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DEPARTMENT OF ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

AD23632 - Framework for Data Visualization and Analytics

Mini Project : Summer Olympics Medal Analysis (1896–2024)

Report submitted by

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Chapter 1: Abstract

This project presents a comprehensive visual analytics study of the Summer Olympics Medal Dataset (1896–2024). The primary objective is to analyze patterns and trends in medal distributions across different countries, sports, genders, and years using Python (Matplotlib, Seaborn, Plotly), Power BI, and Tableau. The project involves data preprocessing, static and interactive visualizations, and dashboard creation to highlight global dominance in sports, athlete achievements, and the evolution of the Olympic Games over time. By leveraging multiple frameworks, this study offers an integrated comparison of visualization tools for sports analytics.

Chapter 2: Introduction

The **Olympic Games** are the world's foremost sporting event, showcasing global athletic excellence since **1896**. Analyzing over a century of Olympic data provides valuable insights into sports trends, gender participation, and national dominance.

This project applies **data visualization and analytics** techniques to uncover these insights and present them through interactive dashboards.

Tools used:

- **Python (Matplotlib, Plotly):** For statistical and exploratory data visualization.
- **Power BI:** For dynamic, business-style dashboards with filters and drill-downs.
- **Tableau:** For aesthetically rich, interactive data exploration.

The outcome provides a data-driven understanding of Olympic history and comparative performance among nations.

Chapter 3: Dataset Description

Dataset Name: Summer Olympics Medals (1896–2024)

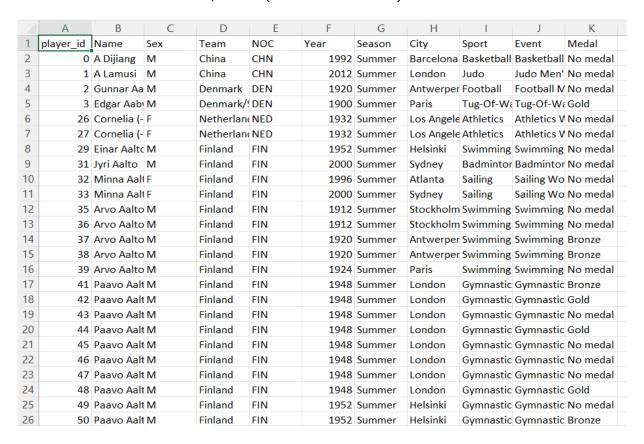
Source: Kaggle / Official Olympic data archive

Data Size: ~271,000 records

Key Columns:

- Year Olympic year
- City Host city
- Sport Category of sport
- Event Specific competition
- Name Athlete name
- Sex Gender (M/F)

- *Team* Country/Team name
- NOC National Olympic Committee code
- Medal Gold/Silver/Bronze
- Season Summer/Winter (filtered for Summer)



Chapter 4: Objectives

To **preprocess and clean** the Olympics dataset for consistency and accuracy.

- 1. To perform **Exploratory Data Analysis (EDA)** using Python libraries.
- 2. To create **static visualizations** using Matplotlib and Seaborn.
- To build interactive visualizations using Plotly for in-depth insights.
- 4. To design **interactive dashboards** in Power BI and Tableau.
- To derive analytical insights about medal trends, athlete dominance, and country performance.
- 6. To compare visualization frameworks based on interactivity, usability, and insight clarity.

Chapter 5: Methodology

1. Data Preprocessing:

- Loaded CSV data into Pandas.
- Filtered Season = "Summer".
- o Cleaned missing values in *Medal* and *NOC*.
- Standardized country codes and gender labels.
- Created derived features like Medal Count and Decade.

2. Data Visualization Frameworks:

- Python: Used Matplotlib, Seaborn, and Plotly for EDA and charts.
- Power BI: Imported dataset, created calculated fields (e.g., Total Medals, Gold Ratio), and developed slicers for Year, Country, and Sport.
- Tableau: Built worksheets for Medal trends, Top Countries, Gender Distribution, and Athlete Rankings.

3. **Comparative Analysis:**

Evaluated dashboards for interactivity, flexibility, and presentation clarity.

