#### DATA SCIENCE COHORT 2 FINAL PROJECT

Capstone Project: Predictive Modelling for COVID-19 Using The Country Wise Dataset

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```
In [3]: import numpy as np
        import pandas as pd
        import seaborn as sns
        import plotly.express as px
        import matplotlib.pyplot as plt
        import warnings
        warnings.filterwarnings('ignore')
         # import the dataset
In [4]:
        data = pd.read csv(r"C:\Users\iwund\Desktop\3MTT Final Project\country wise latest.csv")
In [5]: # Shape of dataset
        # print(f"The dataset has {data.shape[1]} columns and {data.shape[0]} rows")
        print("The number of columns is: ", data.shape[1])
        print("The number of rows is: ", data.shape[0])
        The number of columns is: 15
        The number of rows is: 187
In [6]: # Overview of the dataset
        print(data.info()) # Data types and non-null counts
```

```
<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)							

dtypes: float64(4), int64(9), object(2)

memory usage: 22.0+ KB

None

In [7]: data.describe() # Summary statistics

	Confirmed	Deaths	Recovered	Active	New cases	New deaths	New recovered	Deaths / 100 Cases	/ 100 Cases	100 Recovered	Confirr last w
count	1.870000e+02	187.000000	1.870000e+02	1.870000e+02	187.000000	187.000000	187.000000	187.000000	187.000000	187.00	1.870000e
mean	8.813094e+04	3497.518717	5.063148e+04	3.400194e+04	1222.957219	28.957219	933.812834	3.019519	64.820535	inf	7.868248e
std	3.833187e+05	14100.002482	1.901882e+05	2.133262e+05	5710.374790	120.037173	4197.719635	3.454302	26.287694	NaN	3.382737e
min	1.000000e+01	0.000000	0.000000e+00	0.000000e+00	0.000000	0.000000	0.000000	0.000000	0.000000	0.00	1.000000e
25%	1.114000e+03	18.500000	6.265000e+02	1.415000e+02	4.000000	0.000000	0.000000	0.945000	48.770000	1.45	1.051500e
50%	5.059000e+03	108.000000	2.815000e+03	1.600000e+03	49.000000	1.000000	22.000000	2.150000	71.320000	3.62	5.020000e
75%	4.046050e+04	734.000000	2.260600e+04	9.149000e+03	419.500000	6.000000	221.000000	3.875000	86.885000	6.44	3.708050e
max	4.290259e+06	148011.000000	1.846641e+06	2.816444e+06	56336.000000	1076.000000	33728.000000	28.560000	100.000000	inf	3.834677e

Doaths /

Recovered

### Key Insights from the Analysis

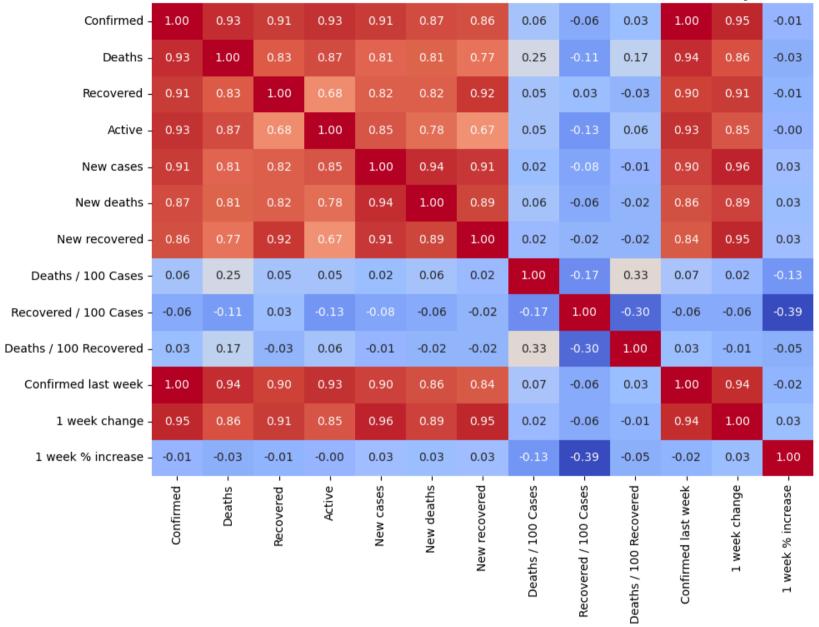
- 1. Top 10 Countries by Confirmed Cases The top 10 countries dominate the total confirmed cases globally, with significant disparities between them. Leading countries such as the U.S., Brazil, and India show much higher confirmed case counts compared to others.
- 2. Distribution of Death Rates Death rates range from 0% to a maximum of approximately 28.5%. The majority of countries have a death rate below 5%, indicating that fatality rates are concentrated within this range. Some outliers with higher fatality rates warrant further investigation, potentially related to healthcare system quality or population demographics.
- 3. Distribution of Recovery Rates Recovery rates vary widely, with most countries having rates above 60%. A significant proportion of countries approach 100% recovery rates, suggesting effective healthcare responses in these regions.
- 4. Total Cases by WHO Region The Americas region accounts for the majority of confirmed cases globally, followed by Europe and South-East Asia. Regions such as Africa and the Western Pacific report relatively lower confirmed case counts, possibly due to lower transmission rates or underreporting.

#### **General Observations**

The dataset highlights stark contrasts in COVID-19 impacts across regions and countries. Higher case and death rates are often concentrated in regions with large populations and dense urban centers (e.g., Americas and Europe). Active cases remain substantial, suggesting the ongoing need for public health measures and vaccinations in many regions. The rate of weekly increase in cases provides insight into countries still experiencing significant outbreaks

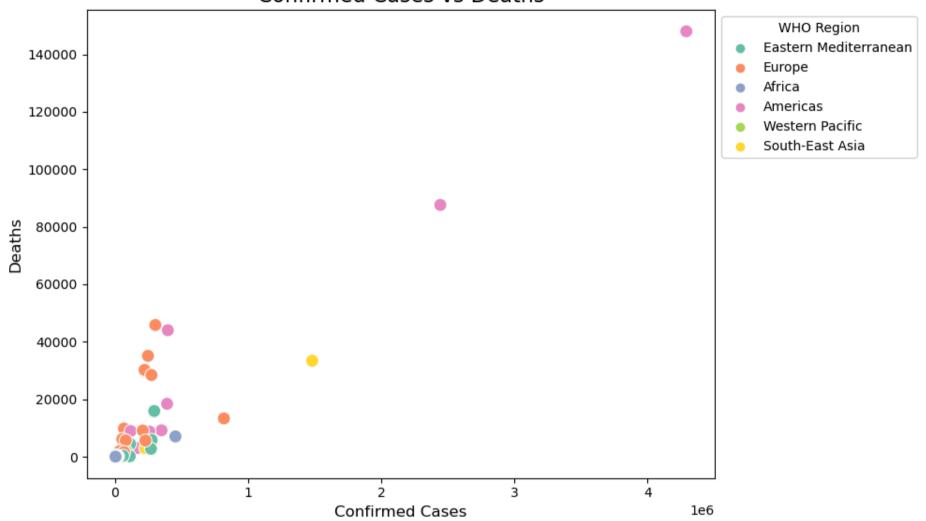
```
In [11]: # Drop non-numeric columns for correlation computation
          numeric data = data.select dtypes(include=["float64", "int64"])
          correlation matrix fixed = numeric data.corr()
          # Replot the heatmap with fixed data
          plt.figure(figsize=(12, 8))
          sns.heatmap(correlation matrix fixed, annot=True, fmt=".2f", cmap="coolwarm", cbar=True)
          plt.title("Correlation Matrix of COVID-19 Metrics (Numeric Data Only)", fontsize=16)
          plt.tight layout()
          plt.show()
          correlation matrix fixed
          # Scatter plot: Confirmed vs Deaths
          plt.figure(figsize=(10, 6))
          sns.scatterplot(x="Confirmed", y="Deaths", data=data, hue="WHO Region", palette="Set2", s=100)
          plt.title("Confirmed Cases vs Deaths", fontsize=16)
          plt.xlabel("Confirmed Cases", fontsize=12)
          plt.ylabel("Deaths", fontsize=12)
          plt.legend(title="WHO Region", loc="upper left", bbox to anchor=(1, 1))
          plt.tight layout()
          plt.show()
          # Scatter plot: Active cases vs Recovery rate
          plt.figure(figsize=(10, 6))
          sns.scatterplot(x="Active", y="Recovered / 100 Cases", data=data, hue="WHO Region", palette="Set1", s=100)
          plt.title("Active Cases vs Recovery Rate", fontsize=16)
          plt.xlabel("Active Cases", fontsize=12)
          plt.ylabel("Recovery Rate (%)", fontsize=12)
          plt.legend(title="WHO Region", loc="upper left", bbox to anchor=(1, 1))
          plt.tight layout()
          plt.show()
```

## Correlation Matrix of COVID-19 Metrics (Numeric Data Only)



1.0 - 0.8 - 0.6 - 0.4 - 0.2 - 0.0 - -0.2

## Confirmed Cases vs Deaths



# Active Cases vs Recovery Rate WHO Region 100 Eastern Mediterranean Europe Africa Americas Western Pacific South-East Asia Recovery Rate (%) 60 40 20 0 0.0 0.5 1.0 1.5 2.0 2.5

Insights from Correlation and Plots

### **Correlation Analysis**

Confirmed Cases: Strong positive correlation with Deaths (0.93) and Recovered (0.91), indicating that higher case counts are associated with both outcomes. Strong correlation with Active Cases (0.93), showing consistent patterns in countries with high case loads. Death Rate (Deaths / 100

1e6

**Active Cases** 

Cases): Weak or negligible correlation with total confirmed cases, suggesting that other factors (e.g., healthcare quality, population demographics) have a more significant impact on fatality outcomes.

### **Scatter Plots**

Confirmed Cases vs Deaths The scatterplot reveals a proportional relationship, with some outliers where deaths are disproportionately high relative to confirmed cases. Regions: The Americas and Europe dominate the higher ends of the spectrum, indicating higher case loads and fatalities.

### Active Cases vs Recovery Rates

Trend: Countries with fewer active cases tend to have higher recovery rates, highlighting the effectiveness of containment and healthcare systems. Outliers: Some countries exhibit high active cases with moderate recovery rates, which may point to ongoing outbreaks or challenges in healthcare management.

In [ ]: