
gender                 0
took_part_in_the_hostilities 0
place_of_residence      0
place_of_residence_district 0
type_of_injury          0
ammunition             5253
killed_by              0
dtype: int64
```

```
In [193]: 1 gender_counts = data['gender'].value_counts()    #Genderwise Count
          2 gender_counts
```

```
Out[193]: M    9701
          F    1423
          Name: gender, dtype: int64
```

```
In [194]: 1 #Mean_Age
          2
          3 # Calculate the mean age
          4 mean_age = data['age'].mean()
          5
          6 # Print the mean age
          7 print(f"Mean Age: {mean_age:.2f}")
          8
```

Mean Age: 26.75

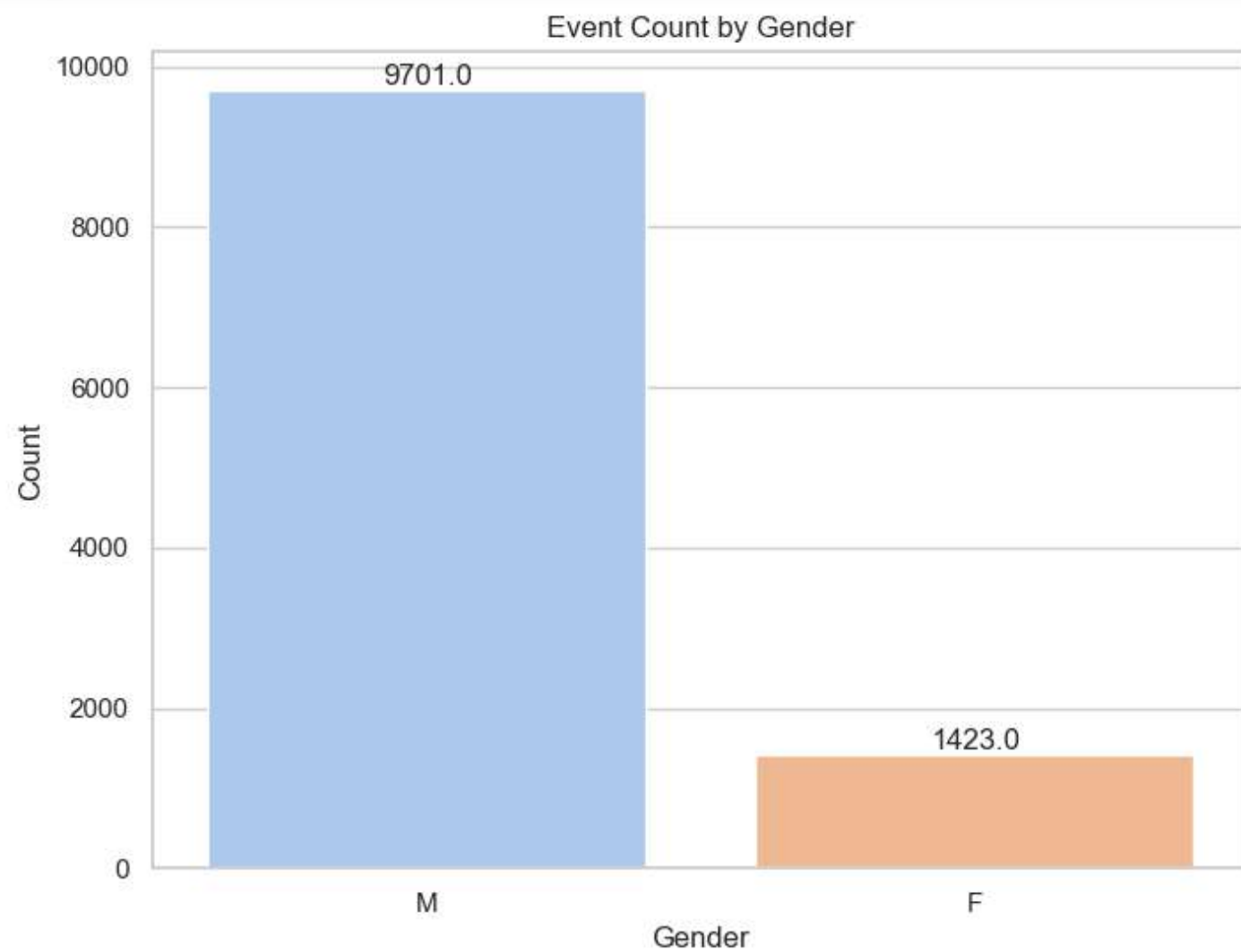
## Data Visualization

In [195]:

```
1 import seaborn as sns
2 import matplotlib.pyplot as plt
3 sns.set(color_codes=True)
4 %matplotlib inline
```

In [196]:

```
1 # bar chart for genderwise event count
2
3 plt.figure(figsize=(8, 6))
4 sns.set(style="whitegrid")
5 ax = sns.countplot(data=data, x='gender', palette='pastel')
6
7 # Annotate the count on top of each bar
8 for p in ax.patches:
9     ax.annotate(f'{p.get_height()}', (p.get_x() + p.get_width() / 2., p.get_height()), ha='center', va='bottom')
10
11 plt.xlabel('Gender')
12 plt.ylabel('Count')
13 plt.title('Event Count by Gender')
14 plt.show()
15
```

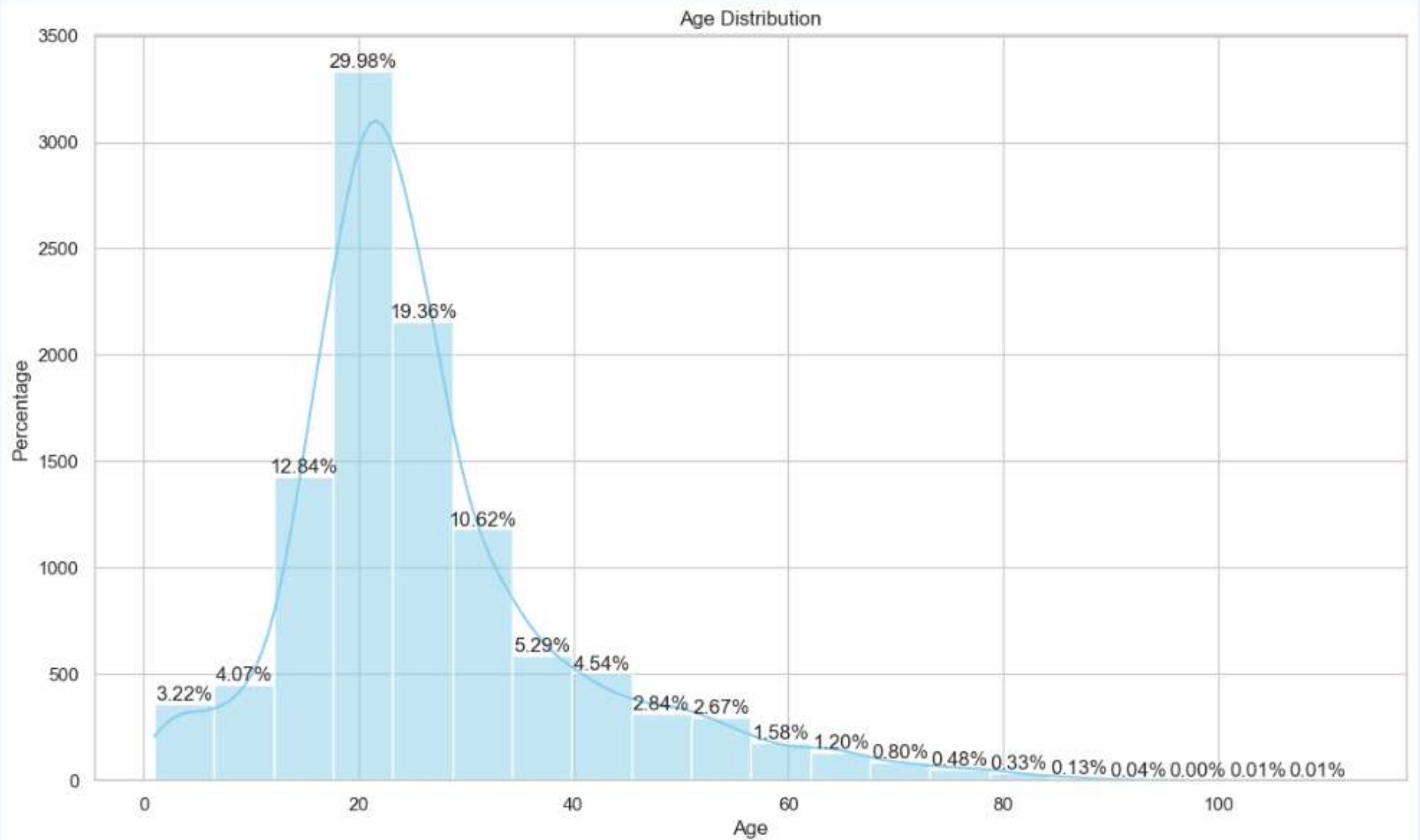


Inference

87% are Male and 13% are female died in the Fatalities.



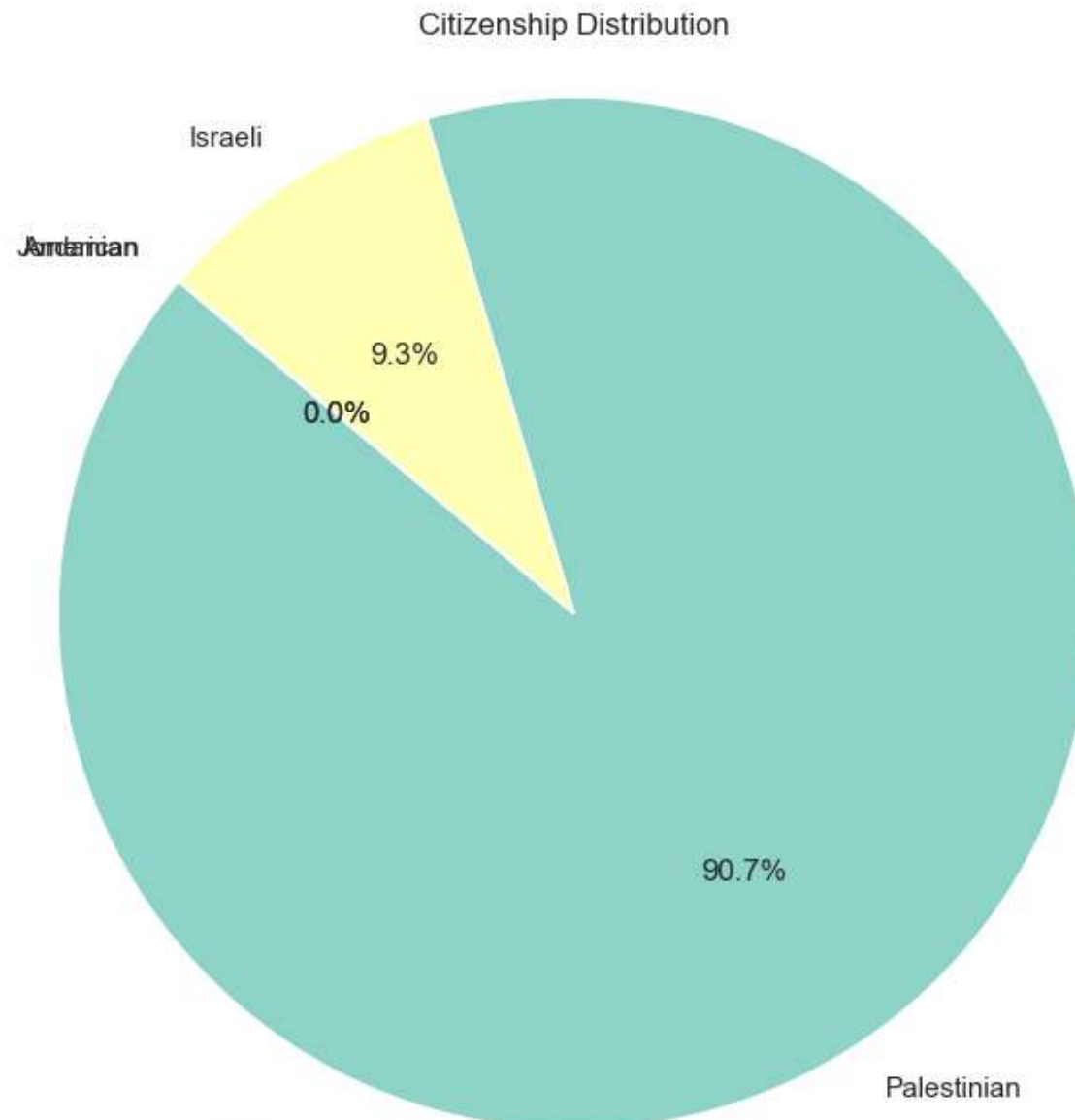
```
In [197]: 1 #Histogram for Age Distribution
2
3 #plt.figure(figsize=(8, 6))
4 #sns.histplot(data=data, x='age', bins=20, kde=True, color='skyblue')
5 #plt.xlabel('Age')
6 #plt.ylabel('Frequency')
7 #plt.title('Age Distribution')
8 #plt.show()
9
10 #OR
11
12 # Create the histogram
13 plt.figure(figsize=(14, 8))
14 sns.set_style("whitegrid")
15 ax = sns.histplot(data=data, x='age', bins=20, kde=True, color='skyblue')
16
17 plt.xlabel('Age')
18 plt.ylabel('Percentage')
19 plt.title('Age Distribution')
20
21 # Annotate each bar with the percentage
22 total_count = len(data['age'])
23 for p in ax.patches:
24     percentage = (p.get_height() / total_count) * 100
25     ax.annotate(f'{percentage:.2f}%', (p.get_x() + p.get_width() / 2., p.get_height()), ha='center', va='bottom')
26
27 plt.show()
28
29
```







```
In [198]: 1 #Pie Chart for Citizenship Distribution
2
3 citizenship_counts = data['citizenship'].value_counts()
4 plt.figure(figsize=(8, 8))
5 plt.pie(citizenship_counts, labels=citizenship_counts.index, autopct='%1.1f%%', startangle=140, colors=sns.color_palette('Set3'))
6 plt.axis('equal')
7 plt.title("Citizenship Distribution")
8 plt.show()
```



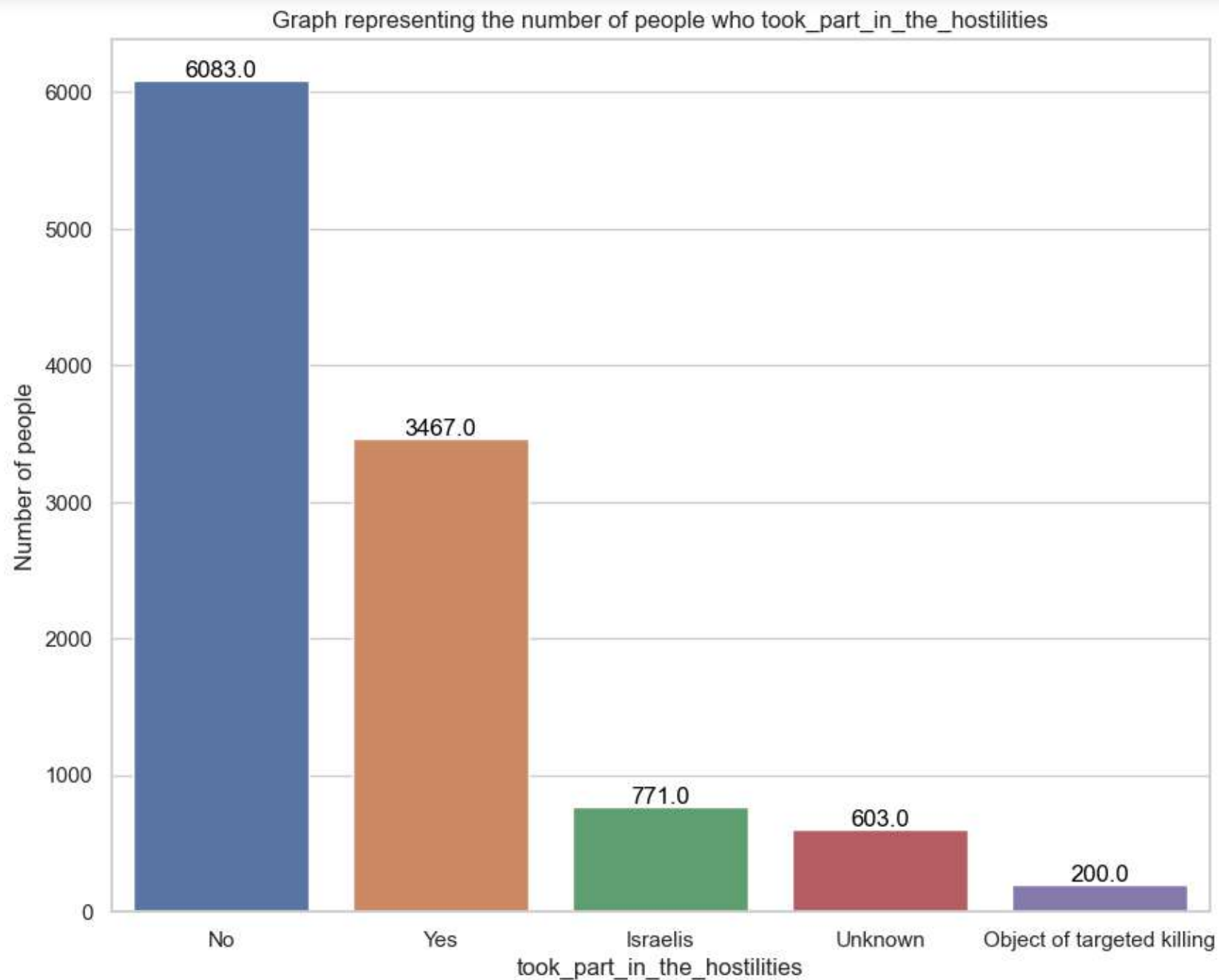


```
In [213]: 1 American_Citizen_Count = data[data['citizenship'] == 'American']['citizenship'].count()
2 Jordanian_Citizen_Count = data[data['citizenship'] == 'Jordanian']['citizenship'].count()
3
4 print("American_Citizen_Count:", American_count)
5 print("Jordanian_Citizen_Countt:", Jordanian_count)
```

American\_Citizen\_Count: 1  
Jordanian\_Citizen\_Countt: 2

```
In [200]: 1 took_part_in_the_hostilities = data.took_part_in_the_hostilities.value_counts().reset_index()
2 took_part_in_the_hostilities.columns = ['took_part_in_the_hostilities', 'count']
```

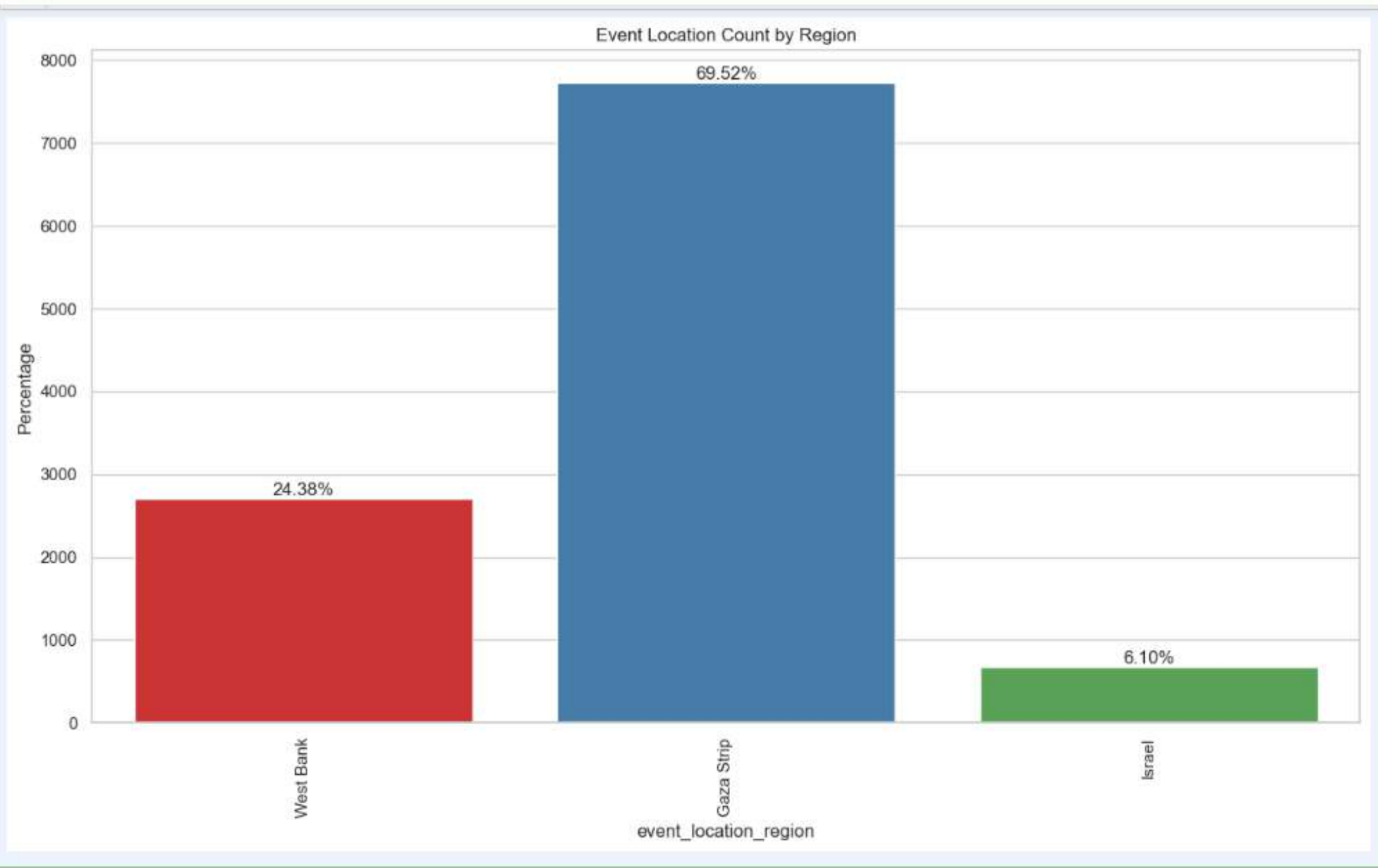
```
In [214]: 1 # Histogram for the 'took_part_in_the_hostilities' column
2
3 #plt.figure(figsize=(10, 6))
4 #sns.histplot(data['took_part_in_the_hostilities'])
5 #plt.title('Hostilities Distribution')
6 #plt.ylabel('Frequency')
7 #plt.show()
8 #for p in ax.patches:
9     # ax.annotate(f'{p.get_height()}', (p.get_x() + p.get_width() / 2., p.get_height()),
10     #             ha='center', va='center', fontsize=12, color='black', xytext=(0, 5), textcoords='offset points')
11 #plt.show()
12
13 #OR
14
15 plt.figure(figsize = (10,8))
16 ax = sns.barplot(x = 'took_part_in_the_hostilities', y = 'count', data = took_part_in_the_hostilities)
17 plt.xlabel('took_part_in_the_hostilities')
18 plt.ylabel('Number of people')
19 plt.title('Graph representing the number of people who took_part_in_the_hostilities')
20 for p in ax.patches:
21     ax.annotate(f'{p.get_height()}', (p.get_x() + p.get_width() / 2., p.get_height()),
22               ha='center', va='center', fontsize=12, color='black', xytext=(0, 5), textcoords='offset points')
23 plt.show()
```



31% are taking part in the Hostilities 54.7% are not taking part in the Hostilities



```
In [216]: 1 #Event Location Count by Region
2
3 # Create the count plot
4 plt.figure(figsize=(15, 8))
5 sns.set_style("whitegrid")
6 ax = sns.countplot(data=data, x='event_location_region', palette='Set1')
7
8 plt.xticks(rotation=90)
9 plt.ylabel('Percentage')
10 plt.title('Event Location Count by Region')
11
12 # Annotate each bar with the percentage
13 total_count = len(data['event_location_region'])
14 for p in ax.patches:
15     percentage = (p.get_height() / total_count) * 100
16     ax.annotate(f'{percentage:.2f}%', (p.get_x() + p.get_width() / 2., p.get_height()), ha='center', va='bottom')
17
18 plt.show()
19
```



Inference

We can say Gaza city face the most number of casualties



Code



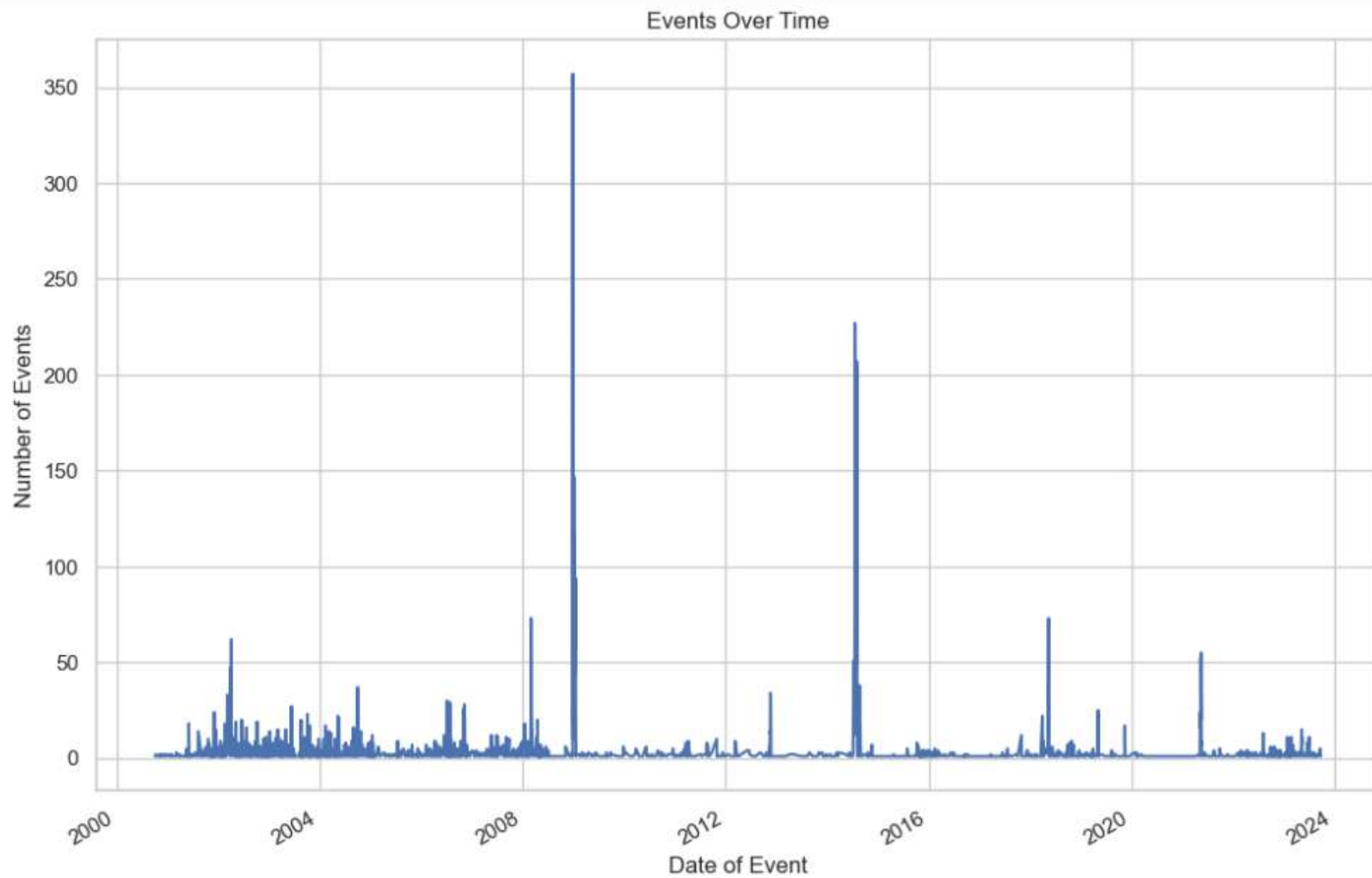
## Time Series Analysis

In [203]:

```
1 # Converting date columns to datetime
2
3 data['date_of_event'] = pd.to_datetime(data['date_of_event'])
4 data['date_of_death'] = pd.to_datetime(data['date_of_death'])
```

In [204]:

```
1 #Calculating the number of events per day
2
3 events_per_day = data.groupby('date_of_event').size()
4
5 plt.figure(figsize=(12, 8))
6 events_per_day.plot()
7 plt.title('Events Over Time')
8 plt.xlabel('Date of Event')
9 plt.ylabel('Number of Events')
10 plt.show()
```



File Edit View Insert Cell Kernel Help

Not Trusted



Python 3 (ipykernel)



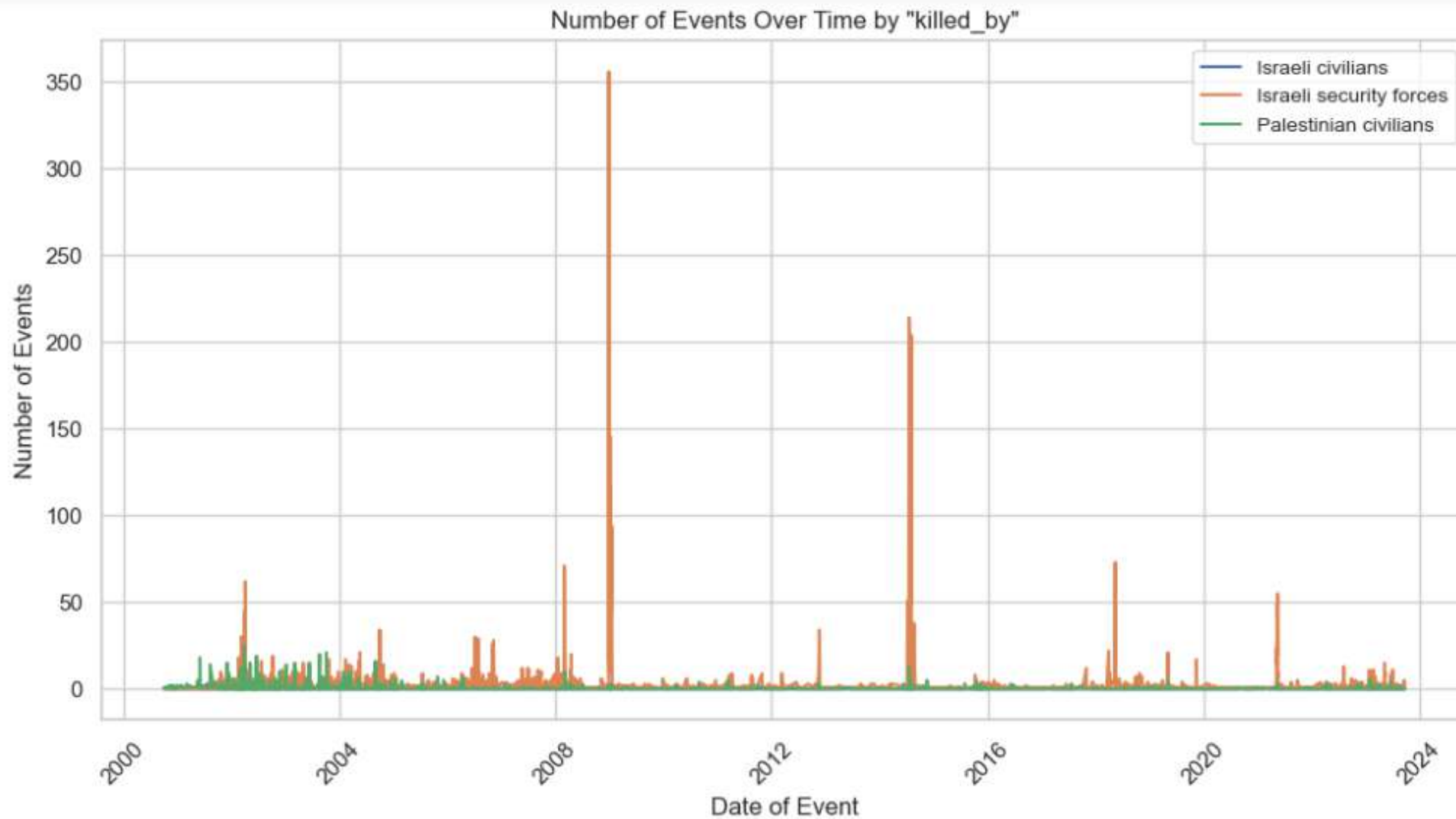
Code



In [205]:

```
1 # Grouping the 'date_of_event' and 'killed_by' column and calculating
2 #the count of events for each combination
3
4 events_by_killed_by = data.groupby(['date_of_event', 'killed_by']).size().unstack().fillna(0)
5
6 plt.figure(figsize=(12, 6))
7 for column in events_by_killed_by.columns:
8     plt.plot(events_by_killed_by.index, events_by_killed_by[column], label=column)
9
10 plt.title('Number of Events Over Time by "killed_by"')
11 plt.xlabel('Date of Event')
12 plt.ylabel('Number of Events')
13 plt.legend(loc='upper right', fontsize='small')
14 plt.xticks(rotation=45)
15 plt.show()
```







## Geological Analysis

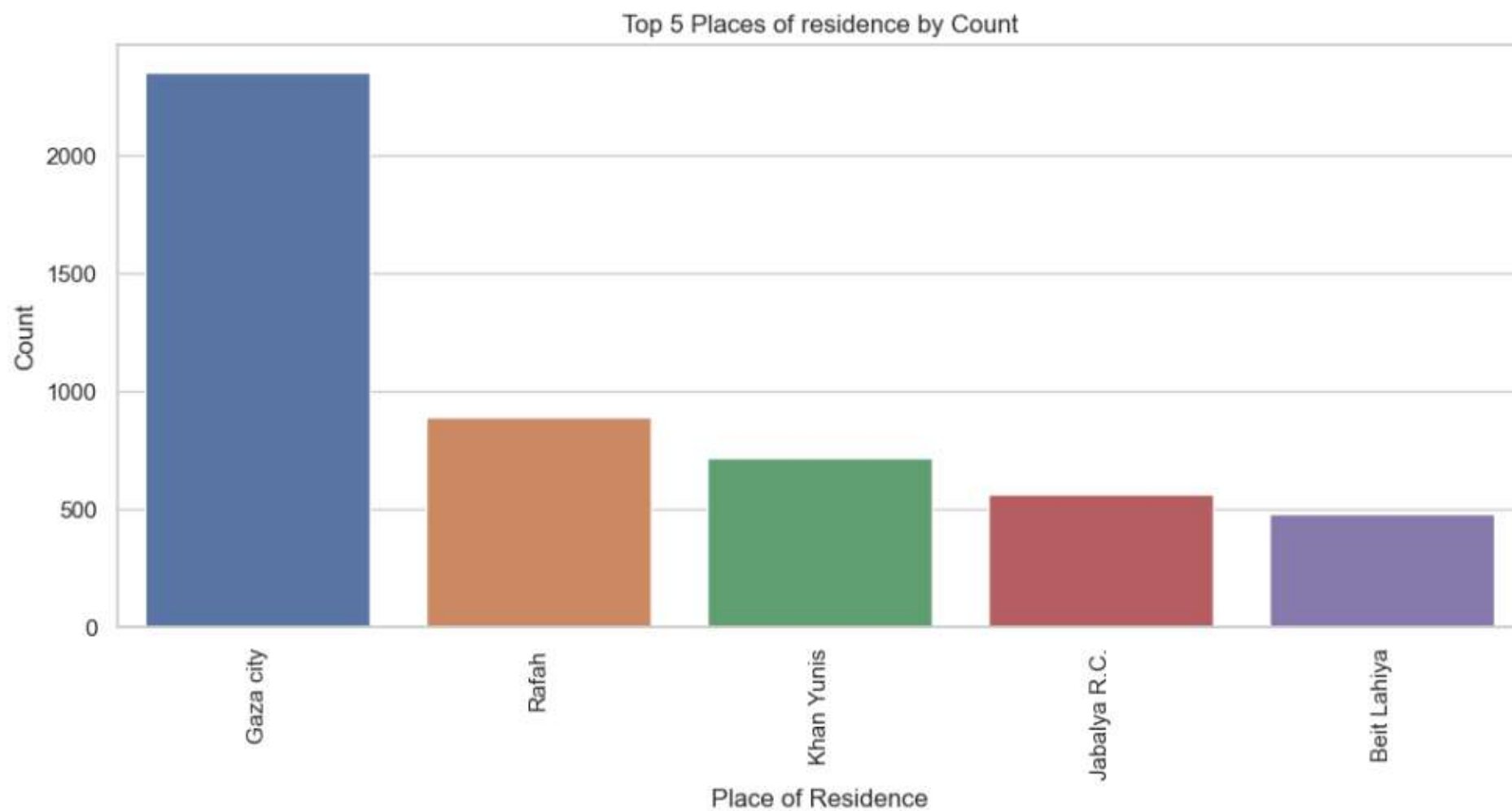
In [237]:

```
1 # Get the top 5 event locations by count
2 top5_event_locations = data['event_location'].value_counts().head(5)
3
4 # Calculate the percentage
5 total_count = len(data['event_location'])
6 percentage = (top5_event_locations / total_count) * 100
7
8 # Create the bar plot
9 plt.figure(figsize=(10, 2))
10 sns.set_style("whitegrid")
11 ax = sns.barplot(x=top5_event_locations.index, y=top5_event_locations.values)
12
13 plt.title('Top 5 Event Locations by Count')
14 plt.xlabel('Event Location')
15 plt.ylabel('Percentage')
16
17 # Annotate each bar with the percentage
18 for i, p in enumerate(ax.patches):
19     ax.annotate(f'{percentage[i]:.2f}%', (p.get_x() + p.get_width() / 2., p.get_height()), ha='center', va='bottom')
20
21 plt.xticks(rotation=90)
22 plt.show()
23
```



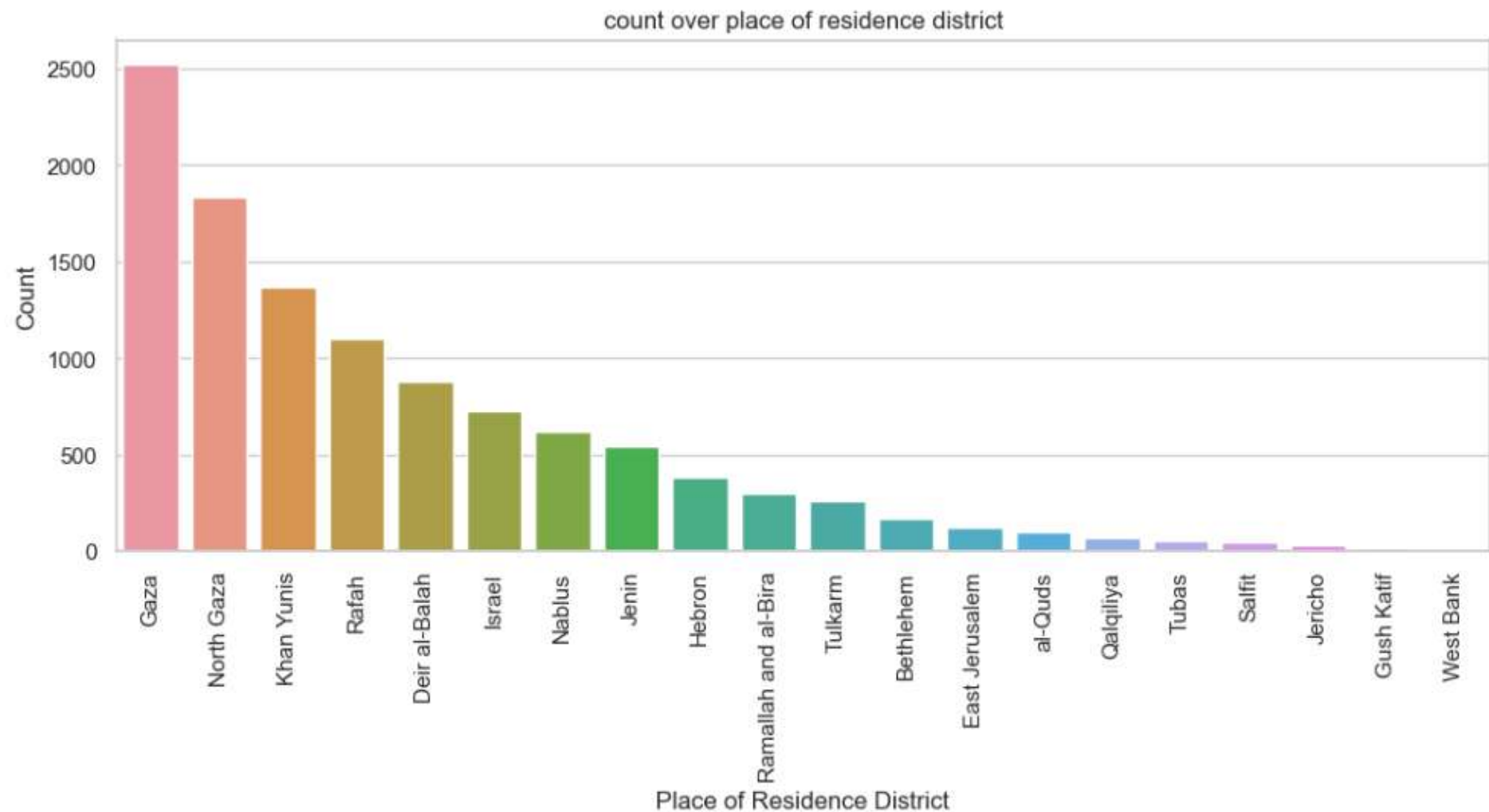


```
In [238]: 1 #Top 5 place of residence by count
2
3 top5_place_of_residence = data['place_of_residence'].value_counts().head(5)
4
5 #Bar plot for the top 5 event locations
6 plt.figure(figsize=(12, 5))
7 sns.barplot(x=top5_place_of_residence.index, y=top5_place_of_residence.values)
8 plt.title("Top 5 Places of residence by Count")
9 plt.xlabel('Place of Residence')
10 plt.ylabel('Count')
11 plt.xticks(rotation=90)
12 plt.show()
```





```
In [242]: 1 #places of residence district by count
2
3 place_of_residence_district = data['place_of_residence_district'].value_counts()
4
5 plt.figure(figsize=(12, 4.5))
6 sns.barplot(x=place_of_residence_district.index, y=place_of_residence_district.values)
7 plt.title('count over place of residence district')
8 plt.xlabel('Place of Residence District')
9 plt.ylabel('Count')
10 plt.xticks(rotation=90)
11 plt.show()
```

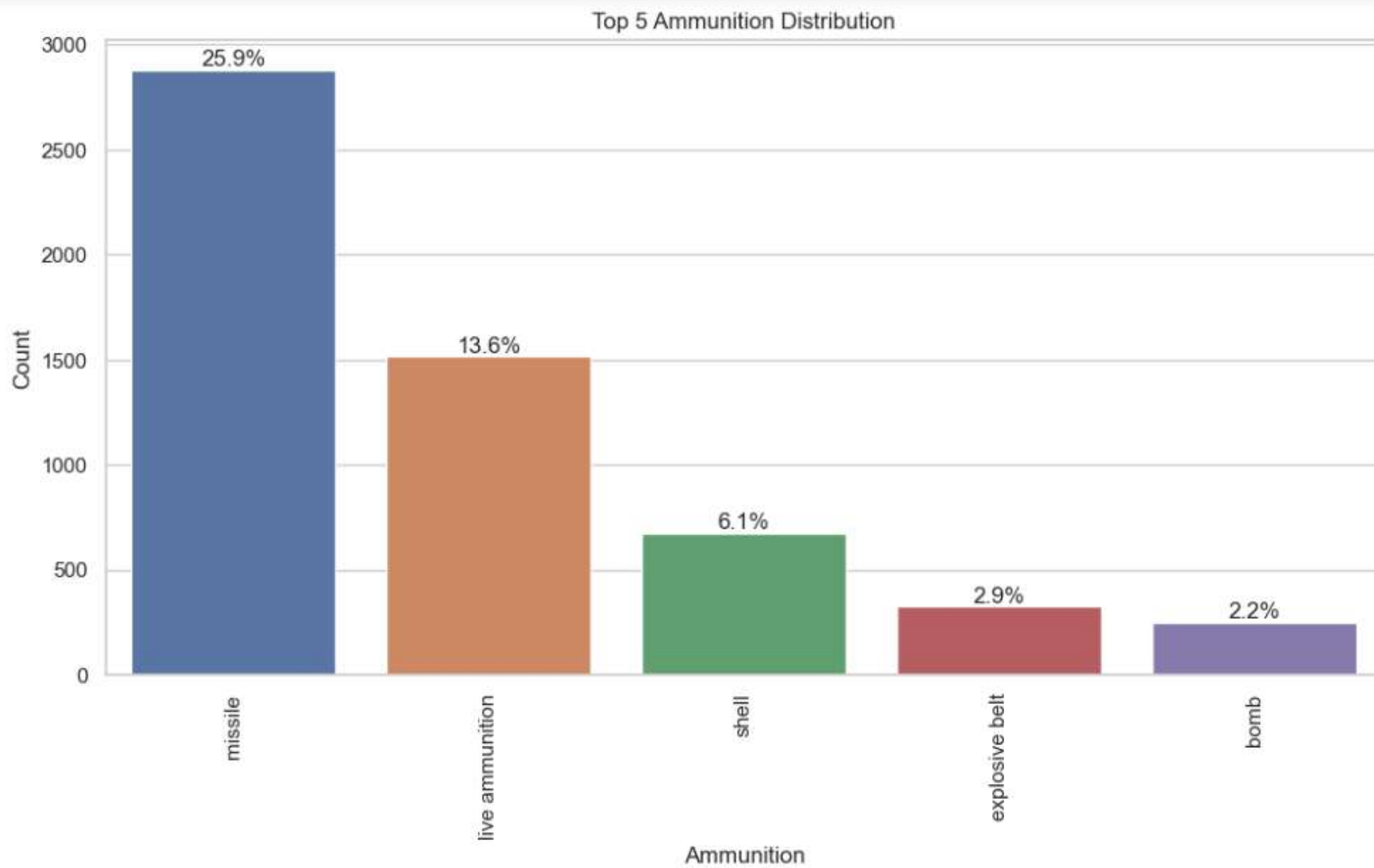




We can say Gaza city face the most number of casualties

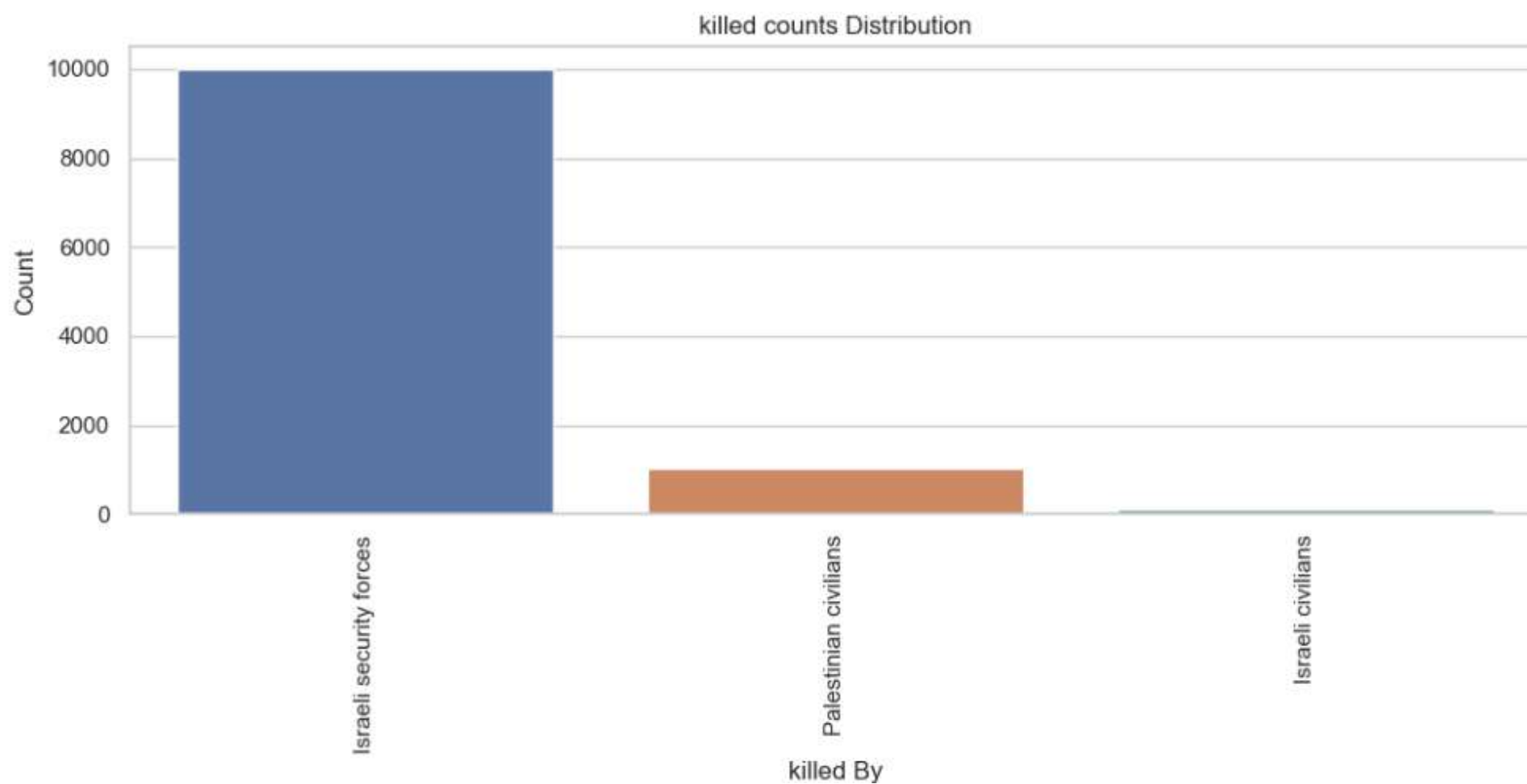
## Injuries and Ammunition

```
In [209]: 1 top5_ammunition_counts = data['ammunition'].value_counts().head(5)
2
3 plt.figure(figsize=(12, 6))
4 ax = sns.barplot(x=top5_ammunition_counts.index, y=top5_ammunition_counts.values)
5 plt.title('Top 5 Ammunition Distribution')
6 plt.xlabel('Ammunition')
7 plt.ylabel('Count')
8 plt.xticks(rotation=90)
9
10 # Add percentage labels above the bars
11 total = len(data['ammunition']) # Total number of data points
12
13 for p in ax.patches:
14     percentage = '{:.1f}%'.format(100 * p.get_height() / total)
15     x = p.get_x() + p.get_width() / 2
16     y = p.get_height()
17     ax.annotate(percentage, (x, y), ha='center', va='bottom')
18
19 plt.show()
```



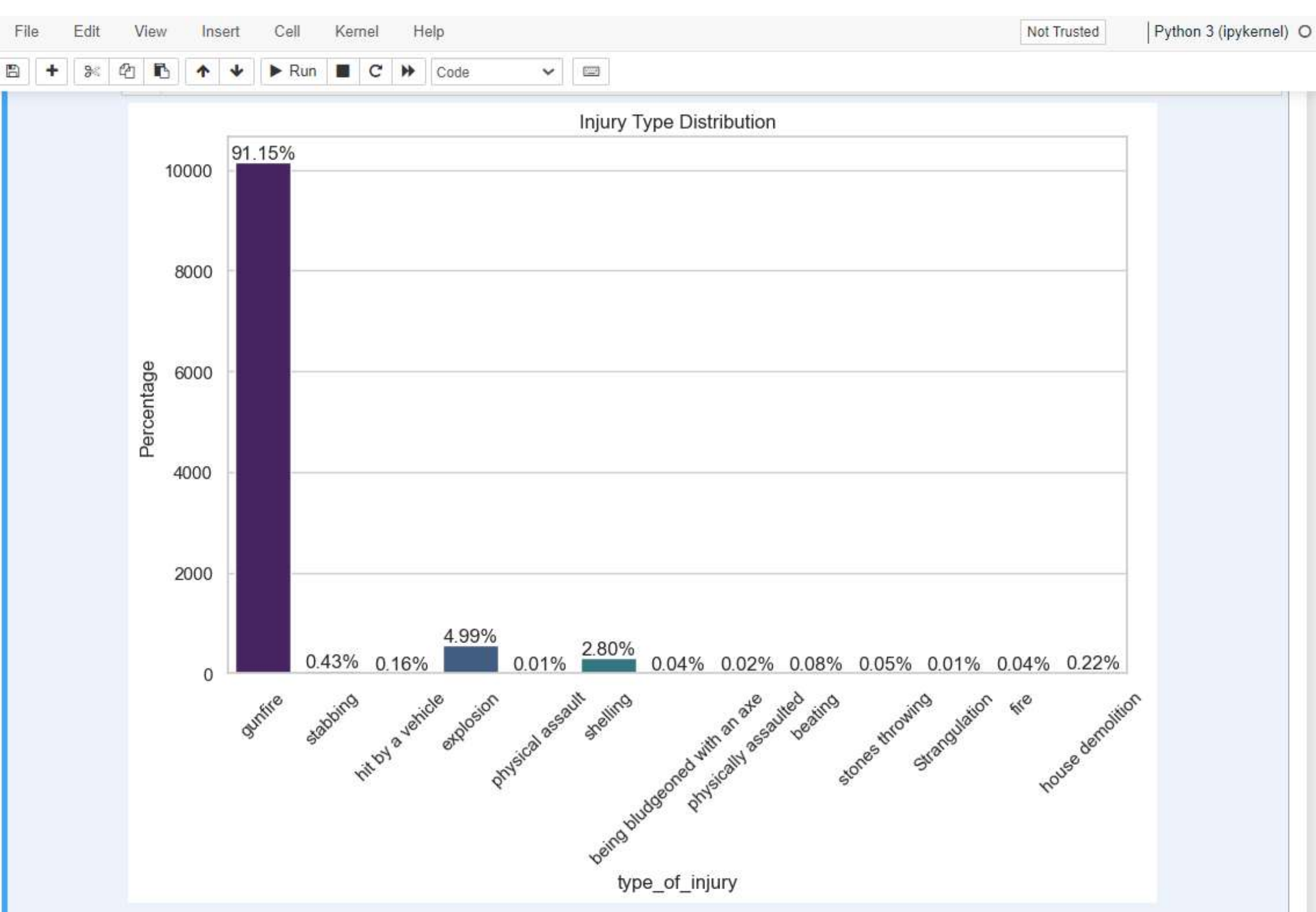
In [243]:

```
1 #Killed Count vs Killed By
2
3 killed_by_counts = data['killed_by'].value_counts()
4 plt.figure(figsize=(12, 4))
5 sns.barplot(x=killed_by_counts.index, y=killed_by_counts.values)
6 plt.title('killed counts Distribution')
7 plt.xlabel('killed By')
8 plt.ylabel('Count')
9 plt.xticks(rotation=90)
10 plt.show()
```



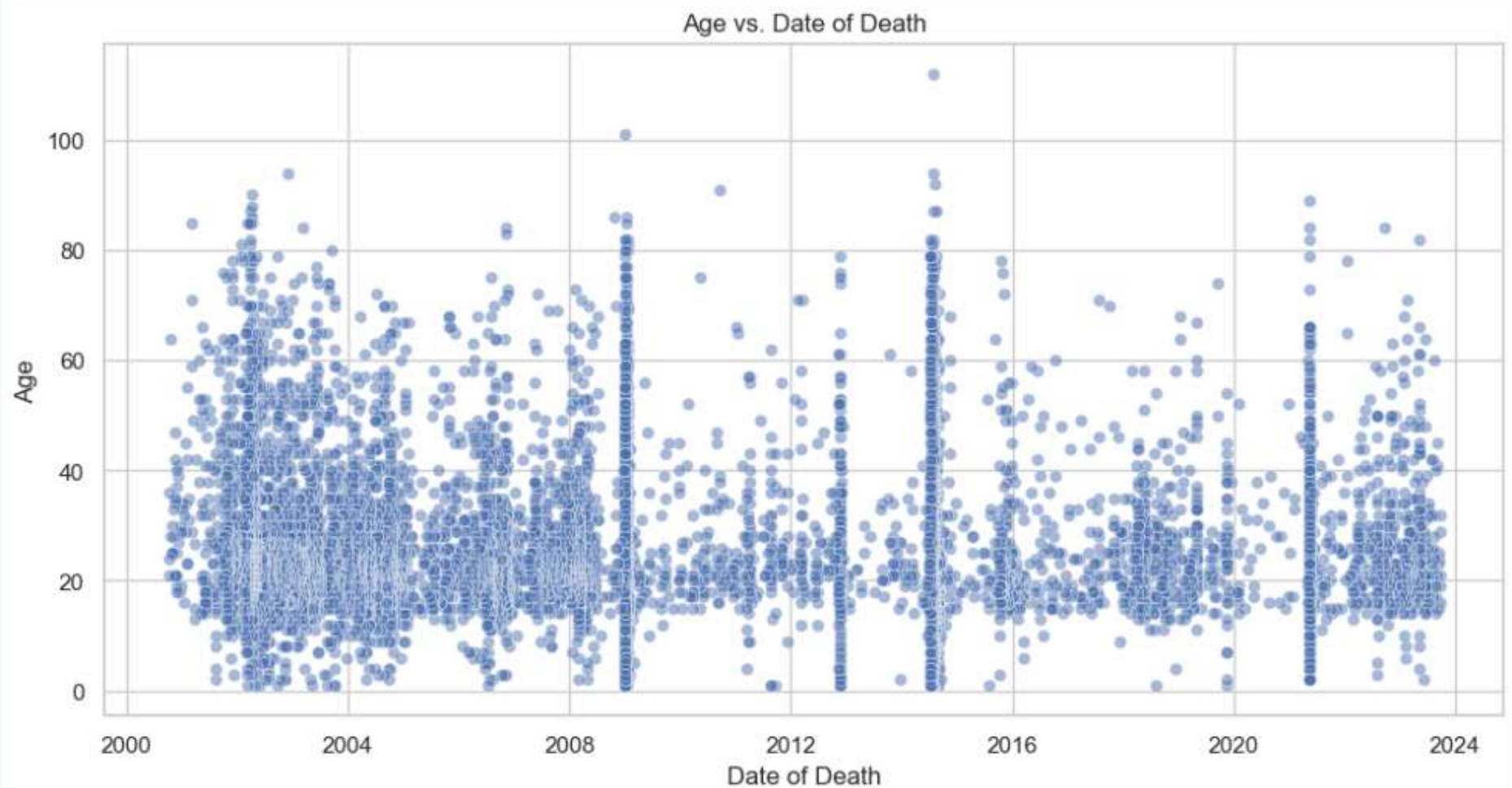
```
In [211]: 1 #Count of Injuries by Type
2
3 plt.figure(figsize=(10, 6))
4 sns.set_style("whitegrid")
5 ax = sns.countplot(data=data, x='type_of_injury', palette='viridis')
6
7 plt.xticks(rotation=45)
8 plt.ylabel('Percentage')
9 plt.title('Injury Type Distribution')
10
11 # Annotate each bar with the percentage
12 total_count = len(data['type_of_injury']) # Assuming 'type_of_injury' is the column representing injury types
13 for p in ax.patches:
14     percentage = (p.get_height() / total_count) * 100
15     ax.annotate(f'{percentage:.2f}%', (p.get_x() + p.get_width() / 2., p.get_height()), ha='center', va='bottom')
16
17 plt.show()
```







```
In [245]: 1 #Scatter Plot for Age Vs Date of Death
2 plt.figure(figsize=(12, 5.8))
3 sns.scatterplot(data=data, x='date_of_death', y='age', alpha=0.5, color='b')
4 plt.xlabel('Date of Death')
5 plt.ylabel('Age')
6 plt.title('Age vs. Date of Death')
7 plt.show()
```



### Inference

Based on the graph, it is evident that individuals in the age group 18 - 40 experienced the highest number of casualties during war from 2000 to 2023. This suggests that this age group was disproportionately affected by the conflict.