

# Economic Growth Analysis and Prediction for G7 and BRICS Nations

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## 1. Introduction

In an interconnected global economy, understanding macroeconomic trends is pivotal for policymakers and businesses. This project analyzes historical economic data (1972–2022) and forecasts GDP growth for **G7** (Canada, France, Germany, Italy, Japan, UK, USA) and **BRICS** (Brazil, Russia, India, China, South Africa) nations through **2023–2030**. By integrating exploratory analysis, machine learning, and interactive visualization, the study deciphers growth drivers and offers actionable insights for strategic decision-making.

## 2. Dataset Overview

- **Source:** [World Bank](#). [For Convenience to download the dataset [click here](#)]
- **Scope:** 12 countries, 15 parameters, including GDP growth, FDI, inflation, population trends, and unemployment.
- **Preprocessing:**
  - **Missing Values:** Converted non-numeric entries (e.g., "-") to NaN, imputed row-wise means, and assigned 0.00 where data was absent.
  - **Outliers:** Removed using IQR (e.g., Japan's 33% gross savings in 1996–1998; Russia's 99% current account balance in 2004).
  - **Reshaping:** Transformed from wide to long format for time-series analysis.

## 3. Methodology

### 3.1 Exploratory Data Analysis (EDA)

- **Correlation Analysis:**
  - **G7:** GDP growth correlated strongly with GDP per capita growth ( $r = 1$ ) and gross savings ( $r = 0.52$ ).
  - **BRICS:** GDP growth linked to FDI inflows ( $r = 0.42$ ) and imports of goods/services ( $r = 0.24$ ).
- **Group Trends:** BRICS showed higher volatility but stronger growth potential compared to G7.

### 3.2 Machine Learning Pipeline

- **Feature Engineering:**
  - Scaled features (e.g., Year) and target variable (GDP growth) using StandardScaler.
  - Split data into **80% training** and **20% testing** sets.
- **Model Selection:**

- **G7: Support Vector Regression (SVR)** outperformed others ( $MSE = 0.65$ ) with hyperparameters  $C=100$ ,  $\epsilon=0.1$ .
- **BRICS: Linear Regression** achieved the lowest error ( $MSE = 0.72$ ), reflecting simpler economic relationships.
- **Validation:** 5-fold cross-validation ensured robustness.

### 3.3 Time-Series Forecasting

- **Predictions (2023–2030):**

**Year   G7 (%)   BRICS (%)**

2023 4.79   6.57

2024 4.77   6.60

2025 4.74   6.63

2026 4.71   6.66

2027 4.69   6.70

2028 4.67   6.73

2029 4.65   6.76

2030 4.65   6.79

- **Trend Drivers:**
  - **G7 Decline:** Aging populations, market saturation.
  - **BRICS Growth:** FDI inflows, trade openness.

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## 4. Interactive Dashboard

Built using **Streamlit** and **Plotly**, the dashboard enables:

1. **Predicted Trends:** Interactive line charts comparing G7 and BRICS forecasts.
2. **Comparative Analysis:** Parameter-specific trends (e.g., inflation) across customizable year ranges.
3. **Country Benchmarking:** Side-by-side comparison of metrics like FDI or unemployment.  
*Example Insight:* BRICS' GDP growth is projected to surpass G7 by 2030.

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## 5. Key Insights

1. **G7 Challenges:**
  - Aging demographics and high GDP per capita correlate with slower growth.
  - **Recommendation:** Boost innovation and savings rates to counter stagnation.

## 2. BRICS Opportunities:

- FDI and trade openness are critical growth drivers.
- **Recommendation:** Stabilize fiscal policies to sustain momentum.

## 3. Global Shift: BRICS' rising influence signals a rebalancing of economic power.

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## 6. Limitations & Future Work

- **Limitations:**

- Assumes no major geopolitical disruptions (e.g., wars, pandemics).
- Relies on historical trends, omitting external factors like climate change.

- **Future Directions:**

- Incorporate climate data, geopolitical indices, or sector-specific metrics.
- Expand country coverage to include emerging economies like Indonesia or Mexico.

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## 7. Conclusion

This project underscores the transformative potential of data science in economics. By systematically preprocessing data, training tailored models, and visualizing results, it provides a replicable framework for forecasting and policy analysis. The interactive dashboard democratizes access to insights, empowering stakeholders to navigate global economic shifts proactively.

**Tools Used:** Python (Pandas, Scikit-learn, Matplotlib), Streamlit.

**Dashboard:** [[Streamlit Link](#)]

**Code:** [[GitHub Repository Link](#)]