

# G20 Countries Economic Indicators

## Dataset description

Country Name: The name of the country being analyzed.

Country Code: The unique code for each country (ISO Alpha-3 code).

Time: The year of the data point.

Time Code: The code for the time period (usually corresponds to the year).

Current account balance (% of GDP) [BN.CAB.XOKA.GD.ZS]: The balance of a country's current account, including trade balance, primary income, and secondary income, as a percentage of GDP.

Exports of goods and services (% of GDP) [NE.EXP.GNFS.ZS]: The value of all goods and services sold abroad, expressed as a percentage of GDP.

Foreign direct investment, net inflows (% of GDP) [BX.KLT.DINV.WD.GD.ZS]: The inflow of foreign investment into a country, as a percentage of GDP.

Foreign direct investment, net outflows (% of GDP) [BM.KLT.DINV.WD.GD.ZS]: The outflow of investment from a country to foreign countries, as a percentage of GDP.

GDP (current US\$) [NY.GDP.MKTP.CD]: The total monetary value of goods and services produced in a country, in current US dollars.

GDP growth (annual %) [NY.GDP.MKTP.KD.ZG]: The annual growth rate of GDP.

GDP per capita (current US\$) [NY.GDP.PCAP.CD]: GDP divided by the midyear population, indicating the economic output per person.

GDP per capita growth (annual %) [NY.GDP.PCAP.KD.ZG]: The annual growth rate of GDP per capita.

General government final consumption expenditure (% of GDP) [NE.CON.GOV.T.ZS]: Total government spending as a percentage of GDP.

Government expenditure on education, total (% of GDP) [SE.XPD.TOTL.GD.ZS]: Total government spending on education as a percentage of GDP.

Government expenditure per student, primary (% of GDP per capita) [SE.XPD.PRIM.PC.ZS]: Government spending per primary school student as a percentage of GDP per capita.

Government expenditure per student, secondary (% of GDP per capita) [SE.XPD.SECO.PC.ZS]: Government spending per secondary school student as a percentage of GDP per capita.

Government expenditure per student, tertiary (% of GDP per capita) [SE.XPD.TERT.PC.ZS]: Government spending per tertiary-level student as a percentage of GDP per capita.

Gross savings (% of GDP) [NY.GNS.ICTR.ZS]: Total savings as a percentage of GDP, reflecting income saved rather than spent.

Imports of goods and services (% of GDP) [NE.IMP.GNFS.ZS]: The value of all goods and services purchased from abroad, expressed as a percentage of GDP.

GNI (current US\$) [NY.GNP.MKTP.CD]: Gross national income in current US dollars.

GNI growth (annual %) [NY.GNP.MKTP.KD.ZG]: The annual growth rate of GNI.

Employment and Labor

Employment to population ratio, 15+, total (%) (national estimate) [SL.EMP.TOTL.SP.NE.ZS]: The ratio of employed individuals aged 15 and older to the total working-age population.

Employment to population ratio, 15+, female (%) (national estimate) [SL.EMP.TOTL.SP.FE.NE.ZS]: The ratio of employed females aged 15 and older to the total working-age female population.

Employment to population ratio, 15+, male (%) (national estimate) [SL.EMP.TOTL.SP.MA.NE.ZS]: The ratio of employed males aged 15 and older to the total working-age male population.

Unemployment, total (% of total labor force) (national estimate) [SL.UEM.TOTL.NE.ZS]: The percentage of the labor force that is unemployed.

Employment in services (% of total employment) (modeled ILO estimate) [SL.SRV.EMPL.ZS]: The percentage of employed persons working in the services sector.

Employment in industry (% of total employment) (modeled ILO estimate) [SL.IND.EMPL.ZS]: The percentage of employed persons working in the industrial sector.

Employment in agriculture (% of total employment) (modeled ILO estimate) [SL.AGR.EMPL.ZS]: The percentage of employed persons working in agriculture.

Poverty headcount ratio at national poverty lines (% of population) [SI.POVT.NAHC]: The percentage of the population living below the national poverty line.

Multidimensional poverty headcount ratio (UNDP) (% of population) [SI.POVT.MPUN]: The percentage of the population experiencing multiple deprivations (education, health, and living standards).

Fixed broadband subscriptions [IT.NET.BBND]: The number of fixed broadband subscriptions.

Fixed telephone subscriptions [IT.MLT.MAIN]: The number of fixed telephone subscriptions.

Mobile cellular subscriptions [IT.CEL.SETS]: The number of mobile cellular subscriptions.

Secure Internet servers [IT.NET.SECR]: The number of secure servers that encrypt web communications.

High-technology exports (current US\$) [TX.VAL.TECH.CD]: The value of high-tech exports, in current US dollars.

High-technology exports (% of manufactured exports) [TX.VAL.TECH.MF.ZS]: The percentage of manufactured exports that are high-technology products.

Research and development expenditure (% of GDP) [GB.XPD.RSDV.GD.ZS]: R&D expenditure as a percentage of GDP.

Scientific and technical journal articles [IP.JRN.ARTC.SC]: The number of scientific and technical articles published.

Air transport, freight (million ton-km) [IS.AIR.GOOD.MT.K1]: Freight transported by air, in million ton-km.

Air transport, passengers carried [IS.AIR.PSGR]: The total number of passengers carried by air.

Rail lines (total route-km) [IS.RRS.TOTL.KM]: The total length of rail routes, in kilometers.

Railways, goods transported (million ton-km) [IS.RRS.GOOD.MT.K6]: Goods transported by rail, in million ton-km.

Railways, passengers carried (million passenger-km) [IS.RRS.PASG.KM]: Passenger travel by rail, in million passenger-km.

Community health workers (per 1,000 people) [SH.MED.CMHW.P3]: The number of community health workers per 1,000 people.

Current health expenditure (% of GDP) [SH.XPD.CHEX.GD.ZS]: Health expenditure as a percentage of GDP.

Current health expenditure per capita (current US\$) [SH.XPD.CHEX.PC.CD]: Health expenditure per person, in current US dollars.

Hospital beds (per 1,000 people) [SH.MED.BEDS.ZS]: The number of hospital beds per 1,000 people.

Physicians (per 1,000 people) [SH.MED.PHYS.ZS]: The number of physicians per 1,000 people.

Life expectancy at birth, total (years) [SP.DYN.LE00.IN]: The average number of years a newborn is expected to live.

Probability of dying among adolescents ages 10-14 years (per 1,000) [SH.DYN.1014]: The probability of death per 1,000 children aged 10-14.

Probability of dying among adolescents ages 15-19 years (per 1,000) [SH.DYN.1519]: The probability of death per 1,000 teenagers aged 15-19.

Probability of dying among children ages 5-9 years (per 1,000) [SH.DYN.0509]: The probability of death per 1,000 children aged 5-9.

Probability of dying among youth ages 20-24 years (per 1,000) [SH.DYN.2024]: The probability of death per 1,000 youth aged 20-24.

Survival to age 65, female (% of cohort) [SP.DYN.TO65.FE.ZS]: The percentage of females expected to survive to age 65.

Survival to age 65, male (% of cohort) [SP.DYN.TO65.MA.ZS]: The percentage of males expected to survive to age 65.

This is a Jupyter notebook with the extension .ipynb, the python version being used in this project is 3.9.12.

In order to run the cells just click [Shift + Enter] or [Ctrl + Enter]

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
```

```
In [2]: sns.set_theme(style="whitegrid")
```

```
In [3]: df = pd.read_csv("WorldEconomicIndicators.csv")
df.head()
```

Out[3]:

	Country Name	Country Code	Time	Time Code	Current account balance (% of GDP)	Exports of goods and services (% of GDP)	Foreign direct investment, net inflows (% of GDP)	Foreign direct investment, net outflows (% of GDP)	GDP (current US\$)	GDP growth (annual %)	...	Current health expenditure per capita (current US\$)	Hospital beds (per 1,000 people)	Physicians (per 1,000 people)	ex
0	Pakistan	PAK	2000	YR2000	-0.085440186	9.629626104	0.309595026	0.011056965	99484802345	4.260088011	...	15.1975832	0.5	0.595	
1	Pakistan	PAK	2001	YR2001	1.933180342	10.62617308	0.389106586	0.026763945	97145618480	3.651350171	...	13.20521545	0.5	0.605	
2	Pakistan	PAK	2002	YR2002	3.935733262	11.26296557	0.8435173	0.031657429	97923302809	2.594816684	...	14.74764633	0.49	..	
3	Pakistan	PAK	2003	YR2003	3.179620139	12.16464391	0.475207712	0.01690814	1.12E+11	5.401310873	...	14.1153574	0.48	..	
4	Pakistan	PAK	2004	YR2004	-0.617928013	11.42297416	0.845585702	0.042354919	1.32E+11	7.83125557	...	15.29145527	0.48	0.682	

5 rows × 55 columns

In [4]: df.shape

Out[4]: (1564, 55)

We do not need the Time Code column, so we will remove it

```
In [5]: try:
    df.drop("Time Code", axis=1, inplace=True)
    print("Column dropped")
except Exception as e:
    print(e)
```

Column dropped

In [6]: df.info()

#	Column	Non-Null Count	Dtype
0	Country Name	1564 non-null	object
1	Country Code	1564 non-null	object
2	Time	1564 non-null	int64
3	Current account balance (% of GDP)	1564 non-null	object
4	Exports of goods and services (% of GDP)	1564 non-null	object
5	Foreign direct investment, net inflows (% of GDP)	1564 non-null	object
6	Foreign direct investment, net outflows (% of GDP)	1564 non-null	object
7	GDP (current US\$)	1564 non-null	object
8	GDP growth (annual %)	1564 non-null	object
9	GDP per capita (current US\$)	1564 non-null	object
10	GDP per capita growth (annual %)	1564 non-null	object
11	General government final consumption expenditure (% of GDP)	1564 non-null	object
12	Government expenditure on education, total (% of GDP)	1564 non-null	object
13	Government expenditure per student, primary (% of GDP per capita)	1564 non-null	object
14	Government expenditure per student, secondary (% of GDP per capita)	1564 non-null	object
15	Government expenditure per student, tertiary (% of GDP per capita)	1564 non-null	object
16	Gross savings (% of GDP)	1564 non-null	object
17	Imports of goods and services (% of GDP)	1564 non-null	object
18	GNI (current US\$)	1564 non-null	object
19	GNI growth (annual %)	1564 non-null	object
20	Employment to population ratio, 15+, total (%) (national estimate)	1564 non-null	object
21	Employment to population ratio, 15+, female (%) (national estimate)	1564 non-null	object
22	Employment to population ratio, 15+, male (%) (national estimate)	1564 non-null	object
23	Unemployment, total (% of total labor force) (national estimate)	1564 non-null	object
24	Employment in services (% of total employment) (modeled ILO estimate)	1564 non-null	object
25	Employment in industry (% of total employment) (modeled ILO estimate)	1564 non-null	object
26	Employment in agriculture (% of total employment) (modeled ILO estimate)	1564 non-null	object
27	Poverty headcount ratio at national poverty lines (% of population)	1564 non-null	object
28	Multidimensional poverty headcount ratio (UNDP) (% of population)	1564 non-null	object
29	Fixed broadband subscriptions	1564 non-null	object
30	Fixed telephone subscriptions	1564 non-null	object
31	Mobile cellular subscriptions	1564 non-null	object
32	Secure Internet servers	1564 non-null	object
33	High-technology exports (current US\$)	1564 non-null	object
34	High-technology exports (% of manufactured exports)	1564 non-null	object
35	Research and development expenditure (% of GDP)	1564 non-null	object
36	Scientific and technical journal articles	1564 non-null	object
37	Air transport, freight (million ton-km)	1564 non-null	object
38	Air transport, passengers carried	1564 non-null	object
39	Rail lines (total route-km)	1564 non-null	object
40	Railways, goods transported (million ton-km)	1564 non-null	object
41	Railways, passengers carried (million passenger-km)	1564 non-null	object
42	Community health workers (per 1,000 people)	1564 non-null	object
43	Current health expenditure (% of GDP)	1564 non-null	object
44	Current health expenditure per capita (current US\$)	1564 non-null	object
45	Hospital beds (per 1,000 people)	1564 non-null	object
46	Physicians (per 1,000 people)	1564 non-null	object
47	Life expectancy at birth, total (years)	1564 non-null	float64
48	Probability of dying among adolescents ages 10-14 years (per 1,000)	1564 non-null	object
49	Probability of dying among adolescents ages 15-19 years (per 1,000)	1564 non-null	object
50	Probability of dying among children ages 5-9 years (per 1,000)	1564 non-null	object
51	Probability of dying among youth ages 20-24 years (per 1,000)	1564 non-null	object
52	Survival to age 65, female (% of cohort)	1564 non-null	float64
53	Survival to age 65, male (% of cohort)	1564 non-null	float64

dtypes: float64(3), int64(1), object(50)

memory usage: 659.9+ KB

Many of the columns which should have been numerical are float, so we will convert them to float

The data types in most of the columns could not be converted because of the presence of .., so we will replace them with NaN and then handle the missing values.

```
In [8]: try:  
        df.replace(.., np.nan, inplace=True)  
        print("Replacement successful")  
    except Exception as e:  
        print(e)
```

Replacement successful

```
In [9]: for i in df.columns[3:]: #the first three columns have the correct Data type
    try:
        df[i] = df[i].astype("float64")
        print(f"Data type for {i} changed successfully")
    except Exception as e:
        print(e)

Data type for Current account balance (% of GDP) changed successfully
Data type for Exports of goods and services (% of GDP) changed successfully
Data type for Foreign direct investment, net inflows (% of GDP) changed successfully
Data type for Foreign direct investment, net outflows (% of GDP) changed successfully
Data type for GDP (current US$) changed successfully
Data type for GDP growth (annual %) changed successfully
Data type for GDP per capita (current US$) changed successfully
Data type for GDP per capita growth (annual %) changed successfully
Data type for General government final consumption expenditure (% of GDP) changed successfully
Data type for Government expenditure on education, total (% of GDP) changed successfully
Data type for Government expenditure per student, primary (% of GDP per capita) changed successfully
Data type for Government expenditure per student, secondary (% of GDP per capita) changed successfully
Data type for Government expenditure per student, tertiary (% of GDP per capita) changed successfully
Data type for Gross savings (% of GDP) changed successfully
Data type for Imports of goods and services (% of GDP) changed successfully
Data type for GNI (current US$) changed successfully
Data type for GNI growth (annual %) changed successfully
Data type for Employment to population ratio, 15+, total (%) (national estimate) changed successfully
Data type for Employment to population ratio, 15+, female (%) (national estimate) changed successfully
Data type for Employment to population ratio, 15+, male (%) (national estimate) changed successfully
Data type for Unemployment, total (% of total labor force) (national estimate) changed successfully
Data type for Employment in services (% of total employment) (modeled ILO estimate) changed successfully
Data type for Employment in industry (% of total employment) (modeled ILO estimate) changed successfully
Data type for Employment in agriculture (% of total employment) (modeled ILO estimate) changed successfully
Data type for Poverty headcount ratio at national poverty lines (% of population) changed successfully
Data type for Multidimensional poverty headcount ratio (UNDP) (% of population) changed successfully
Data type for Fixed broadband subscriptions changed successfully
Data type for Fixed telephone subscriptions changed successfully
Data type for Mobile cellular subscriptions changed successfully
Data type for Secure Internet servers changed successfully
Data type for High-technology exports (current US$) changed successfully
Data type for High-technology exports (% of manufactured exports) changed successfully
Data type for Research and development expenditure (% of GDP) changed successfully
Data type for Scientific and technical journal articles changed successfully
Data type for Air transport, freight (million ton-km) changed successfully
Data type for Air transport, passengers carried changed successfully
Data type for Rail lines (total route-km) changed successfully
Data type for Railways, goods transported (million ton-km) changed successfully
Data type for Railways, passengers carried (million passenger-km) changed successfully
Data type for Community health workers (per 1,000 people) changed successfully
Data type for Current health expenditure (% of GDP) changed successfully
Data type for Current health expenditure per capita (current US$) changed successfully
Data type for Hospital beds (per 1,000 people) changed successfully
Data type for Physicians (per 1,000 people) changed successfully
Data type for Life expectancy at birth, total (years) changed successfully
Data type for Probability of dying among adolescents ages 10-14 years (per 1,000) changed successfully
Data type for Probability of dying among adolescents ages 15-19 years (per 1,000) changed successfully
Data type for Probability of dying among children ages 5-9 years (per 1,000) changed successfully
Data type for Probability of dying among youth ages 20-24 years (per 1,000) changed successfully
Data type for Survival to age 65, female (% of cohort) changed successfully
Data type for Survival to age 65, male (% of cohort) changed successfully
```

```
In [10]: (df.isnull().sum()/len(df))*100
```

Out[10]:	Country Name	0.000000
	Country Code	0.000000
	Time	0.000000
	Current account balance (% of GDP)	10.230179
	Exports of goods and services (% of GDP)	2.493606
	Foreign direct investment, net inflows (% of GDP)	2.941176
	Foreign direct investment, net outflows (% of GDP)	7.544757
	GDP (current US\$)	0.191816
	GDP growth (annual %)	0.063939
	GDP per capita (current US\$)	0.191816
	GDP per capita growth (annual %)	0.063939
	General government final consumption expenditure (% of GDP)	3.452685
	Government expenditure on education, total (% of GDP)	18.286445
	Government expenditure per student, primary (% of GDP per capita)	57.736573
	Government expenditure per student, secondary (% of GDP per capita)	60.549872
	Government expenditure per student, tertiary (% of GDP per capita)	59.974425
	Gross savings (% of GDP)	10.805627
	Imports of goods and services (% of GDP)	2.493606
	GNI (current US\$)	1.662404
	GNI growth (annual %)	21.483376
	Employment to population ratio, 15+, total (%) (national estimate)	21.803069
	Employment to population ratio, 15+, female (%) (national estimate)	23.657289
	Employment to population ratio, 15+, male (%) (national estimate)	23.849105
	Unemployment, total (% of total labor force) (national estimate)	17.583120
	Employment in services (% of total employment) (modeled ILO estimate)	1.470588
	Employment in industry (% of total employment) (modeled ILO estimate)	1.470588
	Employment in agriculture (% of total employment) (modeled ILO estimate)	1.470588
	Poverty headcount ratio at national poverty lines (% of population)	74.488491
	Multidimensional poverty headcount ratio (UNDP) (% of population)	98.145780
	Fixed broadband subscriptions	7.672634
	Fixed telephone subscriptions	0.639386
	Mobile cellular subscriptions	0.191816
	Secure Internet servers	44.437340
	High-technology exports (current US\$)	37.276215
	High-technology exports (% of manufactured exports)	37.276215
	Research and development expenditure (% of GDP)	34.271100
	Scientific and technical journal articles	10.038363
	Air transport, freight (million ton-km)	11.381074
	Air transport, passengers carried	10.613811
	Rail lines (total route-km)	48.976982
	Railways, goods transported (million ton-km)	43.414322
	Railways, passengers carried (million passenger-km)	42.327366
	Community health workers (per 1,000 people)	96.035806
	Current health expenditure (% of GDP)	6.905371
	Current health expenditure per capita (current US\$)	6.969309
	Hospital beds (per 1,000 people)	25.191816
	Physicians (per 1,000 people)	34.079284
	Life expectancy at birth, total (years)	0.000000
	Probability of dying among adolescents ages 10-14 years (per 1,000)	2.941176
	Probability of dying among adolescents ages 15-19 years (per 1,000)	2.941176
	Probability of dying among children ages 5-9 years (per 1,000)	2.941176
	Probability of dying among youth ages 20-24 years (per 1,000)	2.941176
	Survival to age 65, female (% of cohort)	0.000000
	Survival to age 65, male (% of cohort)	0.000000
	dtype: float64	

There are a lot of null values in the dataset, so in order to deal with them I have a set of guidelines:-

- i) if the null values are above 50%, we will drop the column
- ii) if the null values are between 10% and 50%, we will impute the values
- iii) if the null values are less than 10%, we will drop just the null rows

```
In [11]: null_values = list((df.isnull().sum()/len(df))*100)
cols_to_delete = []

for i in range(len(null_values)):
    col = df.columns[i]
    if null_values[i] >= 50:
        cols_to_delete.append(col)
        print(f"{i}: {col} marked for deletion")
    elif 10 <= null_values[i] < 50:
        df[col] = df[col].fillna(df[col].median())
        print(f"Values in column '{col}' imputed with median")
```

Values in column 'Current account balance (% of GDP)' imputed with median  
 Values in column 'Government expenditure on education, total (% of GDP)' imputed with median  
 13: Government expenditure per student, primary (% of GDP per capita) marked for deletion  
 14: Government expenditure per student, secondary (% of GDP per capita) marked for deletion  
 15: Government expenditure per student, tertiary (% of GDP per capita) marked for deletion  
 Values in column 'Gross savings (% of GDP)' imputed with median  
 Values in column 'GNI growth (annual %)' imputed with median  
 Values in column 'Employment to population ratio, 15+, total (%) (national estimate)' imputed with median  
 Values in column 'Employment to population ratio, 15+, female (%) (national estimate)' imputed with median  
 Values in column 'Employment to population ratio, 15+, male (%) (national estimate)' imputed with median  
 Values in column 'Unemployment, total (% of total labor force) (national estimate)' imputed with median  
 27: Poverty headcount ratio at national poverty lines (% of population) marked for deletion  
 28: Multidimensional poverty headcount ratio (UNDP) (% of population) marked for deletion  
 Values in column 'Secure Internet servers' imputed with median  
 Values in column 'High-technology exports (current US\$)' imputed with median  
 Values in column 'High-technology exports (% of manufactured exports)' imputed with median  
 Values in column 'Research and development expenditure (% of GDP)' imputed with median  
 Values in column 'Scientific and technical journal articles' imputed with median  
 Values in column 'Air transport, freight (million ton-km)' imputed with median  
 Values in column 'Air transport, passengers carried' imputed with median  
 Values in column 'Rail lines (total route-km)' imputed with median  
 Values in column 'Railways, goods transported (million ton-km)' imputed with median  
 Values in column 'Railways, passengers carried (million passenger-km)' imputed with median  
 42: Community health workers (per 1,000 people) marked for deletion  
 Values in column 'Hospital beds (per 1,000 people)' imputed with median  
 Values in column 'Physicians (per 1,000 people)' imputed with median

```
In [12]: for i in cols_to_delete:
    df.drop(i, axis=1, inplace=True)
    print(f"Column {i} deleted successfully")
```

Column Government expenditure per student, primary (% of GDP per capita) deleted successfully  
 Column Government expenditure per student, secondary (% of GDP per capita) deleted successfully  
 Column Government expenditure per student, tertiary (% of GDP per capita) deleted successfully  
 Column Poverty headcount ratio at national poverty lines (% of population) deleted successfully  
 Column Multidimensional poverty headcount ratio (UNDP) (% of population) deleted successfully  
 Column Community health workers (per 1,000 people) deleted successfully

Now we will remove all the null values

```
In [13]: df.dropna(inplace=True)
```

In [14]: `(df.isnull().sum()/len(df))*100`

```
Out[14]: Country Name          0.0
Country Code          0.0
Time                  0.0
Current account balance (% of GDP) 0.0
Exports of goods and services (% of GDP) 0.0
Foreign direct investment, net inflows (% of GDP) 0.0
Foreign direct investment, net outflows (% of GDP) 0.0
GDP (current US$)      0.0
GDP growth (annual %)   0.0
GDP per capita (current US$) 0.0
GDP per capita growth (annual %) 0.0
General government final consumption expenditure (% of GDP) 0.0
Government expenditure on education, total (% of GDP) 0.0
Gross savings (% of GDP)      0.0
Imports of goods and services (% of GDP) 0.0
GNI (current US$)          0.0
GNI growth (annual %)       0.0
Employment to population ratio, 15+, total (%) (national estimate) 0.0
Employment to population ratio, 15+, female (%) (national estimate) 0.0
Employment to population ratio, 15+, male (%) (national estimate) 0.0
Unemployment, total (% of total labor force) (national estimate) 0.0
Employment in services (% of total employment) (modeled ILO estimate) 0.0
Employment in industry (% of total employment) (modeled ILO estimate) 0.0
Employment in agriculture (% of total employment) (modeled ILO estimate) 0.0
Fixed broadband subscriptions 0.0
Fixed telephone subscriptions 0.0
Mobile cellular subscriptions 0.0
Secure Internet servers      0.0
High-technology exports (current US$) 0.0
High-technology exports (% of manufactured exports) 0.0
Research and development expenditure (% of GDP) 0.0
Scientific and technical journal articles 0.0
Air transport, freight (million ton-km) 0.0
Air transport, passengers carried 0.0
Rail lines (total route-km)      0.0
Railways, goods transported (million ton-km) 0.0
Railways, passengers carried (million passenger-km) 0.0
Current health expenditure (% of GDP) 0.0
Current health expenditure per capita (current US$) 0.0
Hospital beds (per 1,000 people) 0.0
Physicians (per 1,000 people)    0.0
Life expectancy at birth, total (years) 0.0
Probability of dying among adolescents ages 10-14 years (per 1,000) 0.0
Probability of dying among adolescents ages 15-19 years (per 1,000) 0.0
Probability of dying among children ages 5-9 years (per 1,000) 0.0
Probability of dying among youth ages 20-24 years (per 1,000) 0.0
Survival to age 65, female (% of cohort) 0.0
Survival to age 65, male (% of cohort) 0.0
dtype: float64
```

There are no null values left in the dataset

In [15]: `df.shape`

```
Out[15]: (1245, 48)
```

We approximately deleted 300 rows and 7 columns from our dataset. Now we will handle the outliers

```
In [16]: df1 = df
for i in df.columns[3:]:
    q1 = df[i].quantile(.25)
    q3 = df[i].quantile(.75)
    iqr = q3 - q1
    lower = q1 - 1.5*iqr
    upper = q3 + 1.5*iqr
    df1 = df[(df[i] >= lower) & (df[i] <= upper)]
    print(f"Outliers for {i} deleted successfully")
```

Outliers for Current account balance (% of GDP) deleted successfully  
Outliers for Exports of goods and services (% of GDP) deleted successfully  
Outliers for Foreign direct investment, net inflows (% of GDP) deleted successfully  
Outliers for Foreign direct investment, net outflows (% of GDP) deleted successfully  
Outliers for GDP (current US\$) deleted successfully  
Outliers for GDP growth (annual %) deleted successfully  
Outliers for GDP per capita (current US\$) deleted successfully  
Outliers for GDP per capita growth (annual %) deleted successfully  
Outliers for General government final consumption expenditure (% of GDP) deleted successfully  
Outliers for Government expenditure on education, total (% of GDP) deleted successfully  
Outliers for Gross savings (% of GDP) deleted successfully  
Outliers for Imports of goods and services (% of GDP) deleted successfully  
Outliers for GNI (current US\$) deleted successfully  
Outliers for GNI growth (annual %) deleted successfully  
Outliers for Employment to population ratio, 15+, total (%) (national estimate) deleted successfully  
Outliers for Employment to population ratio, 15+, female (%) (national estimate) deleted successfully  
Outliers for Employment to population ratio, 15+, male (%) (national estimate) deleted successfully  
Outliers for Unemployment, total (% of total labor force) (national estimate) deleted successfully  
Outliers for Employment in services (% of total employment) (modeled ILO estimate) deleted successfully  
Outliers for Employment in industry (% of total employment) (modeled ILO estimate) deleted successfully  
Outliers for Employment in agriculture (% of total employment) (modeled ILO estimate) deleted successfully  
Outliers for Fixed broadband subscriptions deleted successfully  
Outliers for Fixed telephone subscriptions deleted successfully  
Outliers for Mobile cellular subscriptions deleted successfully  
Outliers for Secure Internet servers deleted successfully  
Outliers for High-technology exports (current US\$) deleted successfully  
Outliers for High-technology exports (% of manufactured exports) deleted successfully  
Outliers for Research and development expenditure (% of GDP) deleted successfully  
Outliers for Scientific and technical journal articles deleted successfully  
Outliers for Air transport, freight (million ton-km) deleted successfully  
Outliers for Air transport, passengers carried deleted successfully  
Outliers for Rail lines (total route-km) deleted successfully  
Outliers for Railways, goods transported (million ton-km) deleted successfully  
Outliers for Railways, passengers carried (million passenger-km) deleted successfully  
Outliers for Current health expenditure (% of GDP) deleted successfully  
Outliers for Current health expenditure per capita (current US\$) deleted successfully  
Outliers for Hospital beds (per 1,000 people) deleted successfully  
Outliers for Physicians (per 1,000 people) deleted successfully  
Outliers for Life expectancy at birth, total (years) deleted successfully  
Outliers for Probability of dying among adolescents ages 10-14 years (per 1,000) deleted successfully  
Outliers for Probability of dying among adolescents ages 15-19 years (per 1,000) deleted successfully  
Outliers for Probability of dying among children ages 5-9 years (per 1,000) deleted successfully  
Outliers for Probability of dying among youth ages 20-24 years (per 1,000) deleted successfully  
Outliers for Survival to age 65, female (% of cohort) deleted successfully  
Outliers for Survival to age 65, male (% of cohort) deleted successfully

In [17]: df1.info()

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1176 entries, 5 to 1562
Data columns (total 48 columns):
 #   Column           Non-Null Count Dtype  
 --- 
 0   Country Name    1176 non-null   object  
 1   Country Code    1176 non-null   object  
 2   Time             1176 non-null   int64  
 3   Current account balance (% of GDP) 1176 non-null   float64 
 4   Exports of goods and services (% of GDP) 1176 non-null   float64 
 5   Foreign direct investment, net inflows (% of GDP) 1176 non-null   float64 
 6   Foreign direct investment, net outflows (% of GDP) 1176 non-null   float64 
 7   GDP (current US$) 1176 non-null   float64 
 8   GDP growth (annual %) 1176 non-null   float64 
 9   GDP per capita (current US$) 1176 non-null   float64 
 10  GDP per capita growth (annual %) 1176 non-null   float64 
 11  General government final consumption expenditure (% of GDP) 1176 non-null   float64 
 12  Government expenditure on education, total (% of GDP) 1176 non-null   float64 
 13  Gross savings (% of GDP) 1176 non-null   float64 
 14  Imports of goods and services (% of GDP) 1176 non-null   float64 
 15  GNI (current US$) 1176 non-null   float64 
 16  GNI growth (annual %) 1176 non-null   float64 
 17  Employment to population ratio, 15+, total (%) (national estimate) 1176 non-null   float64 
 18  Employment to population ratio, 15+, female (%) (national estimate) 1176 non-null   float64 
 19  Employment to population ratio, 15+, male (%) (national estimate) 1176 non-null   float64 
 20  Unemployment, total (% of total labor force) (national estimate) 1176 non-null   float64 
 21  Employment in services (% of total employment) (modeled ILO estimate) 1176 non-null   float64 
 22  Employment in industry (% of total employment) (modeled ILO estimate) 1176 non-null   float64 
 23  Employment in agriculture (% of total employment) (modeled ILO estimate) 1176 non-null   float64 
 24  Fixed broadband subscriptions 1176 non-null   float64 
 25  Fixed telephone subscriptions 1176 non-null   float64 
 26  Mobile cellular subscriptions 1176 non-null   float64 
 27  Secure Internet servers 1176 non-null   float64 
 28  High-technology exports (current US$) 1176 non-null   float64 
 29  High-technology exports (% of manufactured exports) 1176 non-null   float64 
 30  Research and development expenditure (% of GDP) 1176 non-null   float64 
 31  Scientific and technical journal articles 1176 non-null   float64 
 32  Air transport, freight (million ton-km) 1176 non-null   float64 
 33  Air transport, passengers carried 1176 non-null   float64 
 34  Rail lines (total route-km) 1176 non-null   float64 
 35  Railways, goods transported (million ton-km) 1176 non-null   float64 
 36  Railways, passengers carried (million passenger-km) 1176 non-null   float64 
 37  Current health expenditure (% of GDP) 1176 non-null   float64 
 38  Current health expenditure per capita (current US$) 1176 non-null   float64 
 39  Hospital beds (per 1,000 people) 1176 non-null   float64 
 40  Physicians (per 1,000 people) 1176 non-null   float64 
 41  Life expectancy at birth, total (years) 1176 non-null   float64 
 42  Probability of dying among adolescents ages 10-14 years (per 1,000) 1176 non-null   float64 
 43  Probability of dying among adolescents ages 15-19 years (per 1,000) 1176 non-null   float64 
 44  Probability of dying among children ages 5-9 years (per 1,000) 1176 non-null   float64 
 45  Probability of dying among youth ages 20-24 years (per 1,000) 1176 non-null   float64 
 46  Survival to age 65, female (% of cohort) 1176 non-null   float64 
 47  Survival to age 65, male (% of cohort) 1176 non-null   float64 
dtypes: float64(45), int64(1), object(2)
memory usage: 450.2+ KB
```

In [18]: df1.describe()

Out[18]:

	Time	Current account balance (% of GDP)	Exports of goods and services (% of GDP)	Foreign direct investment, net inflows (% of GDP)	Foreign direct investment, net outflows (% of GDP)	GDP (current US\$)	GDP growth (annual %)	GDP per capita (current US\$)	GDP per capita growth (annual %)	General government final consumption expenditure (% of GDP)	...	Current health expenditure per capita (current US\$)
<b>count</b>	1176.000000	1176.000000	1176.000000	1176.000000	1176.000000	1.176000e+03	1176.000000	1176.000000	1176.000000	1176.000000	...	1176.000000
<b>mean</b>	2011.312075	0.738546	42.439117	3.727269	2.791240	1.377294e+12	3.195173	25008.143169	1.898861	16.461860	...	2086.616935
<b>std</b>	6.201199	7.433469	30.687337	7.329715	7.312648	3.194280e+12	3.878858	22449.195583	3.736378	4.841612	...	2291.591401
<b>min</b>	2000.000000	-22.939294	8.221612	-40.086346	-42.285842	4.906494e+09	-17.668333	405.424902	-18.854382	5.039343	...	14.414308
<b>25%</b>	2006.000000	-3.125145	24.068339	1.166062	0.300421	1.467500e+11	1.490717	5268.512788	0.422649	12.286122	...	284.519798
<b>50%</b>	2011.000000	-0.780812	33.273210	2.367740	1.153858	3.200000e+11	3.096182	18989.981000	1.974779	16.842715	...	1114.214966
<b>75%</b>	2017.000000	2.952055	52.349917	4.139932	3.363189	1.150000e+12	5.298130	41019.857433	3.951721	19.958276	...	3458.926819
<b>max</b>	2022.000000	45.459992	228.993771	106.498789	104.591831	2.570000e+13	26.170246	108798.451200	23.304693	29.321635	...	12473.791020

