

Global Suicide Rates Overview (1985-2016)

June 23, 2025

```
[1]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[3]: df = pd.read_csv('../data/master.csv')
print(df.head())
print(df.info())
```

	country	year	sex	age	suicides_no	population \
0	Albania	1987	male	15-24 years	21	312900
1	Albania	1987	male	35-54 years	16	308000
2	Albania	1987	female	15-24 years	14	289700
3	Albania	1987	male	75+ years	1	21800
4	Albania	1987	male	25-34 years	9	274300

	suicides/100k pop	country-year	HDI for year	gdp_for_year (\$)	\
0	6.71	Albania1987	NaN	2,156,624,900	
1	5.19	Albania1987	NaN	2,156,624,900	
2	4.83	Albania1987	NaN	2,156,624,900	
3	4.59	Albania1987	NaN	2,156,624,900	
4	3.28	Albania1987	NaN	2,156,624,900	

	gdp_per_capita (\$)	generation
0	796	Generation X
1	796	Silent
2	796	Generation X
3	796	G.I. Generation
4	796	Boomers

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 27820 entries, 0 to 27819
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	country	27820 non-null	object
1	year	27820 non-null	int64
2	sex	27820 non-null	object
3	age	27820 non-null	object
4	suicides_no	27820 non-null	int64

```

5   population                27820 non-null   int64
6   suicides/100k pop         27820 non-null   float64
7   country-year              27820 non-null   object
8   HDI for year              8364 non-null    float64
9   gdp_for_year ($)          27820 non-null   object
10  gdp_per_capita ($)        27820 non-null   int64
11  generation                27820 non-null   object
dtypes: float64(2), int64(4), object(6)
memory usage: 2.5+ MB
None

```

```

[44]: # The data types are ensured to be correct
df['suicides_no'] = pd.to_numeric(df['suicides_no'], errors='coerce')
df['population'] = pd.to_numeric(df['population'], errors='coerce')
df['gdp_per_capita ($)'] = pd.to_numeric(df['gdp_per_capita ($)'],
→errors='coerce')
df['suicides/100k pop'] = pd.to_numeric(df['suicides/100k pop'], errors='coerce')

# Missing values are checked
print(df.isnull().sum())

# Rows with missing essential information (such as population or suicides_no)
→are removed
df = df.dropna(subset=['suicides_no', 'population', 'suicides/100k pop'])

# Missing values in economic or social columns are filled if needed (depending
→on the plan)
# For example, if HDI for year is important, its missing values could be filled
→with the mean
# df['HDI for year'] = df['HDI for year'].fillna(df['HDI for year'].mean())

# Outliers are removed (such as population <= 0)
df = df[df['population'] > 0]
df = df[df['suicides_no'] >= 0]

# Extra spaces are removed from country and category names if present
df['country'] = df['country'].str.strip()
df['age'] = df['age'].str.strip()
df['sex'] = df['sex'].str.strip()

# The data structure after cleaning is reviewed
print(df.info())
print(df.head())

```

```

country      0
year         0
sex          0
age          0

```

```

suicides_no          0
population           0
suicides/100k pop    0
country-year         0
HDI for year        19456
  gdp_for_year ($)   0
gdp_per_capita ($)   0
generation           0
dtype: int64
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 27820 entries, 0 to 27819
Data columns (total 12 columns):
#   Column                Non-Null Count  Dtype
---  -
0   country                27820 non-null  object
1   year                   27820 non-null  int64
2   sex                    27820 non-null  object
3   age                    27820 non-null  object
4   suicides_no            27820 non-null  int64
5   population             27820 non-null  int64
6   suicides/100k pop      27820 non-null  float64
7   country-year           27820 non-null  object
8   HDI for year           8364 non-null   float64
9   gdp_for_year ($)       27820 non-null  object
10  gdp_per_capita ($)      27820 non-null  int64
11  generation             27820 non-null  object
dtypes: float64(2), int64(4), object(6)
memory usage: 2.5+ MB
None
   country  year  sex  age  suicides_no  population \
0  Albania  1987  male  15-24 years      21      312900
1  Albania  1987  male  35-54 years      16      308000
2  Albania  1987  female  15-24 years     14      289700
3  Albania  1987  male   75+ years        1       21800
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   suicides/100k pop  country-year  HDI for year  gdp_for_year ($) \
0              6.71  Albania1987      NaN      2,156,624,900
1              5.19  Albania1987      NaN      2,156,624,900
2              4.83  Albania1987      NaN      2,156,624,900
3              4.59  Albania1987      NaN      2,156,624,900
4              3.28  Albania1987      NaN      2,156,624,900

   gdp_per_capita ($)  generation
0              796  Generation X
1              796      Silent
2              796  Generation X
3              796  G.I. Generation

```

```
[52]: # A dictionary mapping countries to continents is created
continent_mapping = {
    'Albania': 'Europe', 'Algeria': 'Africa', 'Argentina': 'South America',
    'Armenia': 'Asia', 'Australia': 'Oceania', 'Austria': 'Europe',
    'Azerbaijan': 'Asia', 'Bahamas': 'North America', 'Bahrain': 'Asia',
    'Bangladesh': 'Asia', 'Barbados': 'North America', 'Belarus': 'Europe',
    'Belgium': 'Europe', 'Belize': 'North America', 'Benin': 'Africa',
    'Bhutan': 'Asia', 'Bolivia': 'South America', 'Bosnia and Herzegovina': ↵
    ↵ 'Europe',
    'Botswana': 'Africa', 'Brazil': 'South America', 'Bulgaria': 'Europe',
    'Burkina Faso': 'Africa', 'Cambodia': 'Asia', 'Cameroon': 'Africa',
    'Canada': 'North America', 'Chile': 'South America', 'China': 'Asia',
    'Colombia': 'South America', 'Congo': 'Africa', 'Costa Rica': 'North ↵
    ↵ America',
    'Croatia': 'Europe', 'Cuba': 'North America', 'Cyprus': 'Asia',
    'Czech Republic': 'Europe', 'Denmark': 'Europe', 'Dominican Republic': ↵
    ↵ 'North America',
    'Ecuador': 'South America', 'Egypt': 'Africa', 'El Salvador': 'North ↵
    ↵ America',
    'Estonia': 'Europe', 'Finland': 'Europe', 'France': 'Europe',
    'Georgia': 'Asia', 'Germany': 'Europe', 'Greece': 'Europe',
    'Guatemala': 'North America', 'Guyana': 'South America', 'Honduras': 'North ↵
    ↵ America',
    'Hungary': 'Europe', 'Iceland': 'Europe', 'India': 'Asia',
    'Indonesia': 'Asia', 'Ireland': 'Europe', 'Israel': 'Asia',
    'Italy': 'Europe', 'Jamaica': 'North America', 'Japan': 'Asia',
    'Jordan': 'Asia', 'Kazakhstan': 'Asia', 'Kenya': 'Africa',
    'Kuwait': 'Asia', 'Kyrgyzstan': 'Asia', 'Latvia': 'Europe',
    'Lithuania': 'Europe', 'Luxembourg': 'Europe', 'Malta': 'Europe',
    'Mauritius': 'Africa', 'Mexico': 'North America', 'Mongolia': 'Asia',
    'Montenegro': 'Europe', 'Morocco': 'Africa', 'Nepal': 'Asia',
    'Netherlands': 'Europe', 'New Zealand': 'Oceania', 'Nicaragua': 'North ↵
    ↵ America',
    'Norway': 'Europe', 'Pakistan': 'Asia', 'Panama': 'North America',
    'Paraguay': 'South America', 'Philippines': 'Asia', 'Poland': 'Europe',
    'Portugal': 'Europe', 'Qatar': 'Asia', 'Republic of Korea': 'Asia',
    'Romania': 'Europe', 'Russia': 'Europe/Asia', 'Saudi Arabia': 'Asia',
    'Serbia': 'Europe', 'Singapore': 'Asia', 'Slovakia': 'Europe',
    'Slovenia': 'Europe', 'South Africa': 'Africa', 'Spain': 'Europe',
    'Sri Lanka': 'Asia', 'Suriname': 'South America', 'Sweden': 'Europe',
    'Switzerland': 'Europe', 'Thailand': 'Asia', 'Trinidad and Tobago': 'North ↵
    ↵ America',
    'Tunisia': 'Africa', 'Turkey': 'Europe/Asia', 'Turkmenistan': 'Asia',
    'Ukraine': 'Europe', 'United Arab Emirates': 'Asia', 'United Kingdom': ↵
    ↵ 'Europe',
```

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        'United States': 'North America', 'Uruguay': 'South America', 'Uzbekistan': '
        ↪Asia',
        'Venezuela': 'South America', 'Vietnam': 'Asia', 'Zimbabwe': 'Africa'
    }

    # The continent mapping is applied to the dataset
    df['Continent'] = df['country'].map(continent_mapping)

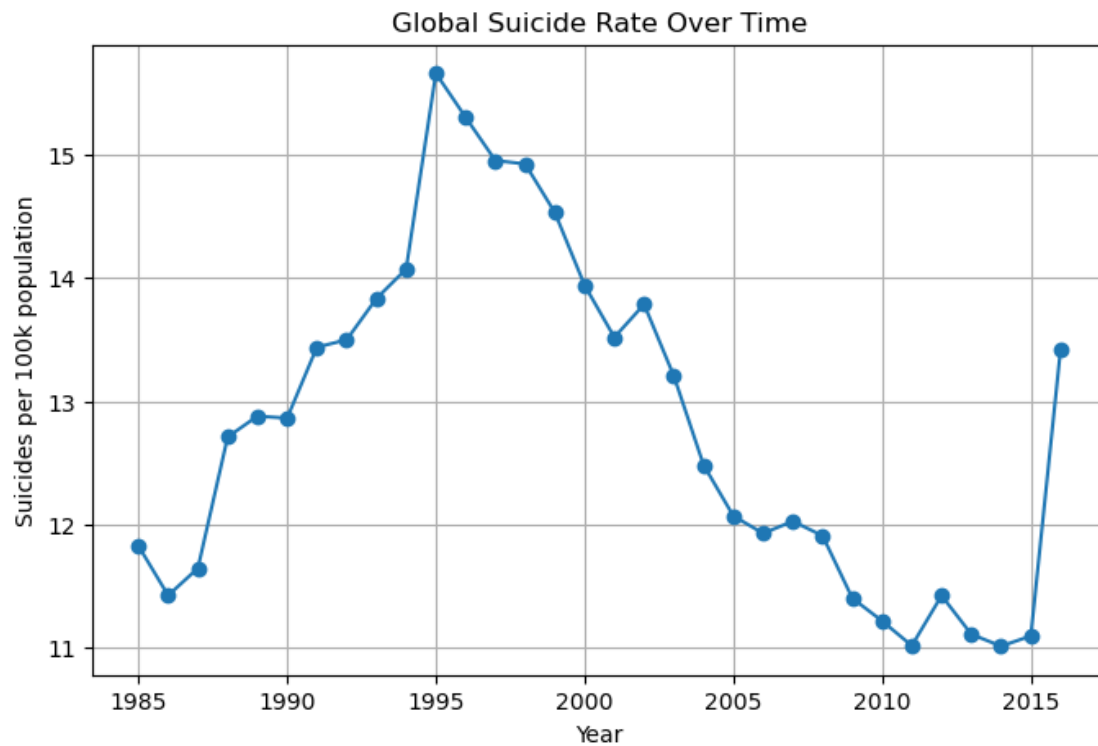
    # Missing values are filled as 'Other'
    df['Continent'] = df['Continent'].fillna('Other')

    # The result is displayed
    print(df[['country', 'Continent']].head())

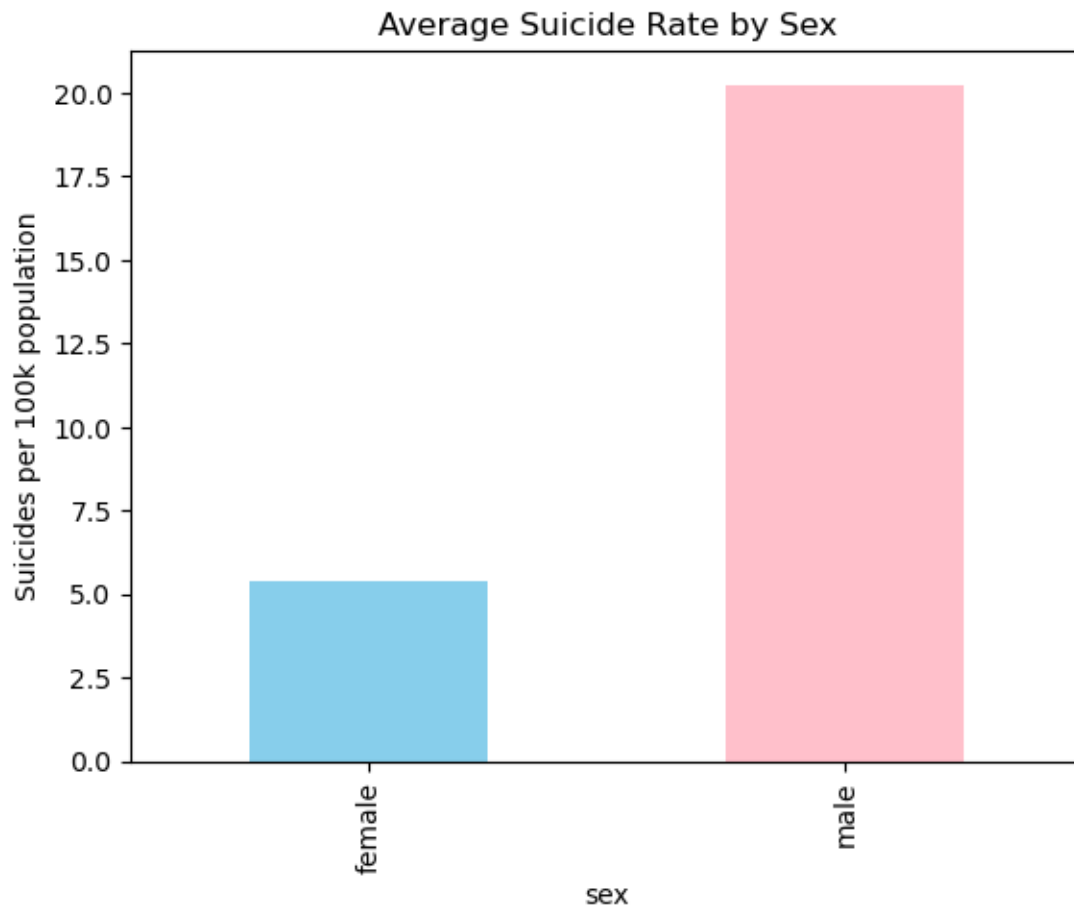
```

	country	Continent
0	Albania	Europe
1	Albania	Europe
2	Albania	Europe
3	Albania	Europe
4	Albania	Europe

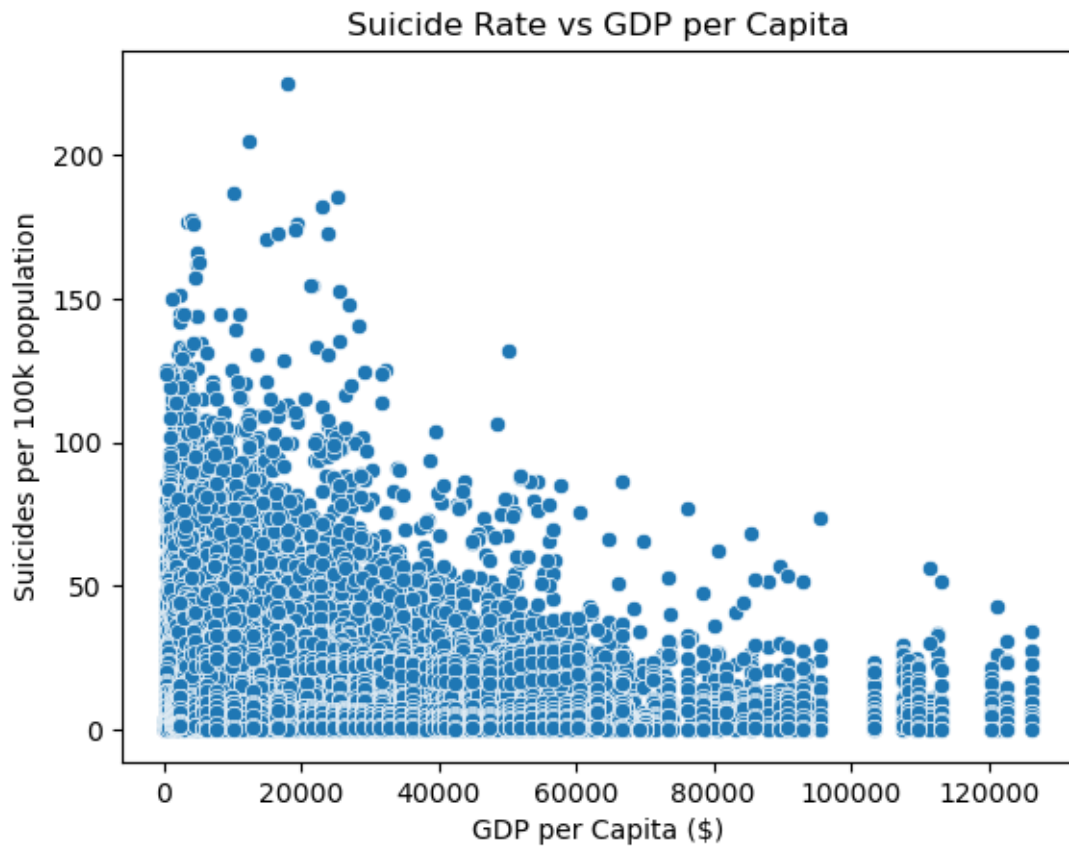
```
[7]: # The global suicide rate trend over the years is analyzed
df.groupby('year')['suicides/100k pop'].mean().plot(figsize=(8,5), marker='o')
plt.title('Global Suicide Rate Over Time')
plt.xlabel('Year')
plt.ylabel('Suicides per 100k population')
plt.grid(True)
plt.show()
```



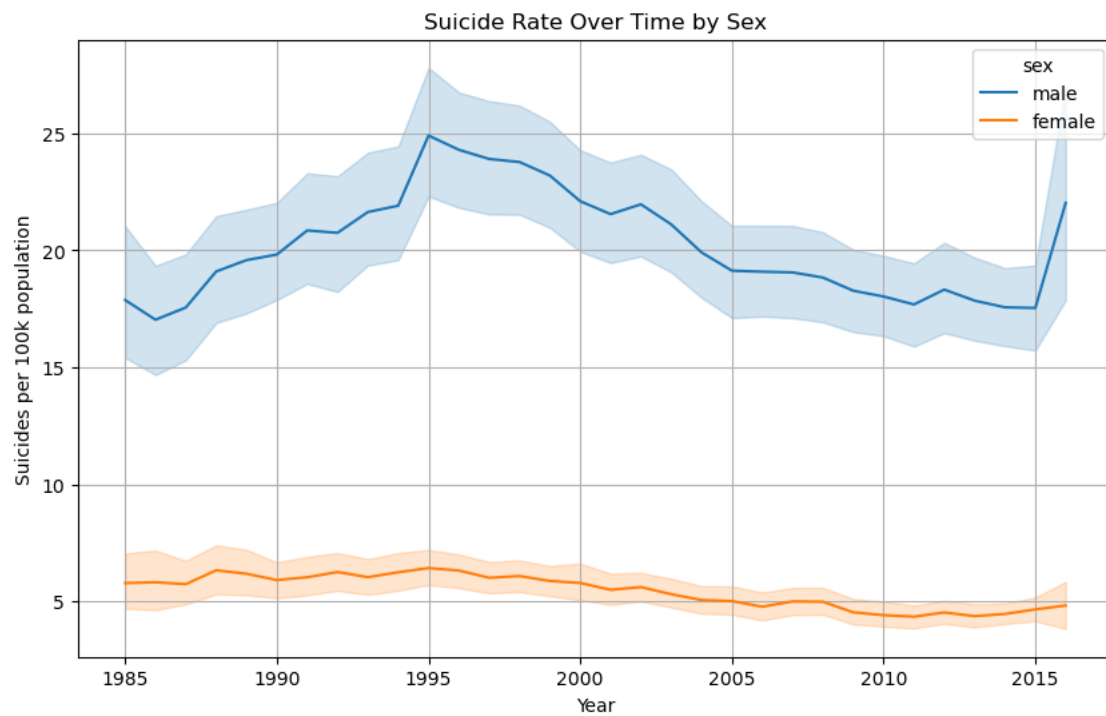
```
[9]: # The suicide rate between genders is compared globally
df.groupby('sex')['suicides/100k pop'].mean().plot(kind='bar', color=['skyblue', 'pink'])
plt.title('Average Suicide Rate by Sex')
plt.ylabel('Suicides per 100k population')
plt.show()
```



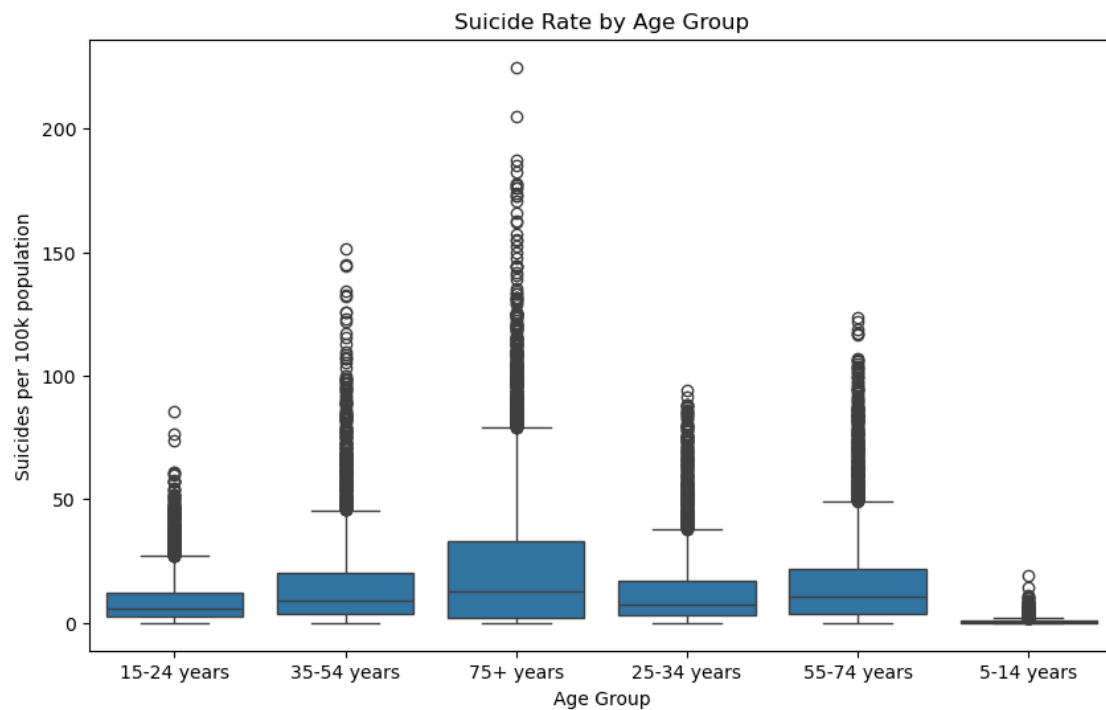
```
[11]: # The relationship between the suicide rate and GDP per capita is analyzed
sns.scatterplot(x='gdp_per_capita ($)', y='suicides/100k pop', data=df)
plt.title('Suicide Rate vs GDP per Capita')
plt.xlabel('GDP per Capita ($)')
plt.ylabel('Suicides per 100k population')
plt.show()
```



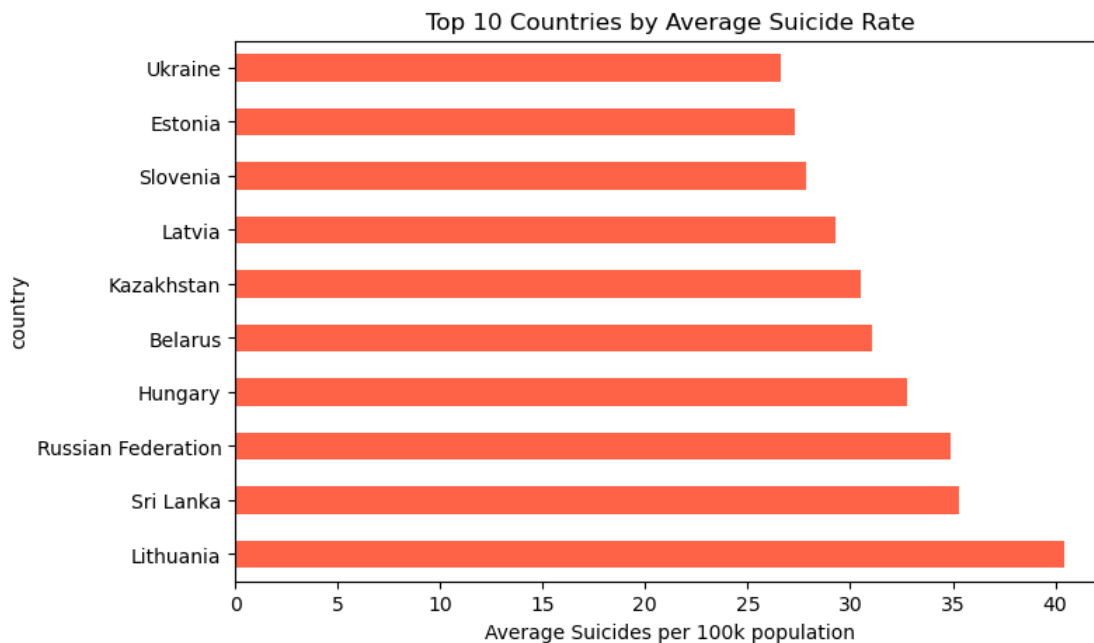

```
[13]: # The trend of the suicide rate by gender over the years is analyzed
plt.figure(figsize=(10,6))
sns.lineplot(x='year', y='suicides/100k pop', hue='sex', data=df)
plt.title('Suicide Rate Over Time by Sex')
plt.ylabel('Suicides per 100k population')
plt.xlabel('Year')
plt.grid(True)
plt.show()
```



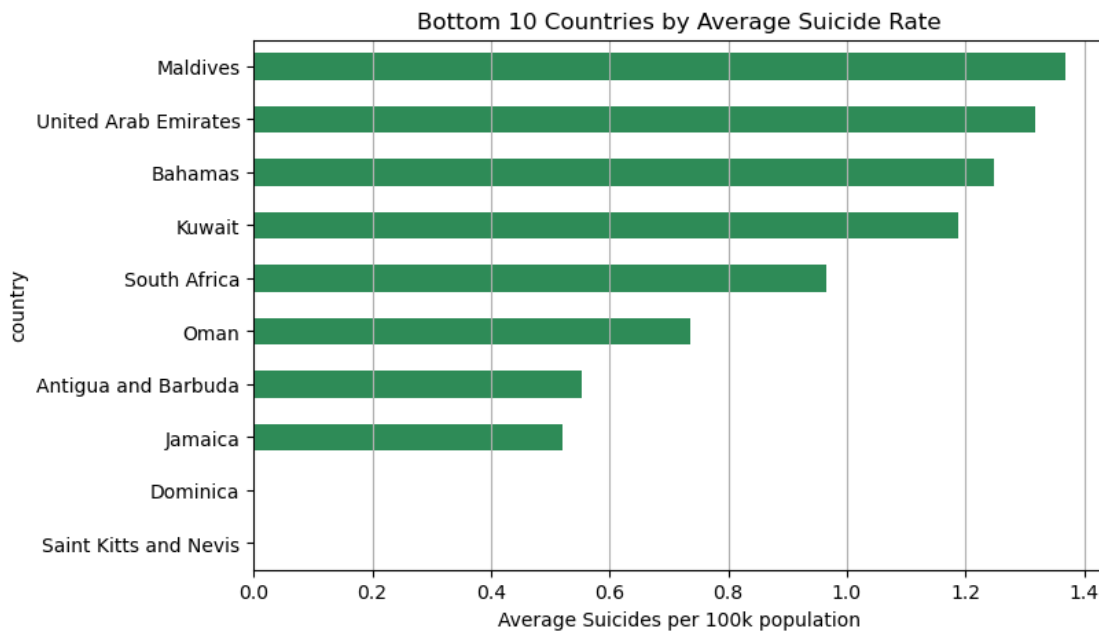
```
[17]: # The suicide rate by age group is compared
plt.figure(figsize=(10,6))
sns.boxplot(x='age', y='suicides/100k pop', data=df)
plt.title('Suicide Rate by Age Group')
plt.ylabel('Suicides per 100k population')
plt.xlabel('Age Group')
plt.show()
```



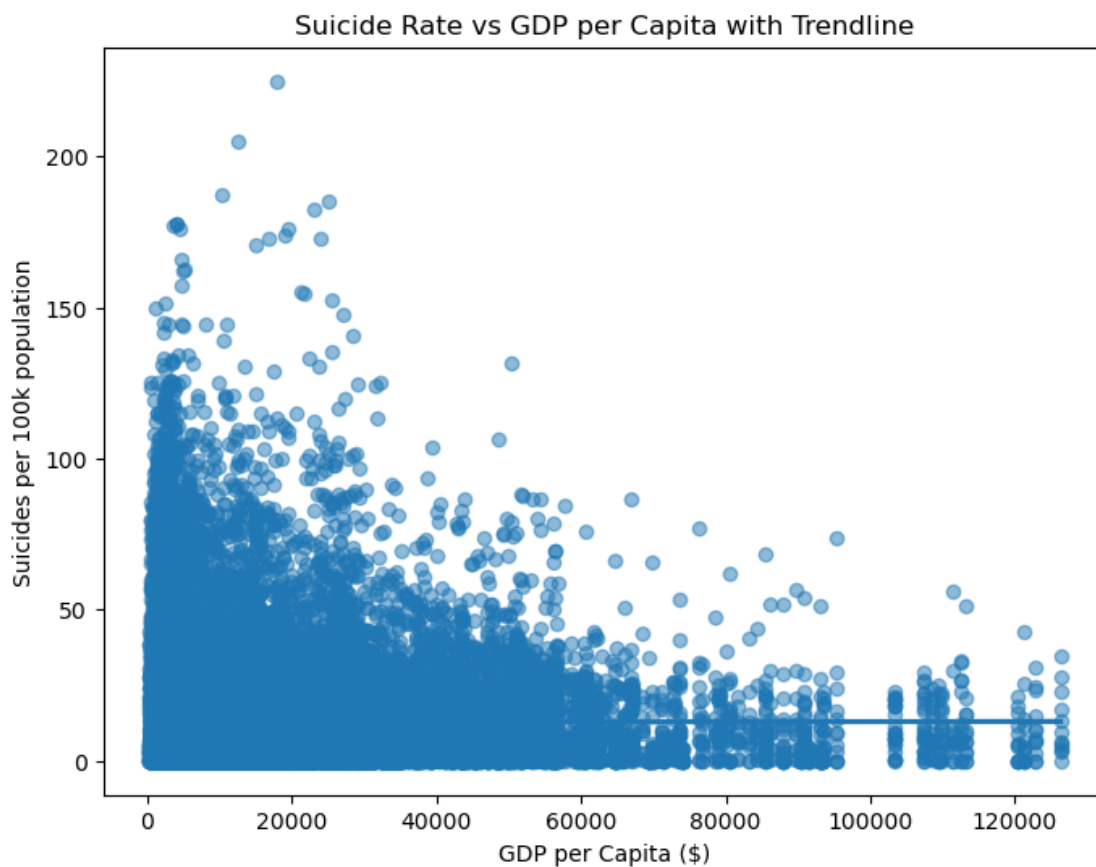
```
[19]: # The top 10 countries with the highest average suicide rate over the years are_
      ↪identified
top10_countries = df.groupby('country')['suicides/100k pop'].mean().
      ↪sort_values(ascending=False).head(10)
top10_countries.plot(kind='barh', figsize=(8,5), color='tomato')
plt.title('Top 10 Countries by Average Suicide Rate')
plt.xlabel('Average Suicides per 100k population')
plt.show()
```



```
[21]: # The bottom 10 countries with the lowest average suicide rate over the years
      ↪ are identified
lowest10_countries = df.groupby('country')['suicides/100k pop'].mean().
      ↪ sort_values().head(10)
lowest10_countries.plot(kind='barh', figsize=(8,5), color='seagreen')
plt.title('Bottom 10 Countries by Average Suicide Rate')
plt.xlabel('Average Suicides per 100k population')
plt.grid(axis='x')
plt.show()
```



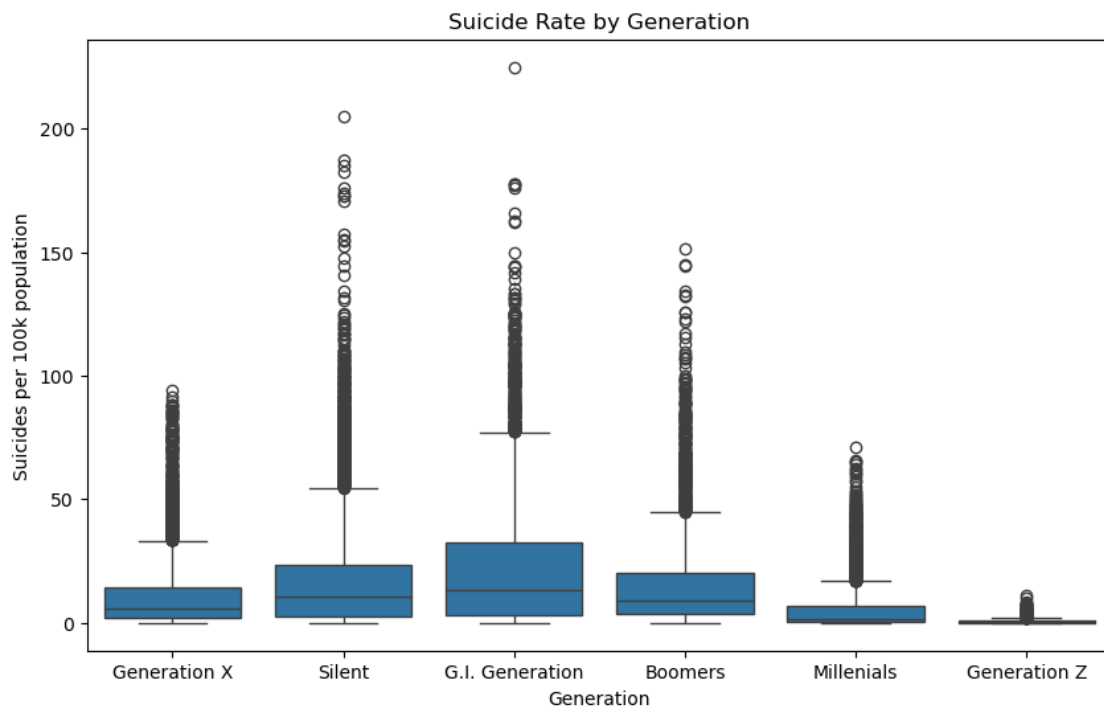
```
[23]: # The relationship between the suicide rate and GDP per capita is visualized
      ↪with a regression line
plt.figure(figsize=(8,6))
sns.regplot(x='gdp_per_capita ($)', y='suicides/100k pop', data=df,
      ↪scatter_kws={'alpha':0.5})
plt.title('Suicide Rate vs GDP per Capita with Trendline')
plt.xlabel('GDP per Capita ($)')
plt.ylabel('Suicides per 100k population')
plt.show()
```



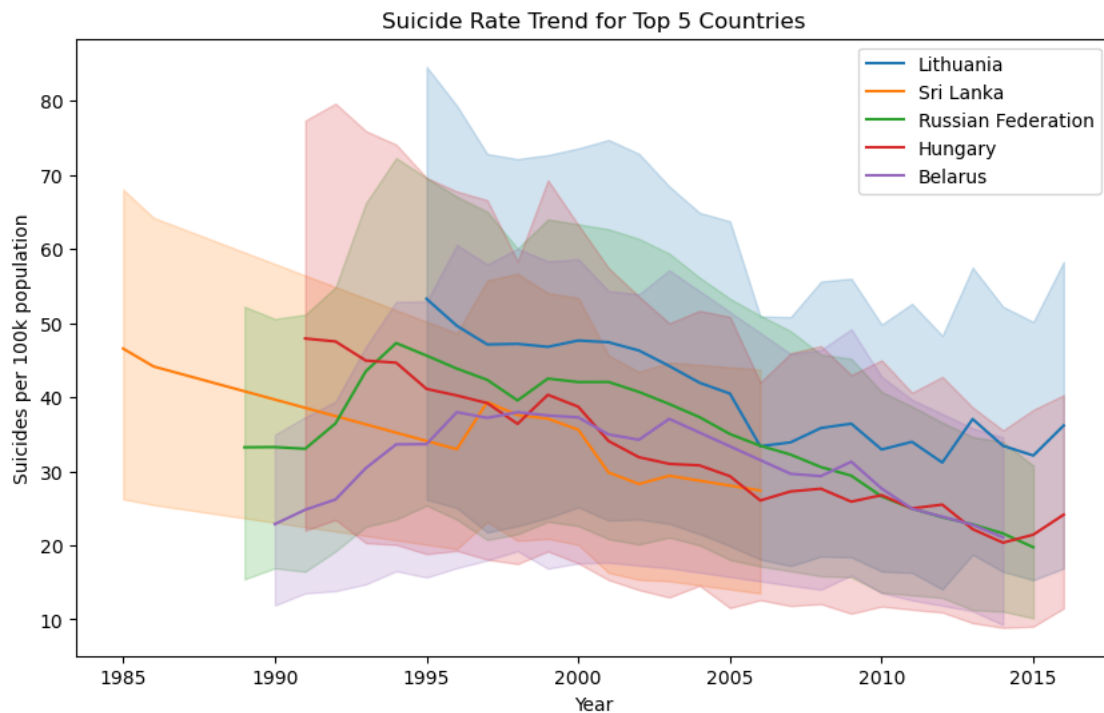
```
[29]: # The suicide rate trend over the years in a specific country (Ireland) is
      ↪ analyzed
country_name = 'Ireland'
ireland_data = df[df['country'] == country_name]
ireland_data.groupby('year')['suicides/100k pop'].mean().plot(figsize=(8,5),
      ↪marker='o')
plt.title(f'Suicide Rate Over Time in {country_name}')
plt.xlabel('Year')
plt.ylabel('Suicides per 100k population')
plt.grid(True)
plt.show()
```



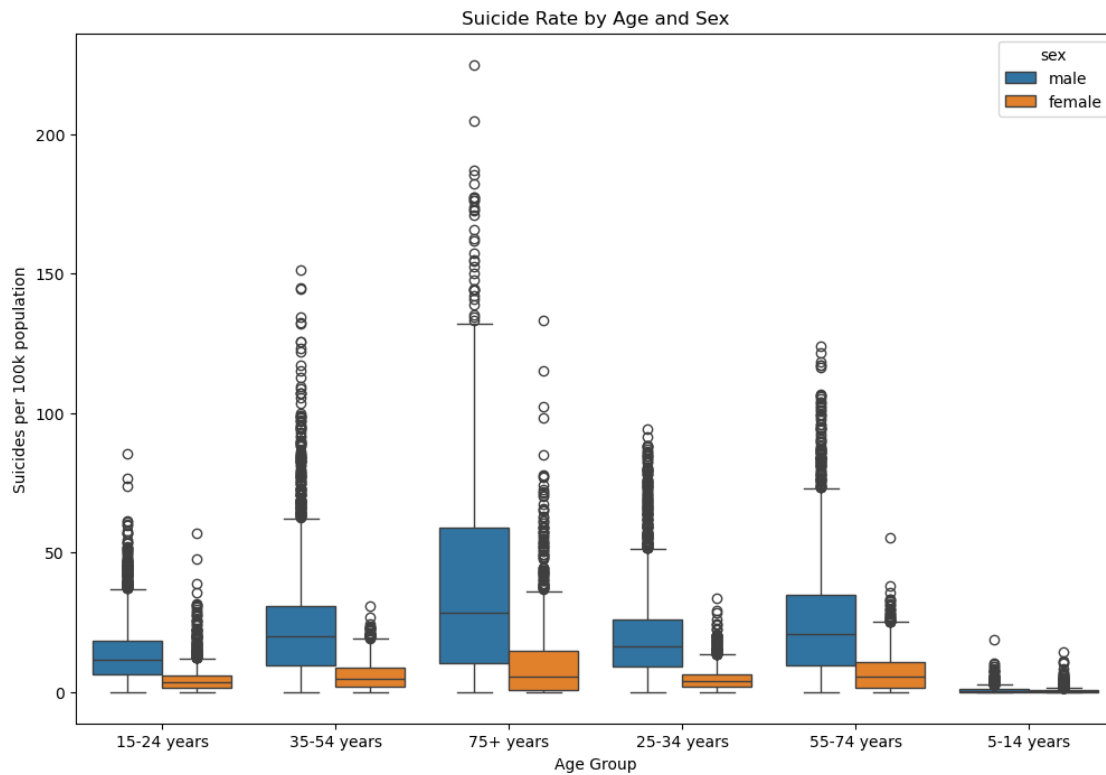
```
[33]: # The suicide rate is analyzed by generation
plt.figure(figsize=(10,6))
sns.boxplot(x='generation', y='suicides/100k pop', data=df)
plt.title('Suicide Rate by Generation')
plt.ylabel('Suicides per 100k population')
plt.xlabel('Generation')
plt.show()
```



```
[38]: # The annual change in suicide rates for the countries with the highest rates is
      ↪ analyzed
top_countries = df.groupby('country')['suicides/100k pop'].mean().
      ↪ sort_values(ascending=False).head(5).index
plt.figure(figsize=(10,6))
for country in top_countries:
    subset = df[df['country'] == country]
    sns.lineplot(x='year', y='suicides/100k pop', data=subset, label=country)
plt.title('Suicide Rate Trend for Top 5 Countries')
plt.ylabel('Suicides per 100k population')
plt.xlabel('Year')
plt.legend()
plt.show()
```




```
[40]: # The relationship between the suicide rate and both gender and age group is
      ↪ analyzed
plt.figure(figsize=(12,8))
sns.boxplot(x='age', y='suicides/100k pop', hue='sex', data=df)
plt.title('Suicide Rate by Age and Sex')
plt.ylabel('Suicides per 100k population')
plt.xlabel('Age Group')
plt.show()
```



```
[50]: # The cumulative suicide rates for the countries are calculated
total_suicides = df.groupby('country')['suicides_no'].sum().
    ↪sort_values(ascending=False)
print(total_suicides.head(25))
```

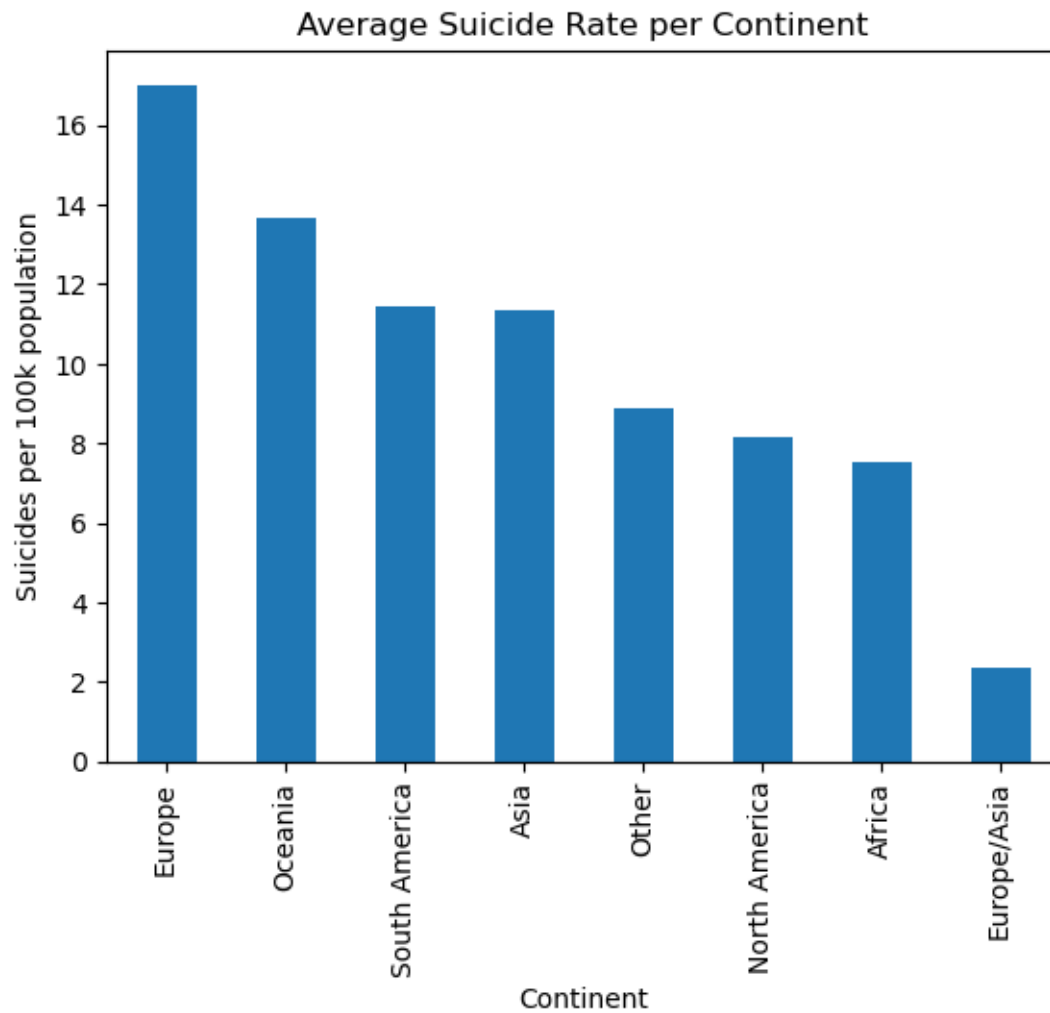
```
country
Russian Federation    1209742
United States         1034013
Japan                 806902
France                329127
Ukraine               319950
Germany               291262
Republic of Korea     261730
Brazil                226613
Poland                139098
United Kingdom        136805
Italy                 132060
Mexico                111139
Thailand              110643
Canada                107561
Kazakhstan            101546
Spain                 100202
Argentina              82219
Hungary               73891
Romania               72777
Australia             70111
Belgium               62761
Belarus               59892
Sri Lanka             55641
Colombia              53080
Netherlands           50833
Name: suicides_no, dtype: int64
```

```
[58]: # The average suicide rate for each continent over the years is calculated
continent_avg = df.groupby('Continent')['suicides/100k pop'].mean().
    ↪sort_values(ascending=False)
print(continent_avg)
```

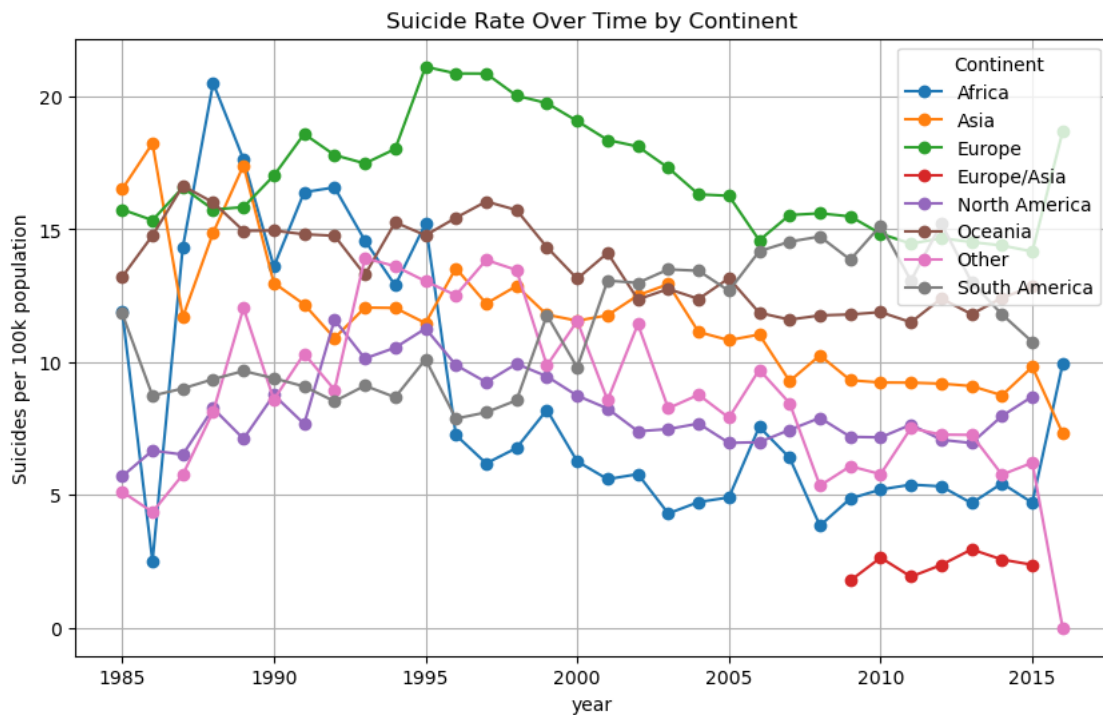
```
Continent
Europe          17.012629
Oceania          13.680410
South America    11.428745
Asia             11.355006
Other             8.856035
North America     8.145457
```

```
Africa          7.549952
Europe/Asia     2.371071
Name: suicides/100k pop, dtype: float64
```

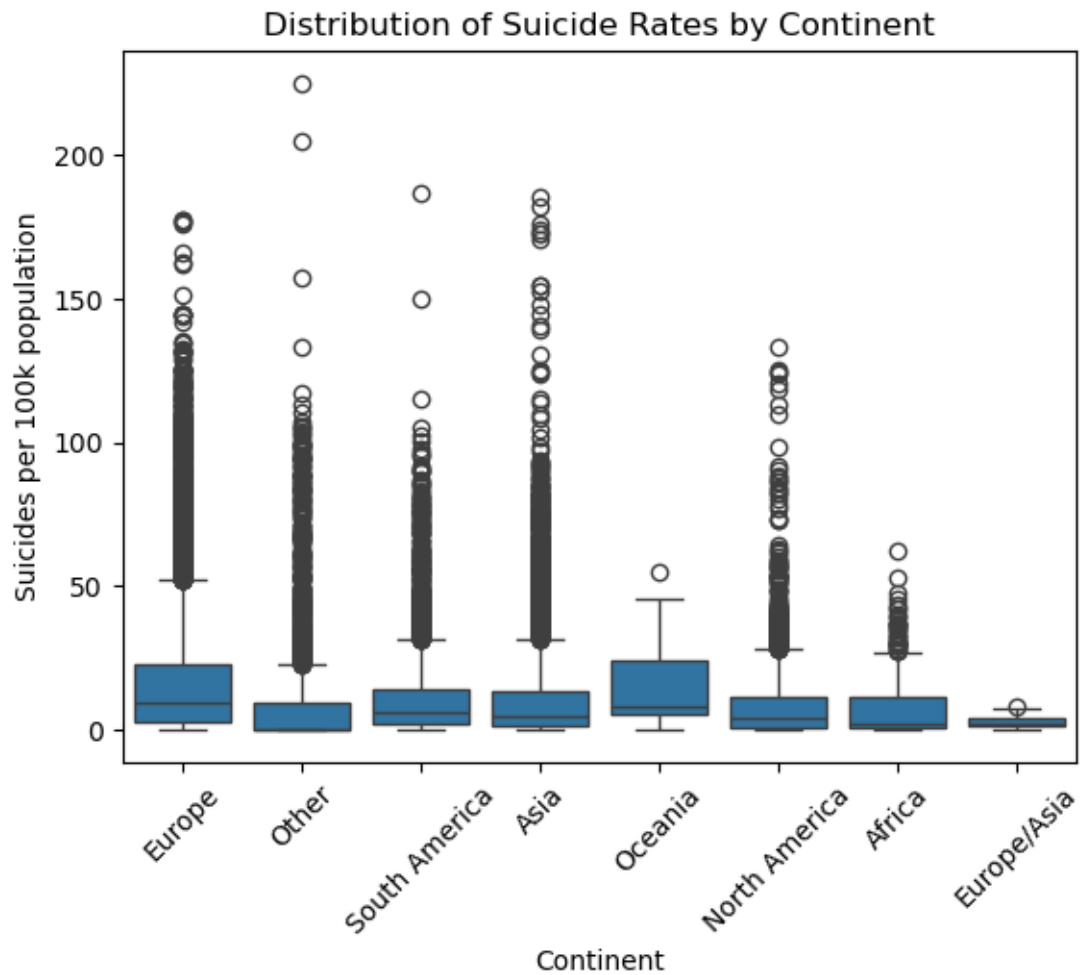
```
[60]: continent_avg.plot(kind='bar', title='Average Suicide Rate per Continent')
plt.ylabel('Suicides per 100k population')
plt.show()
```



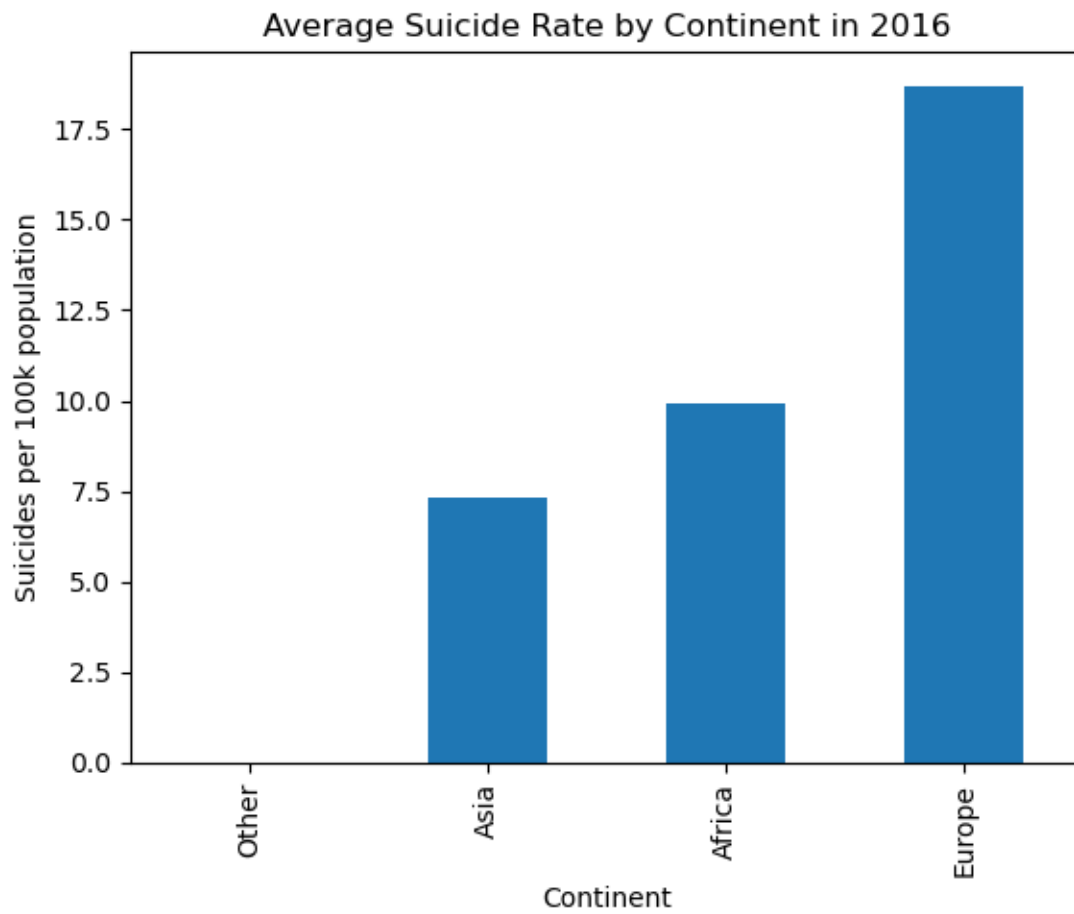
```
[64]: # The suicide rate trend over time for each continent is analyzed
continent_year = df.groupby(['year', 'Continent'])['suicides/100k pop'].mean().
↳unstack()
continent_year.plot(figsize=(10,6), marker='o', title='Suicide Rate Over Time by_
↳Continent')
plt.ylabel('Suicides per 100k population')
plt.grid(True)
plt.show()
```



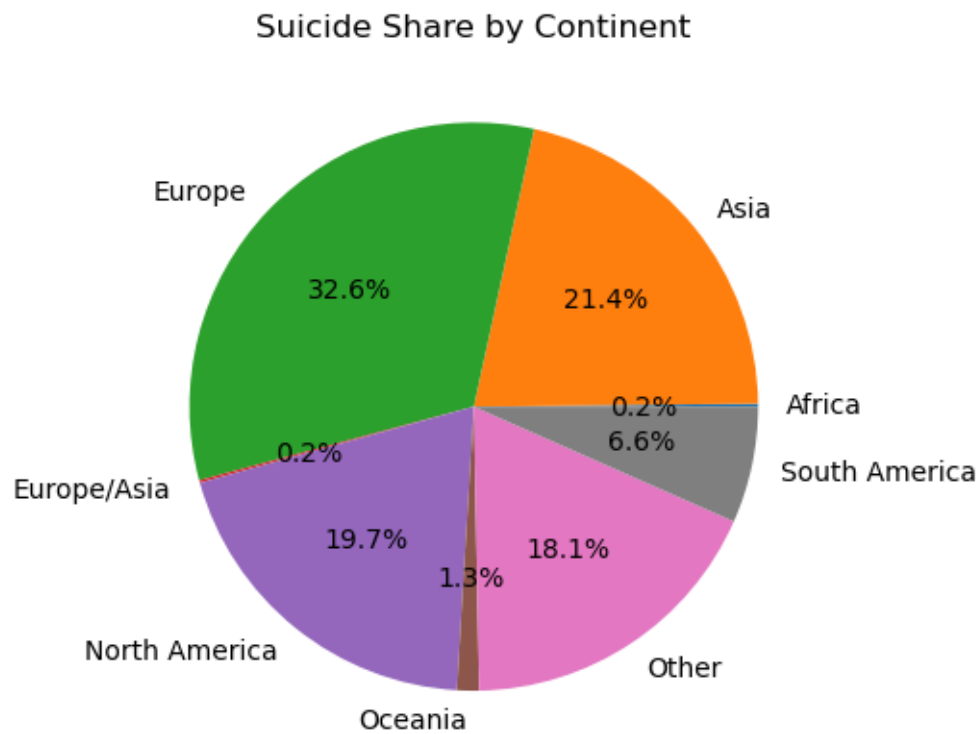
```
[54]: # A boxplot is created to compare suicide rates between continents
sns.boxplot(x='Continent', y='suicides/100k pop', data=df)
plt.title('Distribution of Suicide Rates by Continent')
plt.ylabel('Suicides per 100k population')
plt.xticks(rotation=45)
plt.show()
```



```
[70]: # The highest and lowest continents in terms of suicide rate in the latest year ↴  
      ↪ (2016) are identified  
avg_2016.plot(kind='bar', title='Average Suicide Rate by Continent in 2016')  
plt.ylabel('Suicides per 100k population')  
plt.show()
```



```
[68]: # The contribution percentage of each continent to the total number of suicides
      ↪ is calculated
suicide_share = df.groupby('Continent')['suicides_no'].sum()
suicide_share.plot(kind='pie', autopct='%1.1f%%', title='Suicide Share by
      ↪ Continent')
plt.ylabel('')
plt.show()
```



```
[74]: # The suicide rates by gender across continents are compared
sex_continent = df.groupby(['Continent', 'sex'])['suicides/100k pop'].mean().
↳unstack()
sex_continent.plot(kind='bar', figsize=(10,6), title='Suicide Rate by Sex and_
↳Continent')
plt.ylabel('Suicides per 100k population')
plt.show()
```

