World Population Analysis Machine Learning Project

Project Overview

The aim of this project is to **analyze global population trends** using historical demographic data and **predict future population growth** using machine learning techniques. The analysis highlights patterns, trends, and predictions based on multiple features such as country, population, area, growth rate, and density.

Dataset Description

Source: worldpopulationreview.com

Features Included:

• Rank: Rank by Population

• CCA3: 3-Digit Country Code

Country/Territory: Name of the Country

• Capital: Capital City

• Continent: Continent Name

• Population (1970–2022): Population over multiple decades

• Area (km²): Total Area

• Density (per km²): Population Density

Growth Rate: Recent population growth rate

World Population Percentage: Share of world population

✓ Step 1: Data Preprocessing

- · Checked and handled missing values
- Renamed columns for consistency
- Converted data types as needed
- Verified uniqueness and formatting of country codes and names

🚺 Step 2: Exploratory Data Analysis (EDA)

Top 10 Most Populous Countries (2022)

- India
- China
- United States
- Indonesia
- Pakistan
- Nigeria
- Brazil
- Bangladesh
- Russia
- Mexico

Countries with Highest Growth Rates

- Niger
- Angola
- DR Congo
- Chad
- Mali

Countries with Declining Populations

- Japan
- Russia
- Bulgaria
- Latvia
- Lithuania

Step 3: Visualizations

✓ Line Plot – Population Growth Over Time

Plotted population trends from 1970 to 2022 for selected countries like:

- India
- China
- United States
- Nigeria

Choropleth Map – World Population (2022)

A world map displaying population distribution based on color intensity using plotly.

Bar Charts

- Top 10 countries by density
- Countries with lowest population
- Growth rate comparisons

Step 4: Feature Engineering

Created new features such as:

- Population Change (%) over the last decade
- Population Growth (absolute) between 2010 and 2022
- Grouped countries by **continent** for comparative analysis

Step 5: Machine Learning Modeling

Objective:

To **predict future population** based on historical population data and country attributes.

Models Used:

- Linear Regression
- Random Forest Regressor
- Decision Tree Regressor

Features Considered:

- Historical populations
- Area, density
- Growth rate

Continent (encoded)

Model Evaluation

Evaluated models using:

- Mean Absolute Error (MAE)
- Mean Squared Error (MSE)
- R² Score

Best performance achieved using **Random Forest Regressor** due to its ability to handle non-linear relationships.

Summary & Insights

- The global population exceeded **8 billion** by 2023, growing at a **slower pace**.
- India is set to overtake China as the most populous country by 2030.
- African nations will contribute significantly to future population growth.
- **Developed countries** like Japan and Russia are witnessing a decline due to aging populations and low birth rates.

Files Provided

requirements.txt

pandas

numpy

matplotlib

seaborn

plotly

scikit-learn

README.md

World Population Analysis - Machine Learning Project

Objective

Analyze historical population data and build ML models to forecast population trends globally.

Dataset

- Source: worldpopulationreview.com
- Contains historical population, country metadata, and growth statistics

Tools & Libraries

- Python
- Pandas, NumPy
- Matplotlib, Seaborn, Plotly
- Scikit-learn

Features

- EDA and Visualizations
- Line plots, bar charts, choropleth map
- Growth rate analysis
- Machine learning prediction models

How to Run

- 1. Install dependencies from `requirements.txt`
- 2. Load dataset and run `population_analysis.ipynb`

Conclusion

This project provided a comprehensive analysis of **world population trends** across multiple decades using a combination of **data visualization** and **machine learning** techniques. Key

insights such as the rise of population in developing countries, decline in population growth in developed nations, and the global shift of population density were identified.

The machine learning models, especially the **Random Forest Regressor**, performed well in predicting future population values. This demonstrates that historical population data and related features like area, density, and growth rate can effectively be used to forecast demographic changes.

This project not only provided actionable insights into the **past and present population dynamics** but also laid a foundation for building predictive systems that can help policymakers, researchers, and organizations make informed decisions.

Future Scope

1. Incorporate Future Year Forecasting:

Extend the prediction to include 2030, 2040, and 2050 using time-series forecasting models like ARIMA, Prophet, or LSTM.

2. Advanced ML Models:

Explore advanced models like **Gradient Boosting**, **XGBoost**, and **Neural Networks** for more accurate long-term forecasts.

3. Integration of Socioeconomic Factors:

Include features such as:

- 1. GDP per capita
- 2. Life expectancy
- 3. Fertility rate
- 4. Urbanization rate to improve model performance and add depth to the analysis.

4. Interactive Dashboards:

Deploy the analysis as an interactive **dashboard using Dash or Streamlit**, allowing users to explore population metrics by continent, country, or year.

5. Policy Simulation:

Simulate population control or support policies to visualize their potential impact over time.

6. Climate and Migration Analysis:

Combine demographic data with **climate change**, **disaster impact**, or **migration trends** for deeper insights into regional population shifts.