8 rows × 46 columns

```
In [19]: df = df1.copy()
```

```
In [20]: df['3-Year GDP Growth Avg'] = df.groupby('Country Name')['GDP per capita growth (annual %)'].transform(lambda x: x.rolling(3).mean())
df['5-Year Life Expectancy Avg'] = df.groupby('Country Name')[['Life expectancy at birth, total (years)']].transform(lambda x: x.rolling(5).mean())
df['Balance to GDP Ratio'] = df['Current account balance (% of GDP)'] / (df['GDP growth (annual %)'] + 1e-6)
df['Yearly GDP Growth Change'] = df.groupby('Country Name')['GDP growth (annual %)'].transform(lambda x: x.diff())
df['Yearly Life Expectancy Change'] = df.groupby('Country Name')[['Life expectancy at birth, total (years)']].transform(lambda x: x.pct_change())
df['Cumulative GDP Growth'] = df.groupby('Country Name')['GDP growth (annual %)'].cumsum()
df['Cumulative Life Expectancy'] = df.groupby('Country Name')[['Life expectancy at birth, total (years)']].cumsum()
```

```
In [21]: from sklearn.preprocessing import MinMaxScaler
```

```
scaler = MinMaxScaler()
scaled_columns = list(df.columns[3:])
df[scaled_columns] = scaler.fit_transform(df[scaled_columns])
```

```
In [22]: df.head()
```

Out[22]:

Country Name	Country Code	Time	Current account balance (% of GDP)	Exports of goods and services (% of GDP)	Foreign direct investment, net inflows (% of GDP)	Foreign direct investment, net outflows (% of GDP)	GDP (current US\$)	GDP growth (annual %)	GDP per capita (current US\$)	Probability of dying among youth ages 20-24 years (per 1,000)	Survival to age 65, female (% of cohort)	Survival to age 65, male (% of cohort)	3-Year GDP Growth Avg	5-Year Lif Expectanc Av
5 Pakistan	PAK	2005	0.299065	0.018255	0.283808	0.288109	0.005452	0.569017	0.003942	...	0.508671	0.325652	0.240357	NaN
6 Pakistan	PAK	2006	0.274435	0.019601	0.291476	0.288357	0.006114	0.541075	0.004646	...	0.393064	0.395308	0.275636	NaN
7 Pakistan	PAK	2007	0.269467	0.015416	0.294178	0.288261	0.006970	0.504422	0.005598	...	0.358382	0.413700	0.264447	0.685705
8 Pakistan	PAK	2008	0.222186	0.016552	0.291815	0.288063	0.007670	0.451401	0.006293	...	0.335260	0.429449	0.261218	0.623022
9 Pakistan	PAK	2009	0.304208	0.018885	0.281982	0.288156	0.007087	0.482244	0.005350	...	0.317919	0.428256	0.265656	0.590139

5 rows × 55 columns

```
In [23]: new_nulls = (df.isnull().sum()/len(df))*100
for i in range(len(new_nulls)):
    if new_nulls[i] > 0:
        df[df.columns[i]] = df[df.columns[i]].fillna(df[df.columns[i]].median())
        print(f"Values successfully imputed in {df.columns[i]}")
```

```
Values successfully imputed in 3-Year GDP Growth Avg
Values successfully imputed in 5-Year Life Expectancy Avg
Values successfully imputed in Yearly GDP Growth Change
Values successfully imputed in Yearly Life Expectancy Change
```

In [24]: df.isnull().sum()

```
Out[24]: Country Name          0
Country Code          0
Time                  0
Current account balance (% of GDP) 0
Exports of goods and services (% of GDP) 0
Foreign direct investment, net inflows (% of GDP) 0
Foreign direct investment, net outflows (% of GDP) 0
GDP (current US$)      0
GDP growth (annual %)    0
GDP per capita (current US$) 0
GDP per capita growth (annual %) 0
General government final consumption expenditure (% of GDP) 0
Government expenditure on education, total (% of GDP) 0
Gross savings (% of GDP)      0
Imports of goods and services (% of GDP) 0
GNI (current US$)      0
GNI growth (annual %)    0
Employment to population ratio, 15+, total (%) (national estimate) 0
Employment to population ratio, 15+, female (%) (national estimate) 0
Employment to population ratio, 15+, male (%) (national estimate) 0
Unemployment, total (% of total labor force) (national estimate) 0
Employment in services (% of total employment) (modeled ILO estimate) 0
Employment in industry (% of total employment) (modeled ILO estimate) 0
Employment in agriculture (% of total employment) (modeled ILO estimate) 0
Fixed broadband subscriptions 0
Fixed telephone subscriptions 0
Mobile cellular subscriptions 0
Secure Internet servers      0
High-technology exports (current US$) 0
High-technology exports (% of manufactured exports) 0
Research and development expenditure (% of GDP) 0
Scientific and technical journal articles 0
Air transport, freight (million ton-km) 0
Air transport, passengers carried 0
Rail lines (total route-km) 0
Railways, goods transported (million ton-km) 0
Railways, passengers carried (million passenger-km) 0
Current health expenditure (% of GDP) 0
Current health expenditure per capita (current US$) 0
Hospital beds (per 1,000 people) 0
Physicians (per 1,000 people) 0
Life expectancy at birth, total (years) 0
Probability of dying among adolescents ages 10-14 years (per 1,000) 0
Probability of dying among adolescents ages 15-19 years (per 1,000) 0
Probability of dying among children ages 5-9 years (per 1,000) 0
Probability of dying among youth ages 20-24 years (per 1,000) 0
Survival to age 65, female (% of cohort) 0
Survival to age 65, male (% of cohort) 0
3-Year GDP Growth Avg      0
5-Year Life Expectancy Avg 0
Balance to GDP Ratio      0
Yearly GDP Growth Change 0
Yearly Life Expectancy Change 0
Cumulative GDP Growth     0
Cumulative Life Expectancy 0
dtype: int64
```

In [25]:

```
aggregates = {}
for i in df.columns[3:]:
    if "Cumulative" not in i:
        aggregates[i] = "mean"
    else:
        aggregates[i] = "max"

aggregate_rename = {}
for i in df.columns[3:-7]:
    aggregate_rename[i] = "Average " + i
```

```
In [26]: country_aggregates = df.groupby('Country Name').agg(aggregates).reset_index()
country_aggregates.rename(columns=aggregate_rename, inplace=True)
country_aggregates.head()
```

Out[26]:

Country Name	Average Current account balance (% of GDP)	Average Exports of goods and services (% of GDP)	Average Foreign direct investment, net inflows (% of GDP)	Average Foreign direct investment, net outflows (% of GDP)	Average GDP (current US\$)	Average GDP growth (annual %)	Average GDP per capita (current US\$)	Average GDP per capita growth (annual %)	Average government final consumption expenditure (% of GDP)	Average Probability of dying among youth ages 20-24 years (per 1,000)	Average Survival to age 65, female (% of cohort)	Average Survival to age 65, male (% of cohort)	3-Year GDP Growth Avg
0 Algeria	0.356630	0.104255	0.279670	0.288462	0.006479	0.466302	0.038012	0.470538	0.493519	...	0.132035	0.731982	0.694587
1 Argentina	0.340123	0.045685	0.286121	0.289983	0.015185	0.447864	0.082597	0.469672	0.411391	...	0.204789	0.770789	0.603105
2 Australia	0.286841	0.057104	0.295030	0.294876	0.042701	0.463070	0.437828	0.475612	0.584183	...	0.065607	0.924465	0.887300
3 Austria	0.359772	0.193256	0.288743	0.312446	0.014248	0.434099	0.399327	0.467117	0.602613	...	0.068050	0.906305	0.806401
4 Bahrain	0.380535	0.333603	0.304826	0.304626	0.000825	0.491881	0.193193	0.456015	0.423977	...	0.093862	0.851781	0.848976

5 rows × 53 columns

Now lets make a new column for economic growth index, it will depend on three factors.

**GDP Growth (50%):** The primary indicator for economic growth, reflecting overall economic activity.

**Current Account Balance (30%):** A significant measure of trade and financial stability, contributing to long-term growth potential.

**Life Expectancy (20%):** Reflects the health and well-being of a population, indirectly impacting economic productivity and sustainability.

The proposed weights are arbitrary and could be changed or modified according to the needs

```
In [27]: country_aggregates['Economic Growth Index'] =
    country_aggregates['Average GDP growth (annual %)'] * 0.5 +
    country_aggregates['Average Current account balance (% of GDP)'] * 0.3 +
    country_aggregates['Average Life expectancy at birth, total (years)'] * 0.2
)
country_aggregates.head()
```

Out[27]:

Country Name	Average Current account balance (% of GDP)	Average Exports of goods and services (% of GDP)	Average Foreign direct investment, net inflows (% of GDP)	Average Foreign direct investment, net outflows (% of GDP)	Average GDP (current US\$)	Average GDP growth (annual %)	Average GDP per capita (current US\$)	Average GDP per capita growth (annual %)	Average government final consumption expenditure (% of GDP)	Average Survival to age 65, female (% of cohort)	Average Survival to age 65, male (% of cohort)	3-Year GDP Growth Avg	5-Year Life Expectancy Avg
0 Algeria	0.356630	0.104255	0.279670	0.288462	0.006479	0.466302	0.038012	0.470538	0.493519	...	0.731982	0.694587	0.584894
1 Argentina	0.340123	0.045685	0.286121	0.289983	0.015185	0.447864	0.082597	0.469672	0.411391	...	0.770789	0.603105	0.598248
2 Australia	0.286841	0.057104	0.295030	0.294876	0.042701	0.463070	0.437828	0.475612	0.584183	...	0.924465	0.887300	0.595388
3 Austria	0.359772	0.193256	0.288743	0.312446	0.014248	0.434099	0.399327	0.467117	0.602613	...	0.906305	0.806401	0.581304
4 Bahrain	0.380535	0.333603	0.304826	0.304626	0.000825	0.491881	0.193193	0.456015	0.423977	...	0.851781	0.848976	0.566074

5 rows × 54 columns

In [28]: country\_aggregates.describe(include="all")

Out[28]:

	Country Name	Average Current account balance (% of GDP)	Average Exports of goods and services (% of GDP)	Average Foreign direct investment, net inflows (% of GDP)	Average Foreign direct investment, net outflows (% of GDP)	Average GDP (current US\$)	Average GDP growth (annual %)	Average GDP per capita (current US\$)	Average GDP per capita growth (annual %)	Average General government final consumption expenditure (% of GDP)	... Average Survival to age 65, female (% of cohort)	... Average Survival to age 65, male (% of cohort)	3-Year GDP Growth Avg
<b>count</b>	60	60.000000	60.000000	60.000000	60.000000	60.000000	60.000000	60.000000	60.000000	60.000000	... 60.000000	60.000000	60.000000
<b>unique</b>	60	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	... NaN	NaN	NaN
<b>top</b>	Algeria	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	... NaN	NaN	NaN
<b>freq</b>	1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	... NaN	NaN	NaN
<b>mean</b>	NaN	0.342301	0.148488	0.297849	0.305253	0.048025	0.477986	0.206644	0.492274	0.452508	... 0.750988	0.620929	0.617703
<b>std</b>	NaN	0.087457	0.132396	0.026591	0.022609	0.111780	0.043821	0.193925	0.039610	0.195286	... 0.212838	0.248857	0.055896
<b>min</b>	NaN	0.229249	0.012136	0.275783	0.286403	0.000212	0.392428	0.003957	0.406319	0.019356	... 0.014448	0.004243	0.504339
<b>25%</b>	NaN	0.291397	0.075296	0.285269	0.291015	0.006285	0.441437	0.041417	0.470162	0.286359	... 0.717243	0.490526	0.583206
<b>50%</b>	NaN	0.323133	0.106933	0.291489	0.297108	0.010542	0.479973	0.110275	0.478220	0.432430	... 0.812408	0.690382	0.599209
<b>75%</b>	NaN	0.375067	0.193706	0.298603	0.311461	0.042685	0.501700	0.369291	0.517947	0.600167	... 0.903908	0.827537	0.653949
<b>max</b>	NaN	0.731011	0.843306	0.411999	0.393886	0.637725	0.601045	0.708976	0.639086	0.821999	... 0.946857	0.915748	0.831369

11 rows × 54 columns

In [29]: df.describe(include="all")

Out[29]:

	Country Name	Country Code	Time	Current account balance (% of GDP)	Exports of goods and services (% of GDP)	Foreign direct investment, net inflows (% of GDP)	Foreign direct investment, net outflows (% of GDP)	GDP (current US\$)	GDP growth (annual %)	GDP per capita (current US\$)	Probability of dying among youth ages 20-24 years (per 1,000)	Survival t
												age 65 female (% of cohort)
<b>count</b>	1176	1176	1176.000000	1176.000000	1176.000000	1176.000000	1176.000000	1176.000000	1176.000000	1176.000000	... 1176.000000	1176.000000
<b>unique</b>	60	60	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	... NaN	NaN
<b>top</b>	United Kingdom	GBR	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	... NaN	NaN
<b>freq</b>	23	23	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	... NaN	NaN
<b>mean</b>	NaN	NaN	2011.312075	0.346171	0.154990	0.298895	0.306902	0.053410	0.475917	0.226977	... 0.159054	0.79094
<b>std</b>	NaN	NaN	6.201199	0.108678	0.139000	0.050003	0.049787	0.124315	0.088480	0.207109	... 0.152201	0.16681
<b>min</b>	NaN	NaN	2000.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	... 0.000000	0.000000
<b>25%</b>	NaN	NaN	2006.000000	0.289684	0.071779	0.281423	0.289944	0.005520	0.437036	0.044865	... 0.057803	0.73733
<b>50%</b>	NaN	NaN	2011.000000	0.323958	0.113473	0.289621	0.295754	0.012263	0.473658	0.171455	... 0.109827	0.84370
<b>75%</b>	NaN	NaN	2017.000000	0.378532	0.199882	0.301711	0.310796	0.044565	0.523887	0.374696	... 0.219653	0.90653
<b>max</b>	NaN	NaN	2022.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	... 1.000000	1.000000

11 rows × 55 columns

## Univariate analysis

In [30]: %matplotlib inline

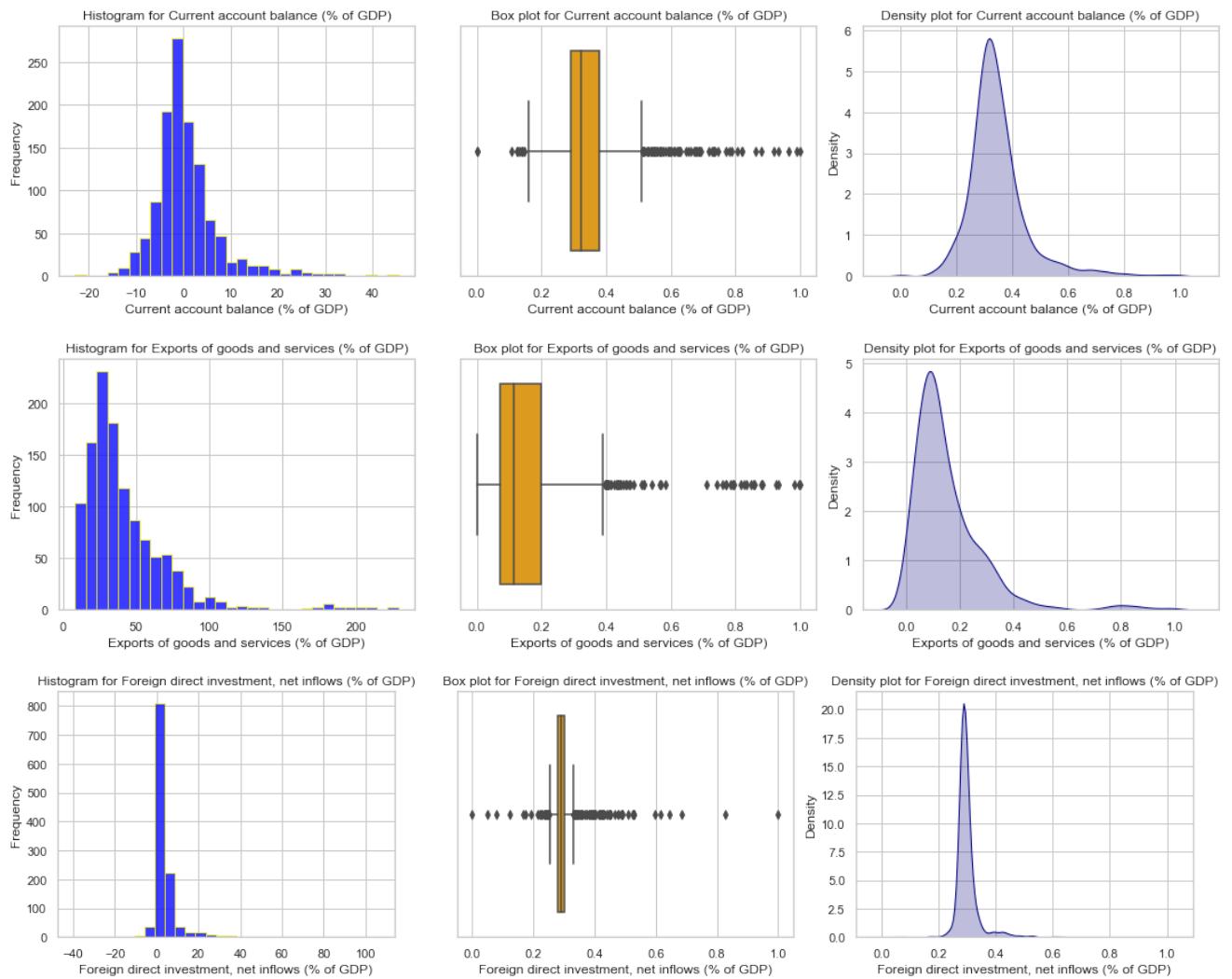
```
In [31]: economic_metrics = list(df1.columns[3:24])
for i in economic_metrics:
    plt.figure(figsize=(15,4))

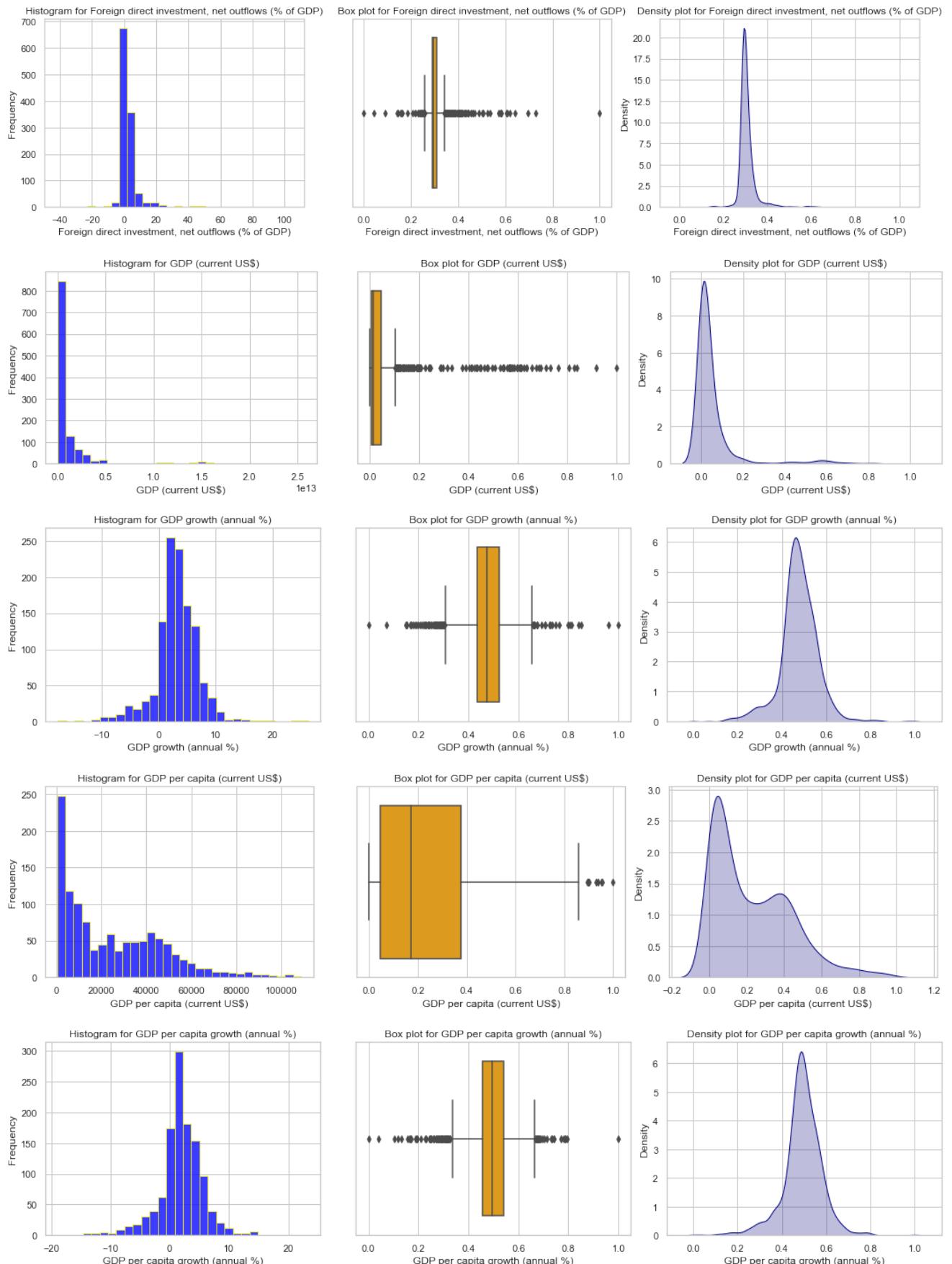
    plt.subplot(1,3,1)
    sns.histplot(data=df1[i], color="blue", edgecolor="yellow", bins=30)
    plt.xlabel(i)
    plt.ylabel("Frequency")
    plt.title(f"Histogram for {i}")

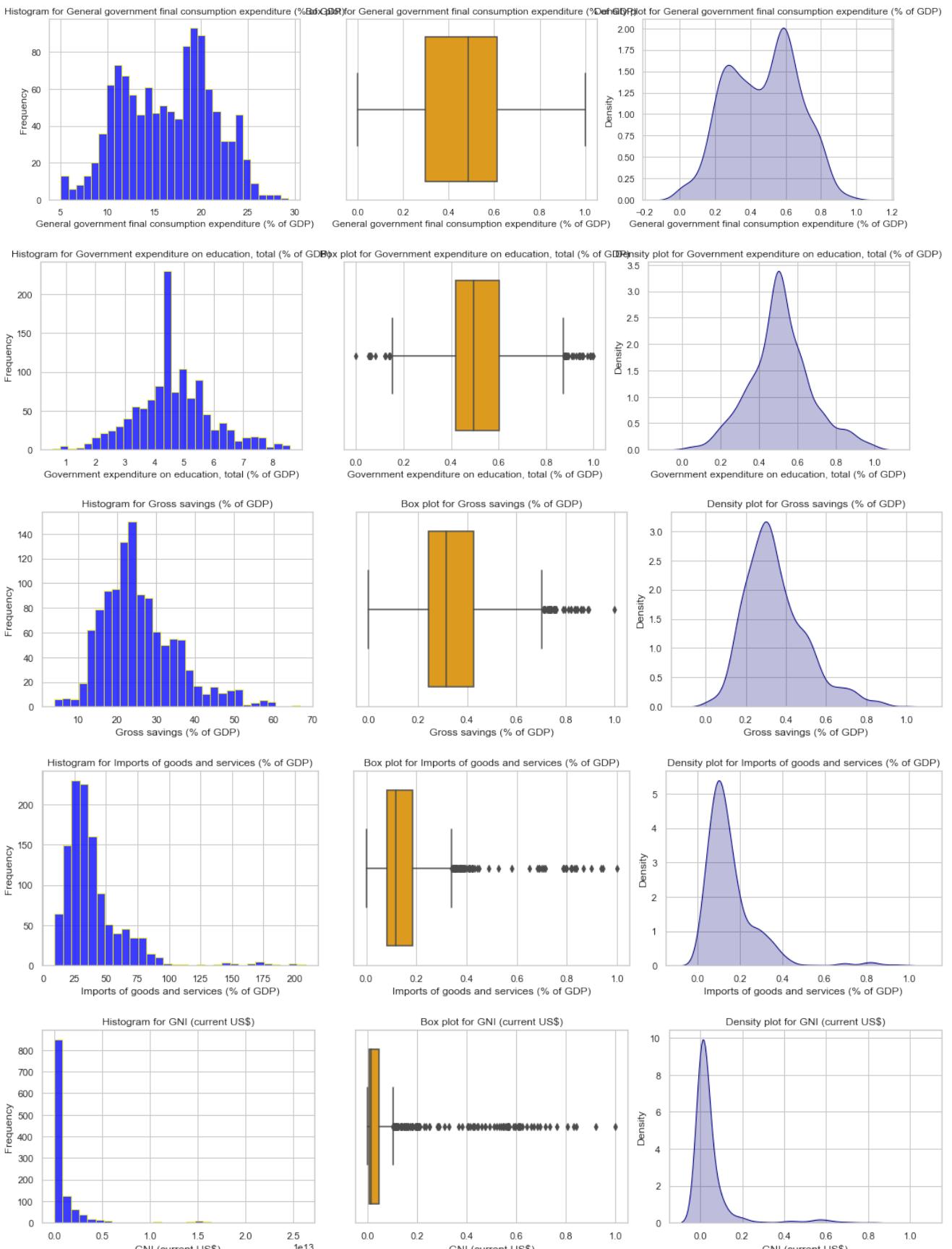
    plt.subplot(1,3,2)
    sns.boxplot(x=df[i], color="orange")
    plt.xlabel(i)
    plt.title(f"Box plot for {i}")

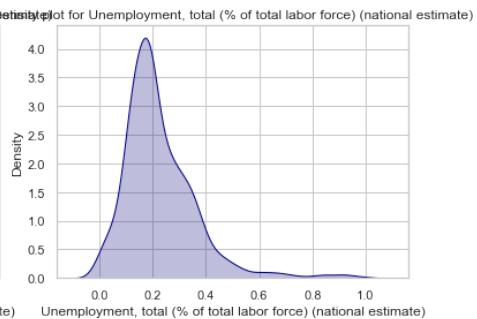
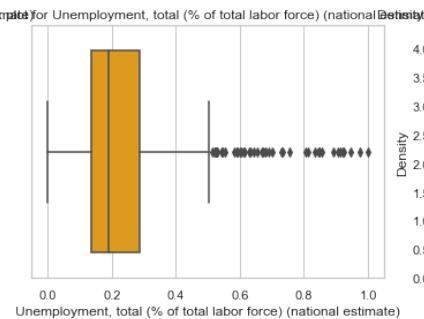
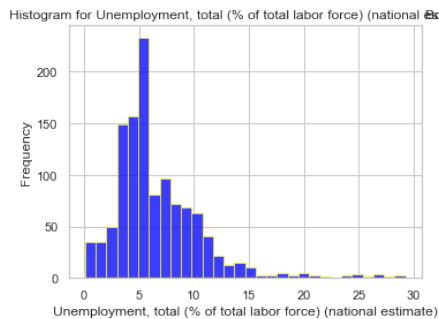
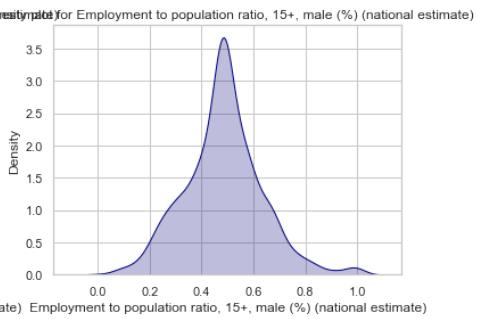
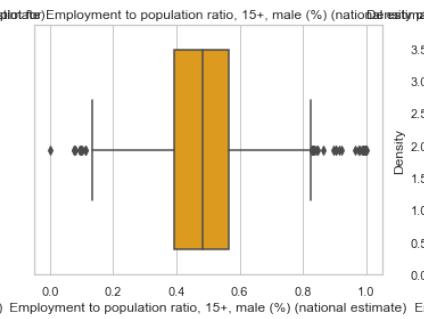
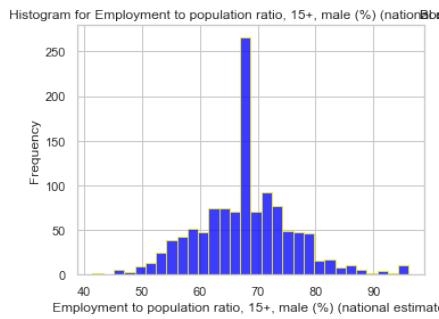
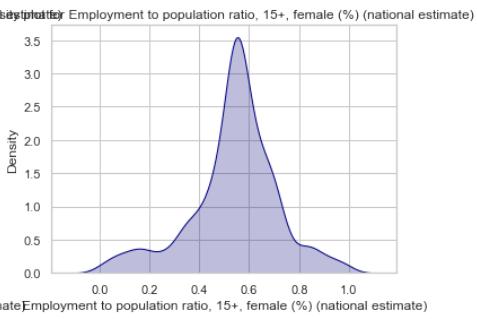
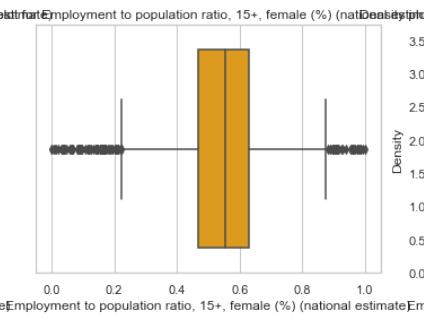
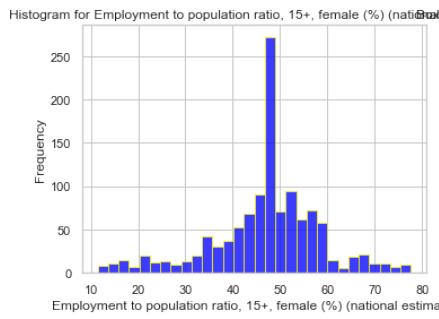
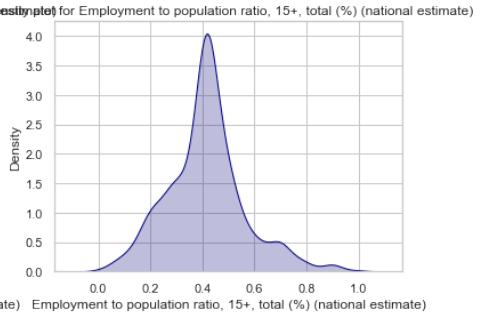
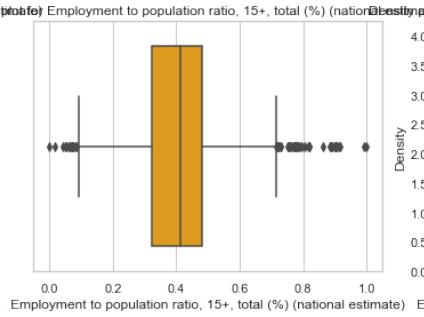
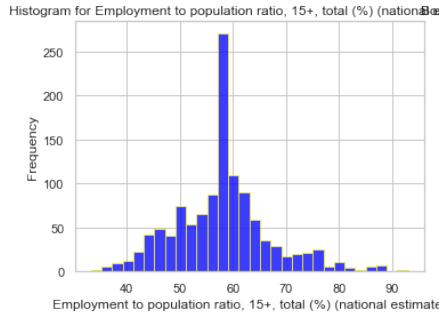
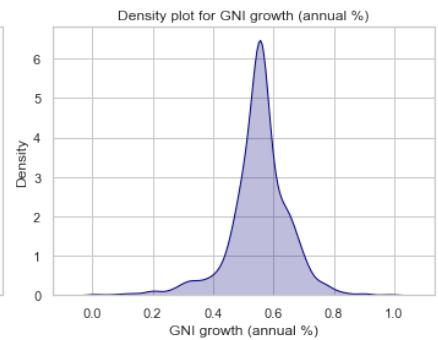
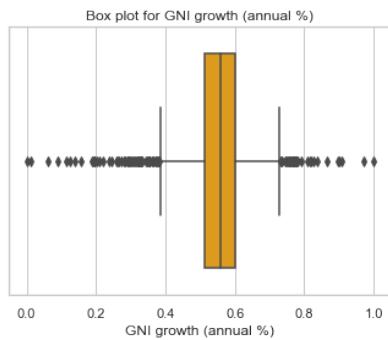
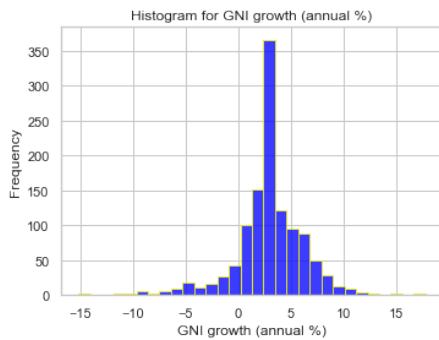
    plt.subplot(1,3,3)
    sns.kdeplot(data=df[i], color="navy", fill=True)
    plt.xlabel(i)
    plt.ylabel("Density")
    plt.title(f"Density plot for {i}")

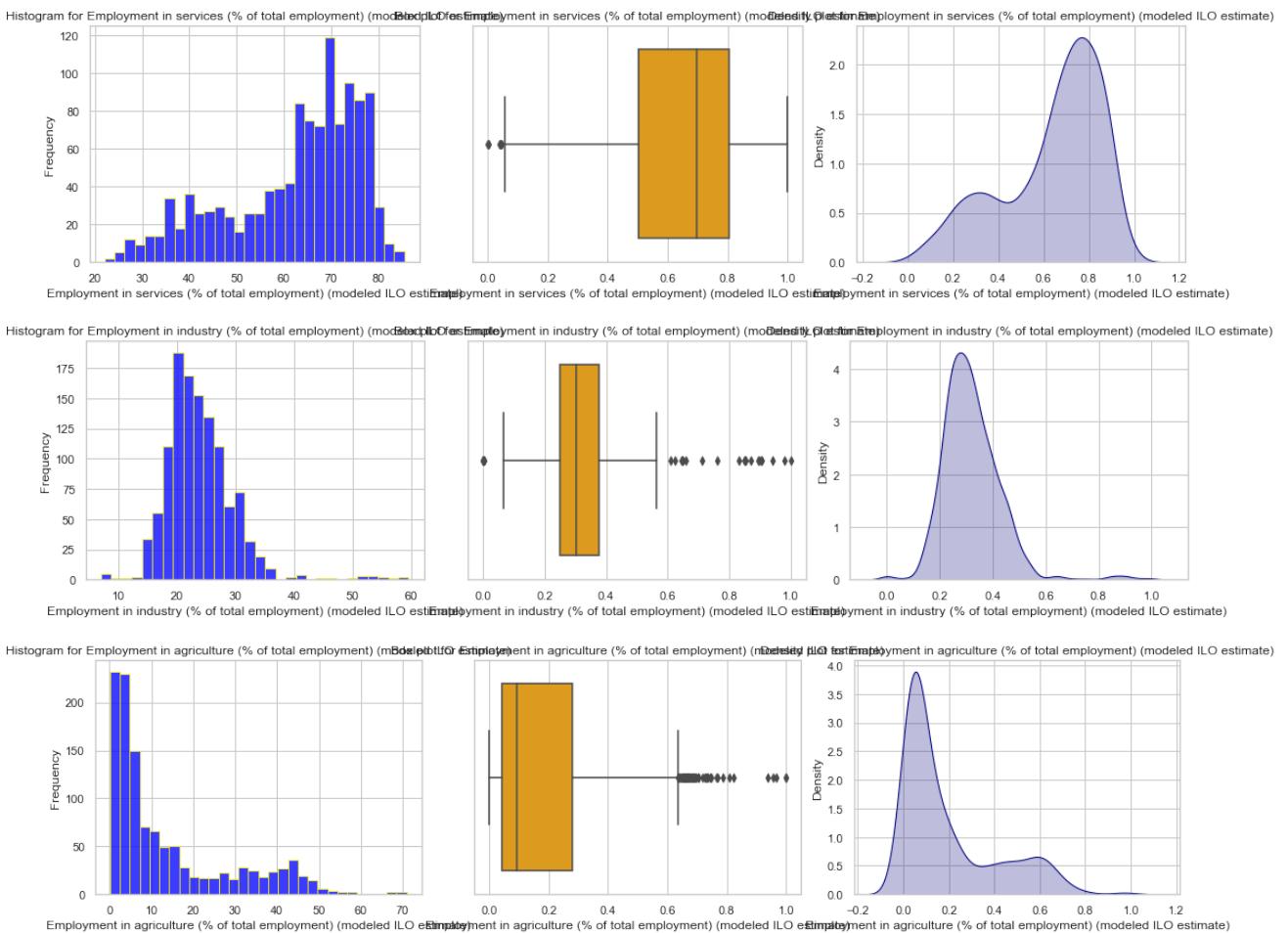
plt.tight_layout()
plt.show()
```