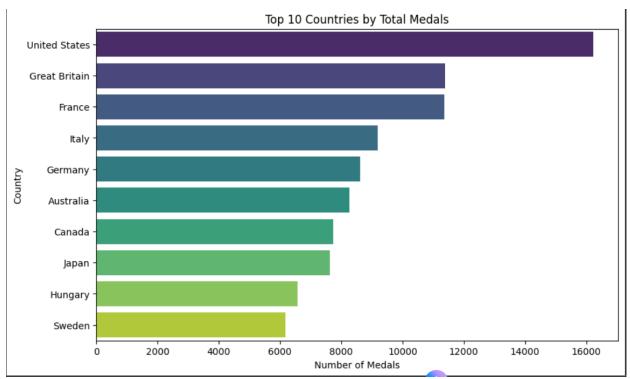
Chapter 6: Python Implementation (Matplotlib & Streamlit)

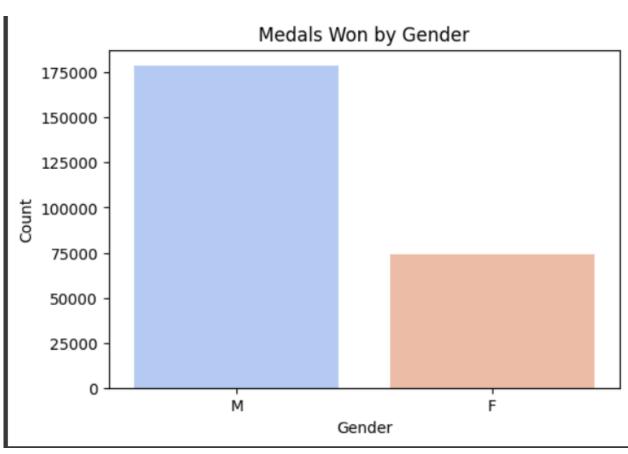
Tools Used:

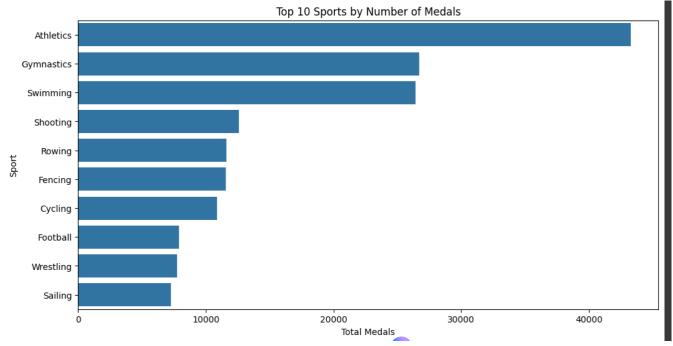
- Pandas for data cleaning
- Matplotlib & Seaborn for static visualization
- Plotly for interactive visualization

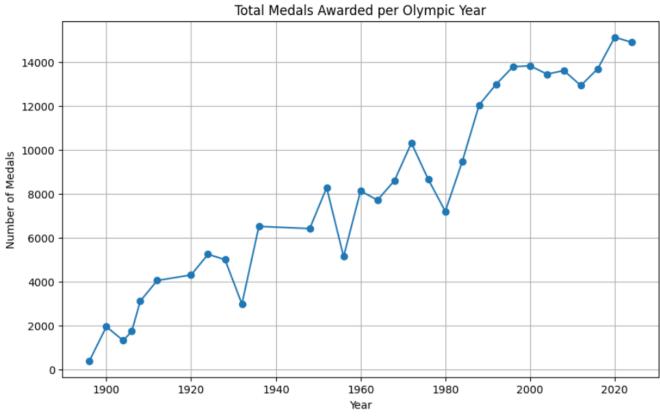
Key Visualizations:

- Medal Distribution by Country (Bar Chart)
- Top 10 Athletes (Horizontal Bar Chart)
- Year-wise Total Medals (Line Chart)
- Gender Participation Growth (Area Chart)
- Sport-wise Medal Trends (Heatmap)

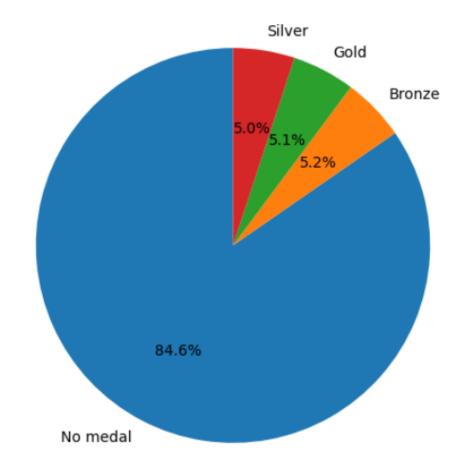


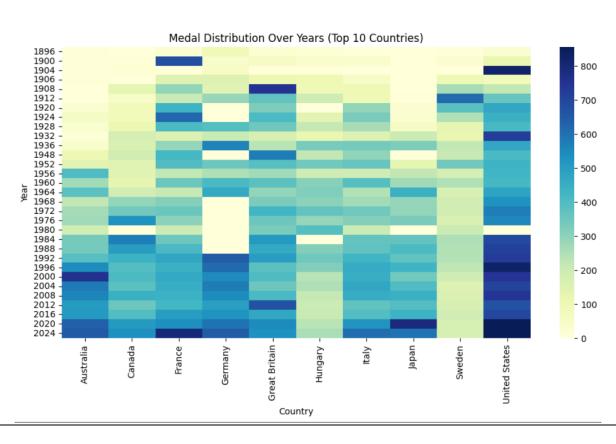






Overall Medal Type Distribution (1994-2024 Summer Olympics)

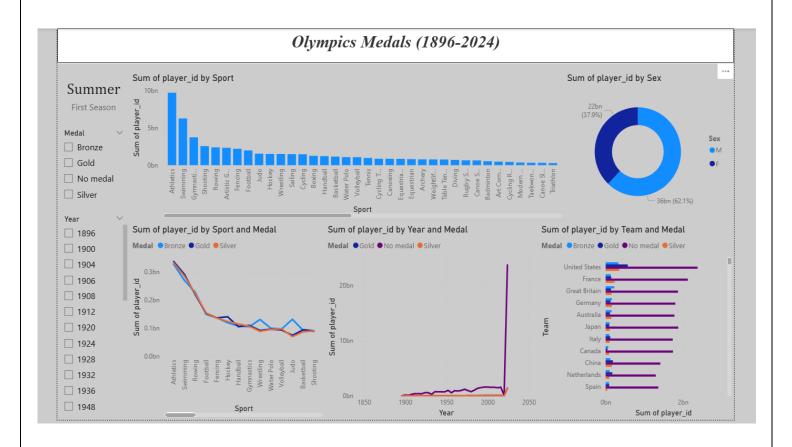




Chapter 7: Power BI Dashboard

Features:

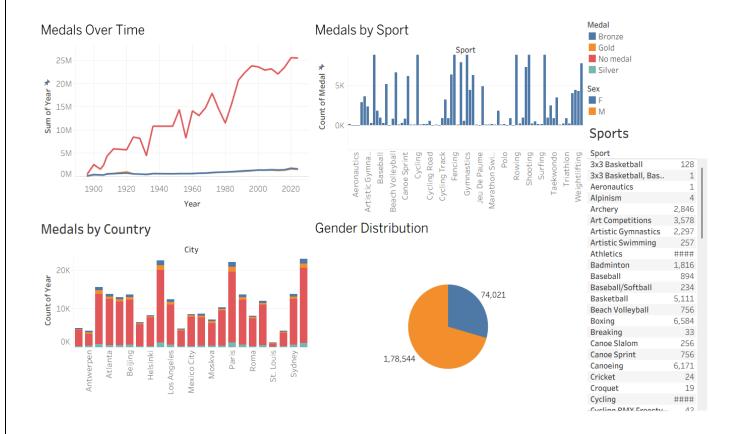
- Filters for *Year*, *Sport*, and *Country*.
- KPI Cards: Total Medals, Gold Medals, Total Athletes.
- Visuals:
 - Medal count by country (Bar Chart)
 - Medal trend by year (Line Chart)
 - Gender participation (Donut Chart)
 - Sport-wise medals (Tree Map)
- Interactive cross-filtering and drill-down to specific games or sports.



Chapter 8: Tableau Dashboard

Features:

- Clean pastel theme with interactive filters.
- Visuals:
 - Country vs Medal Count (Horizontal Bar Chart)
 - Gender Distribution over Time (Line Chart)
 - Top 5 Athletes by Medals (Bar Chart)
 - Map showing Global Medal Spread by Country.
- Hover tooltips and region filters enhance interactivity.



Chapter 9: Analysis & Findings

- 1. USA and Soviet Union are the top-performing nations across all years.
- 2. Athletics and Swimming dominate the overall medal count.
- 3. Female participation has increased steadily since the 1970s.
- 4. Hosting countries often see a spike in total medals won.
- 5. Interactive dashboards make exploring Olympic data more engaging and insightful.

Chapter 10: Conclusion

This project successfully demonstrates how data visualization can transform large historical sports data into actionable insights. Using Python, Power BI, and Tableau, the analysis revealed long-term patterns and allowed intuitive exploration of Olympic history.

Each tool had distinct advantages — Python for customization, Power BI for business-style dashboards, and Tableau for visual storytelling.

Chapter 11: Future Scope

- 1. Integrate machine learning models to predict medal counts for upcoming Olympics.
- 2. Add real-time updates from sports APIs.
- 3. Include Winter Olympics for broader analysis.
- 4. Explore country-level socio-economic factors influencing performance.

Chapter 12: Appendix (Code Snippets)

Key Python code snippets used for data transformation and visualization.

1. Data Loading and Preparation

```
# Import required libraries
import pandas as pd
import numpy as np

# Load the Olympics dataset
df = pd.read_csv("olympics_dataset.csv")

# Display basic info
print(df.head())
print(df.info())

# Filter for Summer Olympics only
df = df[df['Season'] == 'Summer']
```

```
# Drop missing medal values
       df.dropna(subset=['Medal'], inplace=True)
       # Clean and standardize column names
       df.columns = df.columns.str.strip().str.replace(' ', '_')
       # Create additional columns for analysis
       df['Decade'] = (df['Year'] // 10) * 10
       df['Medal\_Count'] = 1
2. Exploratory Data Analysis
       # Check missing values
       print(df.isnull().sum())
       # Summary statistics
       print(df.describe())
       # Top 10 countries by medal count
       top countries = df['NOC'].value counts().head(10)
       print(top countries)
3. Visualization using Matplotlib & Seaborn
       import matplotlib.pyplot as plt
       import seaborn as sns
       plt.style.use('seaborn-v0_8-muted')
       # 1. Top 10 Countries by Medal Count
       plt.figure(figsize=(10,6))
       sns.barplot(x=top_countries.index, y=top_countries.values, palette='coolwarm')
       plt.title("Top 10 Countries by Total Medals (1896–2024)")
       plt.xlabel("Country Code")
       plt.ylabel("Total Medals")
       plt.show()
       # 2. Medal Distribution by Type
       plt.figure(figsize=(8,5))
       sns.countplot(data=df, x='Medal', order=['Gold', 'Silver', 'Bronze'], palette='Set2')
       plt.title("Medal Distribution by Type")
       plt.show()
       # 3. Gender Participation Trend
       gender_trend = df.groupby(['Year','Sex']).size().reset_index(name='Count')
       plt.figure(figsize=(10,5))
       sns.lineplot(data=gender trend, x='Year', y='Count', hue='Sex', marker='o')
       plt.title("Male vs Female Participation Over Time")
       plt.show()
```

4. Interactive Visualization using Plotly

```
import plotly.express as px
# 1. Interactive Medal Trend Over Years
trend = df.groupby('Year')['Medal Count'].sum().reset index()
fig = px.line(trend, x='Year', y='Medal_Count', title='Total Medals Awarded Over the Years')
fig.show()
# 2. Country Medal Count
country_medals = df['NOC'].value_counts().reset_index()
country_medals.columns = ['Country', 'Total Medals']
fig = px.bar(country medals.head(10), x='Country', y='Total Medals',
        title='Top 10 Countries by Total Medals',
        color='Total Medals', color_continuous_scale='Viridis')
fig.show()
# 3. Gender Distribution Pie Chart
gender_count = df['Sex'].value_counts().reset_index()
gender count.columns = ['Gender', 'Count']
fig = px.pie(gender count, names='Gender', values='Count', title='Gender Distribution in Olympics')
fig.show()
# 4. Medal Distribution by Sport
sport medals = df['Sport'].value counts().reset index().head(15)
sport_medals.columns = ['Sport', 'Total Medals']
fig = px.bar(sport medals, x='Sport', v='Total Medals',
        title='Top 15 Sports by Medal Count', color='Total Medals')
fig.update_layout(xaxis_tickangle=-45)
fig.show()
```

5. Saving Cleaned Dataset

Save the cleaned dataset for Power BI and Tableau df.to_csv("olympics_cleaned.csv", index=False) print("Cleaned dataset saved successfully!")







