

✓ Data preprocessing

```
import pandas as pd
import matplotlib.pyplot as plt

df = pd.read_csv("/content/drive/MyDrive/warfare/2023-10-01-2024-08-28-Middle-East-Israel-Palestine.csv")

## CHECK DUPLICATE EVENT IDs HAVE SAME FATALITY REPORTS
df = df.drop_duplicates(subset=['event_id_cnty'], keep='last')

# Convert event string to datetime
df['event_date'] = pd.to_datetime(df['event_date'])

# Sort those values in time
df.sort_values(by='event_date', inplace=True)
```

✓ Daily Fatalities

```
# Sum total fatalities per day in each country
daily_df = df[['event_date', 'fatalities', 'country']].groupby(['event_date', 'country']).sum().reset_index()

# Pivot data to have column-wise fatalities per country (for plotting)
daily_df_pivot = daily_df.pivot(index='event_date', columns='country', values='fatalities')

# Fill any missing dates with zero deaths

# Determine the date range
start_date = daily_df_pivot.index.min()
end_date = daily_df_pivot.index.max()
date_range = pd.date_range(start=start_date, end=end_date, freq='D')

# Reindex and fill any missing values with zero
daily_df_pivot = daily_df_pivot.reindex(date_range).fillna(0)

daily_df_pivot['7-day average'] = daily_df_pivot.rolling(7).mean().sum(axis=1)

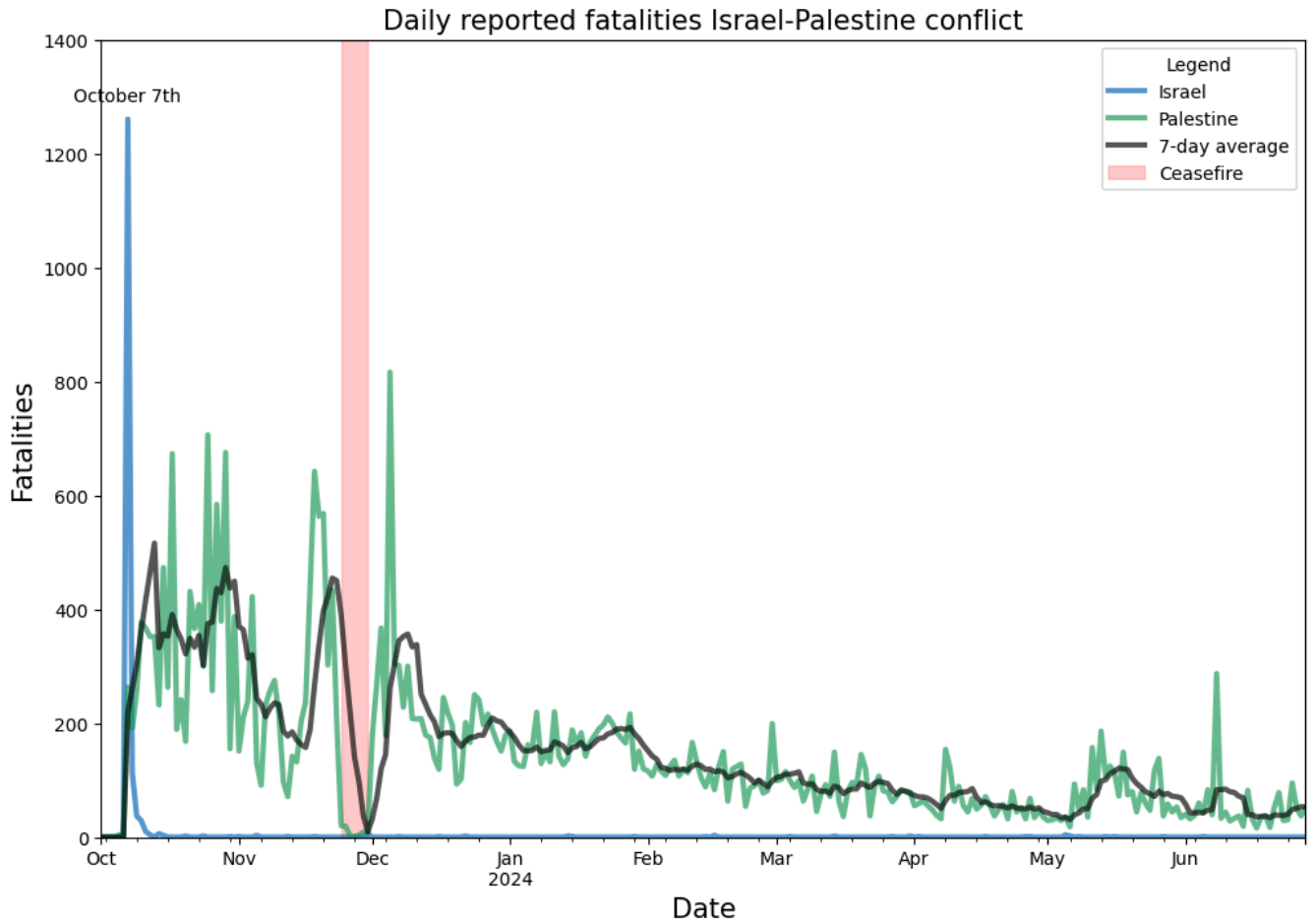
# Plot the result
ceasefire_start = "2023-11-24"
ceasefire_end = "2023-11-30"

fig, ax = plt.subplots(figsize = (12,8))
daily_df_pivot.plot(kind='line', color=['#005EB8', '#149954', '#000000'], ax=ax, alpha=0.65, lw=3)
ax.axvspan(ceasefire_start, ceasefire_end, color='red', alpha=0.2, label='Ceasefire')

# Annotate October 7th
x_coord = daily_df_pivot.idxmax()['Israel']
y_coord = daily_df_pivot.max()['Israel']
label = 'October 7th'
plt.annotate(label, (x_coord, y_coord), textcoords="offset points", xytext=(0,10), ha='center')

plt.xlabel('Date', fontsize = 15)
plt.ylabel('Fatalities', fontsize = 15)
plt.legend(title='Legend')
plt.ylim(0,1400)
plt.title('Daily reported fatalities Israel-Palestine conflict', fontsize = 15)
```

Text(0.5, 1.0, 'Daily reported fatalities Israel-Palestine conflict')



✓ Cumulative fatalities

```
# Sum daily events in per country
df_grouped = df.groupby(['event_date', 'country']).sum().reset_index()
```

```
# Cumulative summation of fatalities across countries
df_grouped['cumulative_fatalities'] = df_grouped[['country', 'fatalities']].groupby(['country']).cumsum()
```

```
# Pivot data to have column-wise fatalities per country (for plotting)
pivot_df_cumsum = df_grouped.pivot(index='event_date', columns='country', values='cumulative_fatalities')
```

```
# Fill any missing dates
```

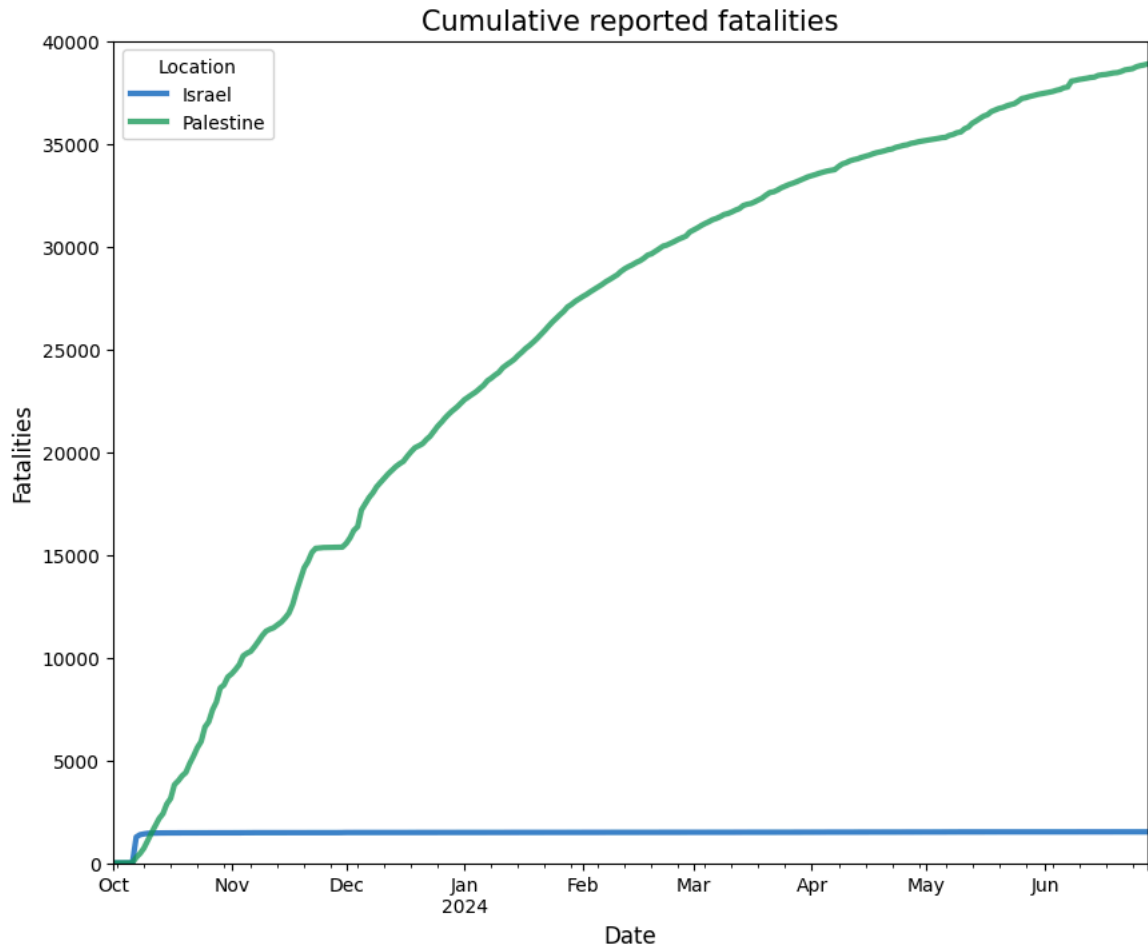
```
# Determine the date range
start_date = pivot_df_cumsum.index.min()
end_date = pivot_df_cumsum.index.max()
date_range = pd.date_range(start=start_date, end=end_date, freq='D')
```

```
# Reindex and fill any missing values with previous day
pivot_df_cumsum = pivot_df_cumsum.reindex(date_range).fillna(method='ffill')
```

<ipython-input-11-30c83c697ff2>:9: FutureWarning: DataFrame.fillna with 'method' is deprecated and will raise in a future version. Use obj.ffill
pivot_df_cumsum = pivot_df_cumsum.reindex(date_range).fillna(method='ffill')

```
# Create a line plot
fig, ax = plt.subplots(figsize = (10,8))
pivot_df_cumsum.plot(kind='line', color=['#005EB8','#149954'], alpha=0.75, lw = 3, ax=ax)
plt.legend(title='Location')
plt.xlabel('Date', fontsize=12)
plt.ylabel('Fatalities', fontsize=12)
plt.ylim(0,40000)
plt.title('Cumulative reported fatalities', fontsize=15)
```

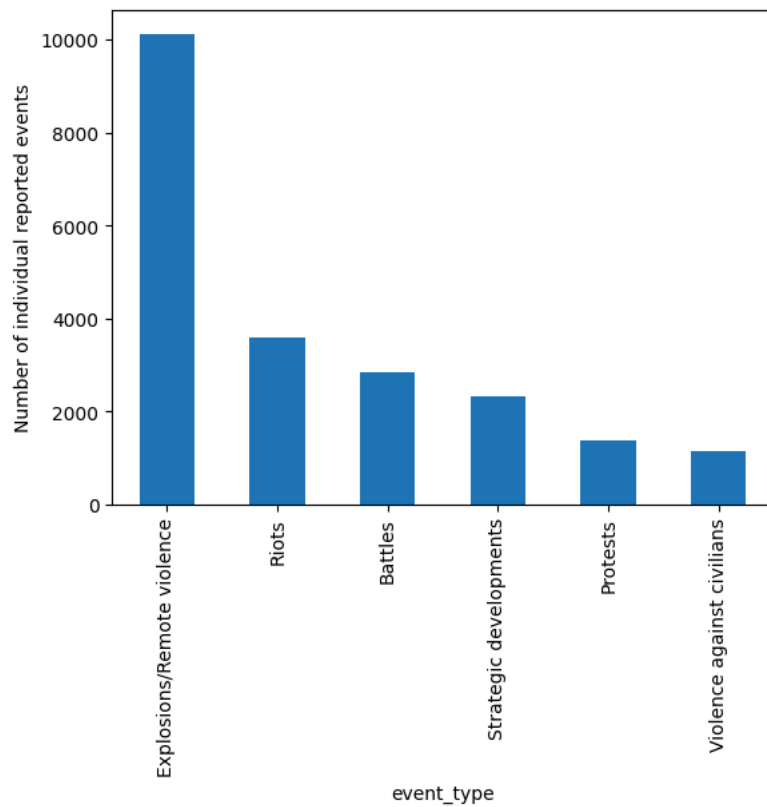
```
Text(0.5, 1.0, 'Cumulative reported fatalities')
```



Types of events

```
df.event_type.value_counts().plot.bar()  
plt.ylabel('Number of individual reported events')
```

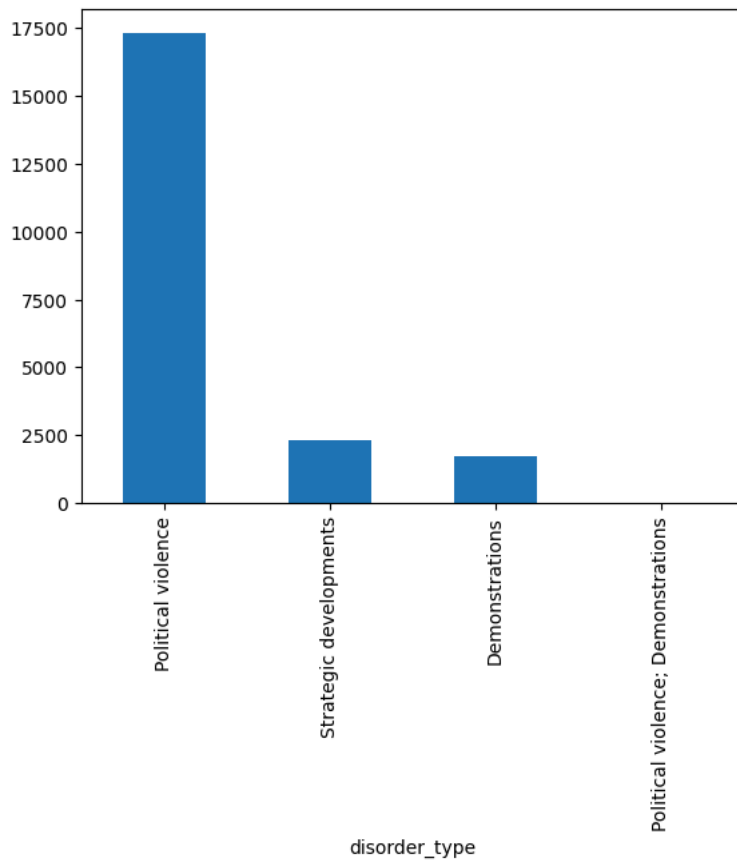
```
Text(0, 0.5, 'Number of individual reported events')
```



```
df.disorder_type.value_counts().plot.bar()
```



<Axes: xlabel='disorder_type'>



✓ Map-wise visualisation (in-development)

```
df_deaths = df[df['fatalities'] > 0]
df_deaths_gp = df[['longitude', 'latitude', 'fatalities']].groupby(['longitude', 'latitude']).sum().reset_index()

unique_loc = df[['longitude', 'latitude', 'country']].drop_duplicates()

import folium
import geopandas as gpd
import pandas as pd
from shapely.geometry import Point

# Create a base map centered around Israel/Palestine
m = folium.Map(location=[31.5, 34.5], tiles="CartoDB Positron", zoom_start=9)

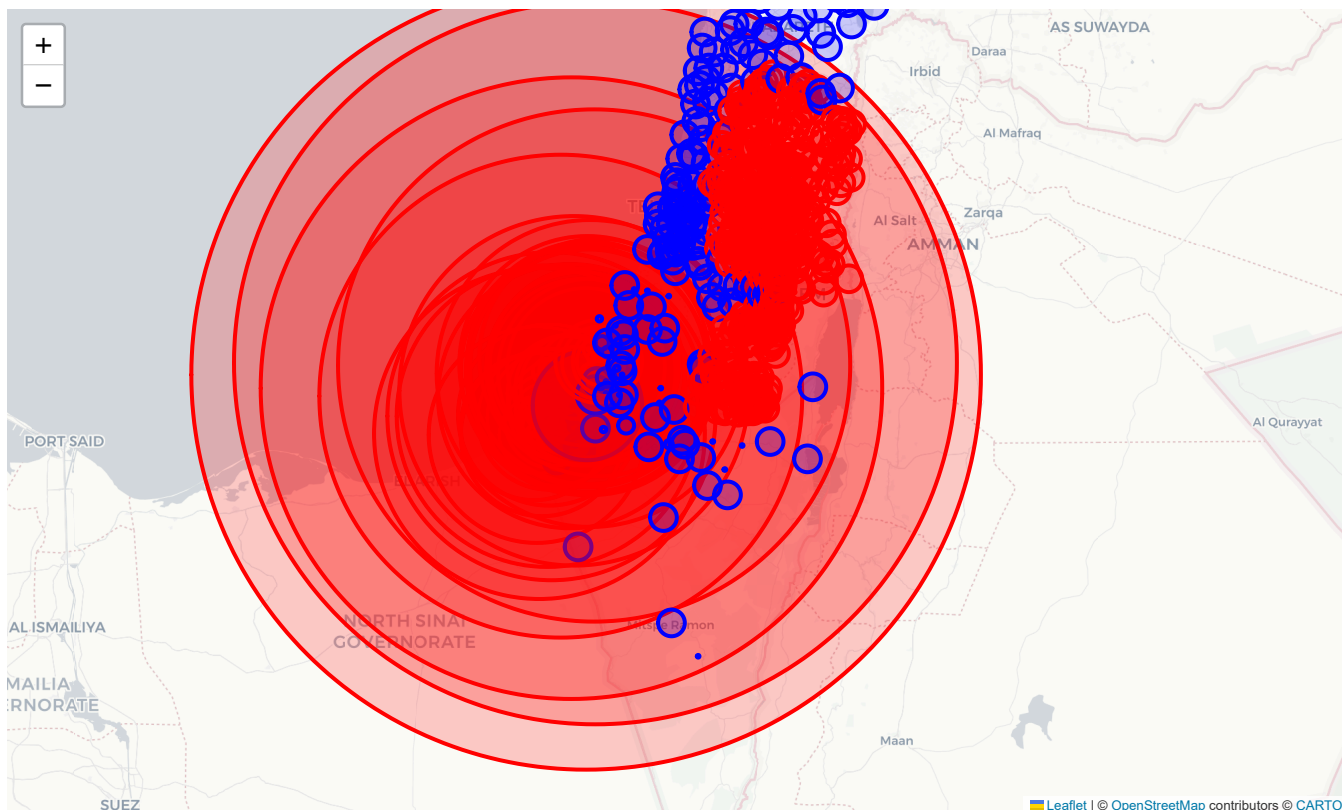
# Create a GeoDataFrame
gdf = gpd.GeoDataFrame(df_deaths_gp, geometry=gpd.points_from_xy(df_deaths_gp.longitude, df_deaths_gp.latitude))

# Function to add points to the folium map
def add_points_to_map(gdf, map_obj):
    for _, row in gdf.iterrows():
        country = unique_loc[(unique_loc['latitude'] == row['latitude']) & (unique_loc['longitude'] == row['longitude'])]['country'].values[0]
        if country == 'Palestine':
            colour_in = 'red'
        else:
            colour_in = 'blue'

        folium.CircleMarker(
            location=[row['latitude'], row['longitude']],
            radius=row['fatalities'] / 10, # Adjust the size as needed
            color=colour_in,
            fill=True,
            fill_color=colour_in
        ).add_to(map_obj)

# Add points to the map
add_points_to_map(gdf, m)
```

...



▼ Visualsing protests

```
df_protests = df[df['event_type'] == 'Protests']
```

```
# Convert tag to numbers
text_to_num = {
    'no report': float('nan'),
    'hundreds': 200,
    'thousands': 2000,
    'dozens': 36,
    'about 20 to 100': 60,
    'over 100': 120,
    'a few dozen': 36,
    '200': 200,
    'tens': 30,
    'several hundred': 400,
    'large': 50,
    'dozens to hundreds': 200,
    'about 200': 200,
    'about 1,000': 1000,
    'dozens to about 100': 84,
    'tens of thousands': 30000,
    'about 400': 400,
    'about 100,000': 100000,
    'about 300': 300,
    'about 150': 150,
    'about 20': 20,
    'about 250': 250,
    'at least 200': 200,
    'about 50': 50,
    'about 2,000': 2000,
    'about 20,000': 20000,
    'about 40': 40,
    'about 100': 100,
    'about 30': 30,
    'around 10,000': 10000,
    'at least 1,000': 1100,
    'about 120,000': 120000,
    'several dozen': 60,
    'at least 2,000': 2000,
    'about 200 to 300': 250,
    'about 5,000': 5000
}
```

```
# Function to convert text to number
def convert_to_number(text):
    for key in text_to_num:
        if key in text:
            return text_to_num[key]
    return None
```

```
df_protests['crowd_size'] = df_protests['tags'].apply(convert_to_number)
```



<ipython-input-20-48b36fd00281>:49: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
df_protests['crowd_size'] = df_protests['tags'].apply(convert_to_number)

```
daily_protests = df_protests[['event_date', 'crowd_size', 'country']].groupby(['event_date', 'country']).sum().reset_index()
```

```
daily_protests_pivot = daily_protests.pivot(index='event_date', columns='country', values='crowd_size')
```

```
# Fill any missing dates with zero deaths
```

```
# Determine the date range
start_date = daily_protests_pivot.index.min()
end_date = daily_protests_pivot.index.max()
date_range = pd.date_range(start=start_date, end=end_date, freq='D')
```

```
# Reindex and fill any missing values with zero
daily_protests_pivot = daily_protests_pivot.reindex(date_range).fillna(0)
```

```
daily_protests.max()
```



0

```
event_date 2024-06-28 00:00:00
country      Palestine
crowd_size   128650.0
dtype: object
```

```
fig, ax = plt.subplots(figsize = (12,8))
daily_protests_pivot.plot(color=['#005EB8','#149954'], alpha=0.75, ax=ax, lw=3)

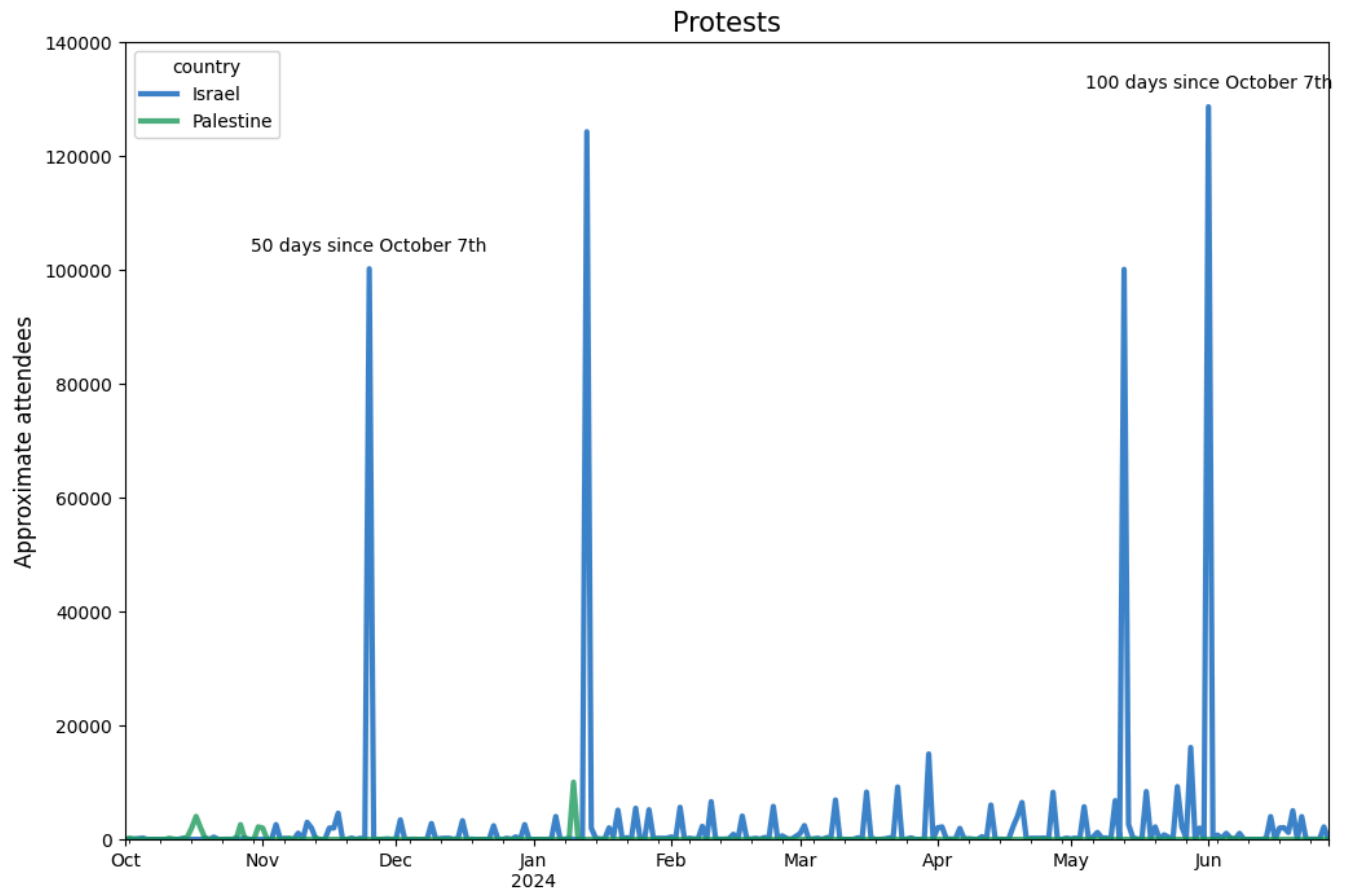
# Global day of action for Palestine
x_coord = daily_protests_pivot.idxmax()['Israel']
y_coord = daily_protests_pivot.max()['Israel']
label = '100 days since October 7th'
plt.annotate(label, (x_coord, y_coord), textcoords="offset points", xytext=(0,10), ha='center')

# Global day of action for Palestine
x_coord = '2023-11-25 00:00:00'
y_coord = 100000
label = '50 days since October 7th'
plt.annotate(label, (x_coord, y_coord), textcoords="offset points", xytext=(0,10), ha='center')

plt.title('Protests', fontsize=15)
plt.ylim([0,140000])
plt.ylabel('Approximate attendees', fontsize=12)
```



Text(0, 0.5, 'Approximate attendees')



Experimental modelling

```
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import PolynomialFeatures
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score# Split the data into training and testing sets

pivot_df_cumsum.sort_index(inplace=True)
pivot_df_cumsum
```

	country	Israel	Palestine
2023-10-01		0.0	0.0
2023-10-02		0.0	0.0
2023-10-03		0.0	0.0
2023-10-04		0.0	0.0
2023-10-05		0.0	4.0
...	
2024-06-24	1512.0	38640.0	
2024-06-25	1512.0	38735.0	
2024-06-26	1512.0	38786.0	
2024-06-27	1512.0	38823.0	
2024-06-28	1512.0	38873.0	

272 rows × 2 columns

다음 단계: [pivot_df_cumsum 변수로 코드 생성](#) [추천 차트 보기](#) [New interactive sheet](#)

```
# Skip to first deaths in the dataset
pivot_df_cumsum = pivot_df_cumsum[4:]
```

```
pivot_df_cumsum.reset_index().index.values
```

```
array([ 0,  1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12,
        13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25,
        26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37, 38,
        39, 40, 41, 42, 43, 44, 45, 46, 47, 48, 49, 50, 51,
        52, 53, 54, 55, 56, 57, 58, 59, 60, 61, 62, 63, 64,
        65, 66, 67, 68, 69, 70, 71, 72, 73, 74, 75, 76, 77,
        78, 79, 80, 81, 82, 83, 84, 85, 86, 87, 88, 89, 90,
        91, 92, 93, 94, 95, 96, 97, 98, 99, 100, 101, 102, 103,
        104, 105, 106, 107, 108, 109, 110, 111, 112, 113, 114, 115, 116,
        117, 118, 119, 120, 121, 122, 123, 124, 125, 126, 127, 128, 129,
        130, 131, 132, 133, 134, 135, 136, 137, 138, 139, 140, 141, 142,
        143, 144, 145, 146, 147, 148, 149, 150, 151, 152, 153, 154, 155,
        156, 157, 158, 159, 160, 161, 162, 163, 164, 165, 166, 167, 168,
        169, 170, 171, 172, 173, 174, 175, 176, 177, 178, 179, 180, 181,
        182, 183, 184, 185, 186, 187, 188, 189, 190, 191, 192, 193, 194,
        195, 196, 197, 198, 199, 200, 201, 202, 203, 204, 205, 206, 207,
        208, 209, 210, 211, 212, 213, 214, 215, 216, 217, 218, 219, 220,
        221, 222, 223, 224, 225, 226, 227, 228, 229, 230, 231, 232, 233,
        234, 235, 236, 237, 238, 239, 240, 241, 242, 243, 244, 245, 246,
        247, 248, 249, 250, 251, 252, 253, 254, 255, 256, 257, 258, 259,
        260, 261, 262, 263, 264, 265, 266, 267])
```

```
X = pivot_df_cumsum.reset_index().index.values.reshape(-1,1)
y = pivot_df_cumsum.reset_index()['Palestine']
```

Polynomial regression

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
# Create polynomial features (degree 2)
poly = PolynomialFeatures(degree=2)
X_train_poly = poly.fit_transform(X_train)
X_test_poly = poly.transform(X_test)
```

```
# Fit a polynomial regression model
poly_reg = LinearRegression()
poly_reg.fit(X_train_poly, y_train)
```

```
# Predict on the test set
y_pred = poly_reg.predict(X_test_poly)
```

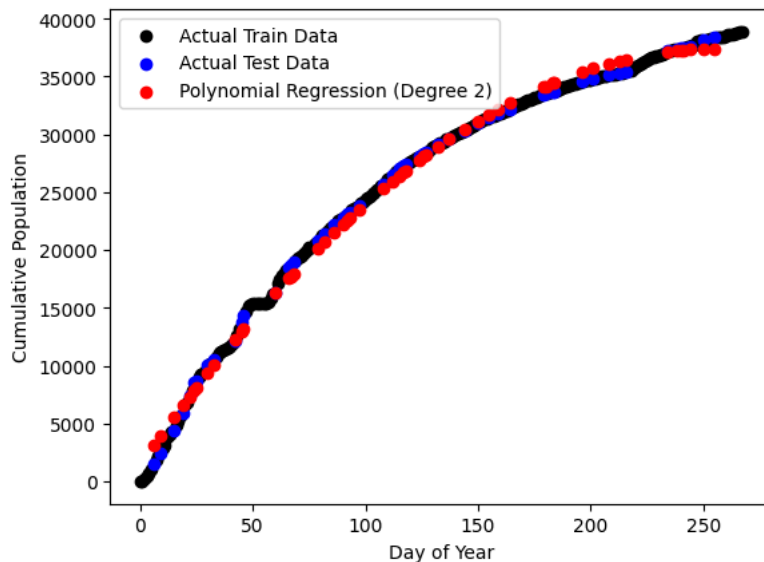


```
# Evaluate the model
mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)

print(f"Mean Squared Error: {mse:.2f}")
print(f"R-squared: {r2:.2f}")

# Plot the original data and the polynomial regression line
plt.scatter(X_train, y_train, color='black', label='Actual Train Data') # Plot the test data
plt.scatter(X_test, y_test, color='blue', label='Actual Test Data') # Plot the test data
plt.scatter(X_test, y_pred, color='red', label='Polynomial Regression (Degree 2)')
plt.xlabel('Day of Year')
plt.ylabel('Cumulative Population')
plt.legend()
plt.show()
```

Mean Squared Error: 489920.39
R-squared: 1.00



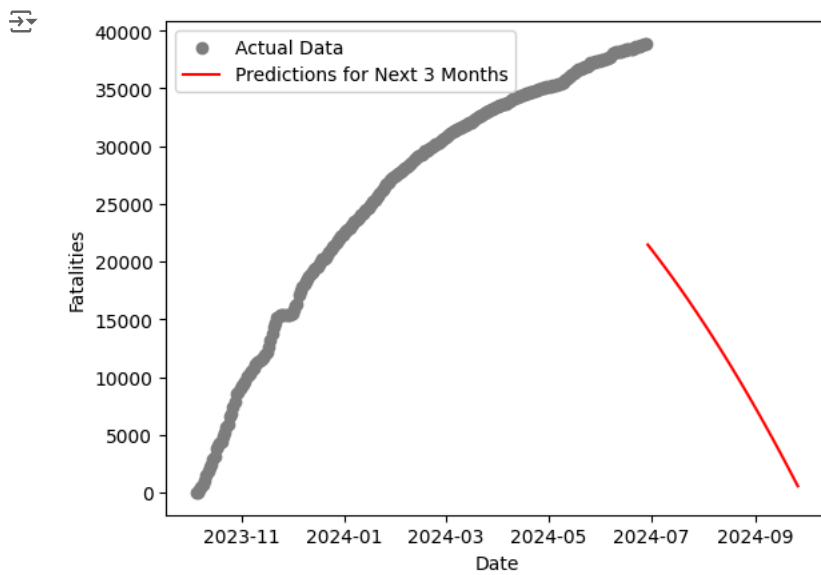
```
# Extend the index to include the next 3 months
last_date = pivot_df_cumsum.index.max()
next_3_months = pd.date_range(start=last_date + pd.DateOffset(days=1), periods=90, freq='D')

# Last index of previous data
last_index = pivot_df_cumsum.reset_index().index.values[-1] - 20 # Where does this 20 come in?

# Create corresponding day-of-year values for the next 3 months
next_3_months_dayofyear = next_3_months.to_series().dt.dayofyear.values.reshape(-1, 1) + last_index

# Use the polynomial regression model to make predictions for the next 3 months
predictions = poly_reg.predict(poly.transform(next_3_months_dayofyear))

# Plot the original data, the polynomial regression line, and the predictions for the next 3 months
plt.scatter(pivot_df_cumsum.index, y, color='grey', label='Actual Data')
# plt.plot(pivot_df_filled.index, poly_reg.predict(X_train_poly), color='red', label='Polynomial Regression (Degree 2)')
plt.plot(next_3_months, predictions, color='red', label='Predictions for Next 3 Months')
plt.xlabel('Date')
plt.ylabel('Fatalities')
plt.legend()
plt.show()
```



Logistic growth model

```
from scipy.optimize import curve_fit
import numpy as np

def logistic_model(x, c, k, t0):
    return c / (1 + np.exp(-k * (x - t0)))

# Prepare the data for fitting
x_data = pivot_df_cumsum.reset_index().index.values
y_data = pivot_df_cumsum['Palestine']

# Initial guess for the parameters
initial_guess = [max(y_data), 1, np.median(x_data)]

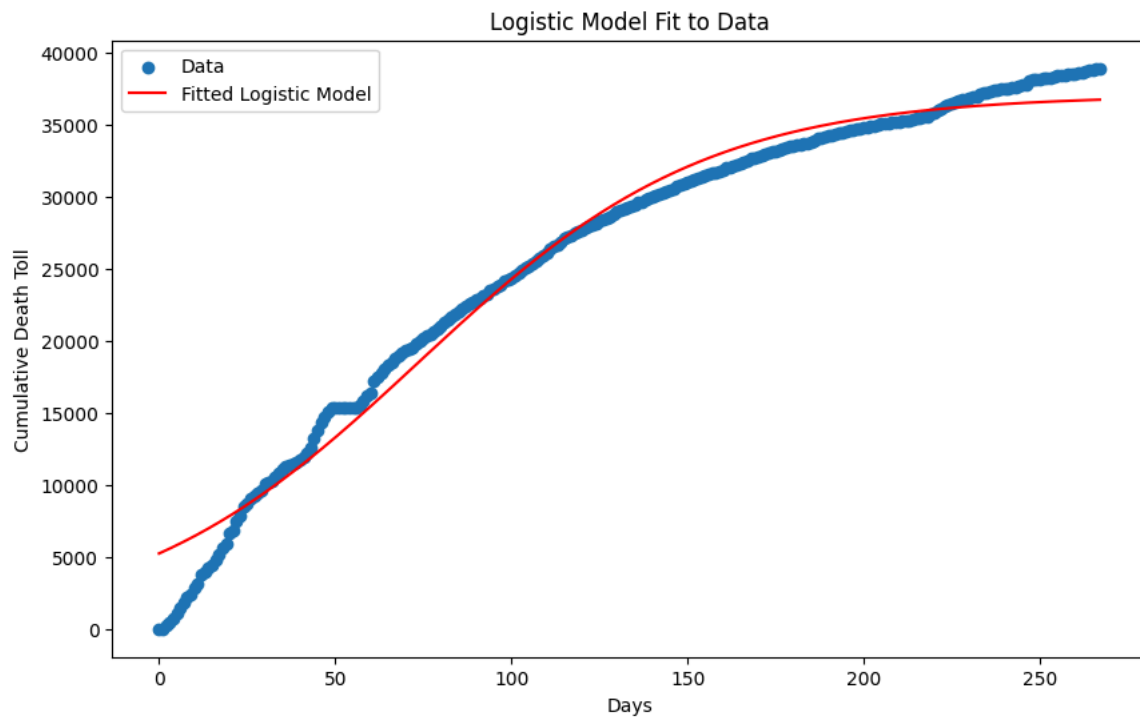
# Use curve_fit to fit the model
params, covariance = curve_fit(logistic_model, x_data, y_data, p0=initial_guess)

# Extract the parameters
c, k, t0 = params

# Generate a sequence of x values
x_model = np.linspace(x_data.min(), x_data.max(), 1000)

# Apply the logistic model
y_model = logistic_model(x_model, c, k, t0)

# Plot the data and the model prediction
plt.figure(figsize=(10, 6))
plt.scatter(x_data, y_data, label='Data')
plt.plot(x_model, y_model, label='Fitted Logistic Model', color='red')
plt.legend()
plt.xlabel('Days')
plt.ylabel('Cumulative Death Toll')
plt.title('Logistic Model Fit to Data')
plt.show()
```



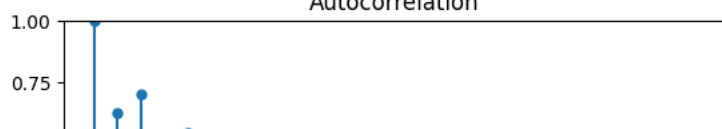
```
# Differencing the series
diff = pivot_df_cumsum.diff(periods=1).dropna()

from statsmodels.tsa.arima.model import ARIMA
from statsmodels.graphics.tsaplots import plot_acf, plot_pacf

plot_acf(diff['Palestine'])
plot_pacf(diff['Palestine'])
plt.show()
```



Autocorrelation



ARIMA