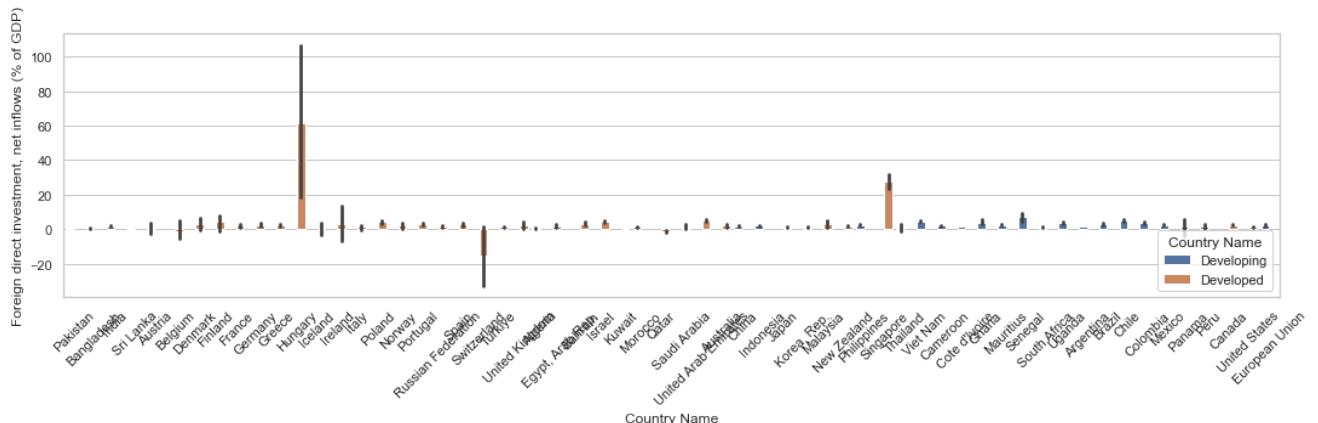
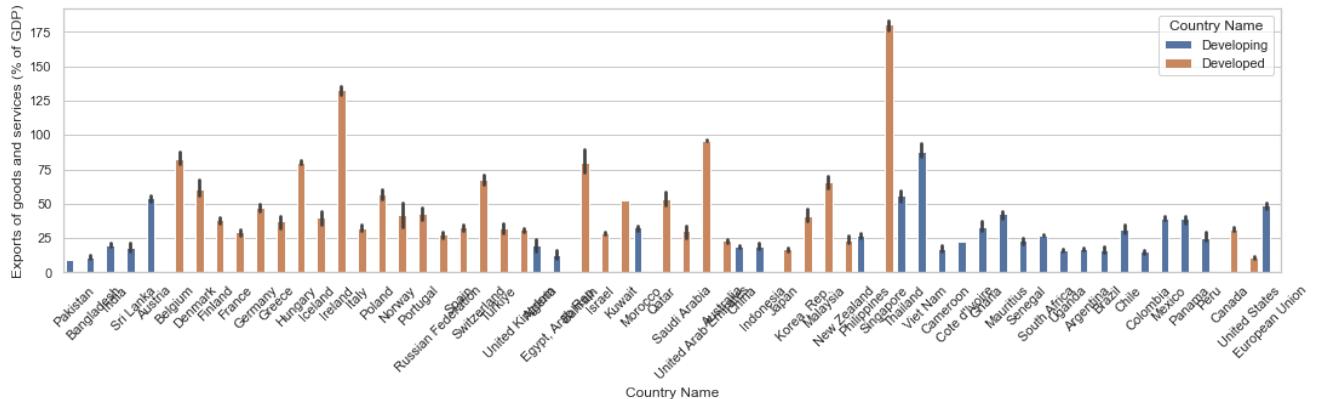
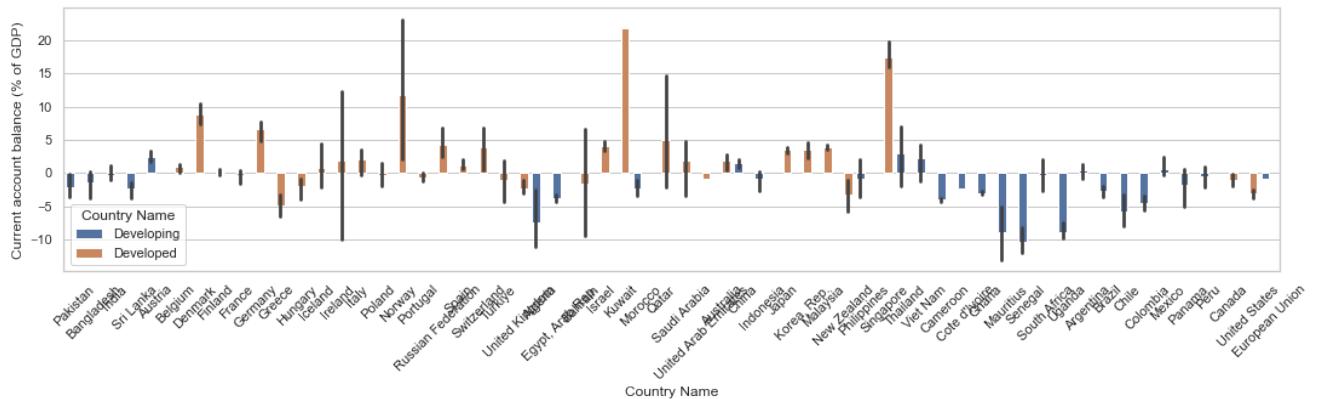


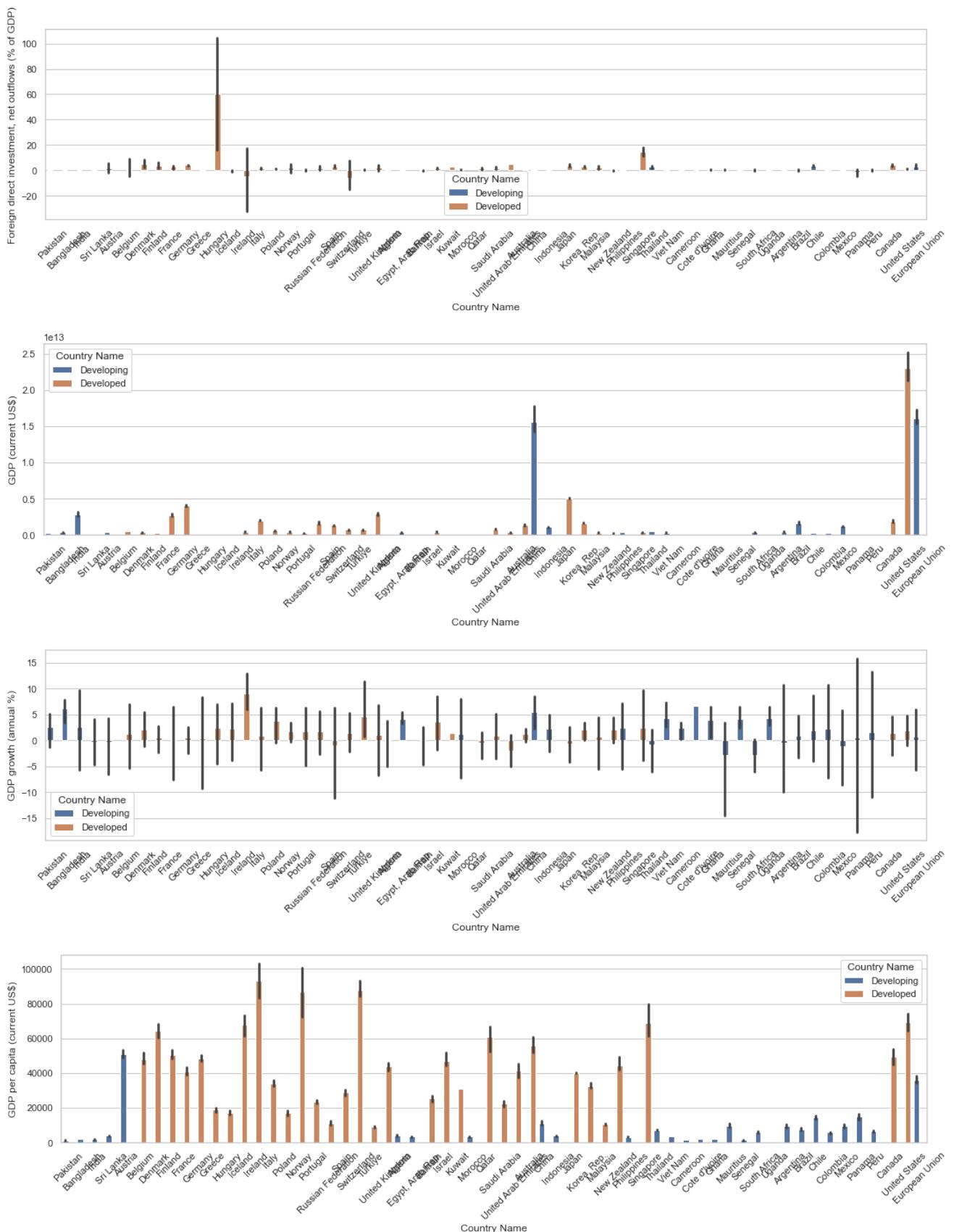


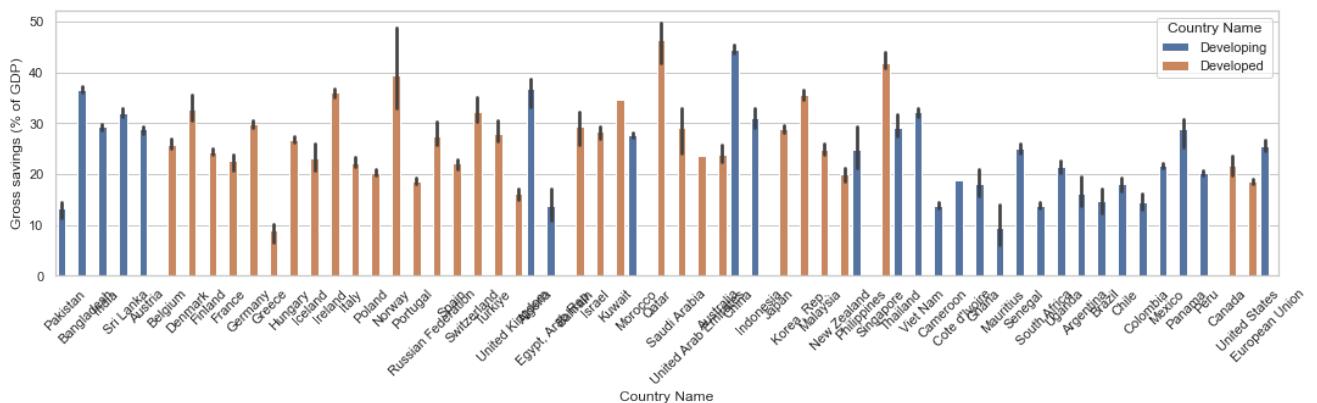
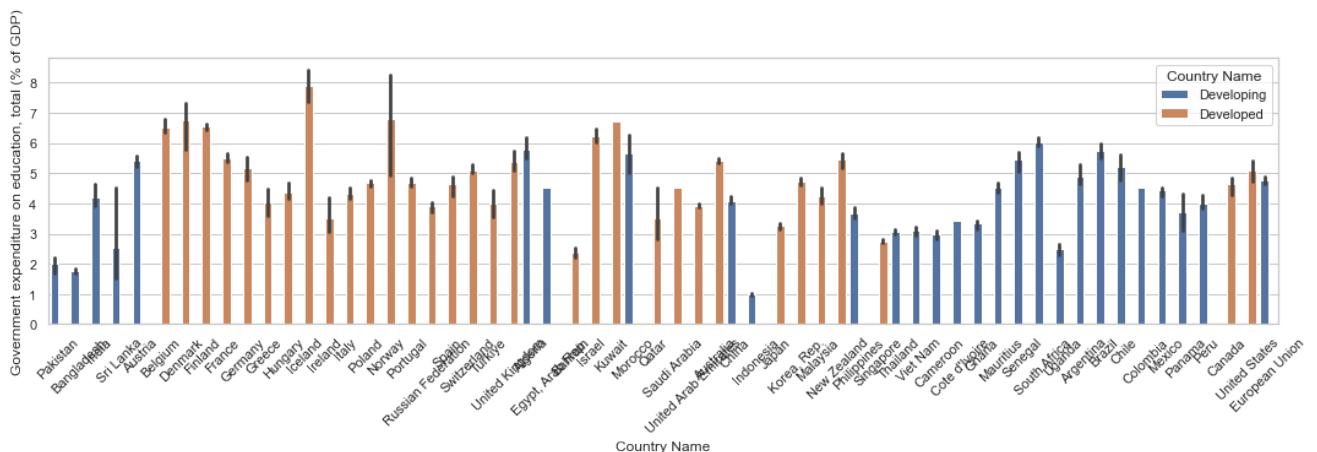
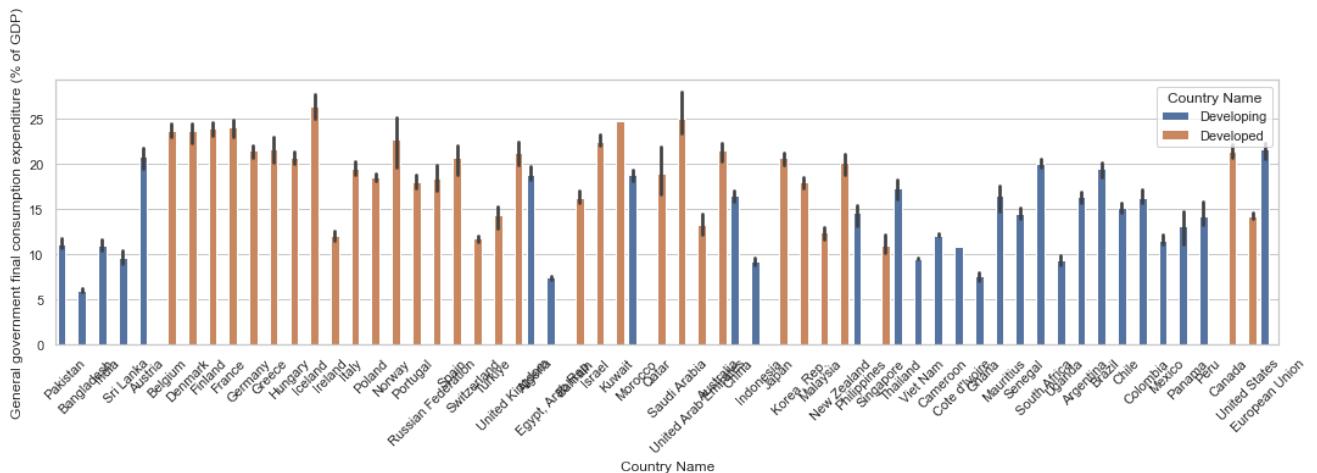
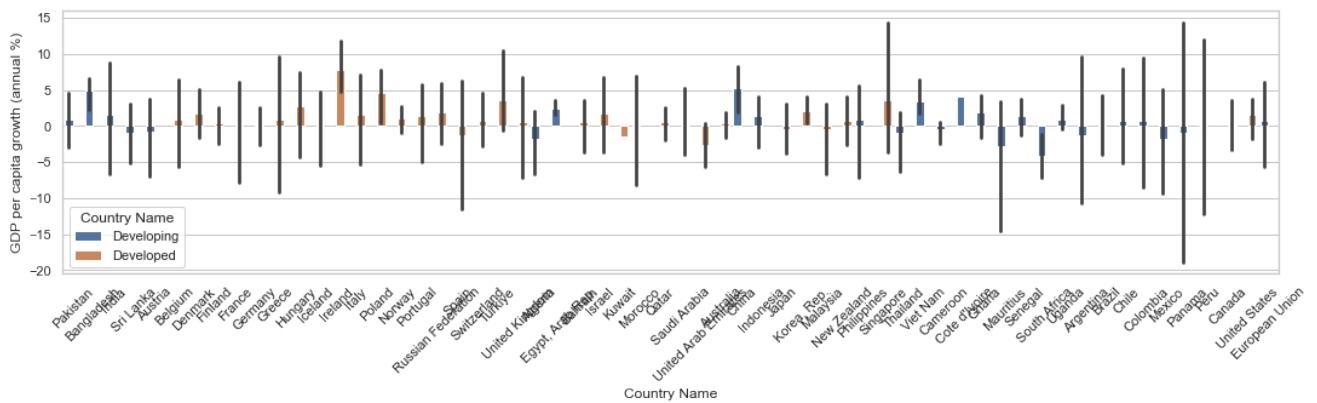


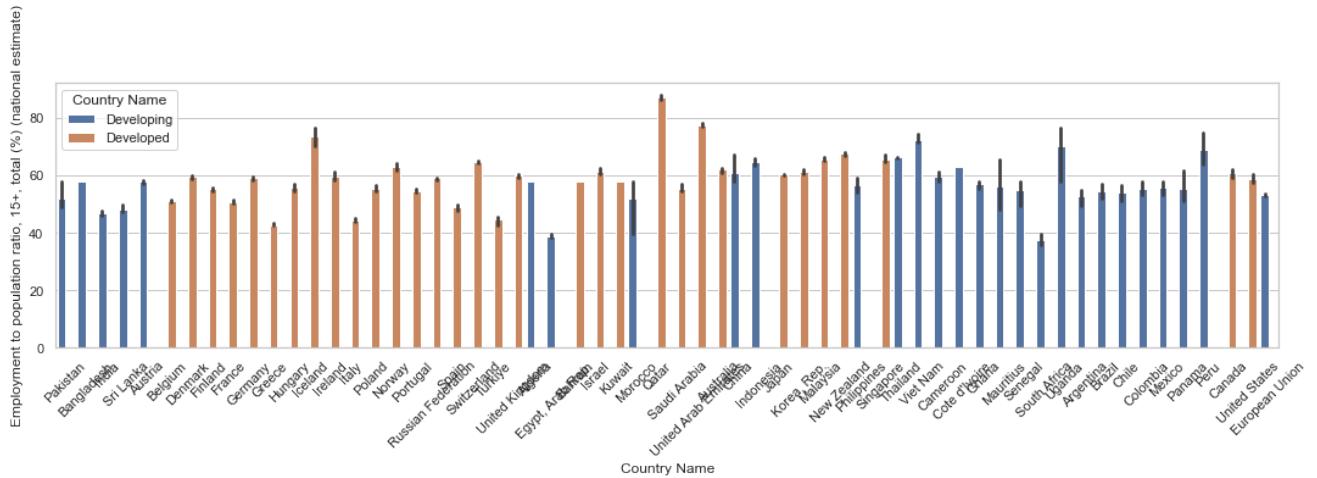
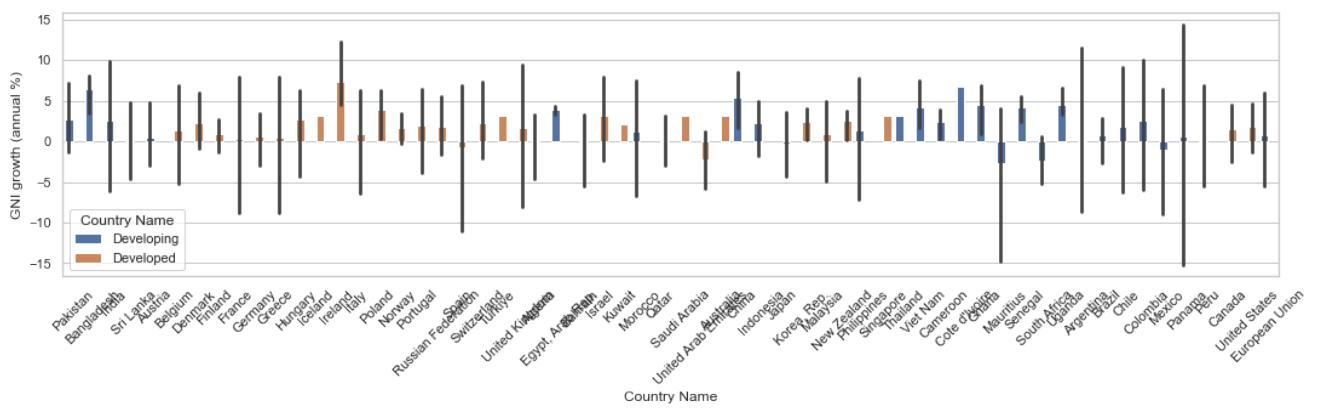
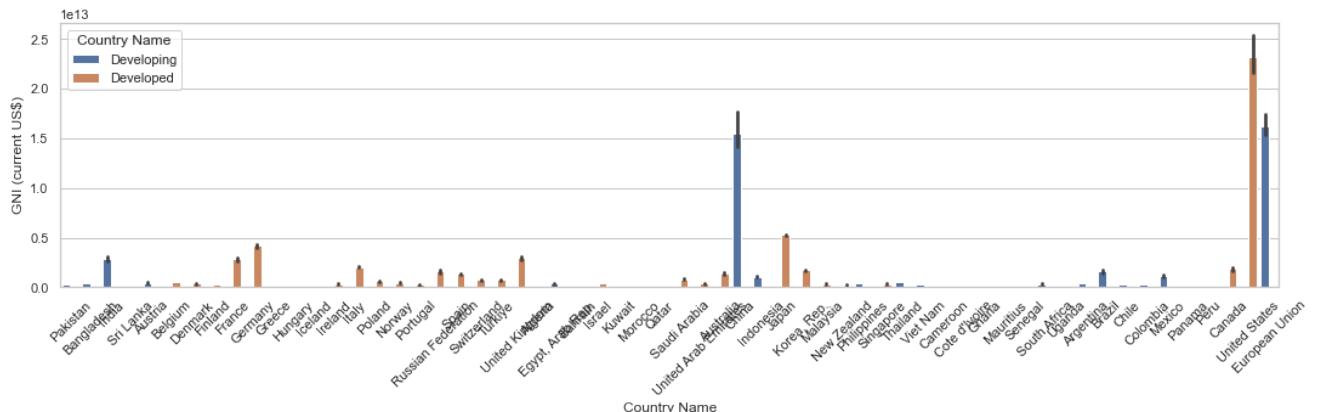
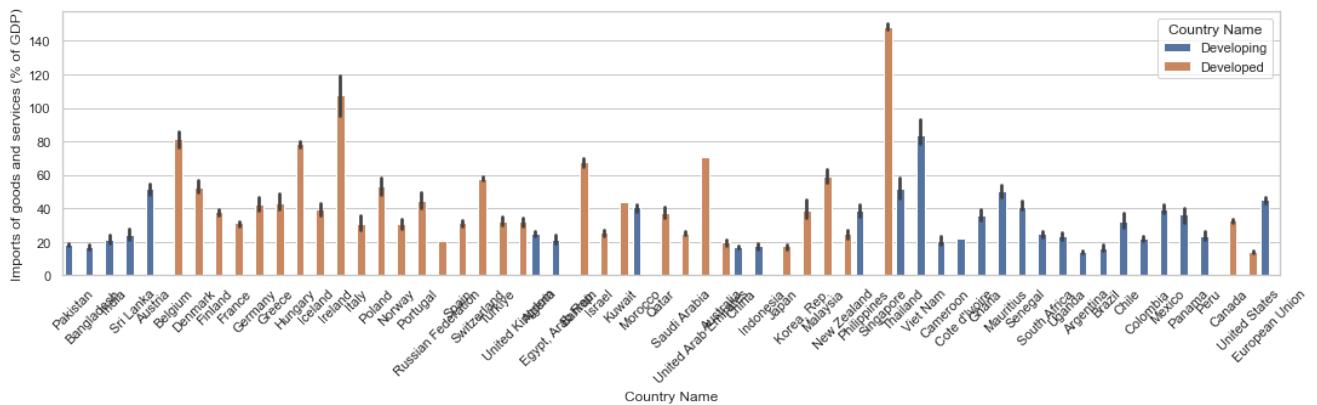
```
In [32]: developed_countries = ['Belgium', 'Denmark', 'Finland', 'France', 'Germany', 'Greece', 'Hungary', 'Iceland', 'Ireland', 'Italy', 'Poland', 'Norway', 'Portugal', 'Russian Federation', 'Spain', 'Switzerland', 'Turkiye', 'United Kingdom', 'Bahrain', 'Israel', 'Kuwait', 'Qatar', 'Saudi Arabia', 'United Arab Emirates', 'Australia', 'Japan', 'Korea, Rep.', 'Malaysia', 'New Zealand', 'Singapore', 'Canada', 'United States']
developing_countries = []
for i in df["Country Name"].unique():
    if i not in developed_countries:
        developing_countries.append(i)

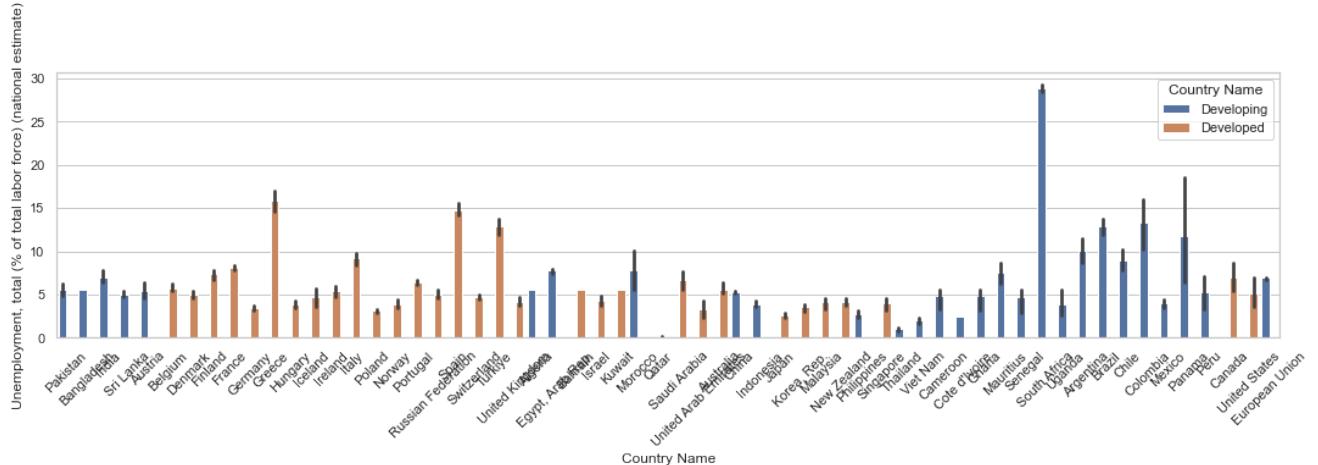
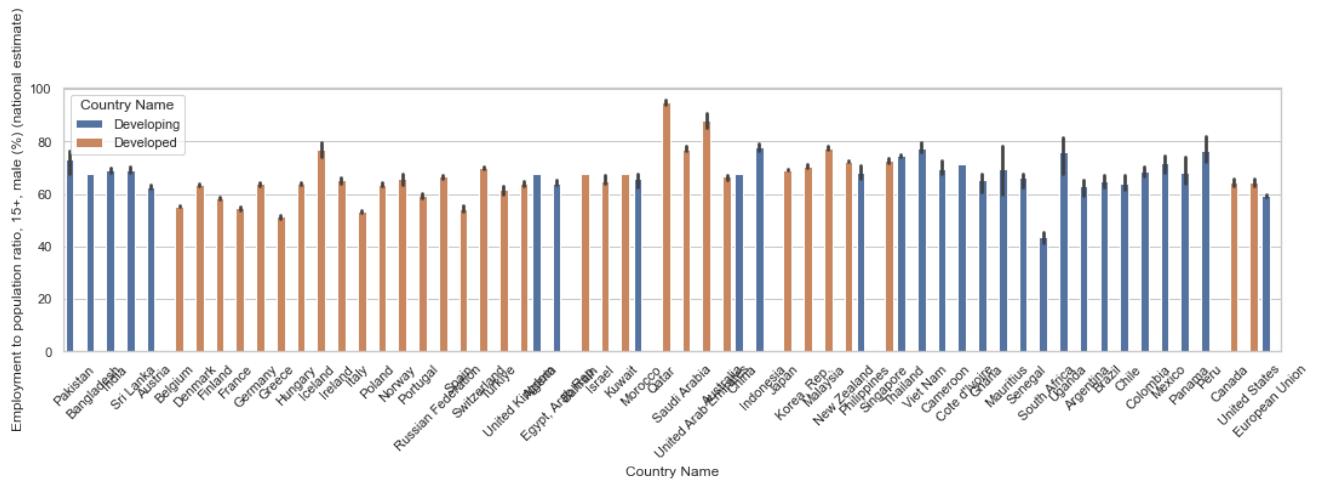
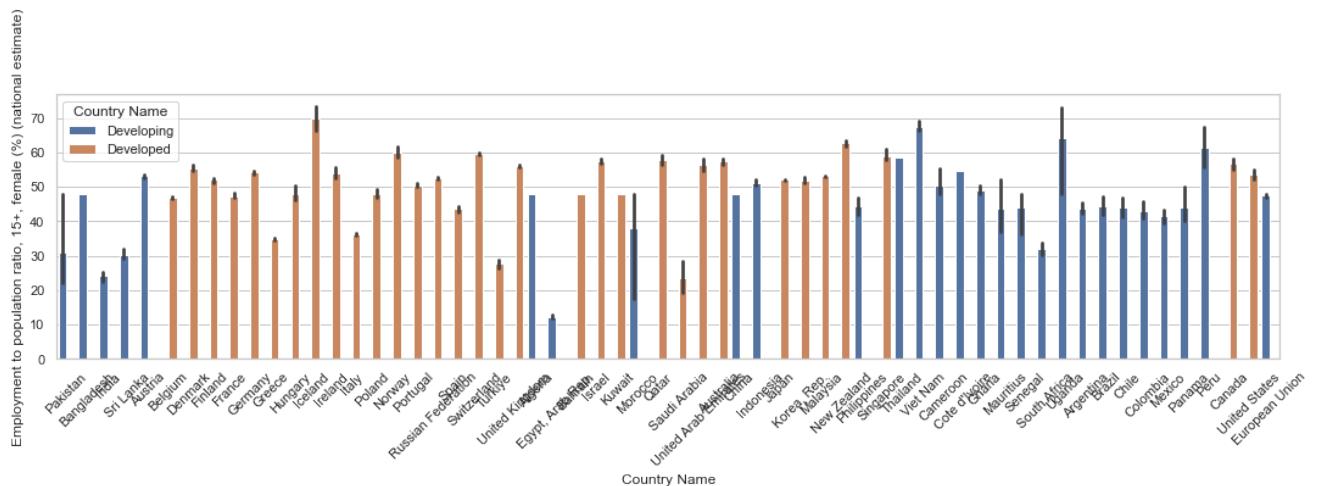
df2 = df1[df1.Time.isin([2019, 2020, 2021, 2022])]
for i in economic_metrics:
    plt.figure(figsize=(18,4))
    sns.barplot(data=df2, x="Country Name", y=i, hue=df2['Country Name'].apply(lambda x: 'Developed' if x in developed_countries
    plt.xticks(rotation=45)
    plt.show()
```

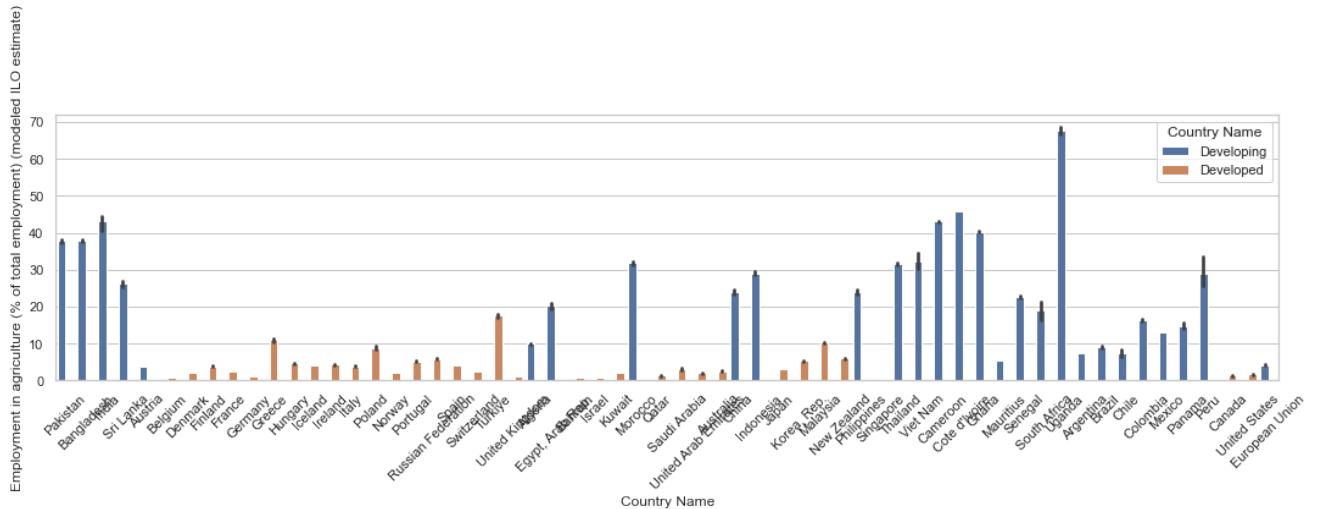
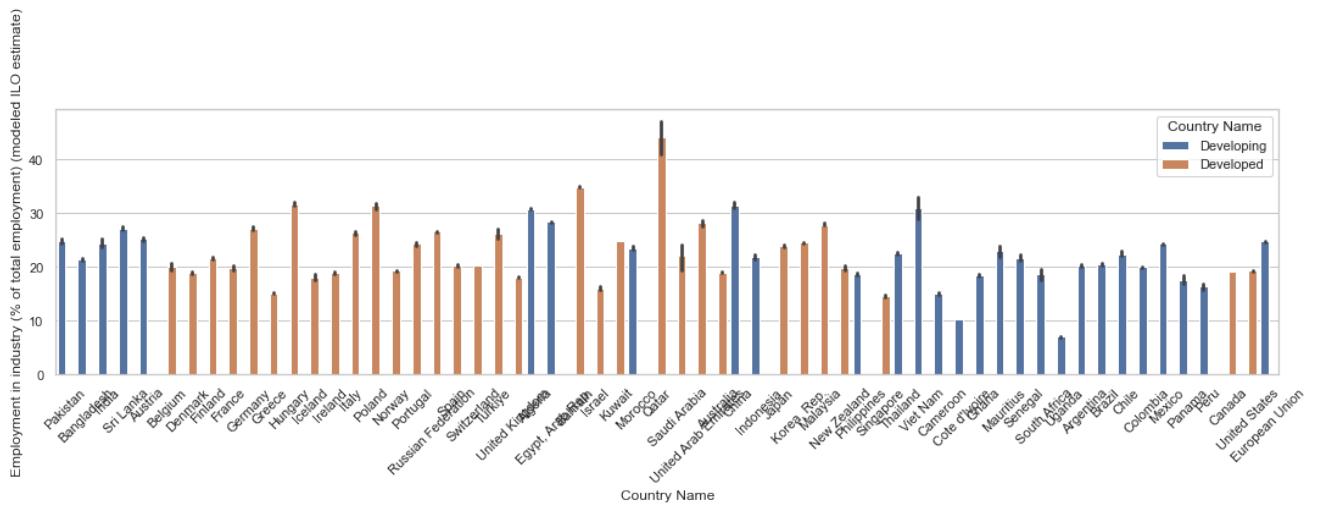
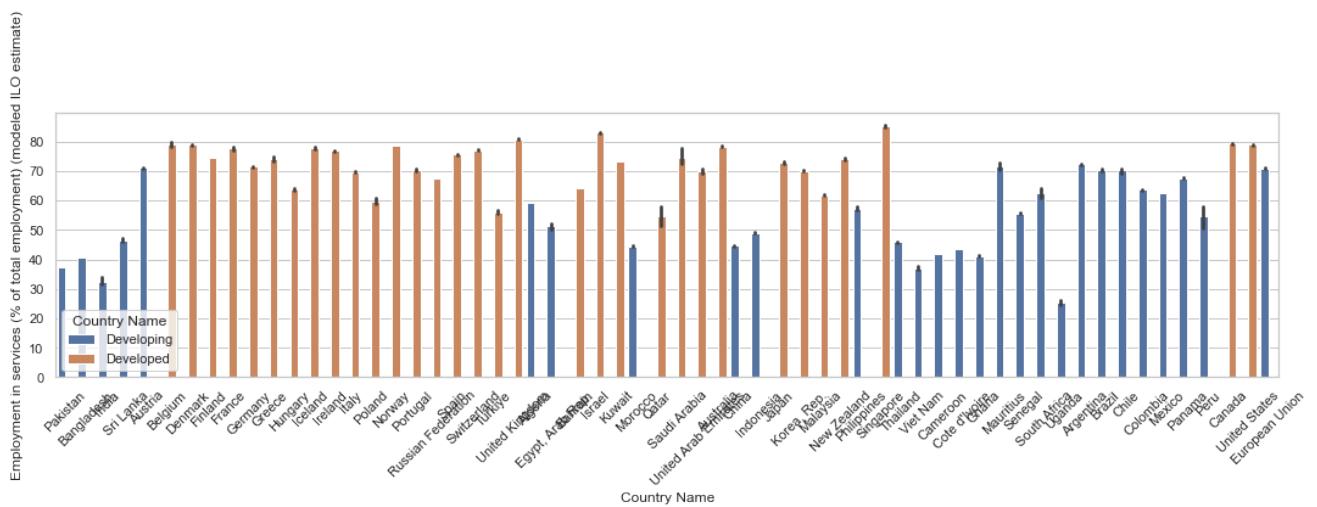












## Bivariate Analysis

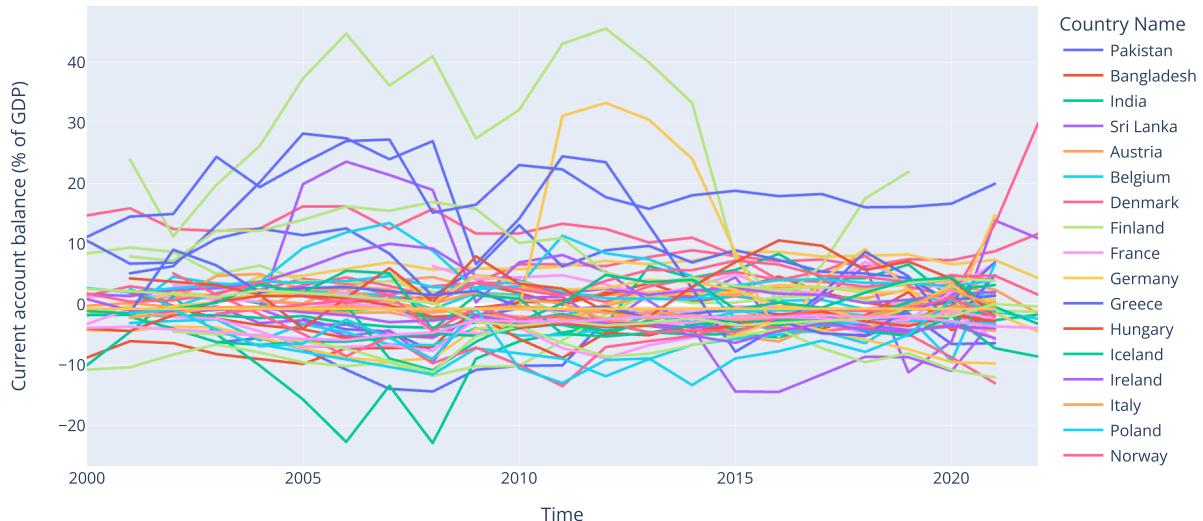
```
In [33]: fig = px.line(df1, x='Time', y='Life expectancy at birth, total (years)', color='Country Name',
                     title="Life Expectancy Trends (2000-2024)")
fig.show()
```

Life Expectancy Trends (2000-2024)

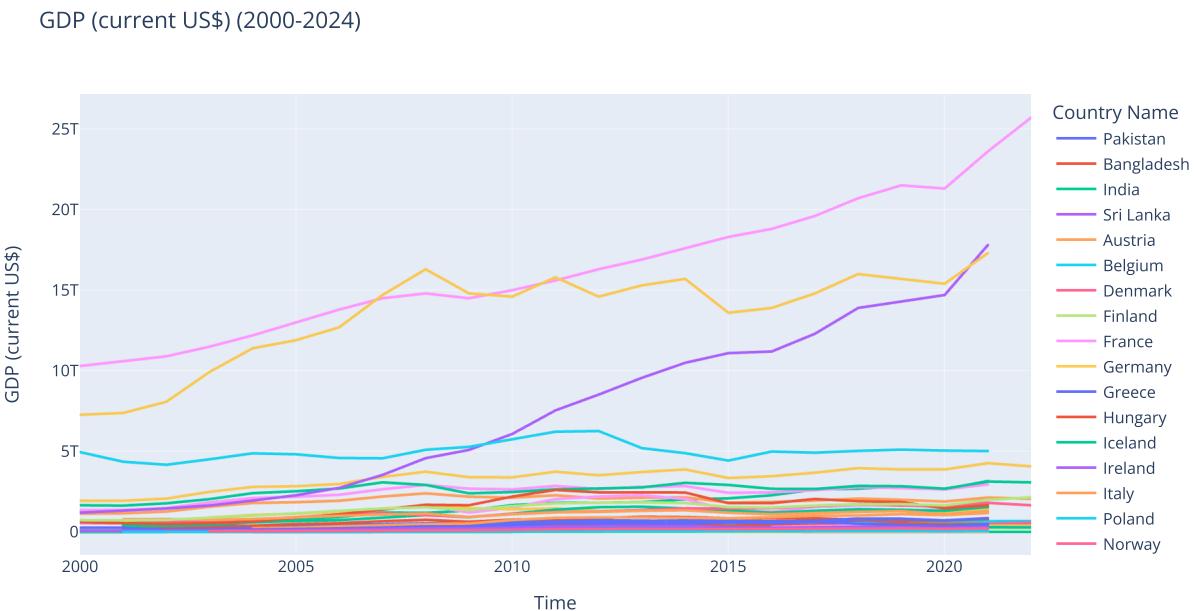


```
In [34]: fig = px.line(df1, x='Time', y='Current account balance (% of GDP)', color='Country Name',
                     title="Current account balance (% of GDP) (2000-2024)")
fig.show()
```

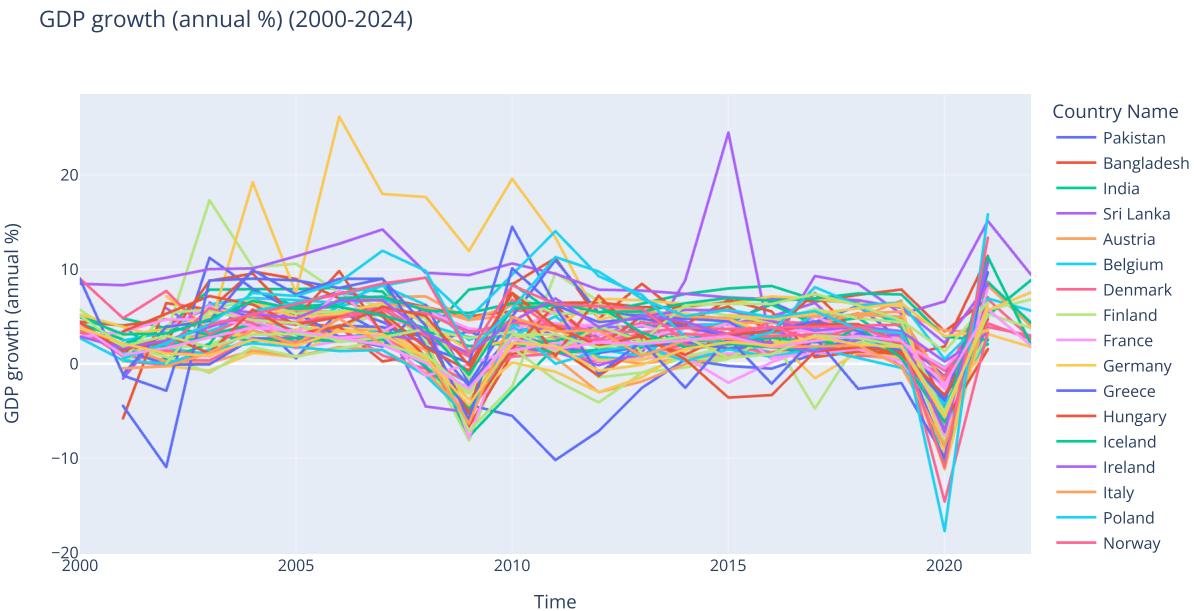
Current account balance (% of GDP) (2000-2024)



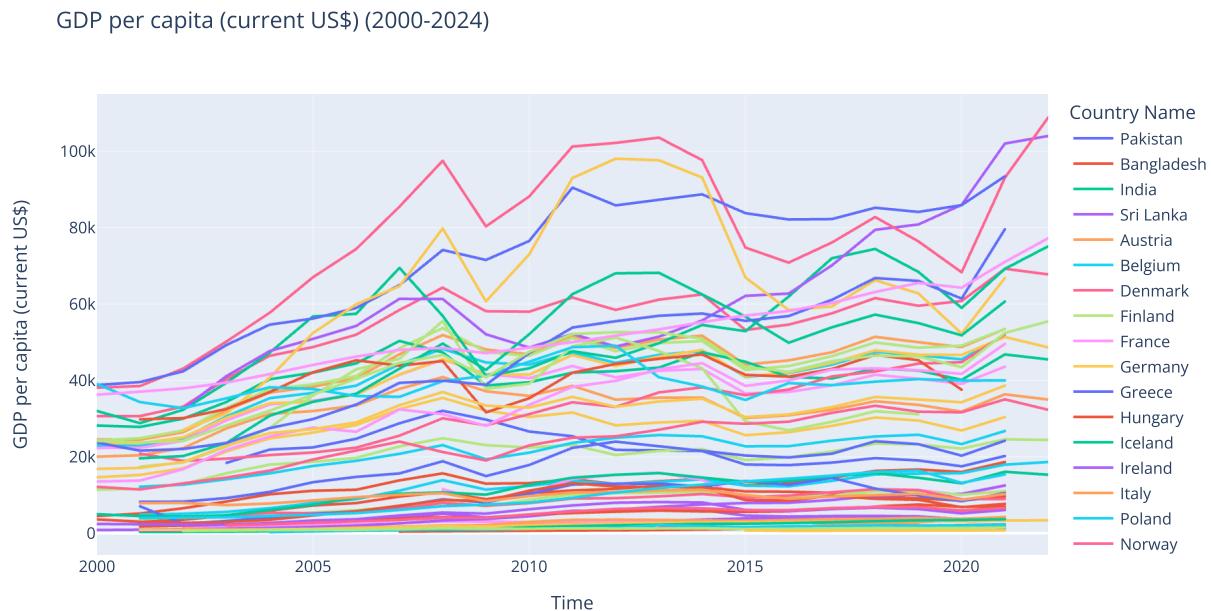
```
In [35]: fig = px.line(df1, x='Time', y='GDP (current US$)', color='Country Name',
                     title="GDP (current US$) (2000-2024)")
fig.show()
```



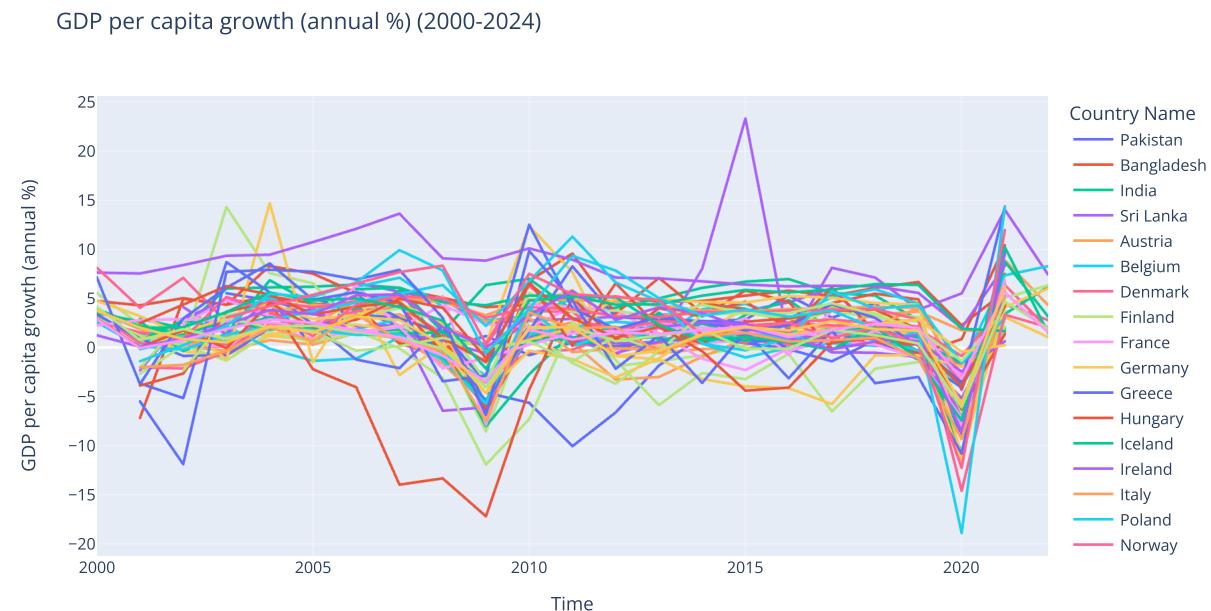
```
In [36]: fig = px.line(df1, x='Time', y='GDP growth (annual %)', color='Country Name',
                     title="GDP growth (annual %) (2000-2024)")
fig.show()
```



```
In [37]: fig = px.line(df1, x='Time', y='GDP per capita (current US$)', color='Country Name',
                     title="GDP per capita (current US$) (2000-2024)")
fig.show()
```

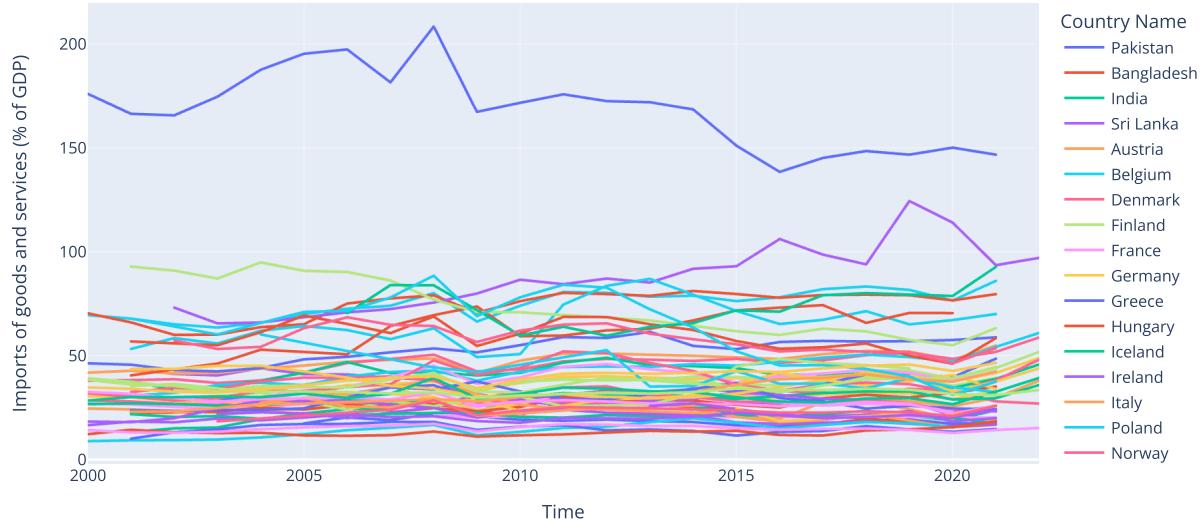


```
In [38]: fig = px.line(df1, x='Time', y='GDP per capita growth (annual %)', color='Country Name',
                     title="GDP per capita growth (annual %) (2000-2024)")
fig.show()
```



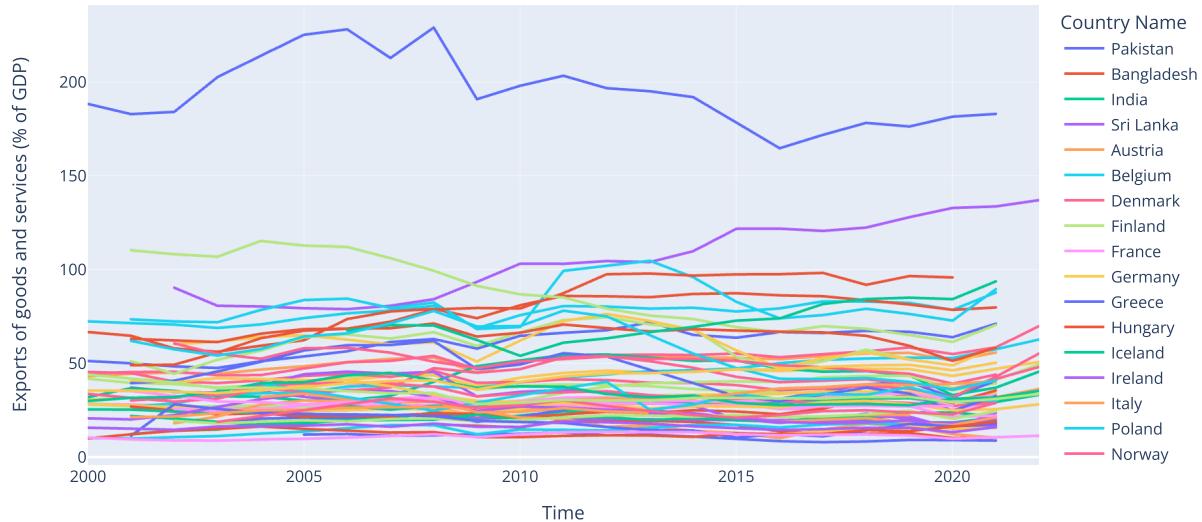
```
In [39]: fig = px.line(df1, x='Time', y='Imports of goods and services (% of GDP)', color='Country Name',
                     title=f"Imports of goods and services (% of GDP) (2000-2024)")
fig.show()
```

Imports of goods and services (% of GDP) (2000-2024)

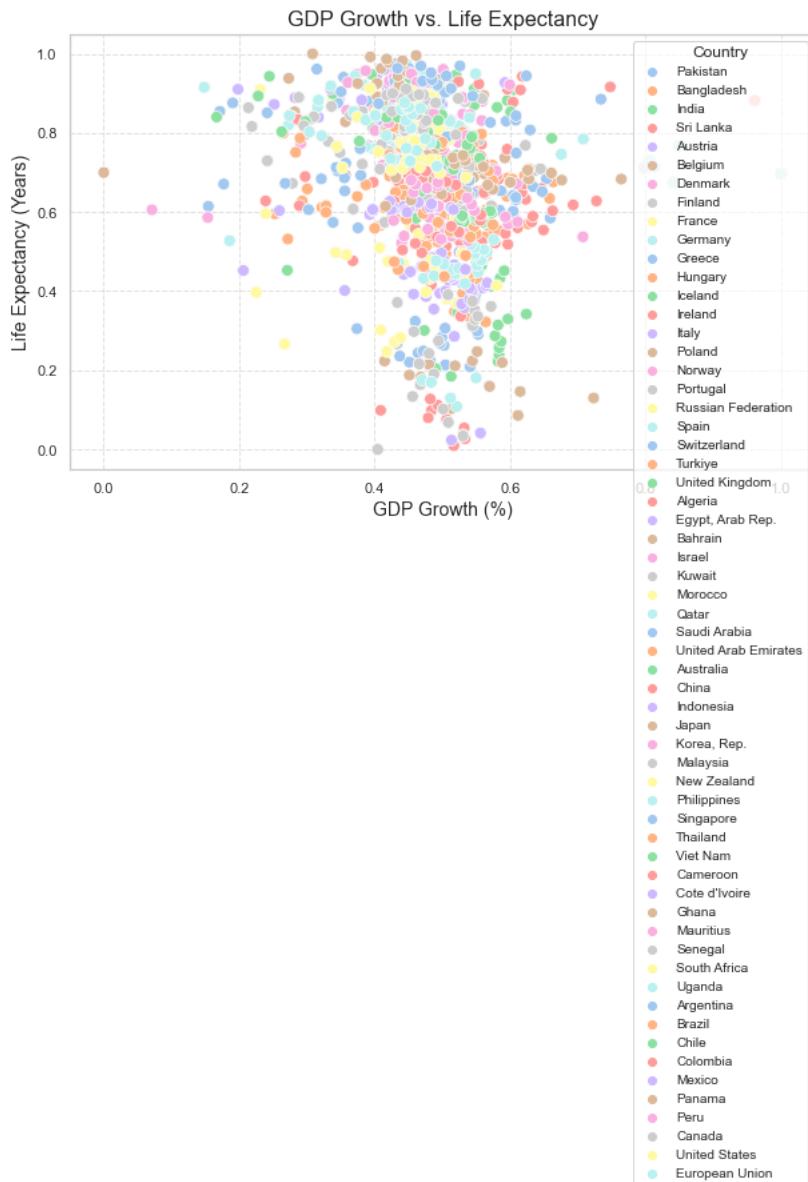


```
In [40]: fig = px.line(df1, x='Time', y='Exports of goods and services (% of GDP)', color='Country Name',
                     title=f"Exports of goods and services (% of GDP) (2000-2024)")
fig.show()
```

Exports of goods and services (% of GDP) (2000-2024)



```
In [44]: plt.figure(figsize=(10, 6))
sns.scatterplot(data=df, x='GDP growth (annual %)', y='Life expectancy at birth, total (years)', hue='Country Name', palette='pastel')
plt.title('GDP Growth vs. Life Expectancy', fontsize=16)
plt.xlabel('GDP Growth (%)', fontsize=14)
plt.ylabel('Life Expectancy (Years)', fontsize=14)
plt.legend(title='Country', fontsize=10)
plt.grid(axis='both', linestyle='--', alpha=0.5)
plt.show()
```



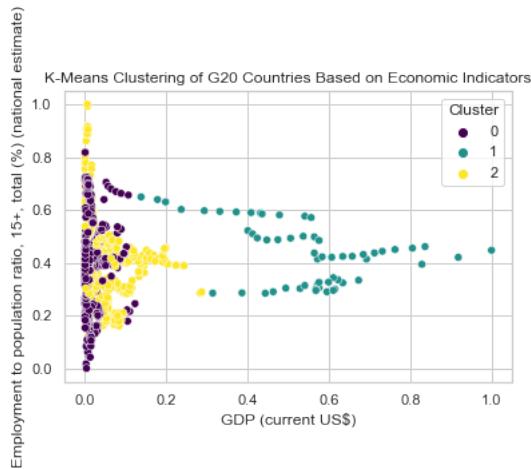
## Multivariate Analysis

```
In [69]: from sklearn.preprocessing import StandardScaler
from sklearn.cluster import KMeans

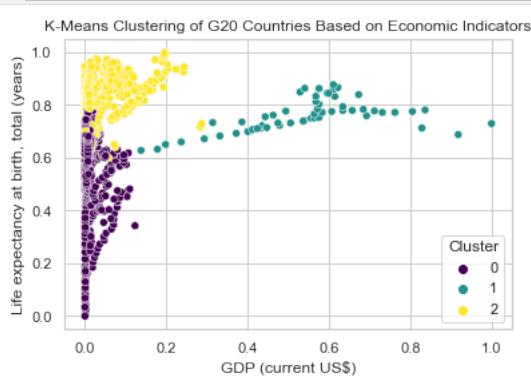
df_numeric = df.select_dtypes(["float64"])
kmeans = KMeans(n_clusters=3, random_state=42)
clusters = kmeans.fit_predict(df_numeric)

df['Cluster'] = clusters

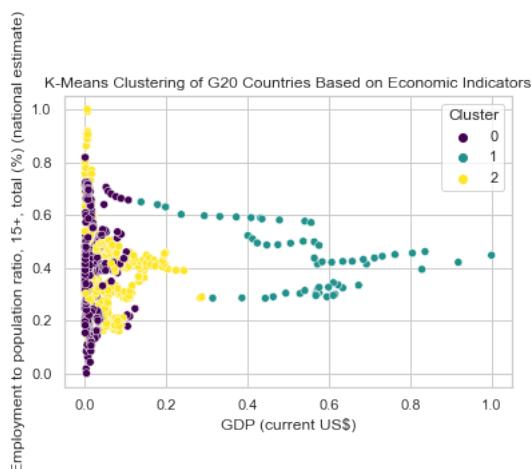
sns.scatterplot(x=df['GDP (current US$)'], y=df['Employment to population ratio, 15+, total (%) (national estimate)'], hue=df['Cluster'])
plt.title('K-Means Clustering of G20 Countries Based on Economic Indicators')
plt.show()
```



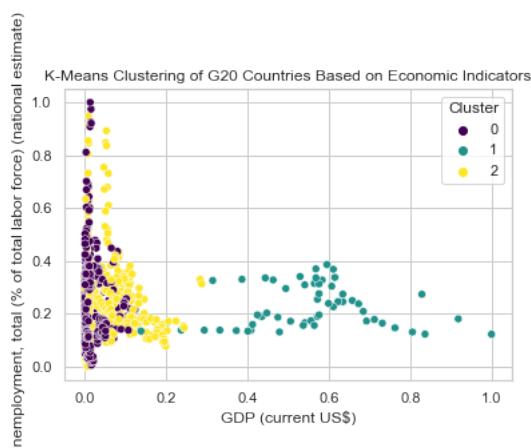
```
In [70]: sns.scatterplot(x=df['GDP (current US$)'], y=df['Life expectancy at birth, total (years)'], hue=df['Cluster'], palette="viridis")
plt.title('K-Means Clustering of G20 Countries Based on Economic Indicators')
plt.show()
```



```
In [71]: sns.scatterplot(x=df['GDP (current US$)'], y=df['Employment to population ratio, 15+, total (%) (national estimate)'], hue=df['Cluster'])
plt.title('K-Means Clustering of G20 Countries Based on Economic Indicators')
plt.show()
```



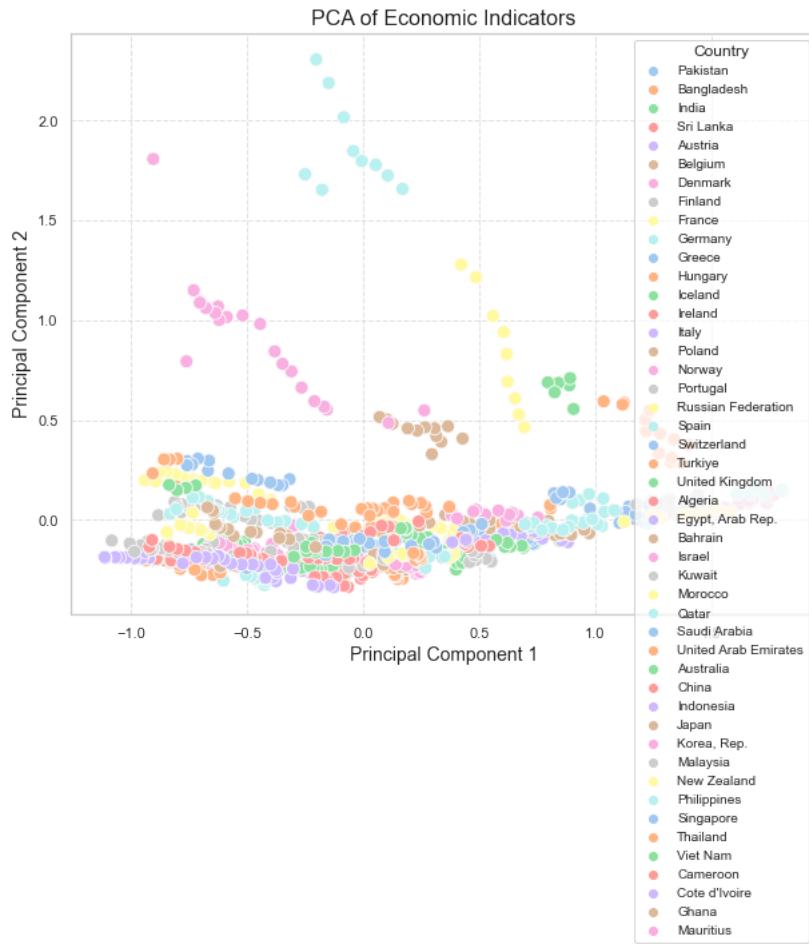
```
In [72]: sns.scatterplot(x=df['GDP (current US$)'], y=df['Unemployment, total (% of total labor force) (national estimate)'], hue=df['Cluster'])
plt.title('K-Means Clustering of G20 Countries Based on Economic Indicators')
plt.show()
```



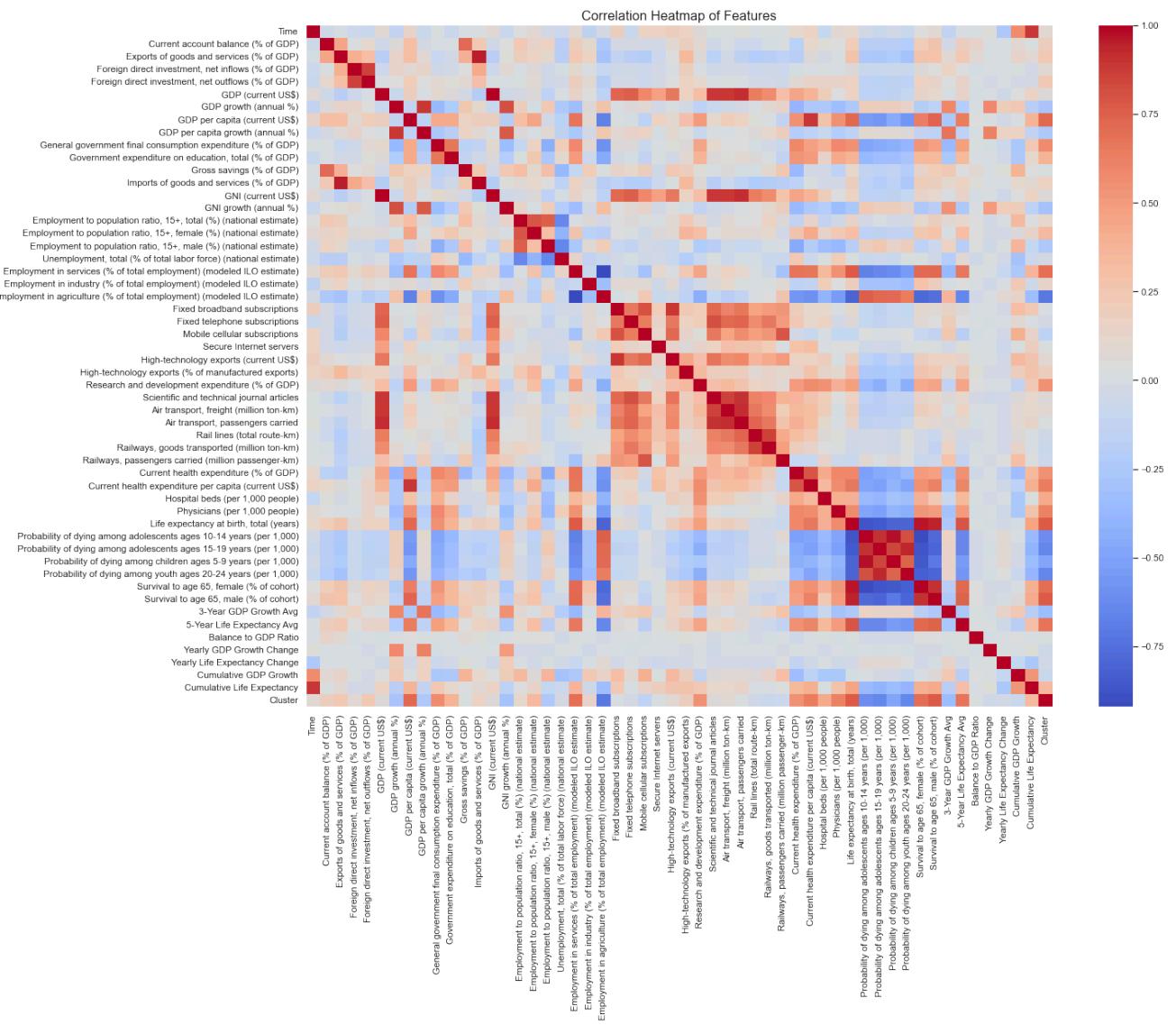
```
In [59]: from sklearn.decomposition import PCA

pca = PCA(n_components=2)
pca_result = pca.fit_transform(df[scaled_columns])
pca_df = pd.DataFrame(pca_result, columns=['PC1', 'PC2'])
pca_df['Country'] = df['Country Name']

plt.figure(figsize=(10, 8))
sns.scatterplot(data=pca_df, x='PC1', y='PC2', hue='Country', palette='pastel', s=100)
plt.title('PCA of Economic Indicators', fontsize=16)
plt.xlabel('Principal Component 1', fontsize=14)
plt.ylabel('Principal Component 2', fontsize=14)
plt.legend(title='Country', fontsize=10)
plt.grid(axis='both', linestyle='--', alpha=0.5)
plt.show()
```



```
In [65]: plt.figure(figsize=(20, 15))
correlation_matrix = df.corr()
sns.heatmap(correlation_matrix, cmap='coolwarm')
plt.title('Correlation Heatmap of Features', fontsize=16)
plt.show()
```



## Case Study: 2008 Global Economic Crisis

The 2007–2008 financial crisis, or the global financial crisis (GFC), was the most severe worldwide economic crisis since the 1929 Wall Street crash that began the Great Depression. Causes of the crisis included predatory lending in the form of subprime mortgages to low-income homebuyers and a resulting housing bubble, excessive risk-taking by global financial institutions, and lack of regulatory oversight, which culminated in a "perfect storm" that triggered the Great Recession, which lasted from late 2007 to mid-2009. The financial crisis began in early 2007, as mortgage-backed securities (MBS) tied to U.S. real estate, as well as a vast web of derivatives linked to those MBS, collapsed in value. Financial institutions worldwide suffered severe damage, reaching a climax with the bankruptcy of Lehman Brothers on September 15, 2008, and a subsequent international banking crisis.

The prerequisites for the crisis were complex. During the 1990s, the U.S. Congress had passed legislation intended to expand affordable housing through looser financing. In 1999, parts of the Glass–Steagall legislation (passed in 1933) were repealed, permitting institutions to mix low-risk operations, such as commercial banking and insurance, with higher-risk operations such as investment banking and proprietary trading. As the Federal Reserve ("Fed") lowered the federal funds rate from 2000 to 2003, institutions increasingly targeted low-income homebuyers, largely belonging to racial minorities, with high-risk loans; this development went unattended by regulators. As interest rates rose from 2004 to 2006, the cost of mortgages rose and the demand for housing fell, causing property values to decline. In early 2007, as more U.S. mortgage holders began defaulting on their repayments, subprime lenders went bankrupt, culminating in April with the bankruptcy of New Century Financial. As demand and prices continued to fall, the contagion spread to worldwide credit markets by August, and central banks began injecting liquidity. By July 2008, Fannie Mae and Freddie Mac, companies which together owned or guaranteed half of the U.S. housing market, were on the verge of collapse; the Housing and Economic Recovery Act enabled the government to take over and cover their combined \$1.6 trillion debt on September 7.

For this analysis we will focus only on the G20 nations

```
In [87]: g20 = ['Argentina', 'Australia', 'Brazil', 'Canada', 'China', 'France', 'Germany', 'India', 'Indonesia', 'Italy', 'Japan', 'Mexico'
for i in g20:
    if i not in df["Country Name"].unique():
        print(i)

years = [2007, 2008, 2009]

df_economic_crisis = df1[(df1["Country Name"].isin(g20)) & (df1["Time"].isin(years))]
df_economic_crisis.head()
```

Out[87]:

	Country Name	Country Code	Time	Current account balance (% of GDP)	Exports of goods and services (% of GDP)	Foreign direct investment, net inflows (% of GDP)	Foreign direct investment, net outflows (% of GDP)	GDP (current US\$)	GDP growth (annual %)	GDP per capita (current US\$)	...	Current health expenditure per capita (current US\$)	Hospital beds (per 1,000 people)	Physicians (per 1,000 people)	expe
53	India	IND	2007	-0.663718	20.799700	2.073394	1.399326	1.220000e+12	7.660815	1022.732467	...	36.000000	1.90	2.289	65.
54	India	IND	2008	-2.583377	24.097357	3.620523	1.606189	1.200000e+12	3.086698	993.503405	...	38.000000	1.87	2.289	66.
55	India	IND	2009	-1.951462	20.400519	2.651590	1.199473	1.340000e+12	7.861889	1096.636136	...	38.000000	1.84	2.289	66.
191	France	FRA	2007	-0.327691	27.853258	3.148834	4.969835	2.660000e+12	2.424736	41557.623560	...	4312.310547	7.35	3.449	81.
192	France	FRA	2008	-0.961095	28.119385	2.320491	4.641131	2.930000e+12	0.254946	45515.961750	...	4807.374512	7.18	3.443	81.

5 rows × 48 columns

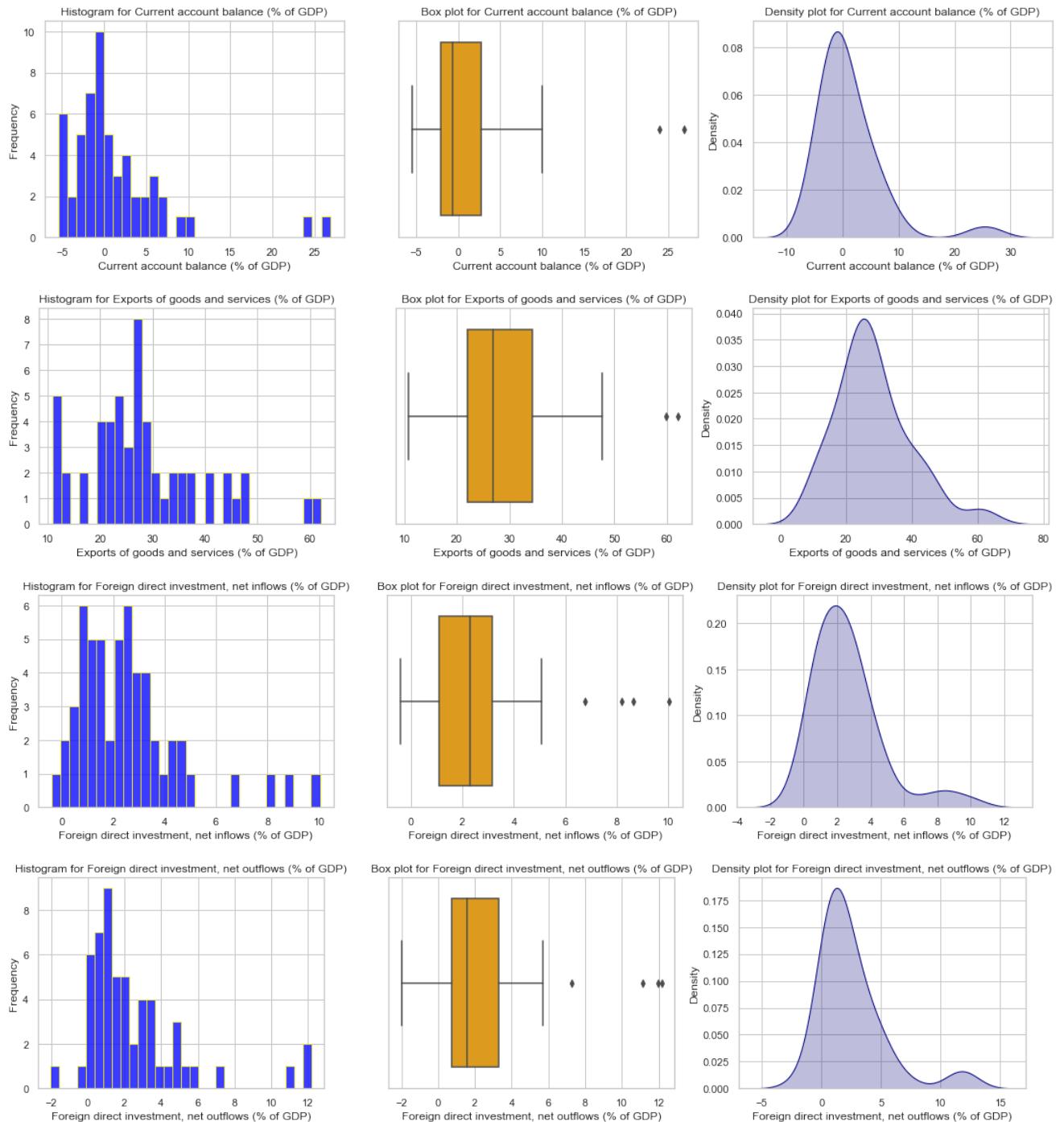
```
In [88]: for i in economic_metrics:
    plt.figure(figsize=(15,4))

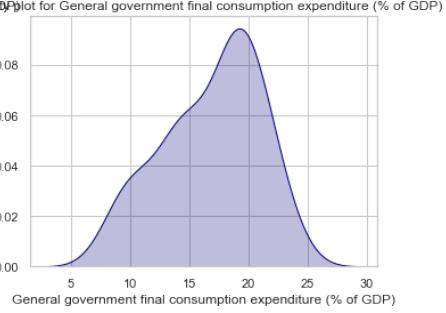
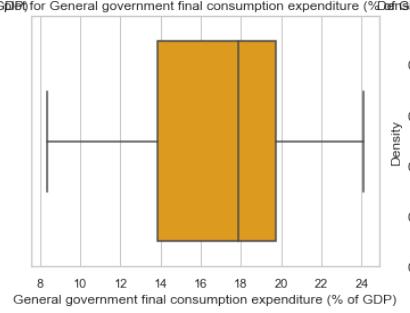
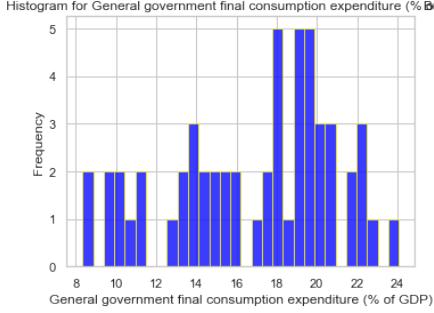
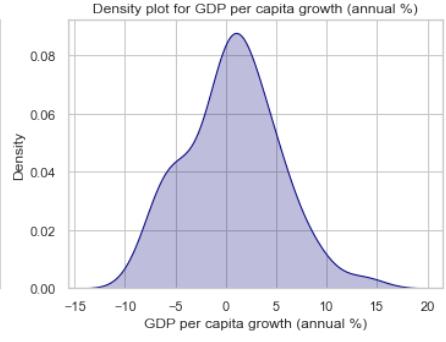
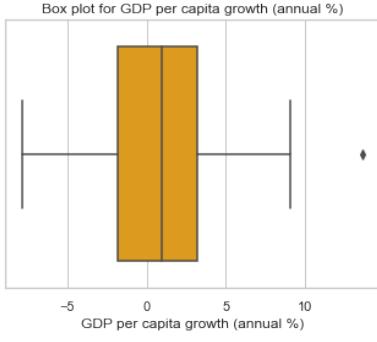
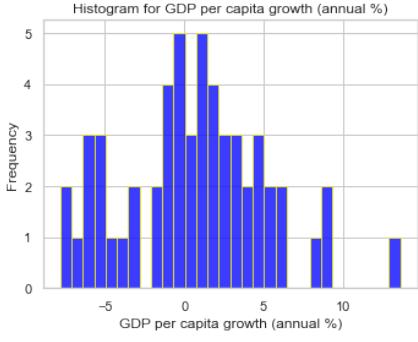
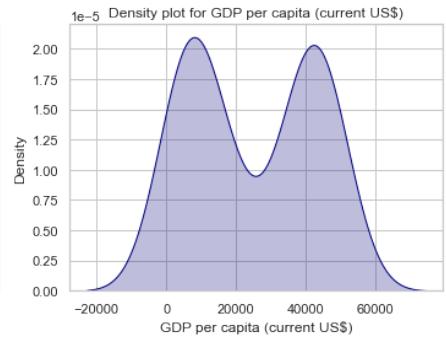
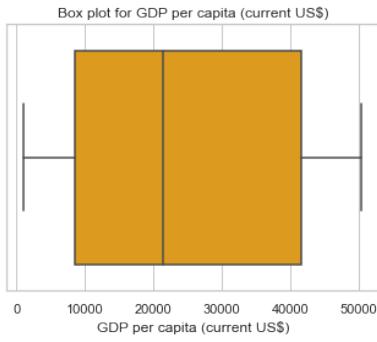
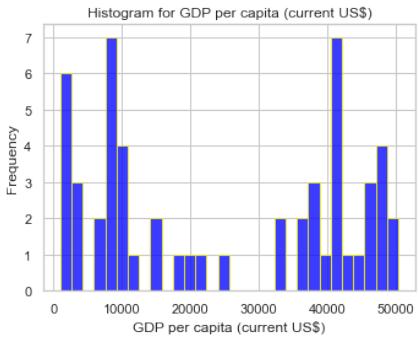
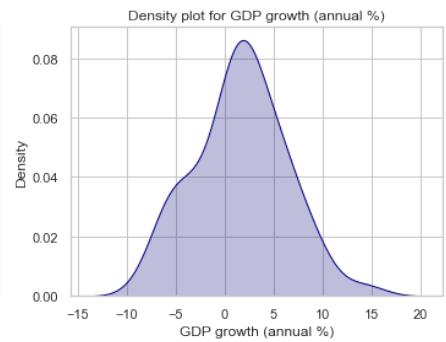
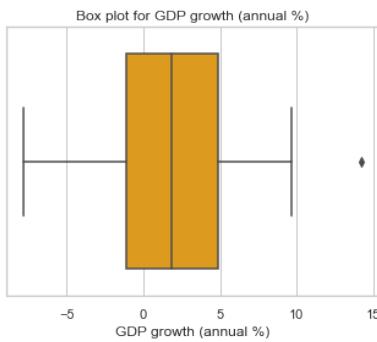
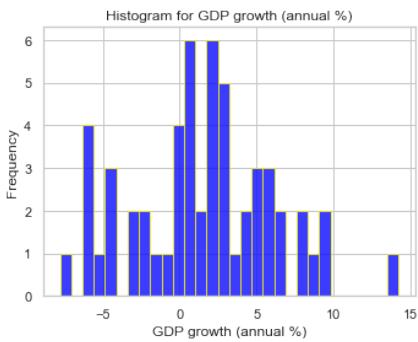
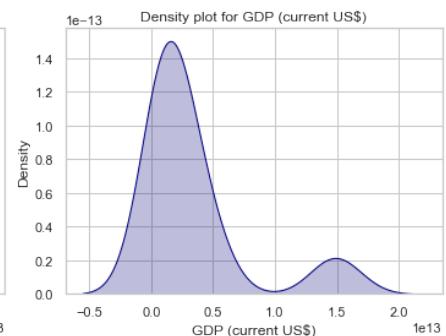
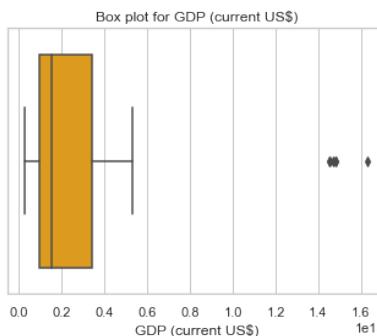
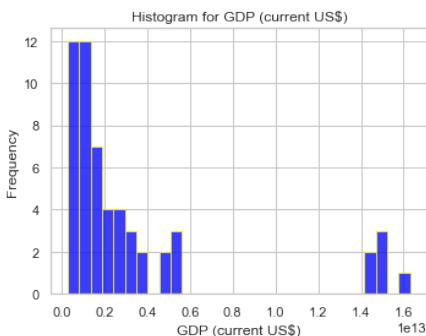
    plt.subplot(1,3,1)
    sns.histplot(data=df_economic_crisis[i], color="blue", edgecolor="yellow", bins=30)
    plt.xlabel(i)
    plt.ylabel("Frequency")
    plt.title(f"Histogram for {i}")

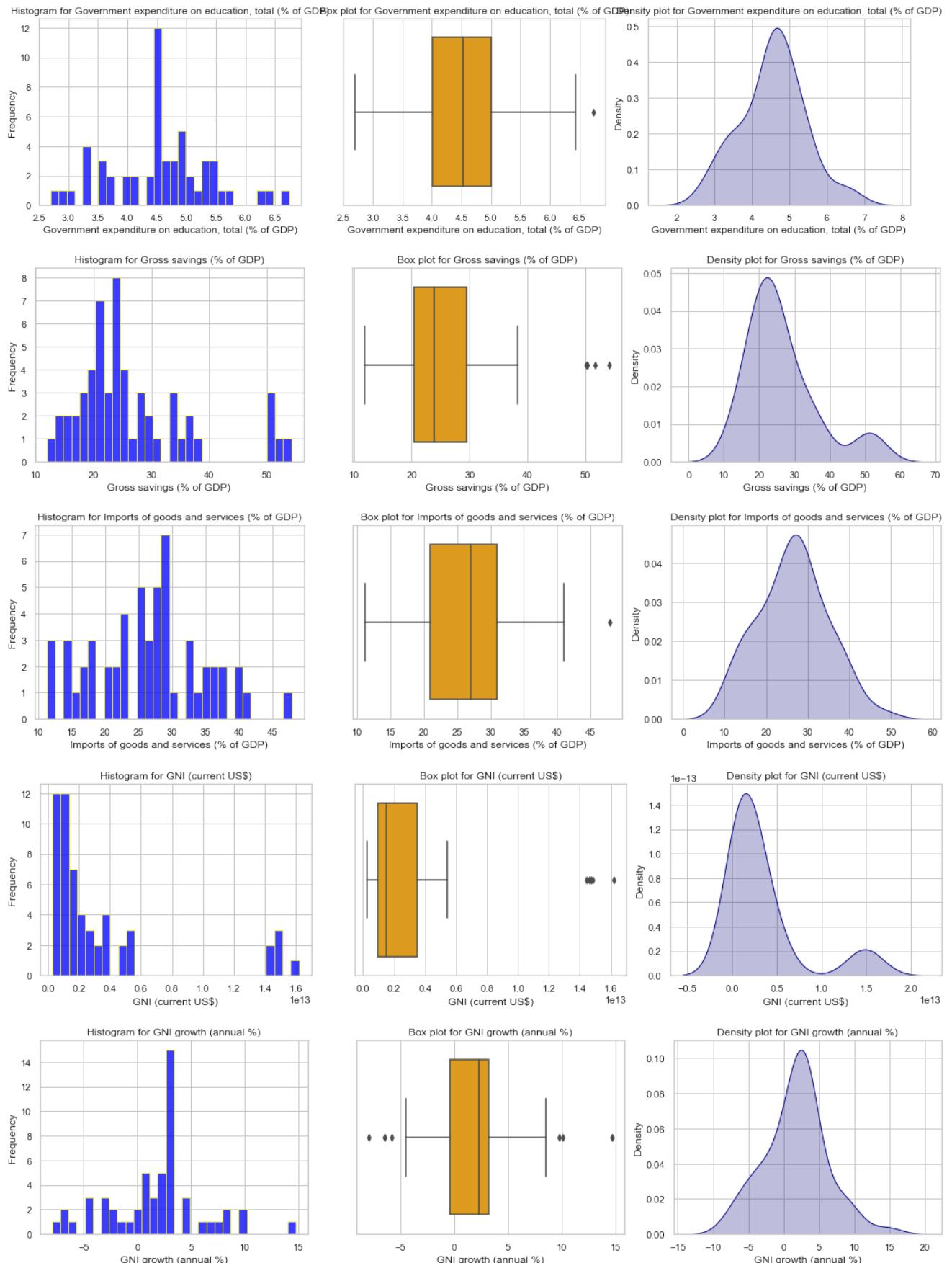
    plt.subplot(1,3,2)
    sns.boxplot(x=df_economic_crisis[i], color="orange")
    plt.xlabel(i)
    plt.title(f"Box plot for {i}")

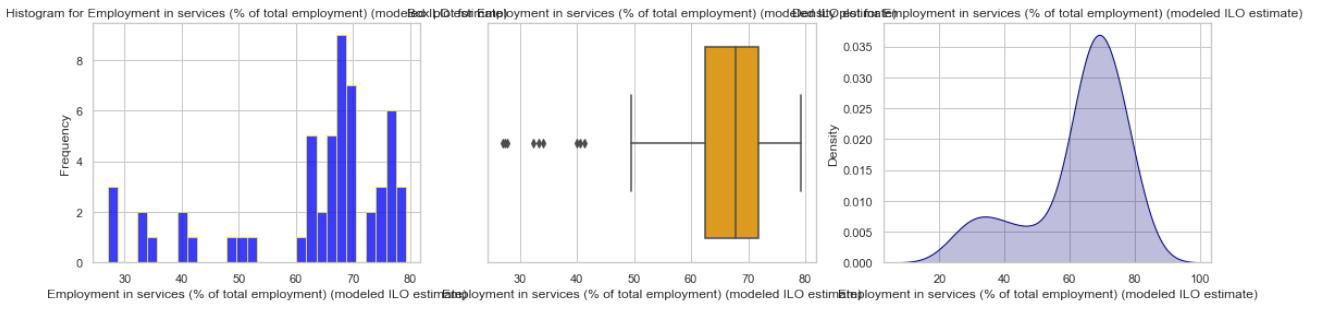
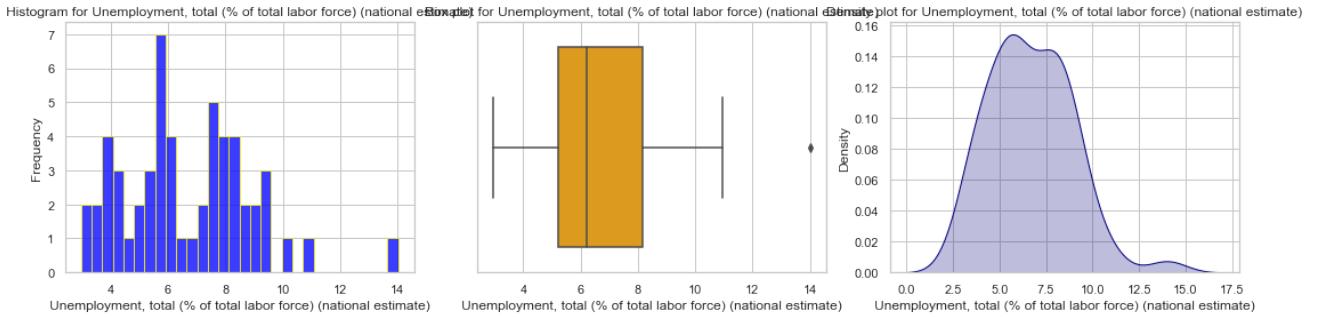
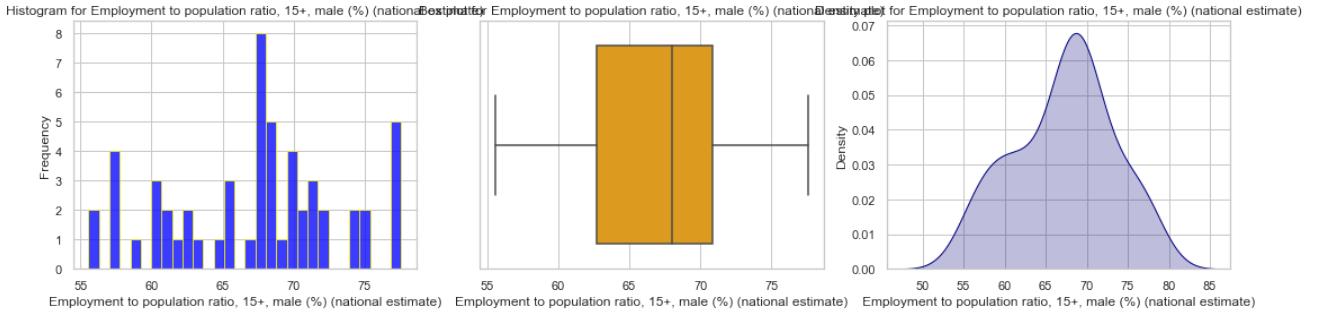
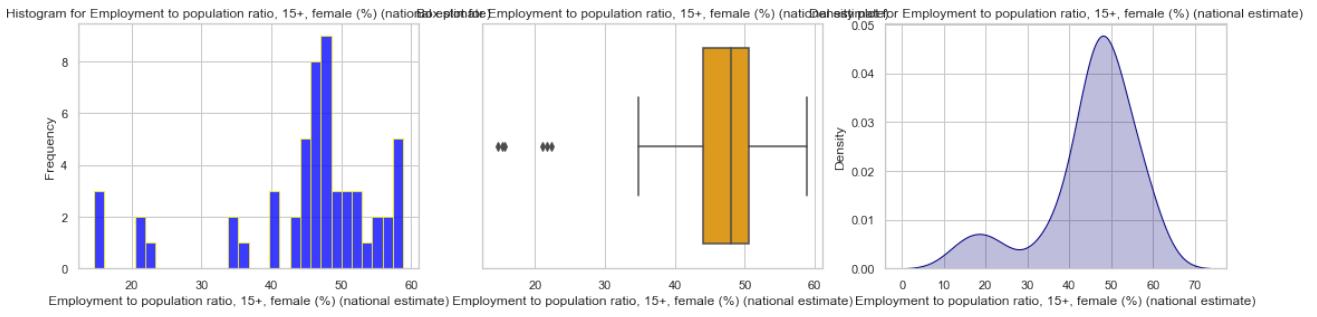
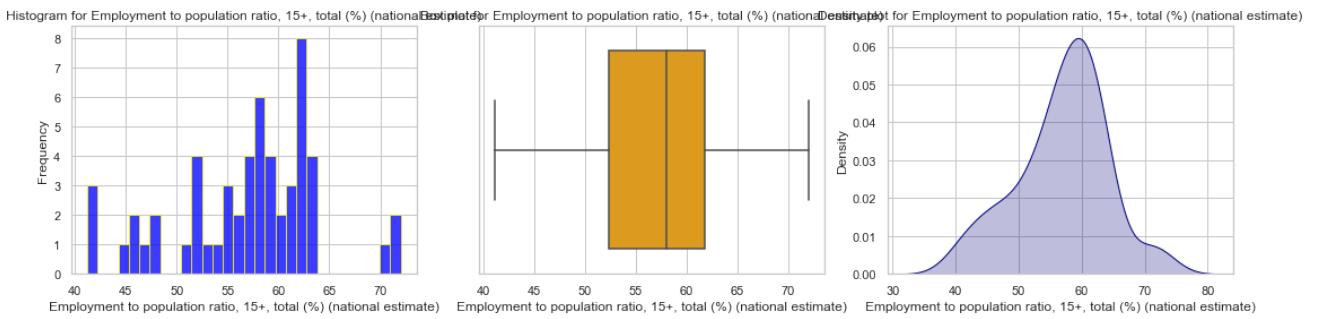
    plt.subplot(1,3,3)
    sns.kdeplot(data=df_economic_crisis[i], color="navy", fill=True)
    plt.xlabel(i)
    plt.ylabel("Density")
    plt.title(f"Density plot for {i}")

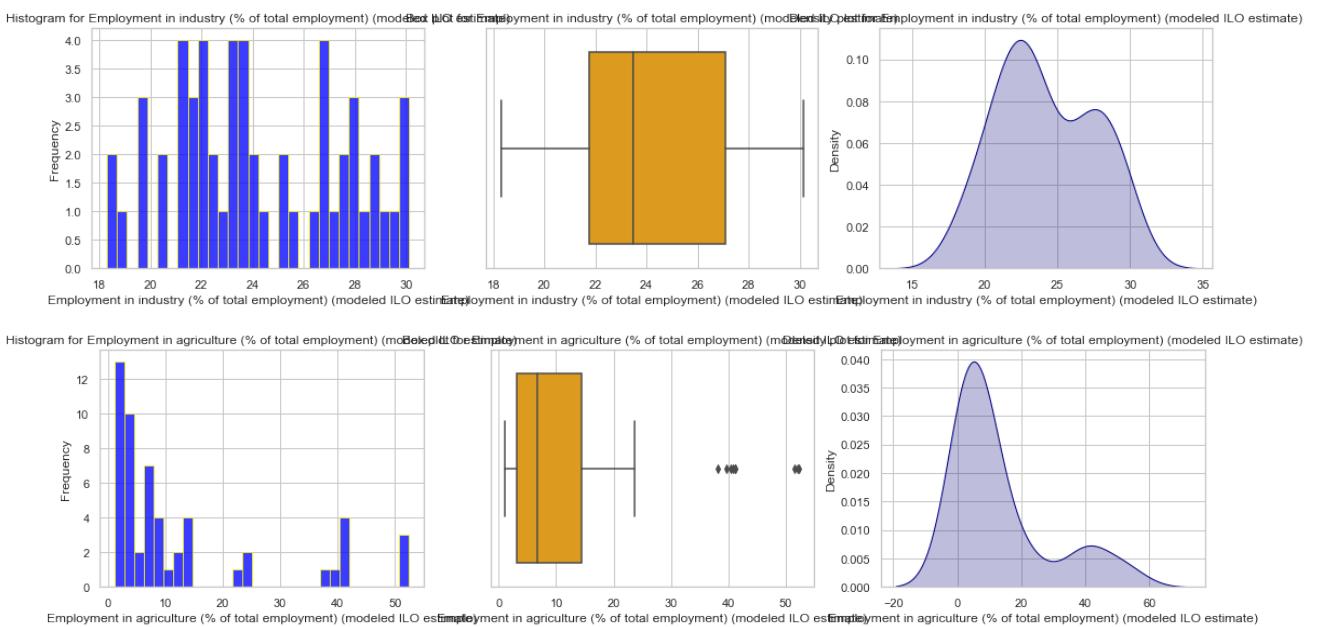
    plt.tight_layout()
    plt.show()
```





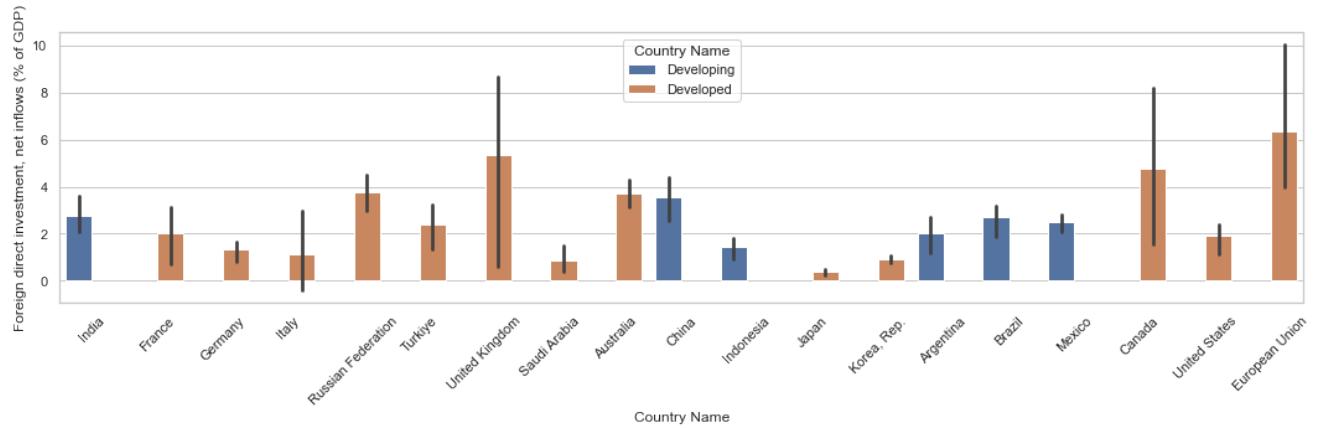
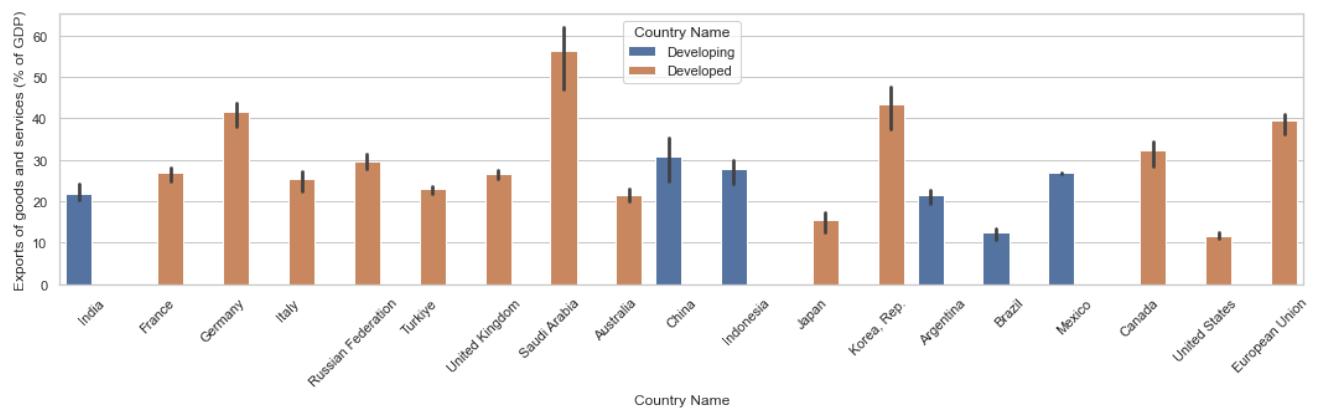
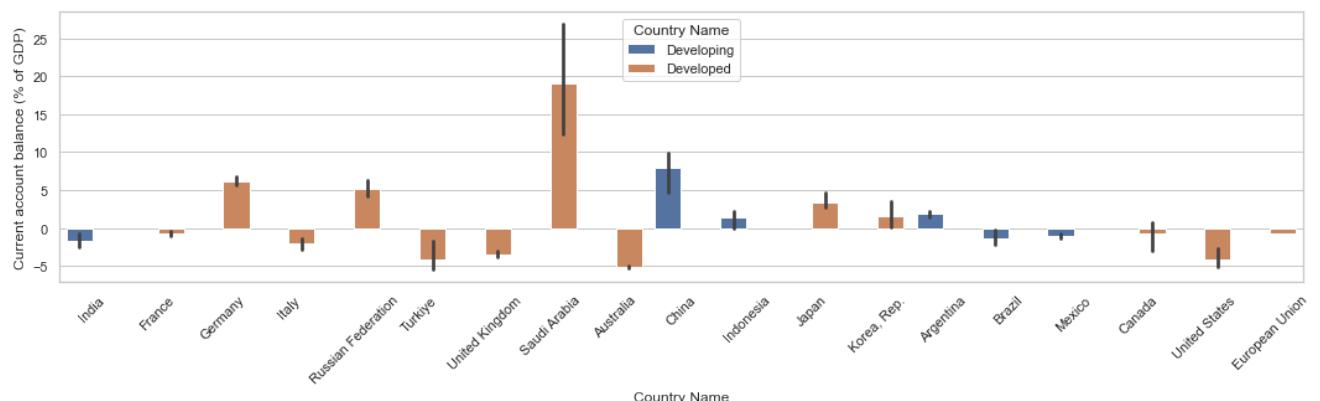


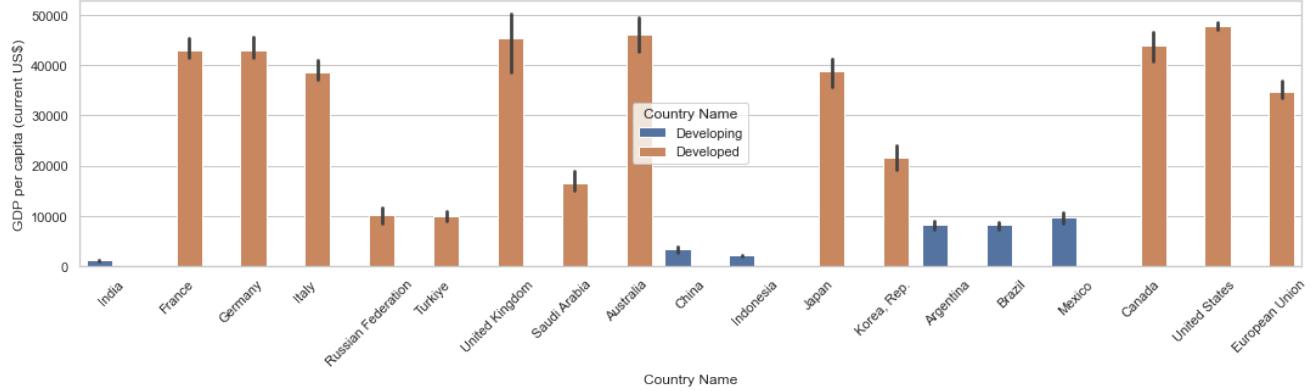
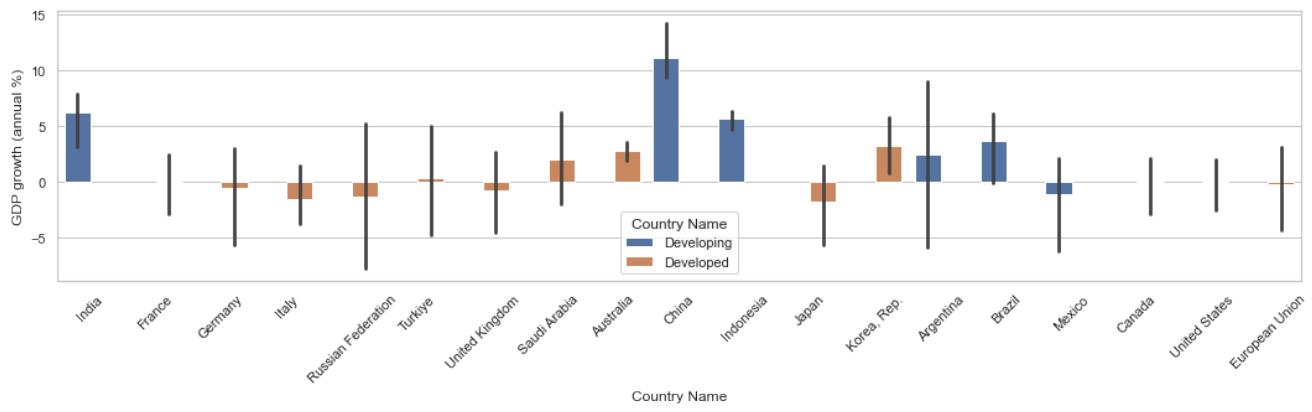
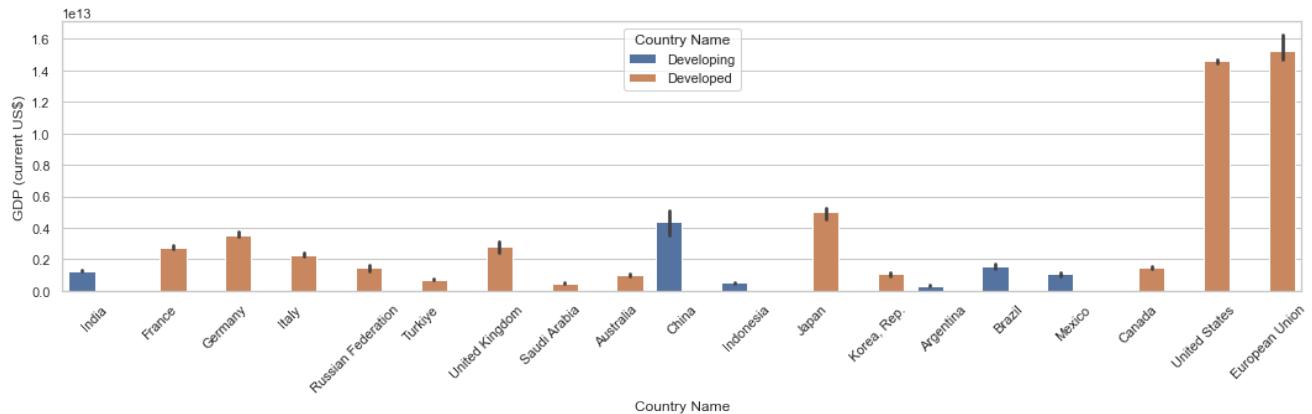
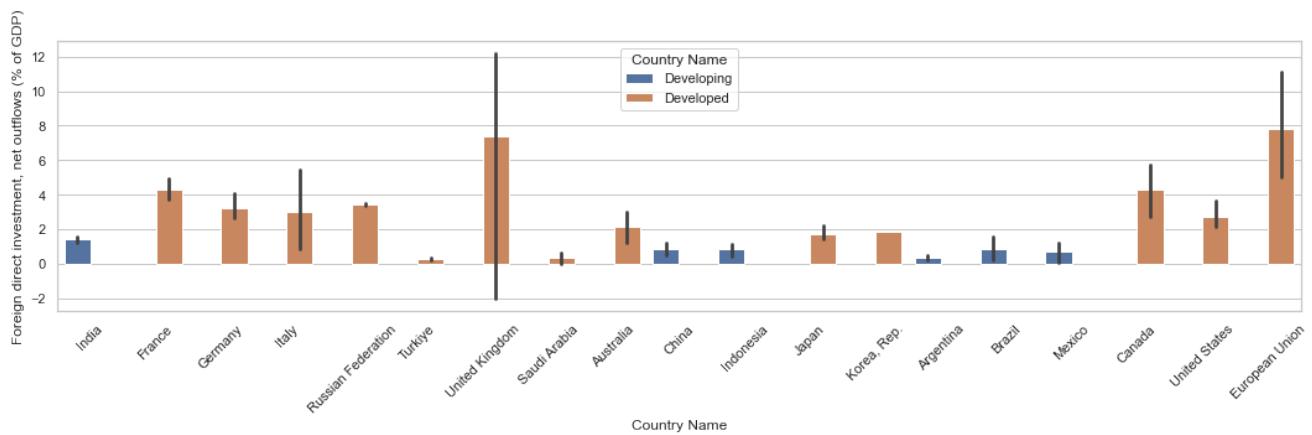


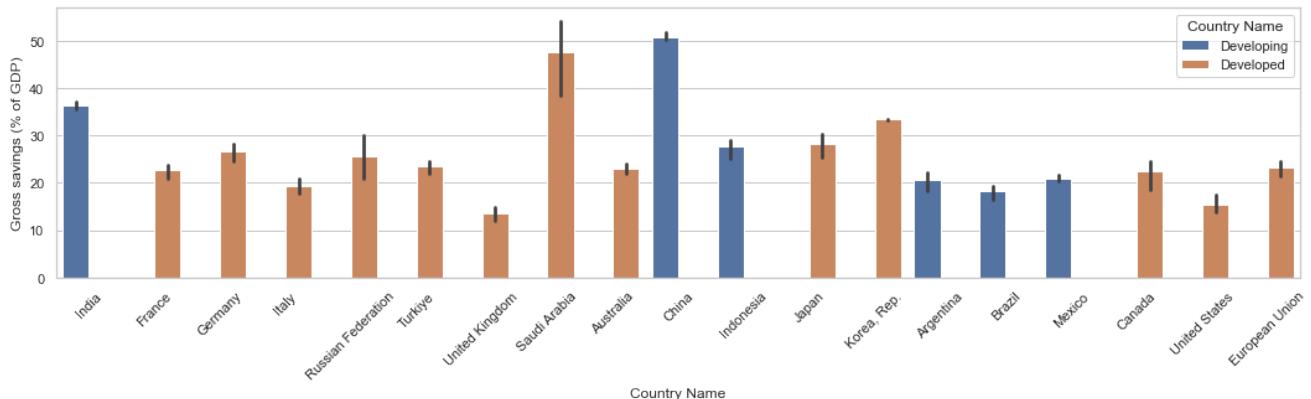
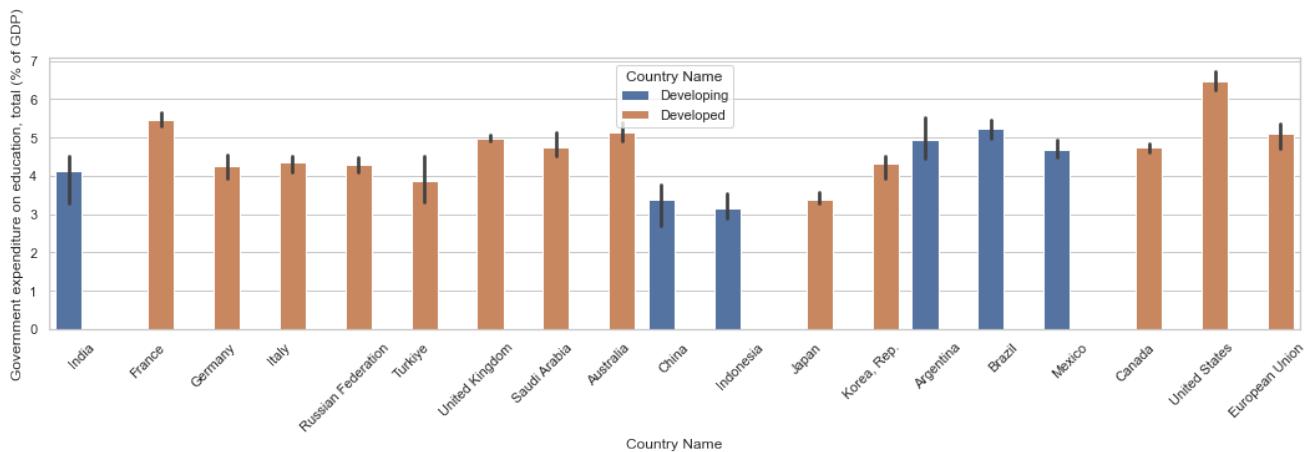
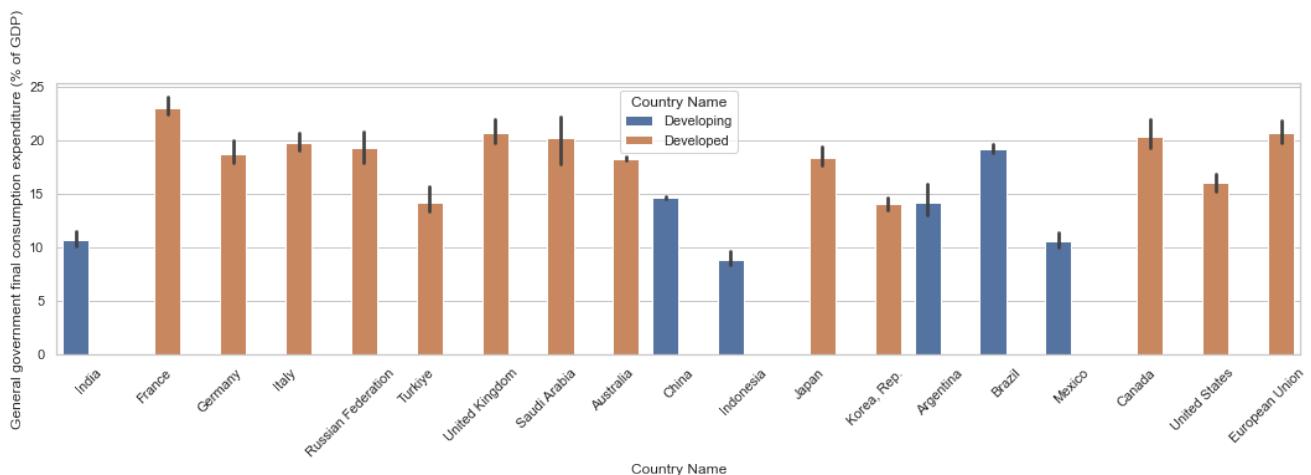
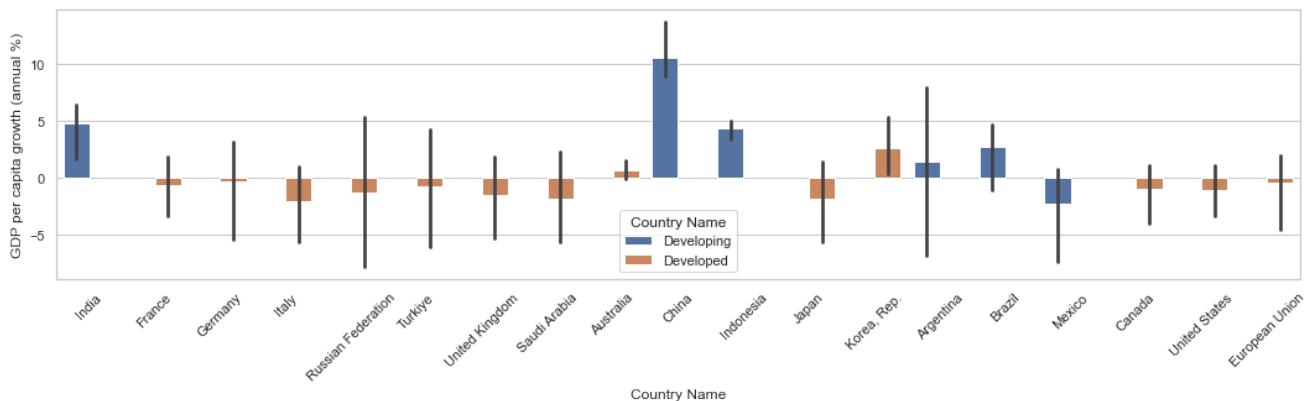


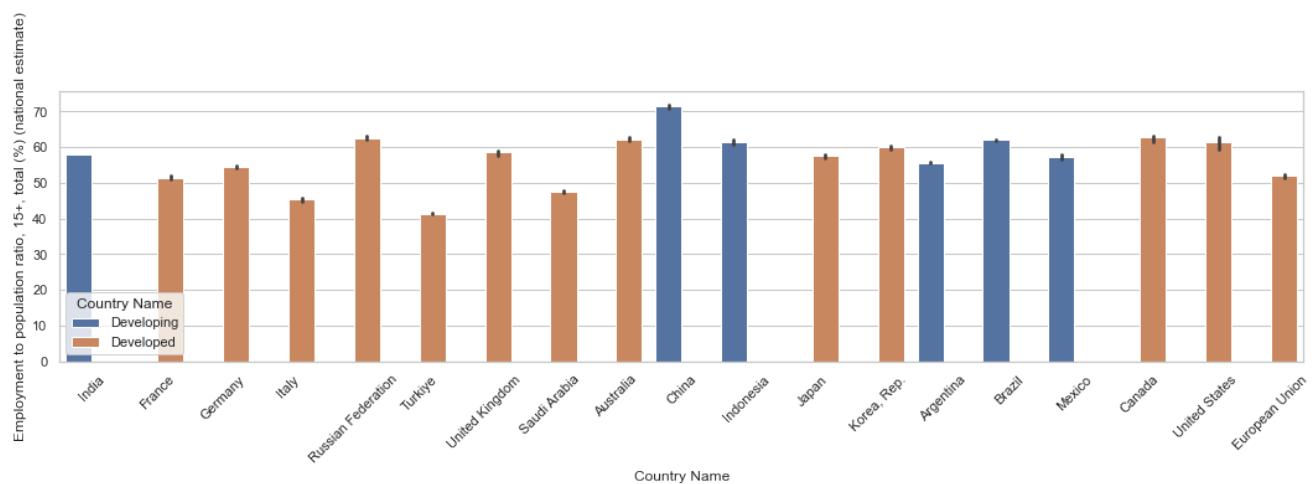
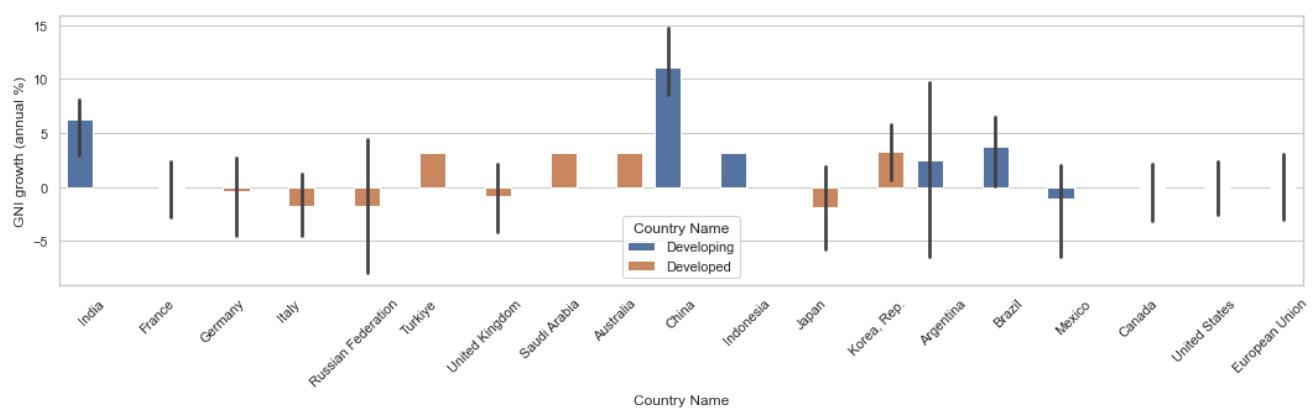
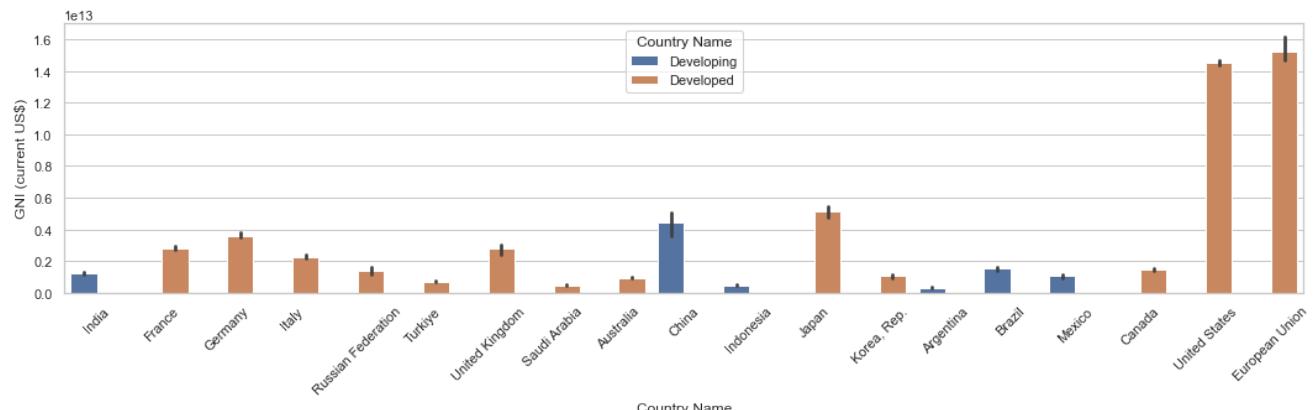
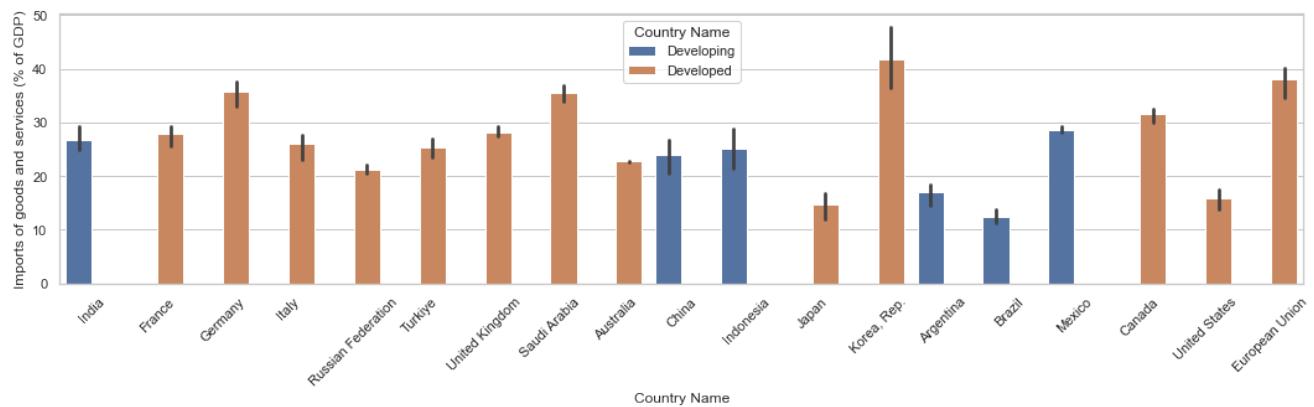
```
In [89]: developed_countries = ['United States', 'United Kingdom', 'Germany', 'Japan', 'Canada', 'Australia', 'France', 'Italy', 'Russia']
developing_countries = ['India', 'Brazil', 'China', 'Mexico', 'South Africa', "Indonesia"]

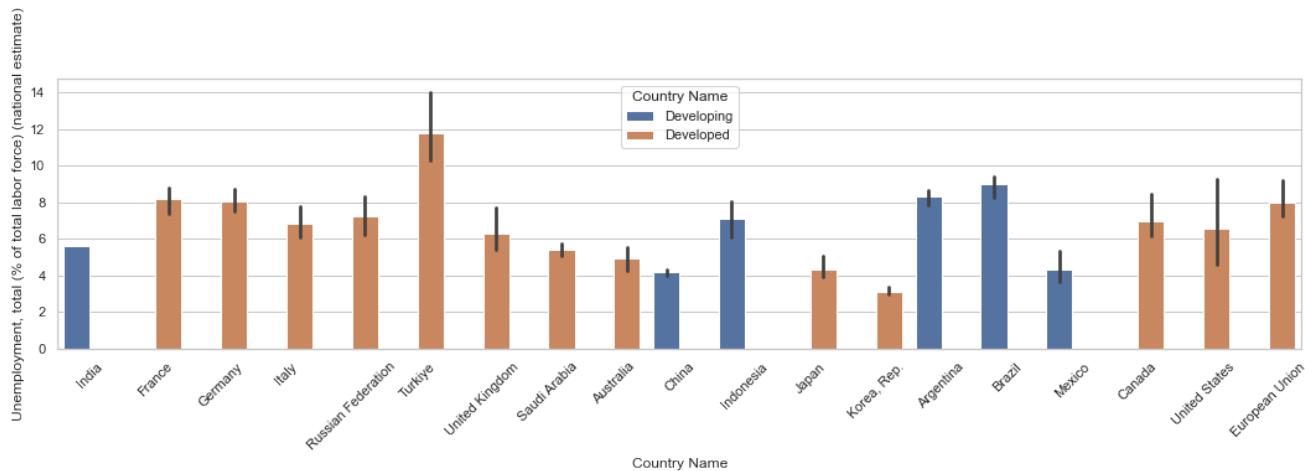
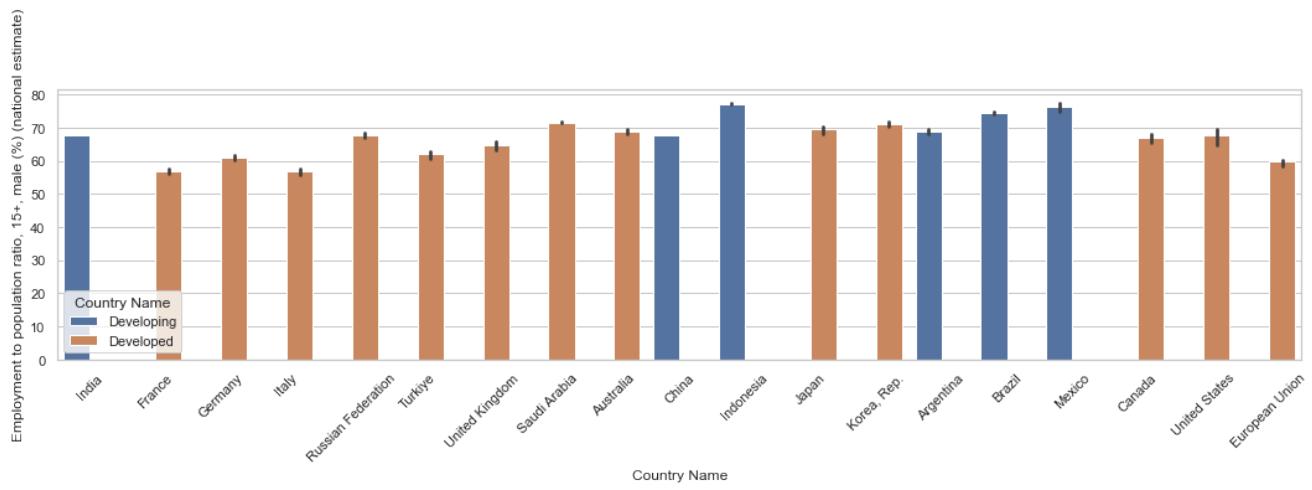
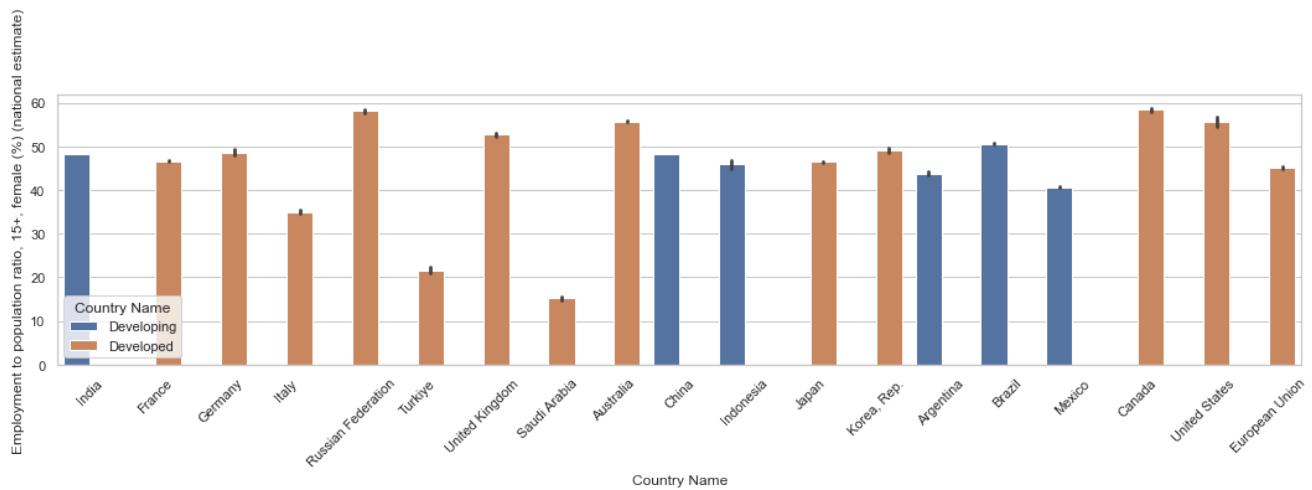
for i in economic_metrics:
    plt.figure(figsize=(18,4))
    sns.barplot(data=df_economic_crisis, x="Country Name", y=i, hue=df_economic_crisis['Country Name'].apply(lambda x: 'Developed' if x in developed_countries else 'Developing'))
    plt.xticks(rotation=45)
    plt.show()
```

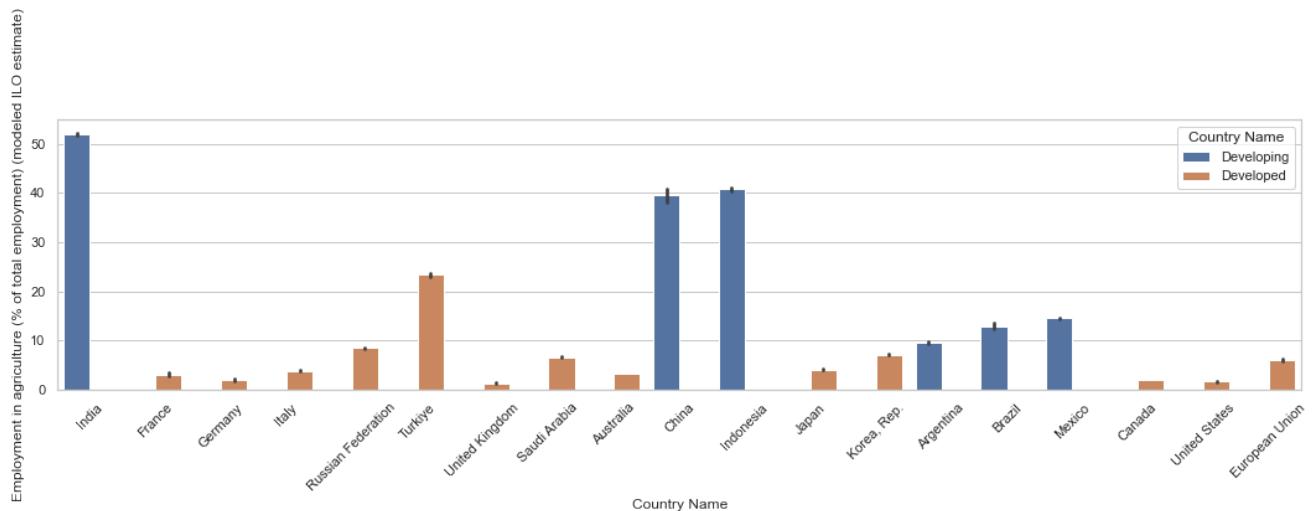
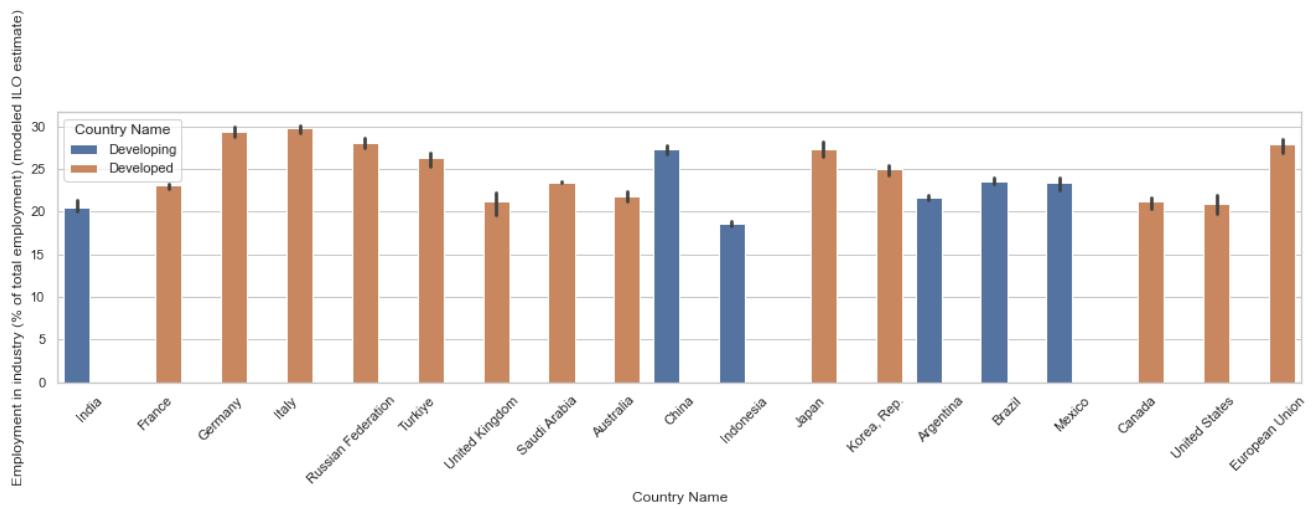
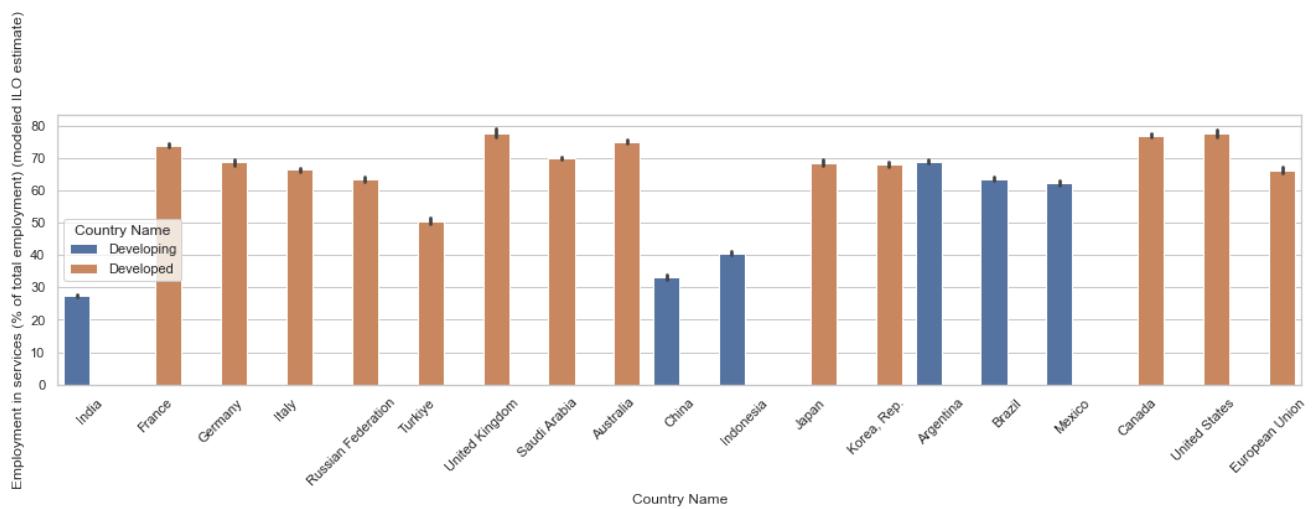






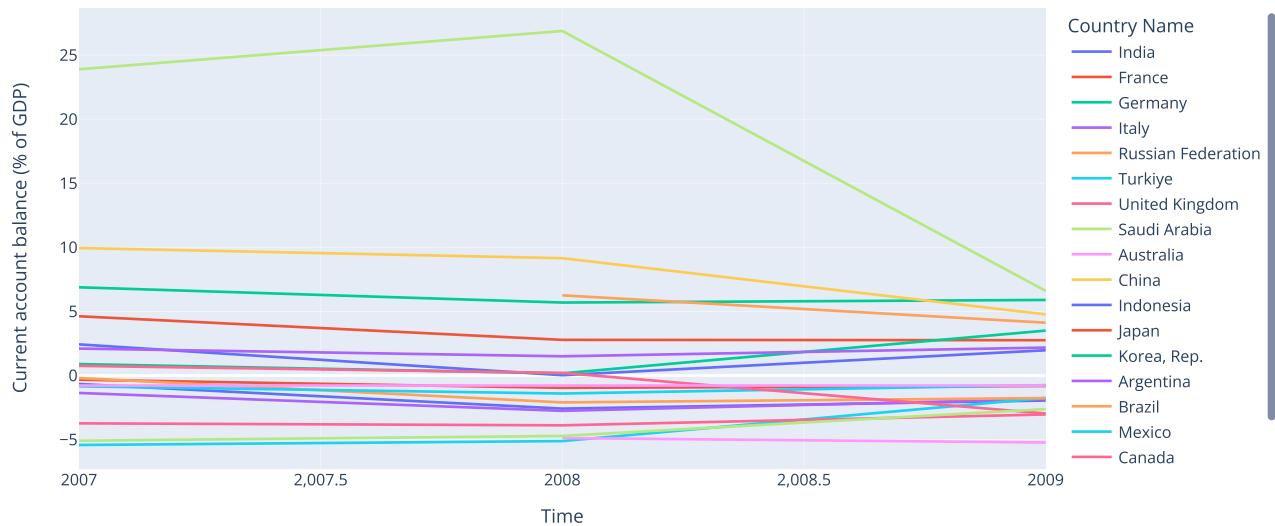






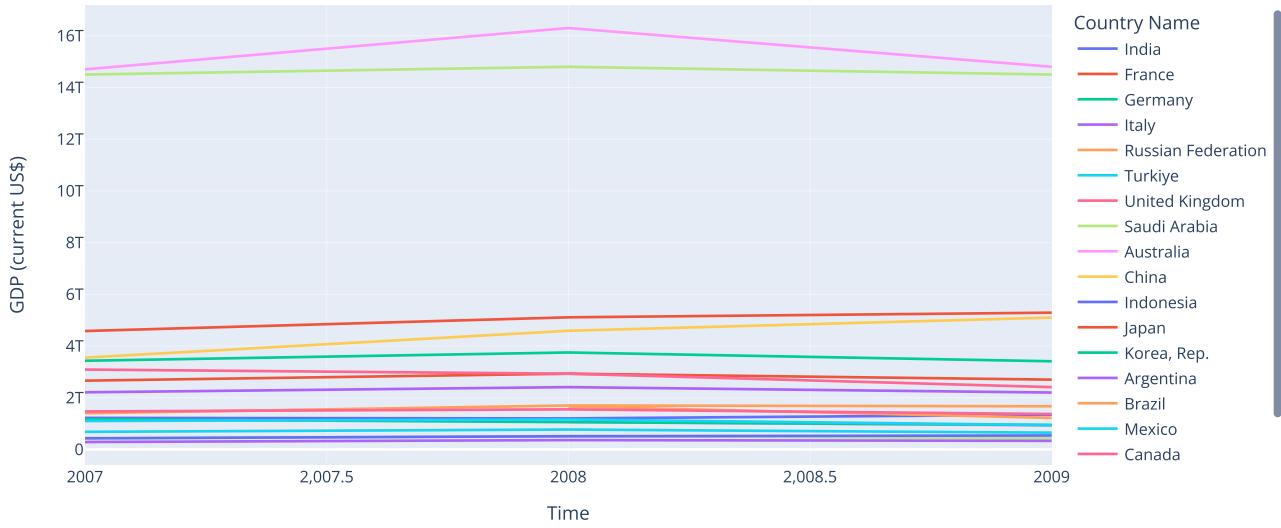
```
In [95]: fig = px.line(df_economic_crisis, x='Time', y='Current account balance (% of GDP)', color='Country Name',
                     title=f"Current account balance (% of GDP) (2007-2009)")
fig.show()
```

Current account balance (% of GDP) (2007-2009)



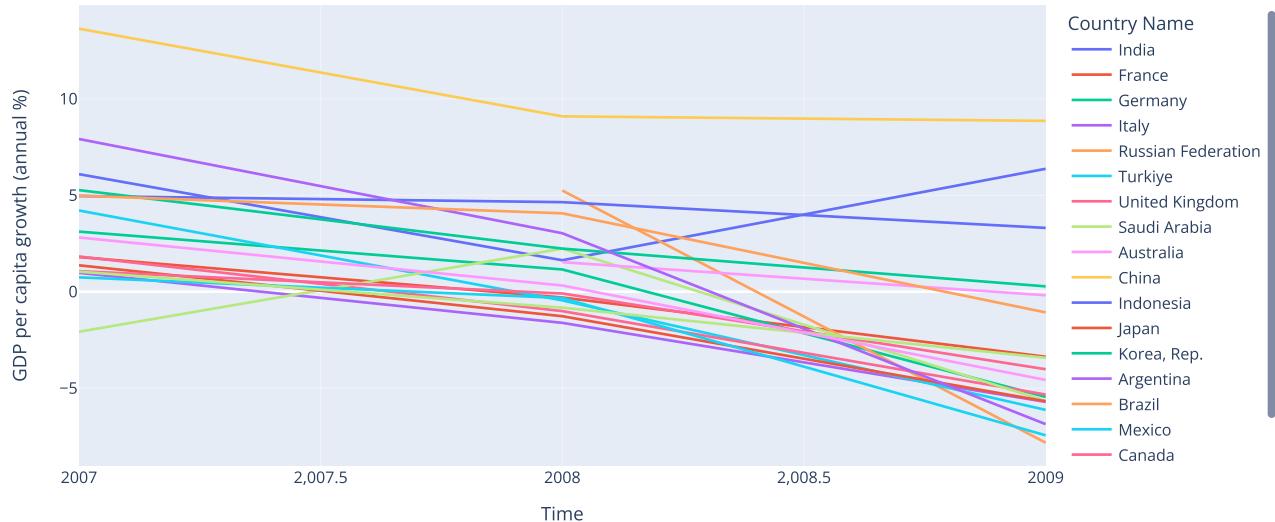
```
In [96]: fig = px.line(df_economic_crisis, x='Time', y='GDP (current US$)', color='Country Name',
                     title=f"GDP (current US$) (2007-2009)")
fig.show()
```

GDP (current US\$) (2007-2009)



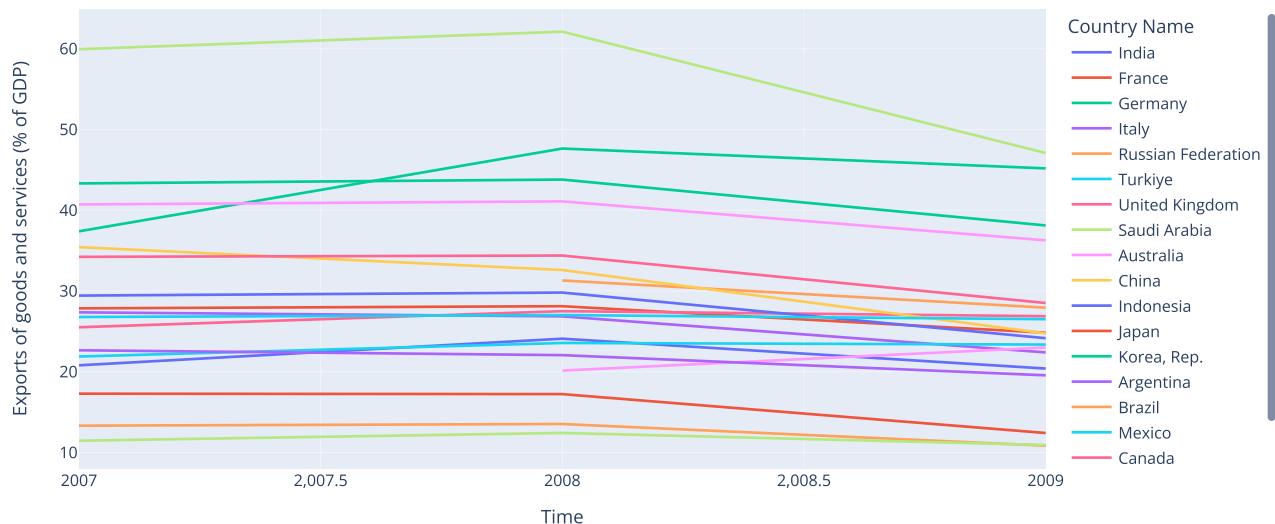
```
In [97]: fig = px.line(df_economic_crisis, x='Time', y='GDP per capita growth (annual %)', color='Country Name',
                     title=f"GDP per capita growth (annual %) (2007-2009)")
fig.show()
```

GDP per capita growth (annual %) (2007-2009)



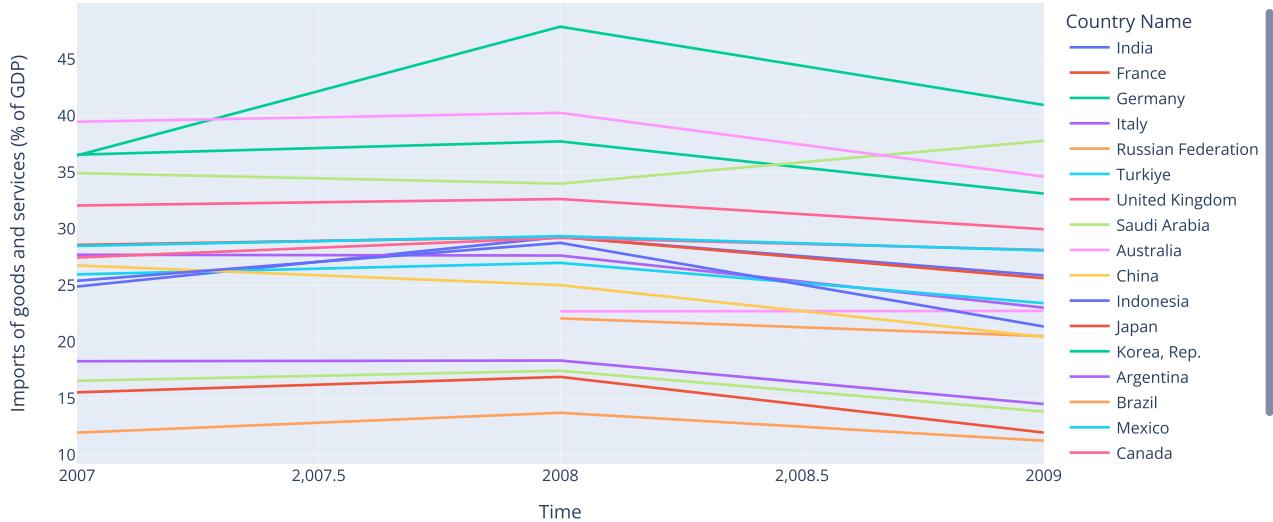
```
In [98]: fig = px.line(df_economic_crisis, x='Time', y='Exports of goods and services (% of GDP)', color='Country Name',
                     title=f"Exports of goods and services (% of GDP) (2007-2009)")
fig.show()
```

Exports of goods and services (% of GDP) (2007-2009)



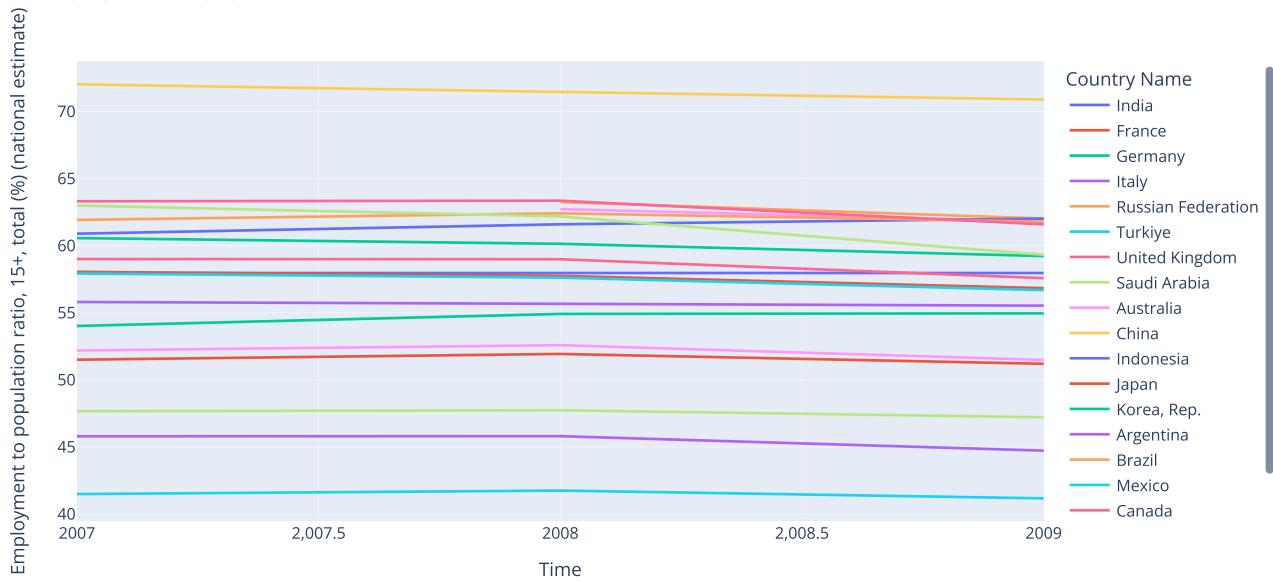
```
In [99]: fig = px.line(df_economic_crisis, x='Time', y='Imports of goods and services (% of GDP)', color='Country Name',
                     title=f"Imports of goods and services (% of GDP) (2007-2009)")
fig.show()
```

Imports of goods and services (% of GDP) (2007-2009)

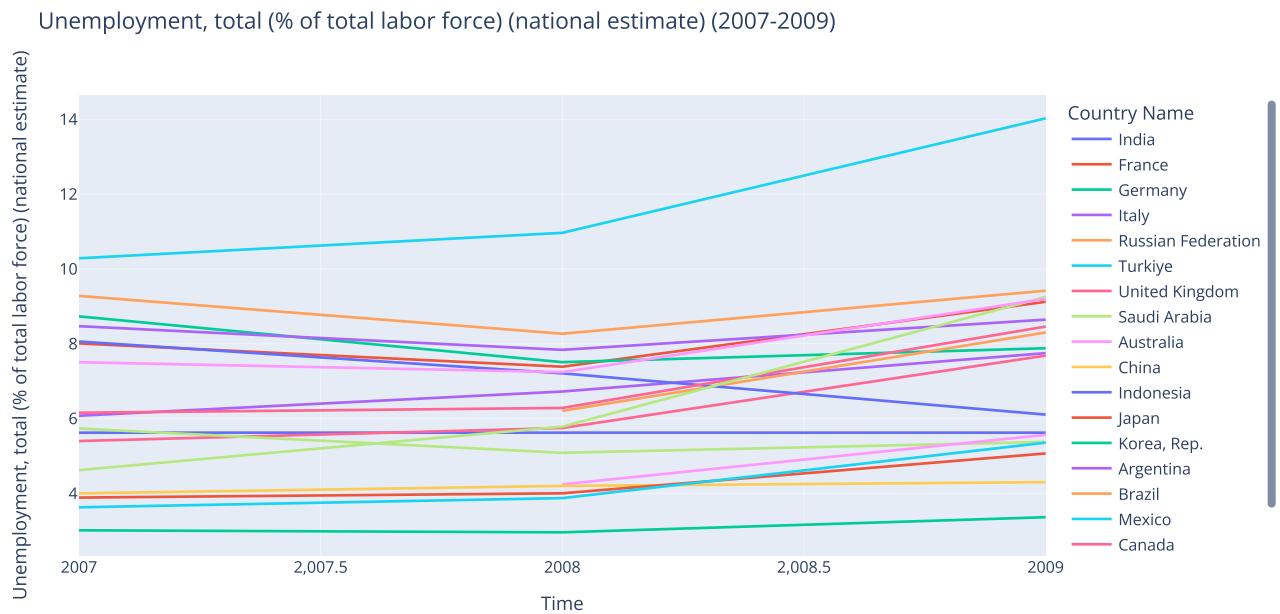


```
In [100]: fig = px.line(df_economic_crisis, x='Time', y='Employment to population ratio, 15+, total (%) (national estimate)', color='Country Name',
                     title=f"Employment to population ratio, 15+, total (%) (national estimate) (2007-2009)")
fig.show()
```

Employment to population ratio, 15+, total (%) (national estimate) (2007-2009)



```
In [101]: fig = px.line(df_economic_crisis, x='Time', y='Unemployment, total (% of total labor force) (national estimate)', color='Country'
                     title=f"Unemployment, total (% of total labor force) (national estimate) (2007-2009)")
fig.show()
```



## Case Study 2: COVID-19 Pandemic

The COVID-19 pandemic (also known as the coronavirus pandemic and COVID pandemic), caused by severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2), began with an outbreak of COVID-19 in Wuhan, China, in December 2019. It spread to other areas of Asia, and then worldwide in early 2020. The World Health Organization (WHO) declared the outbreak a public health emergency of international concern (PHEIC) on 30 January 2020, and assessed the outbreak as having become a pandemic on 11 March.

The pandemic and responses to it damaged the global economy. On 27 February 2020, worries about the outbreak crushed US stock indexes, which posted their sharpest falls since 2008.

Tourism collapsed due to travel restrictions, closing of public places including travel attractions, and advice of governments against travel. Airlines cancelled flights, while British regional airline Flybe collapsed. The cruise line industry was hard hit, and train stations and ferry ports closed. International mail stopped or was delayed.

We will conduct a covid case study on G20 nations

```
In [104]: g20 = ['Argentina', 'Australia', 'Brazil', 'Canada', 'China', 'France', 'Germany', 'India', 'Indonesia', 'Italy', 'Japan', 'Mexico', 'Netherlands', 'New Zealand', 'Norway', 'Pakistan', 'Peru', 'Russia', 'South Africa', 'South Korea', 'Spain', 'Sweden', 'Switzerland', 'United Kingdom', 'United States', 'Uruguay', 'Venezuela']
for i in g20:
    if i not in df["Country Name"].unique():
        print(i)

years = [2020, 2021, 2022]

df_covid_crisis = df1[(df1["Country Name"].isin(g20)) & (df1["Time"].isin(years))]
df_covid_crisis.head()
```

Out[104]:

	Country Name	Country Code	Time	Current account balance (% of GDP)	Exports of goods and services (% of GDP)	Foreign direct investment, net inflows (% of GDP)	Foreign direct investment, net outflows (% of GDP)	GDP (current US\$)	GDP growth (annual %)	GDP per capita (current US\$)	...	Current health expenditure per capita (current US\$)	Hospital beds (per 1,000 people)	Physicians (per 1,000 people)	exp
66	India	IND	2020	1.223621	18.682477	2.406203	0.415824	2.670000e+12	-5.777725	1915.551588	...	64.000000	1.620	0.727	70
67	India	IND	2021	-1.055242	21.399158	1.412171	0.544277	3.170000e+12	9.689592	2250.179018	...	74.000000	1.600	2.289	65
204	France	FRA	2020	-1.563074	27.327736	0.731612	1.085462	2.650000e+12	-7.540459	39179.744260	...	4755.440430	6.000	3.324	82
205	France	FRA	2021	0.333456	30.034994	3.311955	3.994612	2.960000e+12	6.435210	43671.308410	...	5380.875000	2.795	2.289	82
227	Germany	DEU	2020	6.567299	43.476980	4.549376	3.729155	3.890000e+12	-3.826769	46749.476230	...	5936.251465	7.800	4.459	81

5 rows × 48 columns

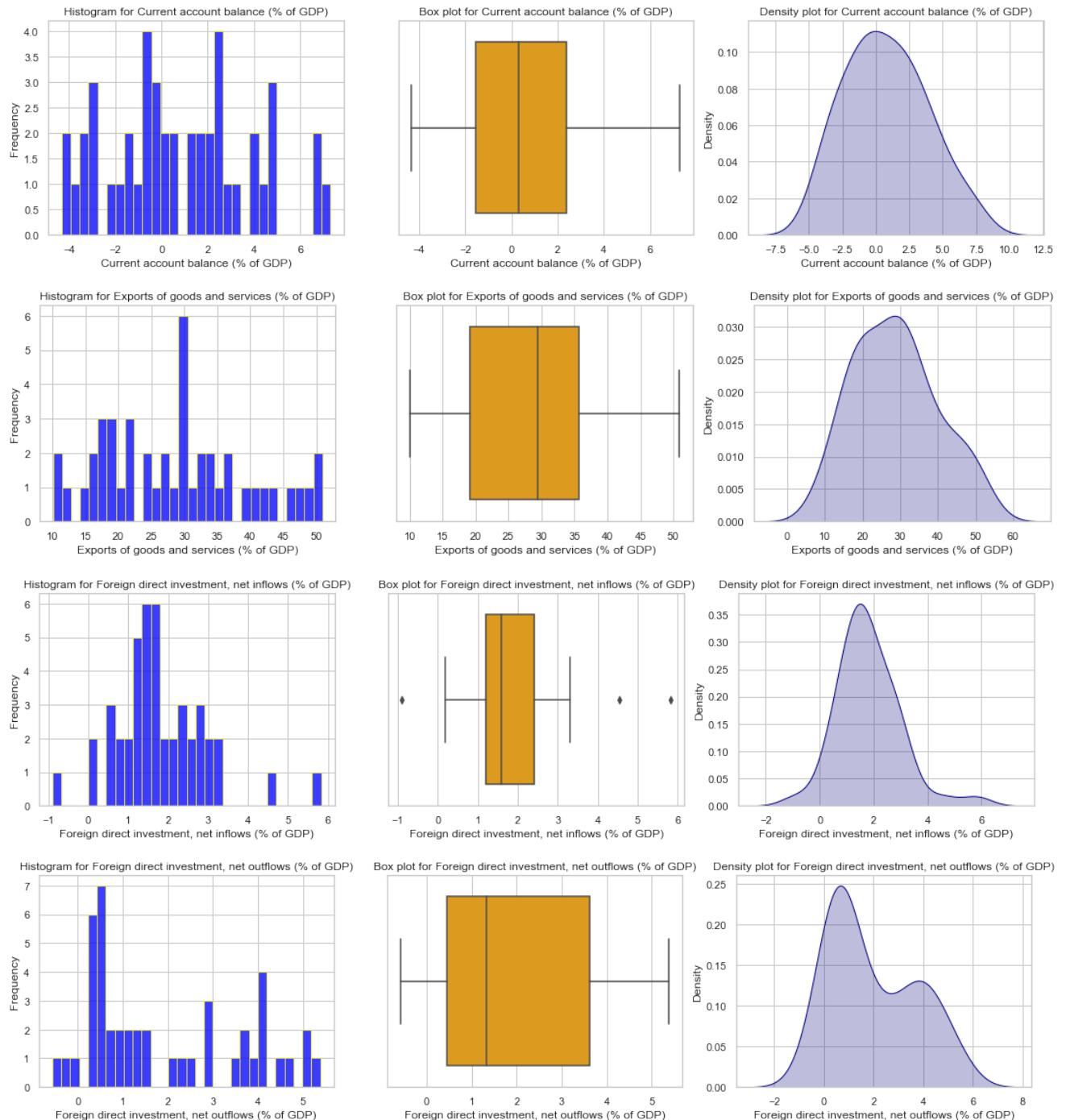
```
In [105]: for i in economic_metrics:
    plt.figure(figsize=(15,4))

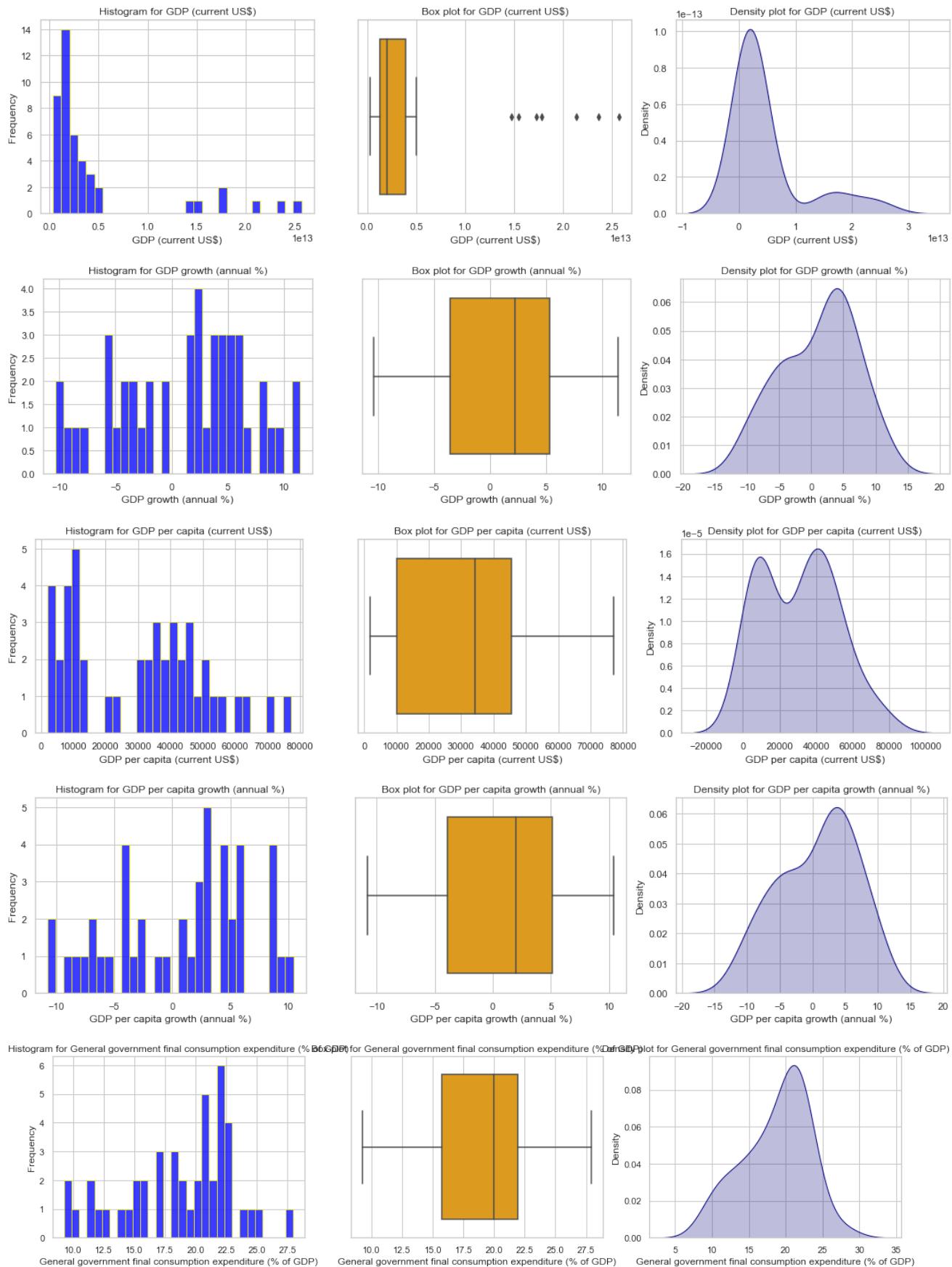
    plt.subplot(1,3,1)
    sns.histplot(data=df_covid_crisis[i], color="blue", edgecolor="yellow", bins=30)
    plt.xlabel(i)
    plt.ylabel("Frequency")
    plt.title(f"Histogram for {i}")

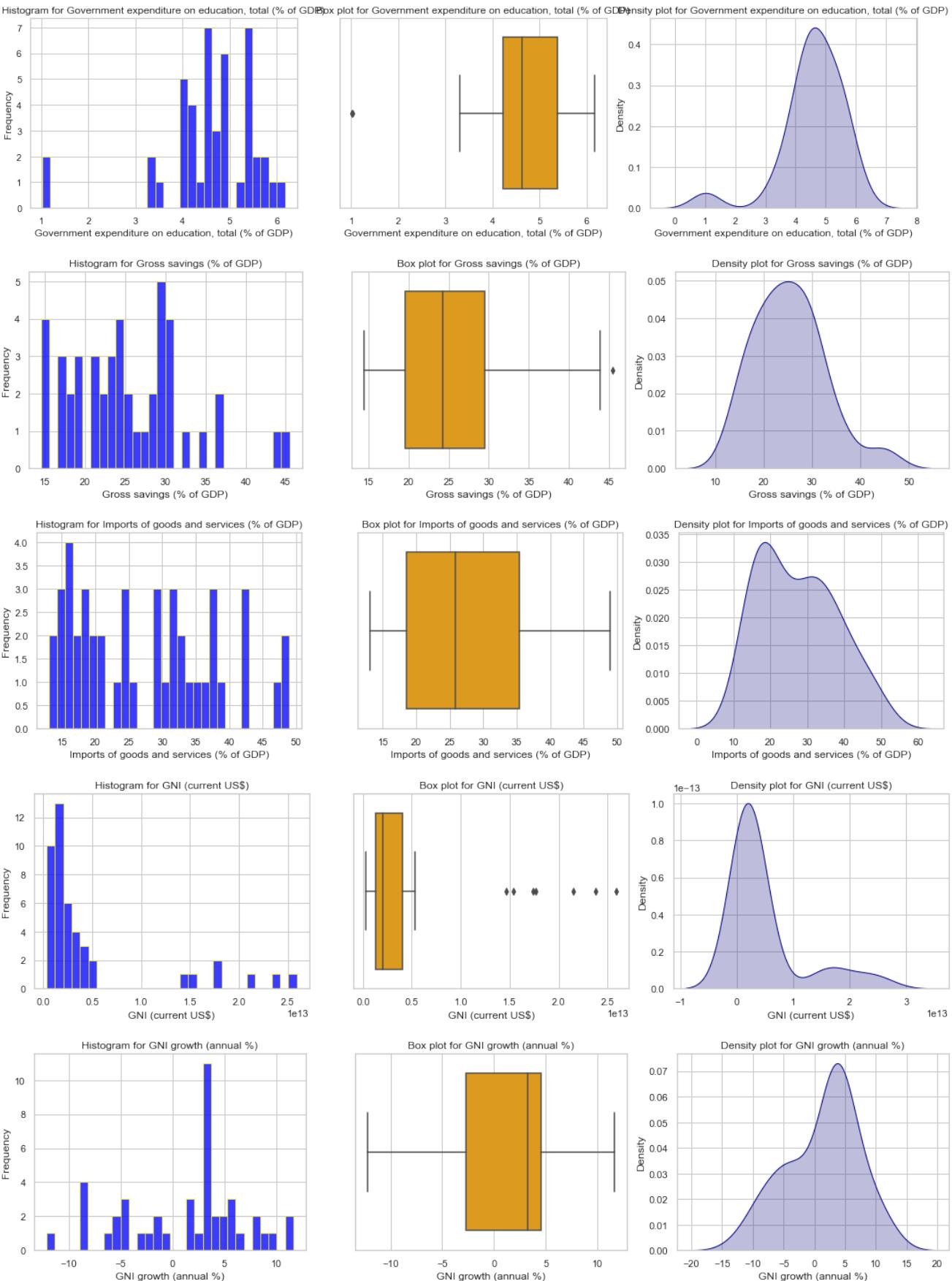
    plt.subplot(1,3,2)
    sns.boxplot(x=df_covid_crisis[i], color="orange")
    plt.xlabel(i)
    plt.title(f"Box plot for {i}")

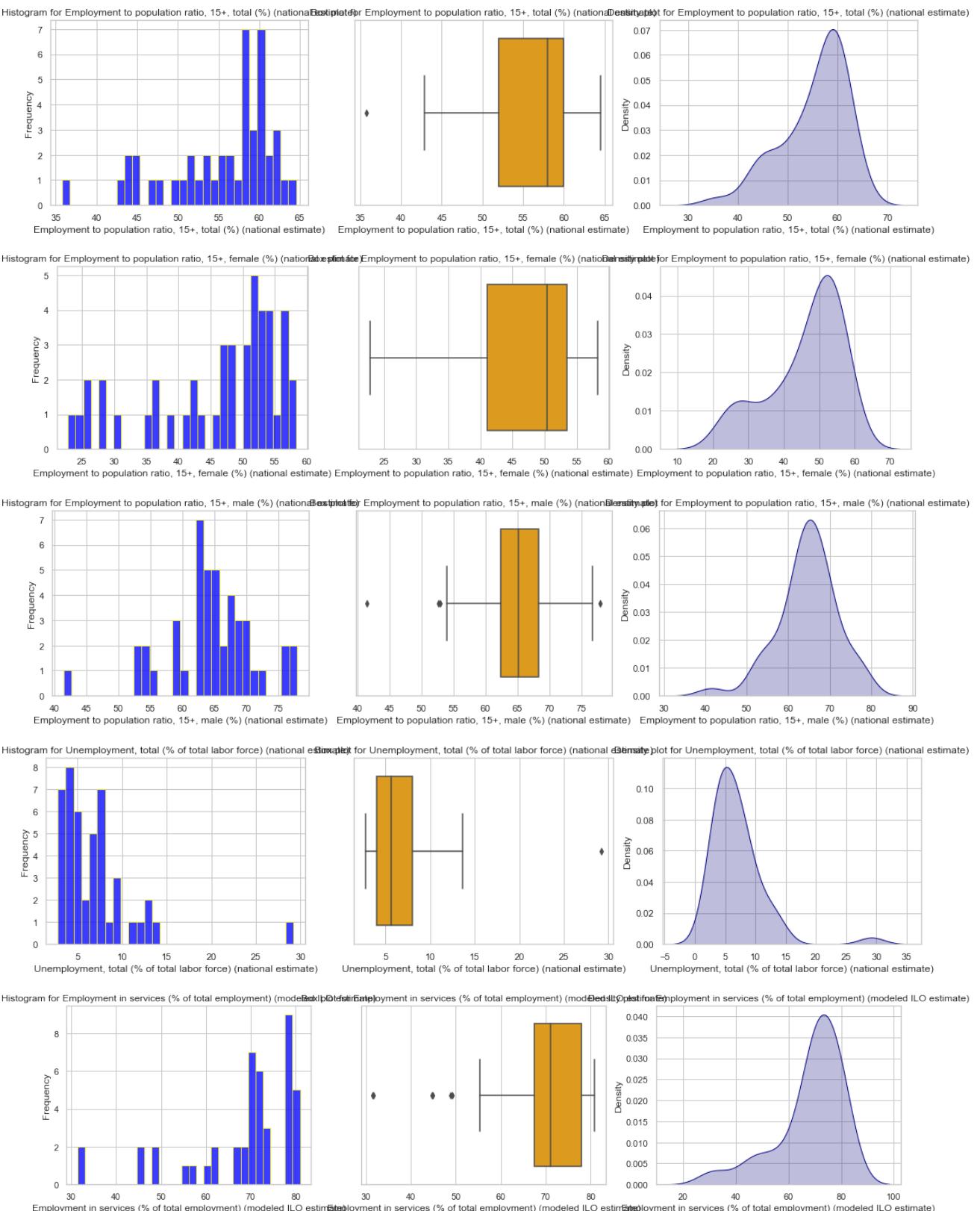
    plt.subplot(1,3,3)
    sns.kdeplot(data=df_covid_crisis[i], color="navy", fill=True)
    plt.xlabel(i)
    plt.ylabel("Density")
    plt.title(f"Density plot for {i}")

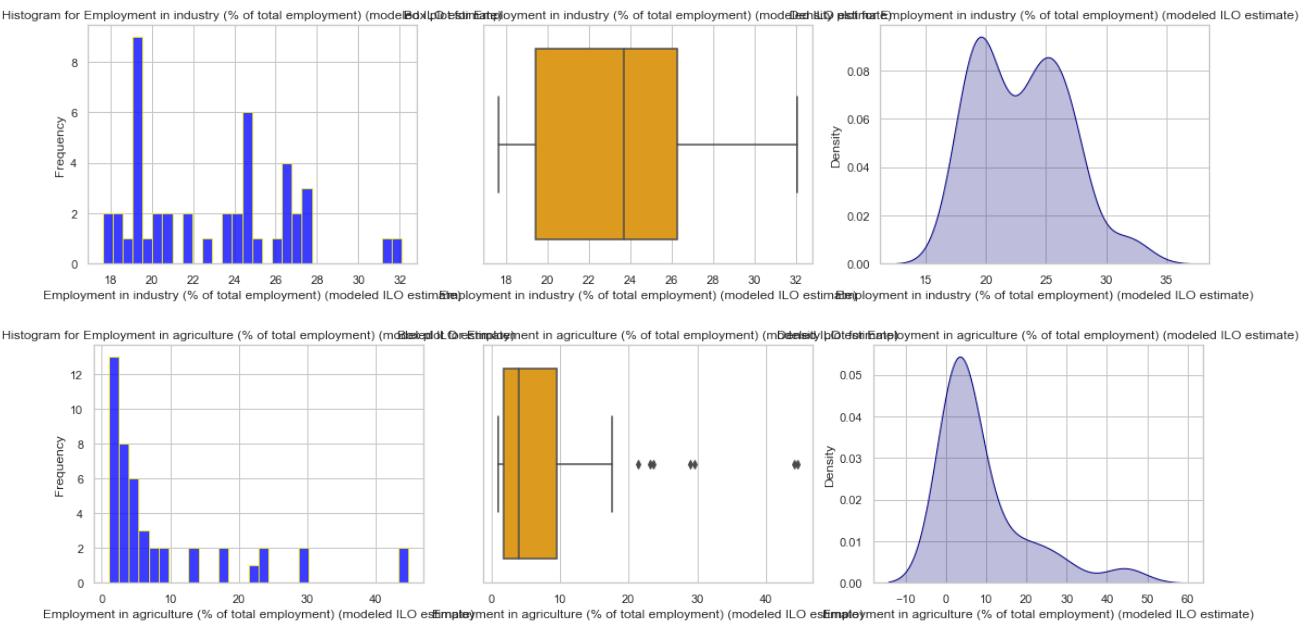
    plt.tight_layout()
    plt.show()
```





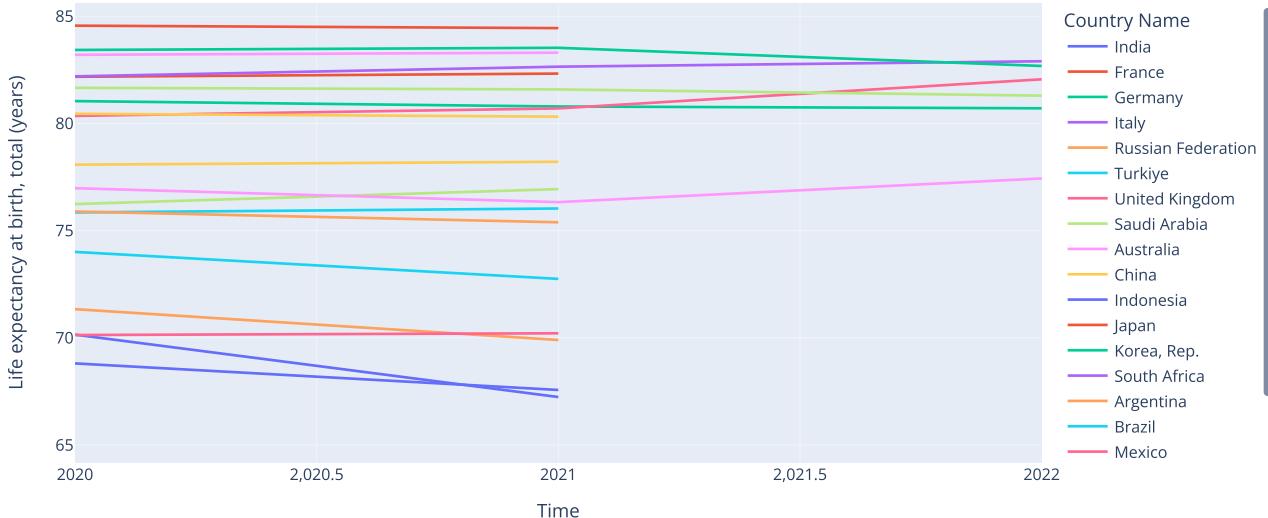






```
In [108]: fig = px.line(df_covid_crisis, x='Time', y='Life expectancy at birth, total (years)', color='Country Name', title="Life Expectancy Trends")
fig.show()
```

### Life Expectancy Trends



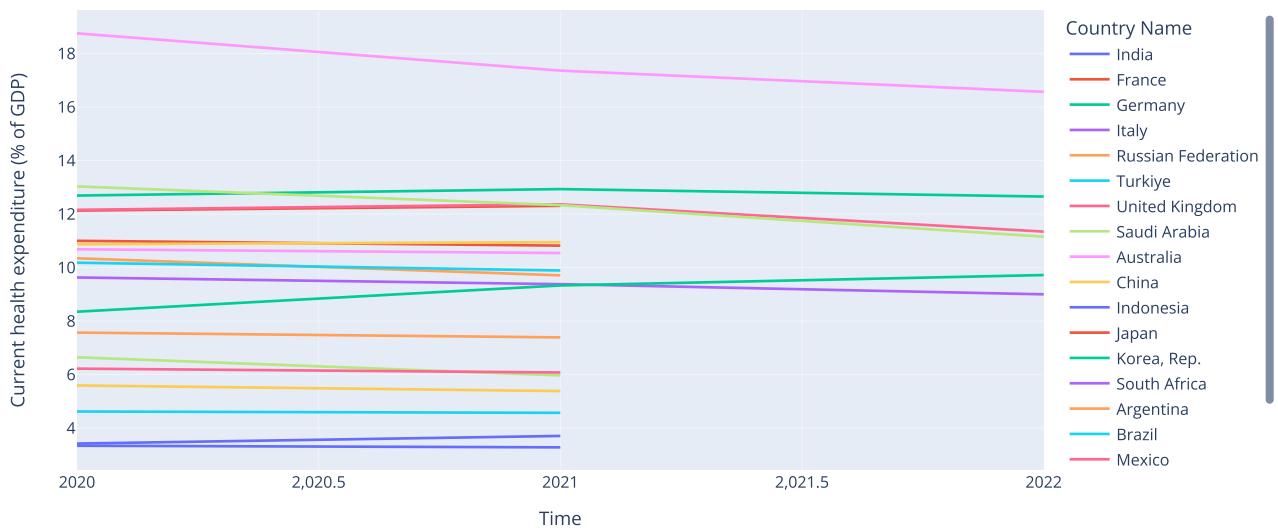
```
In [109]: fig = px.line(df_covid_crisis, x='Time', y='Hospital beds (per 1,000 people)', color='Country Name',
                     title="Hospital beds (per 1,000 people)")
fig.show()
```

Hospital beds (per 1,000 people)



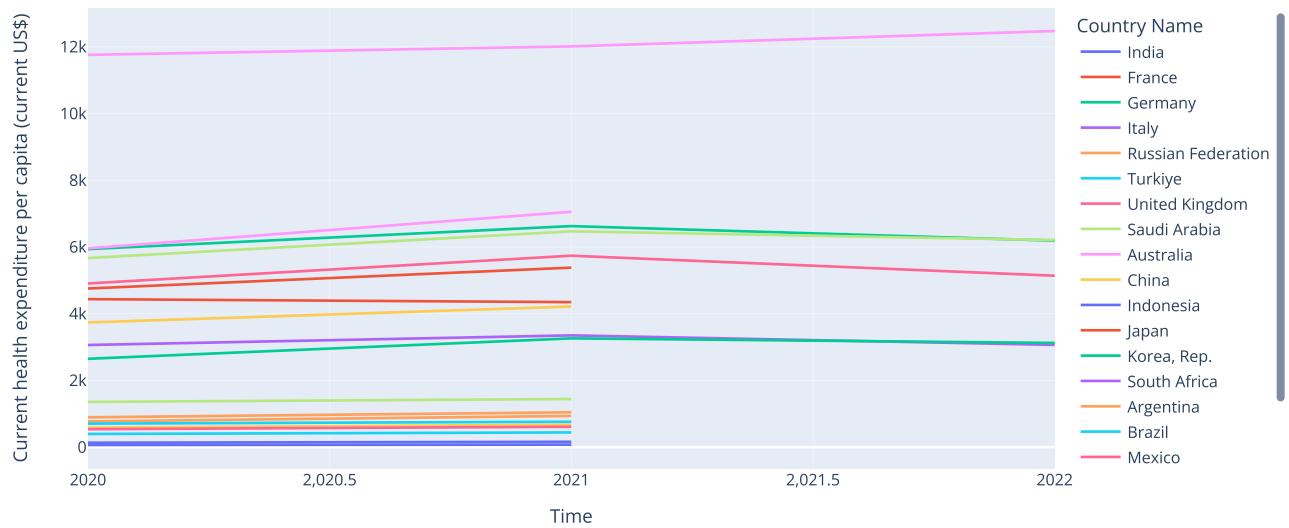
```
In [110]: fig = px.line(df_covid_crisis, x='Time', y='Current health expenditure (% of GDP)', color='Country Name',
                     title="Current health expenditure (% of GDP)")
fig.show()
```

Current health expenditure (% of GDP)



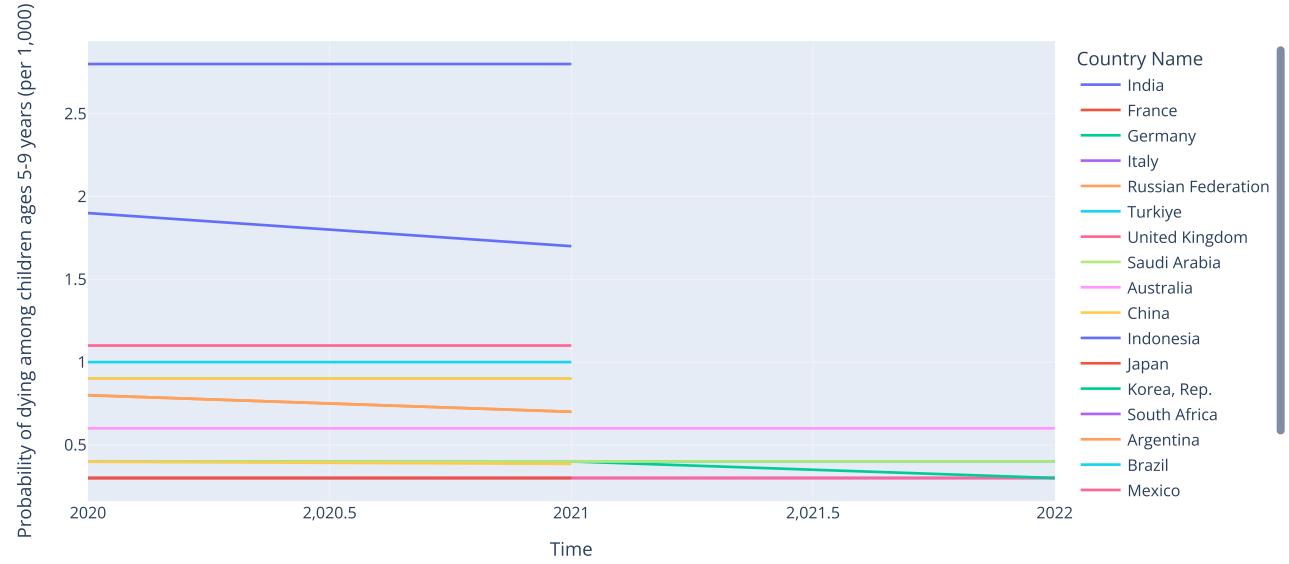
```
In [111]: fig = px.line(df_covid_crisis, x='Time', y='Current health expenditure per capita (current US$)', color='Country Name',
                     title="Current health expenditure per capita (current US$)")
fig.show()
```

Current health expenditure per capita (current US\$)

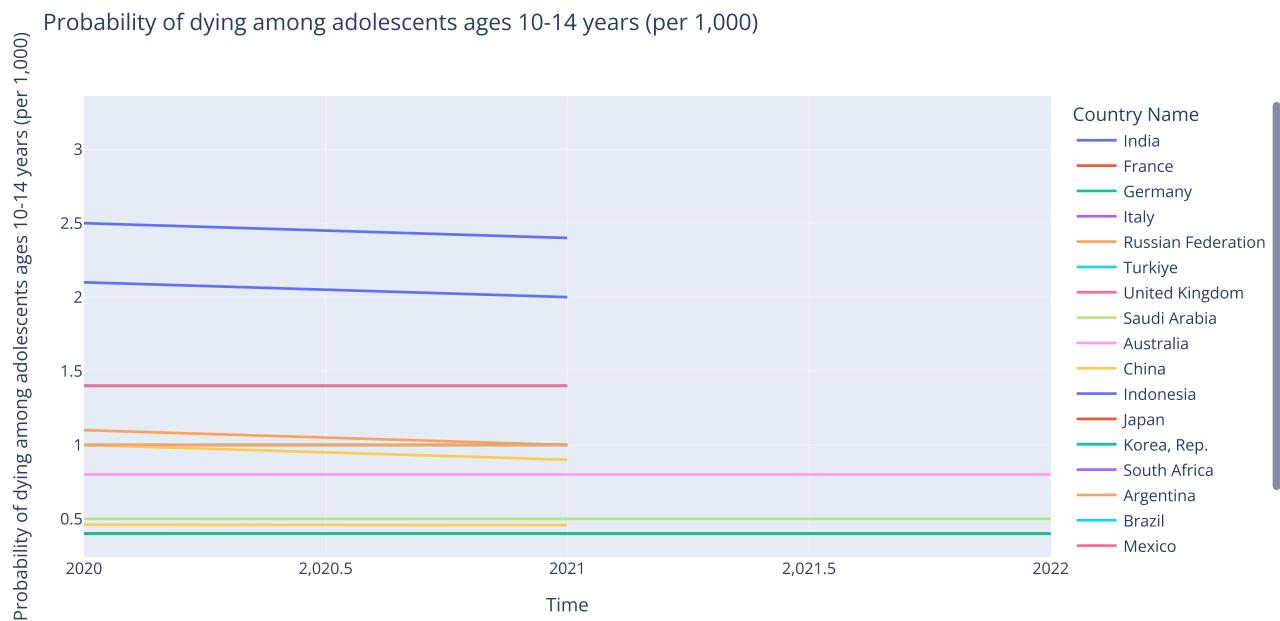


```
In [113]: fig = px.line(df_covid_crisis, x='Time', y='Probability of dying among children ages 5-9 years (per 1,000)', color='Country Name',
                     title="Probability of dying among children ages 5-9 years (per 1,000)")
fig.show()
```

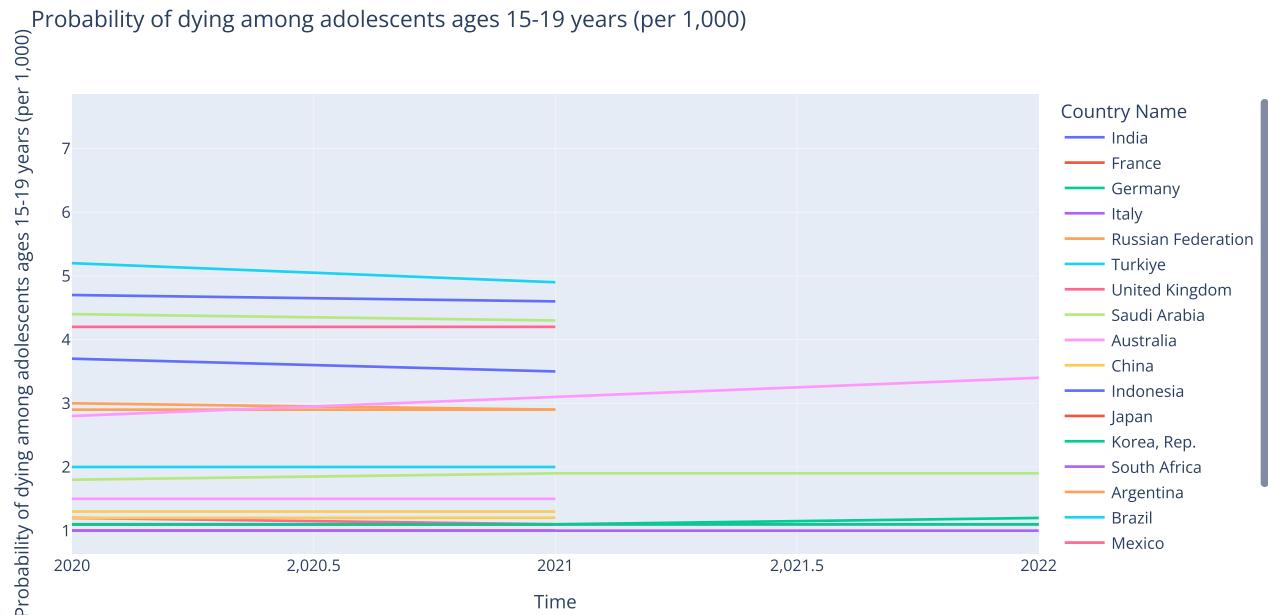
Probability of dying among children ages 5-9 years (per 1,000)



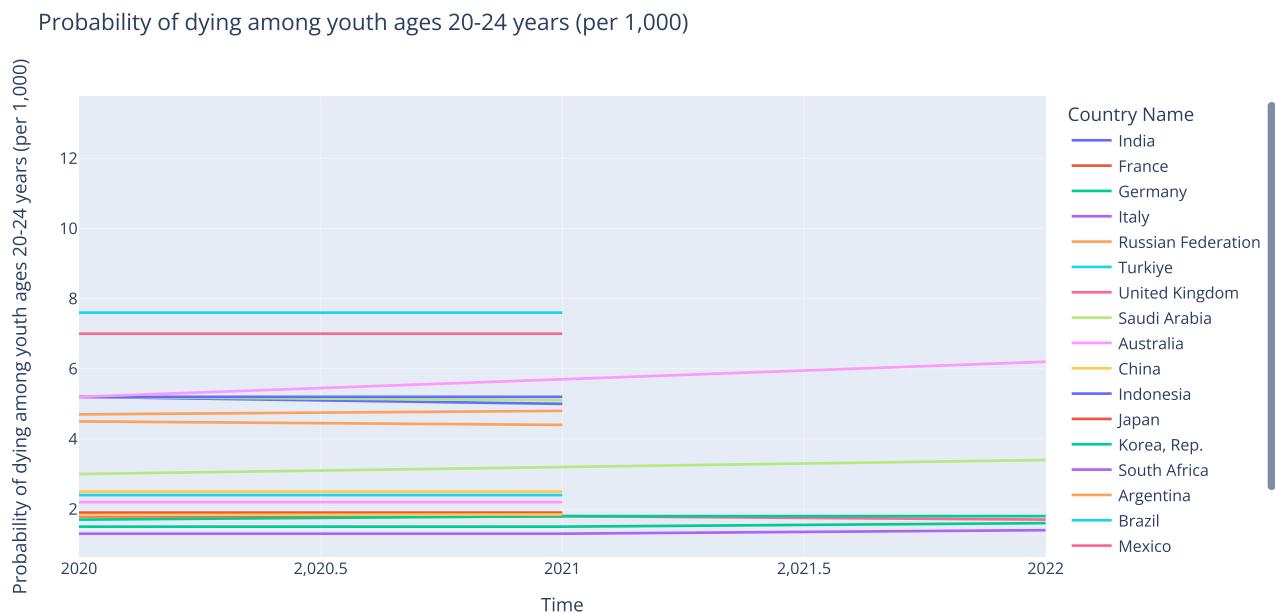
```
In [118]: fig = px.line(df_covid_crisis, x='Time', y='Probability of dying among adolescents ages 10-14 years (per 1,000)', color='Country'
                     title="Probability of dying among adolescents ages 10-14 years (per 1,000)")
fig.show()
```



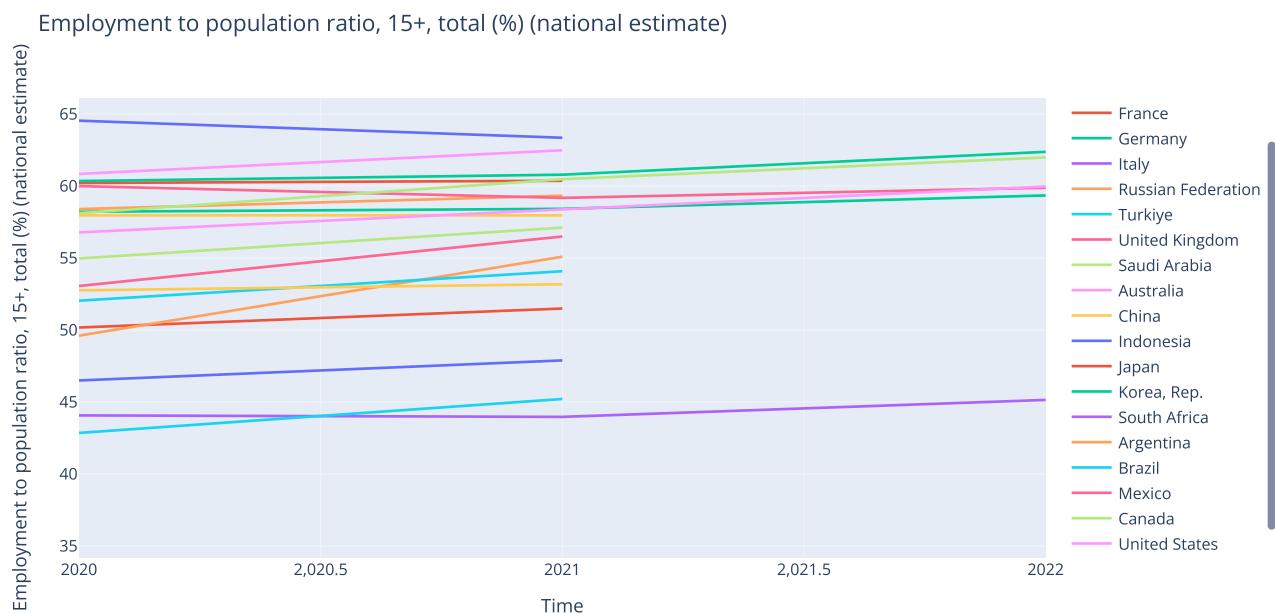
```
In [119]: fig = px.line(df_covid_crisis, x='Time', y='Probability of dying among adolescents ages 15-19 years (per 1,000)', color='Country'
                     title="Probability of dying among adolescents ages 15-19 years (per 1,000)")
fig.show()
```



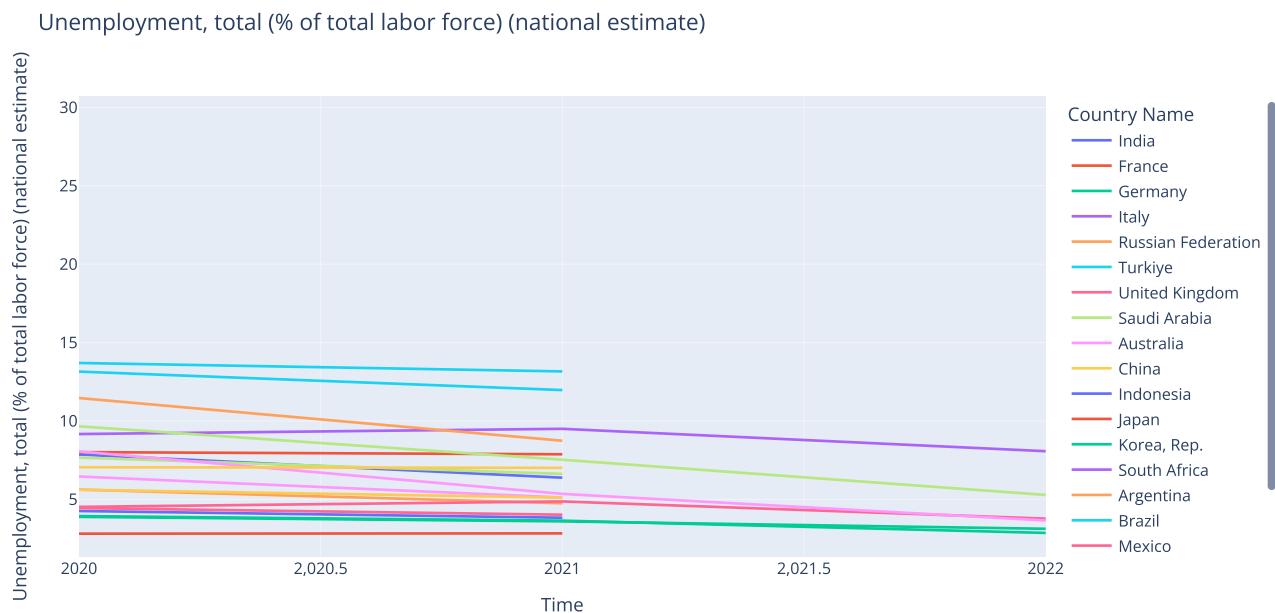
```
In [120]: fig = px.line(df_covid_crisis, x='Time', y='Probability of dying among youth ages 20-24 years (per 1,000)', color='Country Name',
                     title="Probability of dying among youth ages 20-24 years (per 1,000)")
fig.show()
```



```
In [114]: fig = px.line(df_covid_crisis, x='Time', y='Employment to population ratio, 15+, total (%) (national estimate)', color='Country Name',
                     title="Employment to population ratio, 15+, total (%) (national estimate)")
fig.show()
```



```
In [115]: fig = px.line(df_covid_crisis, x='Time', y='Unemployment, total (% of total labor force) (national estimate)', color='Country Name',
                     title="Unemployment, total (% of total labor force) (national estimate)")
fig.show()
```

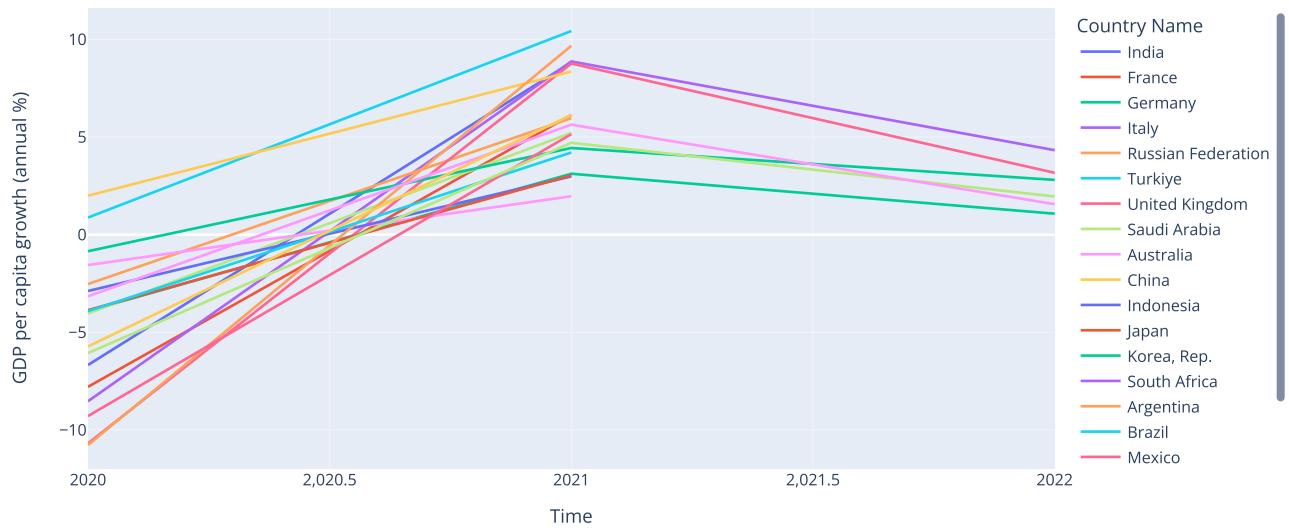


```
In [116]: fig = px.line(df_covid_crisis, x='Time', y='GDP growth (annual %)', color='Country Name',
                     title="GDP growth (annual %)")
fig.show()
```



```
In [117]: fig = px.line(df_covid_crisis, x='Time', y='GDP per capita growth (annual %)', color='Country Name',  
                     title="GDP per capita growth (annual %)")  
fig.show()
```

GDP per capita growth (annual %)



In [ ]: