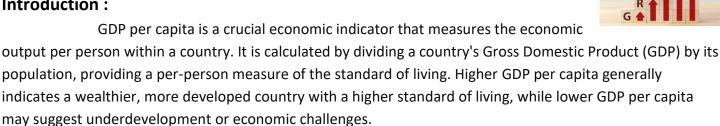
## **Interactive Data Visualization**

Python Data Analysis Project

#### Introduction:



This analysis aims to explore the GDP per Capita trend over time, comparing it across multiple countries for a selected year. By examining the GDP per capita data, we gain valuable insights into how countries are performing economically, allowing us to understand their relative wealth and development.

#### **DataSet Overview: -**

The dataset provides economic data, specifically GDP per Capita for various countries over time.

- **Country:** This column contains the names of countries for which GDP per capita data is provided. It serves as the primary identifier for each country's data.
- Code: This column contains a short code or abbreviation for each country. It is typically used for referencing countries in a compact form (e.g., "USA" for the United States, "GBR" for the United Kingdom).
- Year: This column represents the year for which the GDP per capita is recorded. The dataset includes historical data, spanning across multiple years for each country.
- GDP per capita: This column contains the Gross Domestic Product per Capita for each country and year, which is calculated as the total GDP of a country divided by its population. The value is usually expressed in constant US dollars to account for inflation across different years.

### **Importance of Analysis:**

GDP per capita analysis is crucial for assessing a country's economic development and understanding global wealth distribution. It serves as an indicator of a nation's standard of living, with higher GDP per capita generally reflecting better living conditions, access to healthcare, and overall economic well-being. By comparing GDP per capita across countries, we can identify disparities in economic development and wealth distribution, helping policymakers and international organizations target aid, trade policies, and development programs effectively. It also allows for tracking economic growth, identifying periods of recession or expansion, and evaluating the impact of economic policies.

Furthermore, GDP per capita analysis plays a vital role in global economic planning and forecasting. It helps businesses make informed investment decisions, aids governments in designing long-term development strategies, and supports international organizations in allocating resources effectively. By examining trends



over time, this analysis also provides insights into a country's potential for future growth, poverty alleviation, and social policy formulation, ensuring that economic growth benefits the population and reduces inequalities.

### 1. Import and Load Dataset :-

```
[81]: import pandas as pd

# data Loading
dataset = pd.read_csv('C:/Users/vinay/Downloads/gdp-per-capita-maddison.csv')
print(dataset.head()) # Preview the data

Country Code Year GDP_per_capita
0 Afghanistan AFG 2000 502.37274
1 Afghanistan AFG 2001 489.68200
2 Afghanistan AFG 2002 796.81660
3 Afghanistan AFG 2003 842.80520
4 Afghanistan AFG 2004 869.03925
```

### 2. Preview the Data:-

```
[71]: print(dataset.info()) # Check the structure
        print(dataset.describe()) # Summary statistics
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 1126 entries, 0 to 1125
       Data columns (total 4 columns):
                         Non-Null Count Dtype
        # Column
        0 Country 1126 non-null object
1 Code 1126 non-null object
2 Year 1126 non-null int64
       3 GDP_per_capita 1126 non-null float64
dtypes: float64(1), int64(1), object(2)
       memory usage: 35.3+ KB
       None
                       Year GDP_per_capita
       count 1126.000000 1126.000000
       mean 2011.009769 26399.483603
std 6.631031 23500.919987
              2000.000000 489.682000
2005.000000 9419.550000
        min
        25%
       50% 2011.000000 21841.401500
75% 2017.000000 37291.033250
              2022.000000 160051.230000
```

## 3. Remove missing value :-

```
[71]: print(dataset.info()) # Check the structure
      print(dataset.describe()) # Summary statistics
       <class 'pandas.core.frame.DataFrame'>
      RangeIndex: 1126 entries, 0 to 1125
      Data columns (total 4 columns):
       # Column Non-Null Count Dtype
                           -----
       0 Country 1126 non-null object
1 Code 1126 non-null object
2 Year 1126 non-null int64
       3 GDP_per_capita 1126 non-null float64
      dtypes: float64(1), int64(1), object(2)
      memory usage: 35.3+ KB
      None
                    Year GDP_per_capita
      count 1126.000000 1126.000000
              2011.009769 26399.483603
6.631031 23500.919987
      mean 2011.009769
      std
                           489.682000
9419.550000
      min
             2000.000000
      25%
            2005.000000
            2011.000000 21841.401500
      50%
      75%
             2017.000000
                            37291.033250
             2022.000000 160051.230000
```

### 4. Visualization:

GDP per Capita Trend Visualization

#### GDP per Capita Trend Over Time (Selected Countries)



```
[74]: # Select a specific year (latest year in this case)
latest_year = data['Year'] == latest_year]

# Get the top 10 countries with the highest GDP per capita
top_countries = latest_year_data.nlargest(10, 'GDP_per_capita')

# Create a bar chart
fig = px.bar(
    top_countries,
    x="Country",
    y="GDP_per_capita",
    title=f"Top 10 Countries by GDP per Capita in {latest_year}",
    labels={"GDP_per_capita"; "GDP per Capita (USD)", "Country"; "Country"},
    text="GDP_per_capita",
    template="plotly"
)
fig.update_traces(texttemplate='%{text:.2f}', textposition='outside')
fig.update_layout(xaxis_tickangle=-45)
fig.show()
```

Top 10 Countries by GDP per Capita in 2022



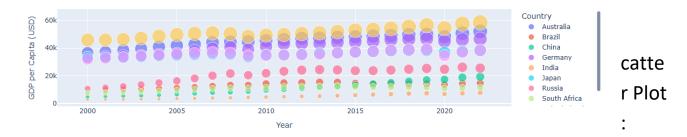
• Bar Chart: Top 10 countries

```
[75]: # List of selected countries
selected_countries = [
    "United States", "China", "India", "Germany", "United Kingdom",
    "Japan", "Brazil", "Russia", "South Africa", "Australia"
]

# Filter data for selected countries
filtered_data = data[data['Country'].isin(selected_countries)]

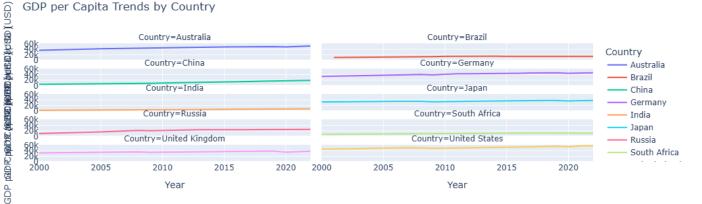
# Create a scatter plot
fig = px.scatter(
    filtered_data,
        x="Year",
        y="GDP_per_capita",
        color="Country",
        size="GDP_per_capita",
        title="GDP per_Capita" Over Time (Selected Countries)",
        labels=("GDP_per_capita": "GDP per Capita (USD)", "Year": "Year"),
        template="plotly"
    )
    fig.show()
```

#### GDP per Capita Over Time (Selected Countries)



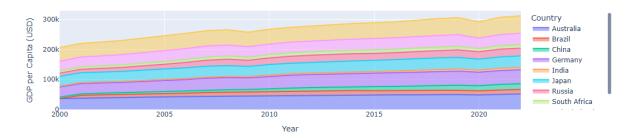
• Area Chart:

#### GDP per Capita Trends by Country



```
[76]: # Create an area chart for selected countries
          fig = px.area(
    filtered_data,
                 x="Year",
                 y="GDP_per_capita",
color="Country",
                title="Cumulative GDP per Capita Trend (Selected Countries)",
labels={"GDP_per_capita": "GDP per Capita (USD)", "Year": "Year"},
template="plotly"
           fig.show()
```

#### Cumulative GDP per Capita Trend (Selected Countries)



## Facet line Chart:

```
# Create a facet line chart for selected countries
 fig = px.line(
filtered_data,
       riftered_data,
x="Year",
y="GDP_per_capita",
color="Country",
facet_col="Country",
        facet_col_wrap=2,
        title="GDP per Capita Trends by Country",
labels={"GDP_per_capita": "GDP per Capita (USD)", "Year": "Year"},
template="plotly"
 fig.show()
```

# 5. Comparisons:-

 Bar Chart : Comparing GDP per capita for countries for a Given year...

```
import plandas as pd
import platly, express as px

# Load the dataset
data = pd.read_csv('C:/Users/vinay/Downloads/gdp-per-capita-maddison.csv')

# Select a specific year (for example, the most recent year)
selected_year = 2020 # You can change this to any year you want

# Filter data for the selected year
year_data = data[data['Year'] == selected_year]

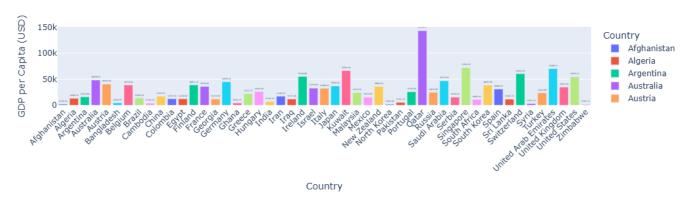
# Create a bar chart
fig = px.bar(
    year_data,
    x="Country",
    y="GOP_per_capita"; "GOP per Capita (USO)", "Country": "Country"),
    color="Country",
    text="GOP_per_capita";

# Customize the chart
fig.update_traces(texttemplate='%(text:.2f)', textposition='outside')
fig.update_layout(xaxis_tickangle=-45)

# Show the plot
fig.show()
```

Box Plot: Comparing GDP per capita Distribution across Countries...

#### GDP per Capita by Country in 2020



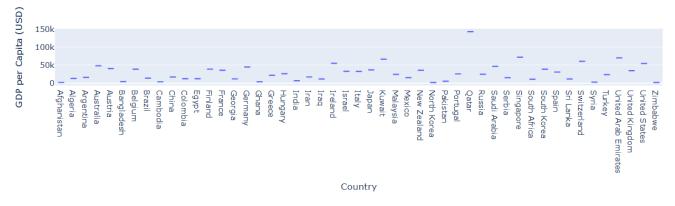
```
[79]: # Select a specific year (for example, 2020)
selected_year = 2020 # You can change this to any year you want

# Filter data for the selected year
year_data = data[data['Year'] == selected_year]

# Create a box plot
fig = px.box(
    year_data,
    x="Country",
    y="GDP_per_capita",
    title=f"GDP per Capita Distribution by Country in {selected_year}",
    labels={"GDP_per_capita": "GDP per Capita (USD)", "Country": "Country"}
)

# Show the plot
fig.show()
```

GDP per Capita Distribution by Country in 2020



## 6. Insights:

The GDP per capita analysis highlights significant disparities in wealth and living standards across countries. High-income nations such as the United States and Switzerland display much higher GDP per capita, reflecting their developed economies, while countries in regions like sub-Saharan Africa and parts of South Asia show much lower figures. These differences often correlate with varying levels of infrastructure, healthcare, and education. This analysis serves as a powerful reminder of the global inequalities that persist and the need for targeted economic policies to address these gaps.

Over time, the analysis also reveals the economic growth trajectories of various nations. Countries like China and India, for example, have shown remarkable progress in GDP per capita due to rapid industrialization, technological advancements, and improved living conditions. This growth offers hope that developing nations can break free from poverty and raise their citizens' standards of living. Conversely, countries facing political instability or economic crises may struggle with stagnant or declining GDP per capita, underscoring the importance of stable governance and sound economic policies in fostering sustainable growth.

Finally, this analysis emphasizes the relationship between economic growth and poverty alleviation. As nations experience growth in GDP per capita, they often see reductions in poverty and improvements in quality of life. However, the analysis

also shows that not all economic growth leads to equitable prosperity. Some countries with high GDP per capita still face significant social challenges, such as income inequality. Therefore, while economic growth is essential, it must be accompanied by social policies that ensure the benefits are shared more broadly to improve the overall well-being of citizens.

# **Summary**

The GDP per capita analysis reveals the stark economic inequalities that exist between countries, with wealthier nations like the U.S. and Switzerland showing significantly higher figures, while developing regions such as sub-Saharan Africa face much lower levels of prosperity. The analysis also highlights the impressive growth in countries like China and India, where industrialization and technological advances have lifted millions out of poverty. However, it also underscores that growth alone isn't enough; despite rising GDP per capita, many countries still struggle with issues like income inequality. This shows that while economic growth is crucial, it must be accompanied by policies that ensure the benefits reach all citizens, improving overall living standards and reducing disparities.

Vinay Upadhyay | LinkedIn