

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

```
In [2]: data=pd.read_csv("country_wise_latest.csv")
```

```
In [3]: data.head()
```

Out[3]:

	Country/Region	Confirmed	Deaths	Recovered	Active	New cases	New deaths	New recovered	Deaths / 100 Cases
0	Afghanistan	27532	546	7660	19326	658	42	1502	1.98
1	Albania	1788	39	1086	663	66	1	9	2.18
2	Algeria	11385	811	8078	2496	117	12	135	7.12
3	Andorra	855	52	792	11	1	0	1	6.08
4	Angola	166	8	64	94	11	1	0	4.82

```
In [4]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 187 entries, 0 to 186
Data columns (total 15 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Country/Region                        187 non-null    object
1   Confirmed                            187 non-null    int64
2   Deaths                              187 non-null    int64
3   Recovered                            187 non-null    int64
4   Active                               187 non-null    int64
5   New cases                            187 non-null    int64
6   New deaths                           187 non-null    int64
7   New recovered                         187 non-null    int64
8   Deaths / 100 Cases                   187 non-null    float64
9   Recovered / 100 Cases                 187 non-null    float64
10  Deaths / 100 Recovered               187 non-null    float64
11  Confirmed last week                   187 non-null    int64
12  1 week change                         187 non-null    int64
13  1 week % increase                     187 non-null    float64
14  WHO Region                            187 non-null    object
dtypes: float64(4), int64(9), object(2)
memory usage: 22.0+ KB
```

```
In [5]: data.shape
```

Out[5]: (187, 15)

```
In [6]: data = data.rename(columns={'Country/Region': 'Country'})
data.head()
```

Out[6]:

	Country	Confirmed	Deaths	Recovered	Active	New cases	New deaths	New recovered	Deaths / 100 Cases	Re
0	Afghanistan	27532	546	7660	19326	658	42	1502	1.98	
1	Albania	1788	39	1086	663	66	1	9	2.18	
2	Algeria	11385	811	8078	2496	117	12	135	7.12	
3	Andorra	855	52	792	11	1	0	1	6.08	
4	Angola	166	8	64	94	11	1	0	4.82	

```
In [7]: countries_by_deaths = data.sort_values(by='Deaths', ascending=True)
print(countries_by_deaths[['Country', 'Deaths', 'WHO Region']])
```

	Country	Deaths	WHO Region
142	Saint Vincent and the Grenadines	0	Americas
75	Holy See	0	Europe
130	Papua New Guinea	0	Western Pacific
118	Namibia	0	Africa
30	Cambodia	0	Western Pacific
..	...	...	...
61	France	29606	Europe
85	Italy	34514	Europe
177	United Kingdom	42373	Europe
23	Brazil	47748	Americas
173	US	118434	Americas

[187 rows x 3 columns]

```
In [8]: data.isnull().sum()
```

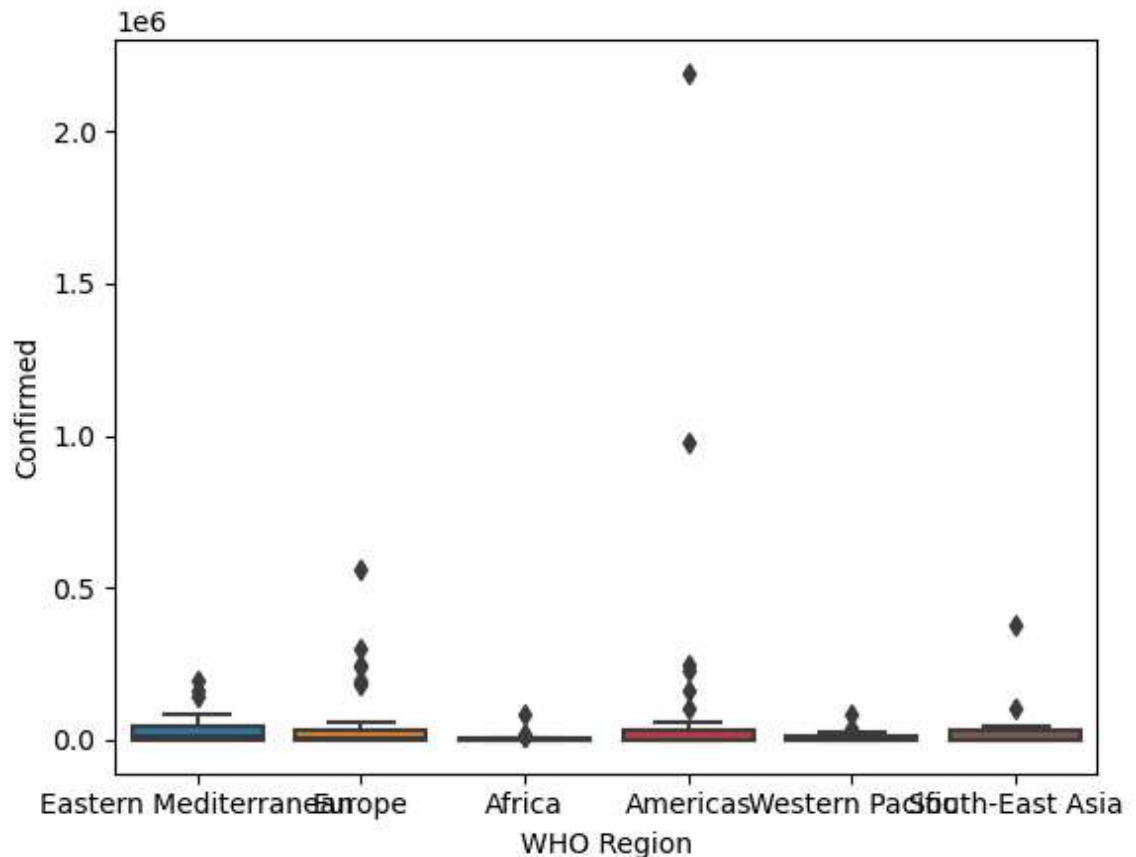
```
Out[8]: Country          0
Confirmed          0
Deaths            0
Recovered          0
Active            0
New cases          0
New deaths         0
New recovered      0
Deaths / 100 Cases 0
Recovered / 100 Cases 0
Deaths / 100 Recovered 0
Confirmed last week 0
1 week change      0
1 week % increase  0
WHO Region         0
dtype: int64
```

```
In [9]: data.columns
```

```
Out[9]: Index(['Country', 'Confirmed', 'Deaths', 'Recovered', 'Active', 'New case  
s',  
            'New deaths', 'New recovered', 'Deaths / 100 Cases',  
            'Recovered / 100 Cases', 'Deaths / 100 Recovered',  
            'Confirmed last week', '1 week change', '1 week % increase',  
            'WHO Region'],  
            dtype='object')
```

```
In [10]: sns.boxplot(x="WHO Region",y="Confirmed",data=data)
```

```
Out[10]: <Axes: xlabel='WHO Region', ylabel='Confirmed'>
```

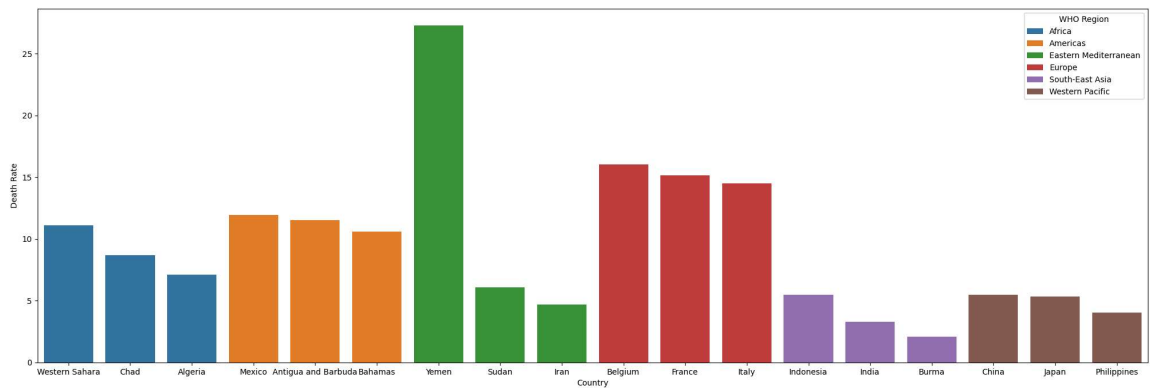


## finding death rate of top 3 countries in each region

```
In [11]: data['Death Rate'] = (data['Deaths'] / data['Confirmed']) * 100  
top_countries_per_region = data.groupby('WHO Region').apply(  
    lambda x: x.nlargest(3, 'Death Rate')  
) .reset_index(drop=True)
```

```
In [12]: plt.figure(figsize=(25, 8))
sns.barplot(x='Country', y='Death Rate', hue='WHO Region', data=top_countri
```

```
Out[12]: <Axes: xlabel='Country', ylabel='Death Rate'>
```

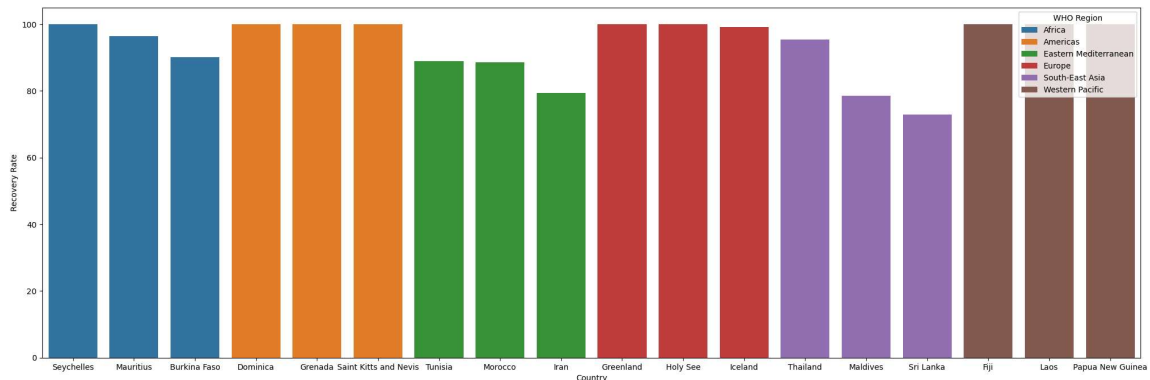


## finding recovery rate of top 3 countries in each region¶

```
In [13]: data['Recovery Rate'] = (data['Recovered'] / data['Confirmed']) * 100
top_countries_per_region = data.groupby('WHO Region').apply(
    lambda x: x.nlargest(3, 'Recovery Rate')
).reset_index(drop=True)
```

```
In [14]: plt.figure(figsize=(25, 8))
sns.barplot(x='Country', y='Recovery Rate', hue='WHO Region', data=top_coun
```

```
Out[14]: <Axes: xlabel='Country', ylabel='Recovery Rate'>
```



## finding countries with most cases in each region top 3

```
In [15]: top_countries_most_cases_per_region = data.groupby('WHO Region').apply(
          lambda x: x.nlargest(3, 'Confirmed')
        ).reset_index(drop=True)
print(top_countries_most_cases_per_region[['WHO Region', 'Country', 'Confirmed']])
```

	WHO Region	Country	Confirmed
0	Africa	South Africa	83890
1	Africa	Nigeria	18480
2	Africa	Ghana	12929
3	Americas	US	2191052
4	Americas	Brazil	978142
5	Americas	Peru	244388
6	Eastern Mediterranean	Iran	197647
7	Eastern Mediterranean	Pakistan	165062
8	Eastern Mediterranean	Saudi Arabia	145991
9	Europe	Russia	560321
10	Europe	United Kingdom	301935
11	Europe	Spain	245268
12	South-East Asia	India	380532
13	South-East Asia	Bangladesh	102292
14	South-East Asia	Indonesia	42762
15	Western Pacific	China	84494
16	Western Pacific	Singapore	41473
17	Western Pacific	Philippines	27799

## comparision by region

```
In [16]: regional_data = data.groupby('WHO Region').agg({
          'Confirmed': 'sum',
          'Deaths': 'sum',
          'Recovered': 'sum'
        }).reset_index()
```

```
In [17]: regional_data['Death Rate (%)'] = (regional_data['Deaths'] / regional_data[
          regional_data['Recovery Rate (%)'] = (regional_data['Recovered'] / regional_data['Confirmed']) * 100)
```

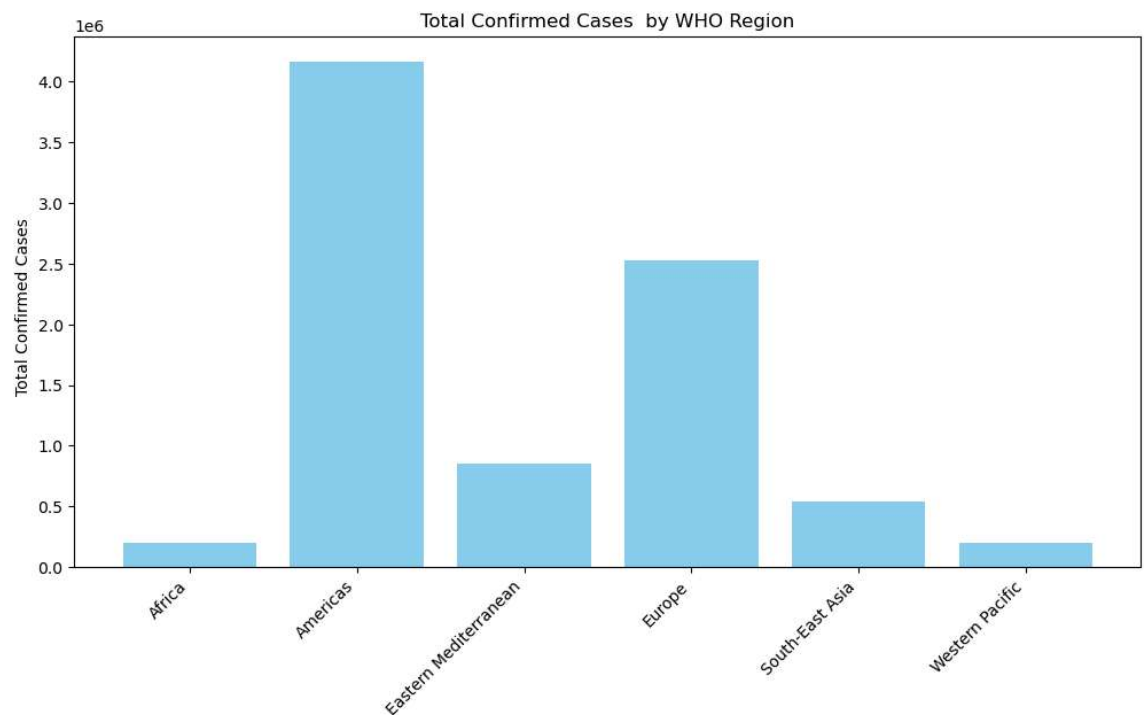
```
In [18]: print(regional_data)
```

	WHO Region	Confirmed	Deaths	Recovered	Death Rate (%) \
0	Africa	199343	4591	95773	2.303066
1	Americas	4164325	215632	1675117	5.178078
2	Eastern Mediterranean	856105	19036	514839	2.223559
3	Europe	2525014	191096	1372022	7.568116
4	South-East Asia	541036	16360	269166	3.023828
5	Western Pacific	202412	7250	163465	3.581803

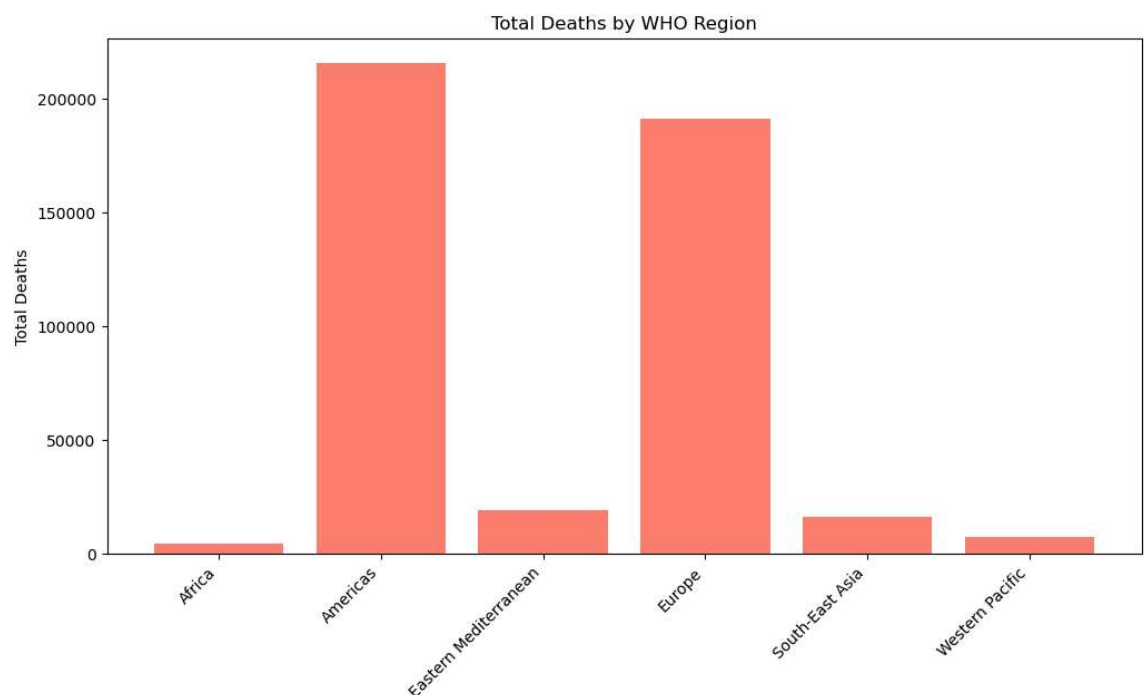
  

	Recovery Rate (%)
0	48.044326
1	40.225415
2	60.137366
3	54.337204
4	49.750109
5	80.758552

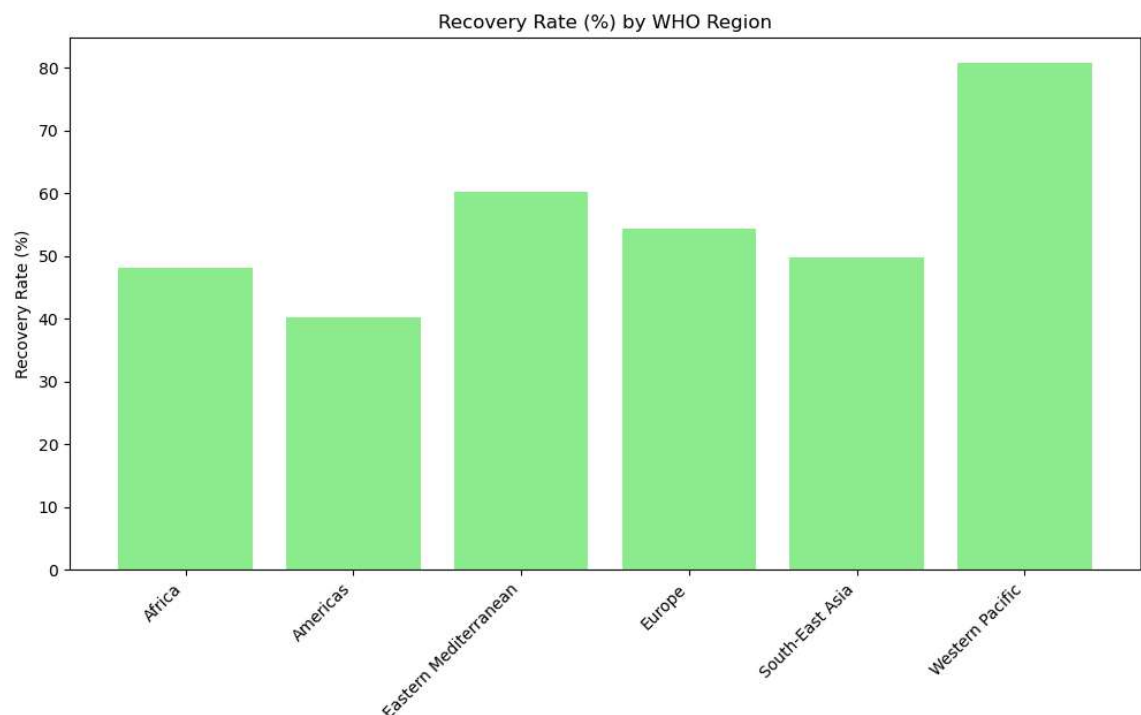
```
In [19]: plt.figure(figsize=(12, 6))
plt.bar(regional_data['WHO Region'], regional_data['Confirmed'], color='sky')
plt.title('Total Confirmed Cases by WHO Region')
plt.ylabel('Total Confirmed Cases ')
plt.xticks(rotation=45, ha='right')
plt.show()
```



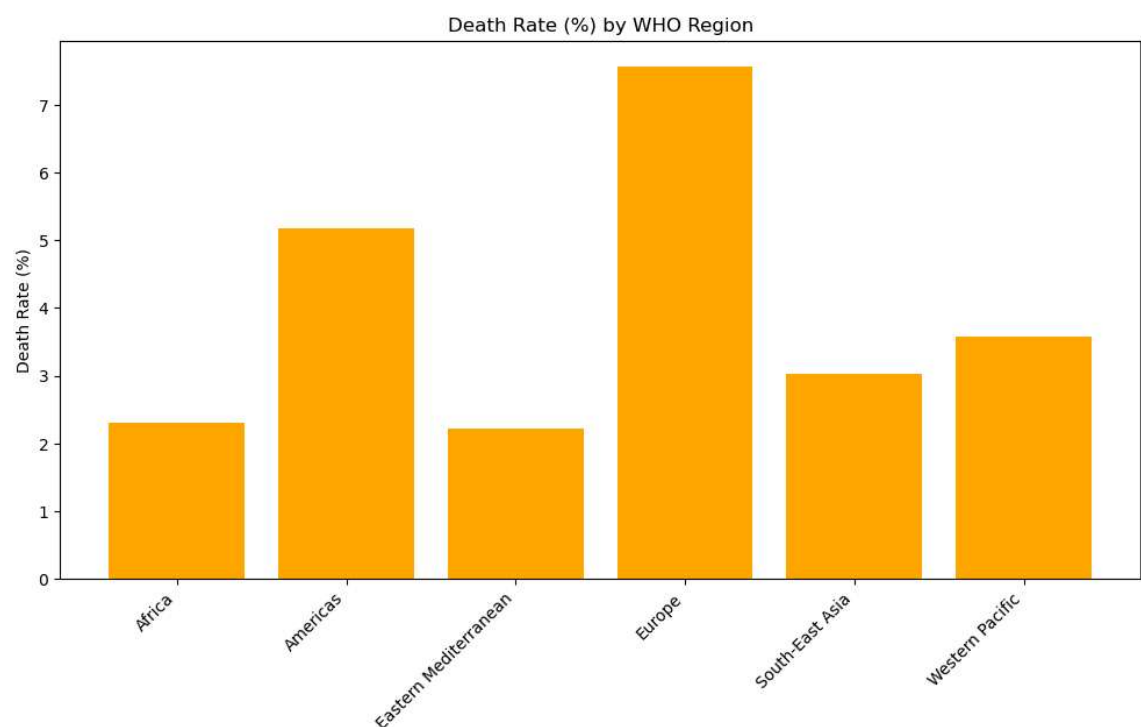
```
In [20]: plt.figure(figsize=(12, 6))
plt.bar(regional_data['WHO Region'], regional_data['Deaths'], color='salmon')
plt.title('Total Deaths by WHO Region')
plt.ylabel('Total Deaths')
plt.xticks(rotation=45, ha='right')
plt.show()
```



```
In [21]: plt.figure(figsize=(12, 6))
plt.bar(regional_data['WHO Region'], regional_data['Recovery Rate (%)'], color='green')
plt.title('Recovery Rate (%) by WHO Region')
plt.ylabel('Recovery Rate (%)')
plt.xticks(rotation=45, ha='right')
plt.show()
```



```
In [22]: plt.figure(figsize=(12, 6))
plt.bar(regional_data['WHO Region'], regional_data['Death Rate (%)'], color='orange')
plt.title('Death Rate (%) by WHO Region')
plt.ylabel('Death Rate (%)')
plt.xticks(rotation=45, ha='right')
plt.show()
```



## Situation According to Death & Recovery Rate

```
In [27]: data['Recovery Rate (%)'] = (data['Recovered'] / data['Confirmed']) * 100
data['Death Rate (%)'] = (data['Deaths'] / data['Confirmed']) * 100

recovery_criteria = data['Recovery Rate (%)'].mean()
death_criteria = data['Death Rate (%)'].mean()

data['Situation'] = data.apply(
    lambda row: 'Critical' if (row['Recovery Rate (%)'] < recovery_criteria
    else 'Not Critical', axis=1
)

critical_countries = data[data['Critical'] == 'Critical']

print(critical_countries[['Country', 'Recovery Rate (%)', 'Death Rate (%)'],
```

	Country	Recovery Rate (%)	Death Rate (%)	Situation
4	Angola	38.554217	4.819277	Critical
16	Belgium	27.712600	16.045271	Critical
23	Brazil	54.652596	4.881500	Critical
25	Bulgaria	52.830702	5.171475	Critical
32	Canada	0.000000	8.207021	Critical
37	Colombia	37.343100	3.417713	Critical
51	Ecuador	49.133348	8.324338	Critical
52	Egypt	26.821579	3.842417	Critical
61	France	37.901491	15.161416	Critical
67	Greece	42.578246	5.825844	Critical
70	Guatemala	19.295585	3.783283	Critical
73	Guyana	55.737705	6.557377	Critical
80	Indonesia	39.282541	5.469810	Critical
82	Iraq	44.068126	3.328538	Critical
98	Liberia	46.125461	6.088561	Critical
109	Mauritania	22.689769	4.001650	Critical
112	Moldova	55.333435	3.387761	Critical
120	Netherlands	0.361470	12.312197	Critical
125	North Macedonia	39.365352	4.631218	Critical
133	Philippines	25.504515	4.014533	Critical
134	Poland	49.385781	4.243108	Critical
149	Sierra Leone	55.817610	4.009434	Critical
159	Sudan	36.982544	6.072319	Critical
161	Sweden	0.000000	9.016291	Critical
163	Syria	0.000000	3.743316	Critical
166	Tanzania	35.952849	4.125737	Critical
173	US	27.343714	5.405349	Critical
177	United Kingdom	0.434862	14.033815	Critical
184	Yemen	30.033003	27.282728	Critical

In [ ]: