

```

import pandas as pd
import plotly.express as px
import plotly.graph_objects as go
import matplotlib.pyplot as plt
import seaborn as sns

data = pd.read_csv("transformed_data.csv")
data2 = pd.read_csv("raw_data.csv")
data3 = pd.read_csv("countries-table.csv")
data4 = pd.read_csv("owid-covid-data.csv")

pd.plotting.register_matplotlib_converters()
%matplotlib inline
data3 = data3.dropna()
data3.head()

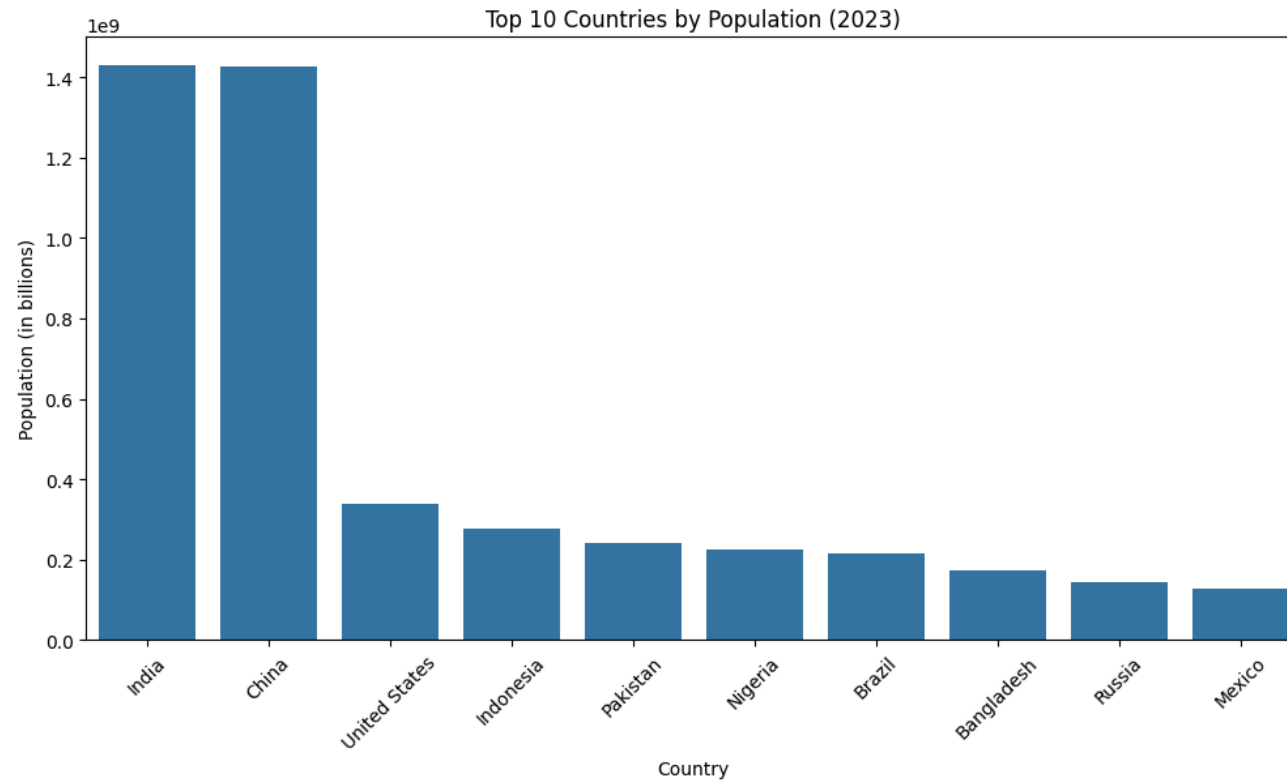
```

| | country | rank | area | landAreaKm | cca2 | cca3 | netChange | growthRate | worldPercentage | density | densityMi | place | pop1980 |
|---|---------------|------|-----------|------------|------|------|-----------|------------|-----------------|----------|-----------|-------|----------|
| 0 | India | 1 | 3287590.0 | 2973190.0 | IN | IND | 0.4184 | 0.0081 | 0.1785 | 480.5033 | 1244.5036 | 356 | 69682838 |
| 1 | China | 2 | 9706961.0 | 9424702.9 | CN | CHN | -0.0113 | -0.0002 | 0.1781 | 151.2696 | 391.7884 | 156 | 98237246 |
| 2 | United States | 3 | 9372610.0 | 9147420.0 | US | USA | 0.0581 | 0.0050 | 0.0425 | 37.1686 | 96.2666 | 840 | 22314001 |
| 3 | Indonesia | 4 | 1904569.0 | 1877519.0 | ID | IDN | 0.0727 | 0.0074 | 0.0347 | 147.8196 | 382.8528 | 360 | 14817709 |

```

top_10_countries = data3.nlargest(10, 'pop2023')
plt.figure(figsize=(12, 6))
sns.barplot(x='country', y='pop2023', data=top_10_countries)
plt.title('Top 10 Countries by Population (2023)')
plt.xlabel('Country')
plt.ylabel('Population (in billions)')
plt.xticks(rotation=45)
plt.show()

```



```
data4.head()
```

| | iso_code | continent | location | date | total_cases | new_cases | new_cases_smoothed | total_deaths | new_deaths | new_deaths_smoothed |
|---|----------|-----------|-------------|------------|-------------|-----------|--------------------|--------------|------------|---------------------|
| 0 | AFG | Asia | Afghanistan | 2020-02-24 | 5.0 | 5.0 | NaN | NaN | NaN | NaN |
| 1 | AFG | Asia | Afghanistan | 2020-02-25 | 5.0 | 0.0 | NaN | NaN | NaN | NaN |
| 2 | AFG | Asia | Afghanistan | 2020-02-26 | 5.0 | 0.0 | NaN | NaN | NaN | NaN |
| 3 | AFG | Asia | Afghanistan | 2020-02-27 | 5.0 | 0.0 | NaN | NaN | NaN | NaN |
| 4 | AFG | Asia | Afghanistan | 2020-02-28 | 5.0 | 0.0 | NaN | NaN | NaN | NaN |

5 rows × 11 columns

```
data4.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 107104 entries, 0 to 107103
Data columns (total 67 columns):
```

| # | Column | Non-Null Count | Dtype |
|----|--|-----------------|---------|
| 0 | iso_code | 107104 non-null | object |
| 1 | continent | 100984 non-null | object |
| 2 | location | 107104 non-null | object |
| 3 | date | 107103 non-null | object |
| 4 | total_cases | 105696 non-null | float64 |
| 5 | new_cases | 105586 non-null | float64 |
| 6 | new_cases_smoothed | 104292 non-null | float64 |
| 7 | total_deaths | 95345 non-null | float64 |
| 8 | new_deaths | 95344 non-null | float64 |
| 9 | new_deaths_smoothed | 93946 non-null | float64 |
| 10 | total_cases_per_million | 104938 non-null | float64 |
| 11 | new_cases_per_million | 104828 non-null | float64 |
| 12 | new_cases_smoothed_per_million | 103540 non-null | float64 |
| 13 | total_deaths_per_million | 94600 non-null | float64 |
| 14 | new_deaths_per_million | 94599 non-null | float64 |
| 15 | new_deaths_smoothed_per_million | 93207 non-null | float64 |
| 16 | reproduction_rate | 80413 non-null | float64 |
| 17 | icu_patients | 15074 non-null | float64 |
| 18 | icu_patients_per_million | 15074 non-null | float64 |
| 19 | hosp_patients | 15640 non-null | float64 |
| 20 | hosp_patients_per_million | 15640 non-null | float64 |
| 21 | weekly_icu_admissions | 4178 non-null | float64 |
| 22 | weekly_icu_admissions_per_million | 4178 non-null | float64 |
| 23 | weekly_hosp_admissions | 7413 non-null | float64 |
| 24 | weekly_hosp_admissions_per_million | 7413 non-null | float64 |
| 25 | new_tests | 41668 non-null | float64 |
| 26 | total_tests | 42489 non-null | float64 |
| 27 | total_tests_per_thousand | 42489 non-null | float64 |
| 28 | new_tests_per_thousand | 41668 non-null | float64 |
| 29 | new_tests_smoothed | 53677 non-null | float64 |
| 30 | new_tests_smoothed_per_thousand | 53677 non-null | float64 |
| 31 | positive_rate | 49226 non-null | float64 |
| 32 | tests_per_case | 48821 non-null | float64 |
| 33 | tests_units | 55573 non-null | object |
| 34 | total_vaccinations | 28794 non-null | float64 |
| 35 | people_vaccinated | 27843 non-null | float64 |
| 36 | people_fully_vaccinated | 25973 non-null | float64 |
| 37 | total_boosters | 10348 non-null | float64 |
| 38 | new_vaccinations | 23717 non-null | float64 |
| 39 | new_vaccinations_smoothed | 53716 non-null | float64 |
| 40 | total_vaccinations_per_hundred | 28794 non-null | float64 |
| 41 | people_vaccinated_per_hundred | 27843 non-null | float64 |
| 42 | people_fully_vaccinated_per_hundred | 25973 non-null | float64 |
| 43 | total_boosters_per_hundred | 10348 non-null | float64 |
| 44 | new_vaccinations_smoothed_per_million | 53716 non-null | float64 |
| 45 | new_people_vaccinated_smoothed | 53020 non-null | float64 |
| 46 | new_people_vaccinated_smoothed_per_hundred | 53020 non-null | float64 |
| 47 | stringency_index | 84724 non-null | float64 |
| 48 | population | 106345 non-null | float64 |
| 49 | population_density | 97112 non-null | float64 |
| 50 | median_age | 87646 non-null | float64 |
| 51 | aged_65_older | 87646 non-null | float64 |
| 52 | aged_70_older | 87646 non-null | float64 |

```
data4[data4['location']=='United States']
```

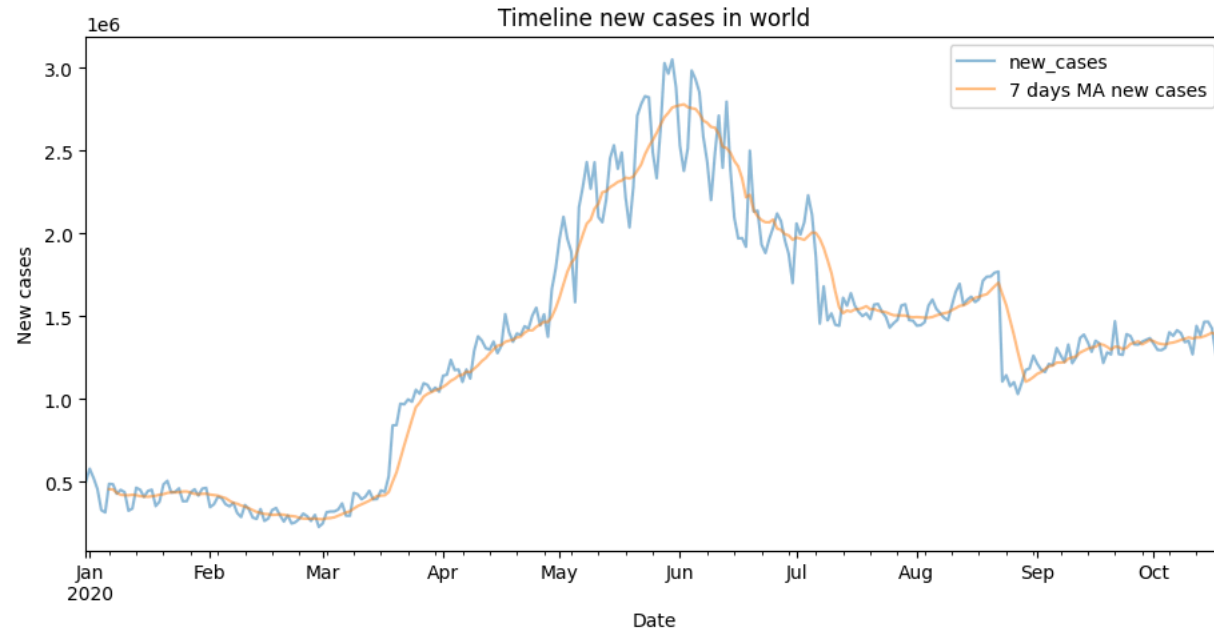
| iso_code | continent | location | date | total_cases | new_cases | new_cases_smoothed | total_deaths | new_deaths | new_deaths_smoothed |
|----------|-----------|----------|------|-------------|-----------|--------------------|--------------|------------|---------------------|
|----------|-----------|----------|------|-------------|-----------|--------------------|--------------|------------|---------------------|

0 rows × 67 columns

```
data5=data4.copy()
data5 = data4.copy()
data5.date = pd.to_datetime(data2['date'])
data5 = data5.groupby('date').sum()
data5['7 days MA new cases'] = 0
data5['7 days MA new cases'] = data5['new_cases'].rolling(7).mean() #Moving average of new cases with window=10
data5['7 days MA new deaths'] = 0
data5['7 days MA new deaths'] = data5['new_deaths'].rolling(7).mean() #Moving average of new deaths with window=10
```

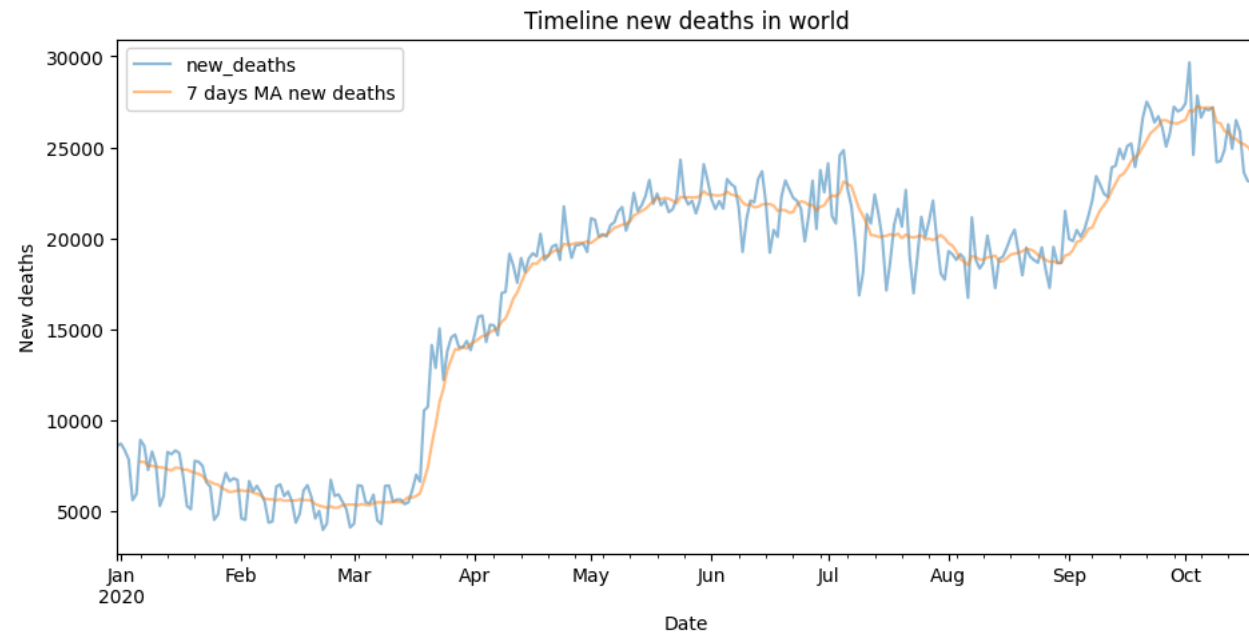
```
data5[['new_cases', '7 days MA new cases']].plot(figsize = (11, 5), alpha = 0.5)
plt.title('Timeline new cases in world')
plt.xlabel('Date')
plt.ylabel('New cases')
```

Text(0, 0.5, 'New cases')



```
data5[['new_deaths', '7 days MA new deaths']].plot(figsize = (11, 5), alpha = 0.5)
plt.title('Timeline new deaths in world')
plt.xlabel('Date')
plt.ylabel('New deaths')
```

```
Text(0, 0.5, 'New deaths')
```



```
print(data)
```

| | CODE | COUNTRY | DATE | HDI | TC | TD | STI \ |
|-------|------|-------------|------------|-------|----------|----------|----------|
| 0 | AFG | Afghanistan | 2019-12-31 | 0.498 | 0.000000 | 0.000000 | 0.000000 |
| 1 | AFG | Afghanistan | 2020-01-01 | 0.498 | 0.000000 | 0.000000 | 0.000000 |
| 2 | AFG | Afghanistan | 2020-01-02 | 0.498 | 0.000000 | 0.000000 | 0.000000 |
| 3 | AFG | Afghanistan | 2020-01-03 | 0.498 | 0.000000 | 0.000000 | 0.000000 |
| 4 | AFG | Afghanistan | 2020-01-04 | 0.498 | 0.000000 | 0.000000 | 0.000000 |
| ... | ... | ... | ... | ... | ... | ... | ... |
| 50413 | ZWE | Zimbabwe | 2020-10-15 | 0.535 | 8.994048 | 5.442418 | 4.341855 |
| 50414 | ZWE | Zimbabwe | 2020-10-16 | 0.535 | 8.996528 | 5.442418 | 4.341855 |
| 50415 | ZWE | Zimbabwe | 2020-10-17 | 0.535 | 8.999496 | 5.442418 | 4.341855 |
| 50416 | ZWE | Zimbabwe | 2020-10-18 | 0.535 | 9.000853 | 5.442418 | 4.341855 |
| 50417 | ZWE | Zimbabwe | 2020-10-19 | 0.535 | 9.005405 | 5.442418 | 4.341855 |

| | POP | GDPCAP |
|-------|-----------|----------|
| 0 | 17.477233 | 7.497754 |
| 1 | 17.477233 | 7.497754 |
| 2 | 17.477233 | 7.497754 |
| 3 | 17.477233 | 7.497754 |
| 4 | 17.477233 | 7.497754 |
| ... | ... | ... |
| 50413 | 16.514381 | 7.549491 |
| 50414 | 16.514381 | 7.549491 |
| 50415 | 16.514381 | 7.549491 |
| 50416 | 16.514381 | 7.549491 |
| 50417 | 16.514381 | 7.549491 |

```
[50418 rows x 9 columns]
```

```
print(data.head())
```

| | CODE | COUNTRY | DATE | HDI | TC | TD | STI | POP | GDPCAP |
|---|------|-------------|------------|-------|-----|-----|-----|-----------|----------|
| 0 | AFG | Afghanistan | 2019-12-31 | 0.498 | 0.0 | 0.0 | 0.0 | 17.477233 | 7.497754 |
| 1 | AFG | Afghanistan | 2020-01-01 | 0.498 | 0.0 | 0.0 | 0.0 | 17.477233 | 7.497754 |
| 2 | AFG | Afghanistan | 2020-01-02 | 0.498 | 0.0 | 0.0 | 0.0 | 17.477233 | 7.497754 |
| 3 | AFG | Afghanistan | 2020-01-03 | 0.498 | 0.0 | 0.0 | 0.0 | 17.477233 | 7.497754 |
| 4 | AFG | Afghanistan | 2020-01-04 | 0.498 | 0.0 | 0.0 | 0.0 | 17.477233 | 7.497754 |

```
print(data2.head())
```

| | iso_code | location | date | total_cases | total_deaths | \ |
|---|----------|-------------|------------|-------------|--------------|---|
| 0 | AFG | Afghanistan | 2019-12-31 | 0.0 | 0.0 | |
| 1 | AFG | Afghanistan | 2020-01-01 | 0.0 | 0.0 | |
| 2 | AFG | Afghanistan | 2020-01-02 | 0.0 | 0.0 | |
| 3 | AFG | Afghanistan | 2020-01-03 | 0.0 | 0.0 | |
| 4 | AFG | Afghanistan | 2020-01-04 | 0.0 | 0.0 | |

| | stringency_index | population | gdp_per_capita | human_development_index | \ |
|---|------------------|------------|----------------|-------------------------|---|
| 0 | 0.0 | 38928341 | 1803.987 | 0.498 | |
| 1 | 0.0 | 38928341 | 1803.987 | 0.498 | |
| 2 | 0.0 | 38928341 | 1803.987 | 0.498 | |
| 3 | 0.0 | 38928341 | 1803.987 | 0.498 | |
| 4 | 0.0 | 38928341 | 1803.987 | 0.498 | |

| | Unnamed: 9 | Unnamed: 10 | Unnamed: 11 | Unnamed: 12 | Unnamed: 13 |
|---|------------|-------------|-------------|-------------|-------------|
| 0 | #NUM! | #NUM! | #NUM! | 17.477233 | 7.497754494 |
| 1 | #NUM! | #NUM! | #NUM! | 17.477233 | 7.497754494 |
| 2 | #NUM! | #NUM! | #NUM! | 17.477233 | 7.497754494 |
| 3 | #NUM! | #NUM! | #NUM! | 17.477233 | 7.497754494 |
| 4 | #NUM! | #NUM! | #NUM! | 17.477233 | 7.497754494 |

```
data["COUNTRY"].value_counts()
```

```
COUNTRY
Afghanistan    294
Indonesia      294
Macedonia      294
Luxembourg     294
Lithuania      294
...
Tajikistan     172
Comoros        171
Lesotho        158
Hong Kong      51
Solomon Islands 4
Name: count, Length: 210, dtype: int64
```

```
data["COUNTRY"].value_counts().mode()
```

```
0    294
Name: count, dtype: int64
```

```
# Aggregating the data
```

```
code = data["CODE"].unique().tolist()
country = data["COUNTRY"].unique().tolist()
hdi = []
tc = []
td = []
sti = []
population = data["POP"].unique().tolist()
gdp = []
```

```
for i in country:
```

```
    hdi.append((data.loc[data["COUNTRY"] == i, "HDI"]).sum()/294)
    tc.append((data2.loc[data2["location"] == i, "total_cases"]).sum())
    td.append((data2.loc[data2["location"] == i, "total_deaths"]).sum())
    sti.append((data.loc[data["COUNTRY"] == i, "STI"]).sum()/294)
    population.append((data2.loc[data2["location"] == i, "population"]).sum()/294)
```

```
aggregated_data = pd.DataFrame(list(zip(code, country, hdi, tc, td, sti, population)),
                                columns = ["Country Code", "Country", "HDI",
                                             "Total Cases", "Total Deaths",
                                             "Stringency Index", "Population"])
```

```
print(aggregated_data.head())
```

| | Country Code | Country | HDI | Total Cases | Total Deaths \ |
|---|--------------|-------------|----------|-------------|----------------|
| 0 | AFG | Afghanistan | 0.498000 | 5126433.0 | 165875.0 |
| 1 | ALB | Albania | 0.600765 | 1071951.0 | 31056.0 |
| 2 | DZA | Algeria | 0.754000 | 4893999.0 | 206429.0 |
| 3 | AND | Andorra | 0.659551 | 223576.0 | 9850.0 |
| 4 | AGO | Angola | 0.418952 | 304005.0 | 11820.0 |

| | Stringency Index | Population |
|---|------------------|------------|
| 0 | 3.049673 | 17.477233 |
| 1 | 3.005624 | 14.872537 |
| 2 | 3.195168 | 17.596309 |
| 3 | 2.677654 | 11.254996 |
| 4 | 2.965560 | 17.307957 |

```
# Sorting Data According to Total Cases
```

```
data = aggregated_data.sort_values(by=["Total Cases"], ascending=False)
print(data.head())
```

| | Country Code | Country | HDI | Total Cases | Total Deaths \ |
|-----|--------------|---------------|---------|-------------|----------------|
| 200 | USA | United States | 0.92400 | 746014098.0 | 26477574.0 |
| 27 | BRA | Brazil | 0.75900 | 425704517.0 | 14340567.0 |
| 90 | IND | India | 0.64000 | 407771615.0 | 7247327.0 |
| 157 | RUS | Russia | 0.81600 | 132888951.0 | 2131571.0 |
| 150 | PER | Peru | 0.59949 | 74882695.0 | 3020038.0 |

| | Stringency Index | Population |
|-----|------------------|------------|
| 200 | 3.350949 | 19.617637 |
| 27 | 3.136028 | 19.174732 |
| 90 | 3.610552 | 21.045353 |
| 157 | 3.380088 | 18.798668 |
| 150 | 3.430126 | 17.311165 |

Top 10 Countries with Highest Covid Cases

```
data = data.head(10)
print(data)
```

| | Country Code | Country | HDI | Total Cases | Total Deaths | \ |
|-----|--------------|----------------|----------|-------------|--------------|---|
| 200 | USA | United States | 0.924000 | 746014098.0 | 26477574.0 | |
| 27 | BRA | Brazil | 0.759000 | 425704517.0 | 14340567.0 | |
| 90 | IND | India | 0.640000 | 407771615.0 | 7247327.0 | |
| 157 | RUS | Russia | 0.816000 | 132888951.0 | 2131571.0 | |
| 150 | PER | Peru | 0.599490 | 74882695.0 | 3020038.0 | |
| 125 | MEX | Mexico | 0.774000 | 74347548.0 | 7295850.0 | |
| 178 | ESP | Spain | 0.887969 | 73717676.0 | 5510624.0 | |
| 175 | ZAF | South Africa | 0.608653 | 63027659.0 | 1357682.0 | |
| 42 | COL | Colombia | 0.581847 | 60543682.0 | 1936134.0 | |
| 199 | GBR | United Kingdom | 0.922000 | 59475032.0 | 7249573.0 | |

| | Stringency Index | Population |
|-----|------------------|------------|
| 200 | 3.350949 | 19.617637 |
| 27 | 3.136028 | 19.174732 |
| 90 | 3.610552 | 21.045353 |
| 157 | 3.380088 | 18.798668 |
| 150 | 3.430126 | 17.311165 |
| 125 | 3.019289 | 18.674802 |
| 178 | 3.393922 | 17.660427 |
| 175 | 3.364333 | 17.898266 |
| 42 | 3.357923 | 17.745037 |
| 199 | 3.353883 | 18.033340 |

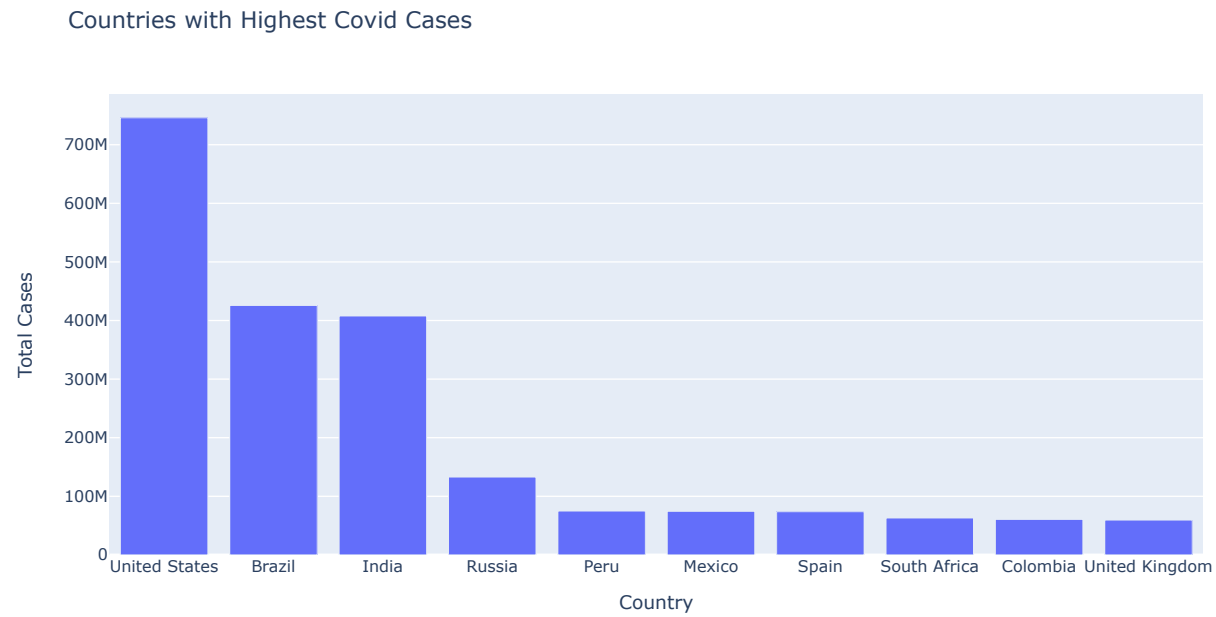
```
data["GDP Before Covid"] = [65279.53, 8897.49, 2100.75,
                             11497.65, 7027.61, 9946.03,
                             29564.74, 6001.40, 6424.98, 42354.41]
data["GDP During Covid"] = [63543.58, 6796.84, 1900.71,
                             10126.72, 6126.87, 8346.70,
                             27057.16, 5090.72, 5332.77, 40284.64]
```

```
print(data)
```

| | Country Code | Country | HDI | Total Cases | Total Deaths | \ |
|-----|--------------|----------------|----------|-------------|--------------|---|
| 200 | USA | United States | 0.924000 | 746014098.0 | 26477574.0 | |
| 27 | BRA | Brazil | 0.759000 | 425704517.0 | 14340567.0 | |
| 90 | IND | India | 0.640000 | 407771615.0 | 7247327.0 | |
| 157 | RUS | Russia | 0.816000 | 132888951.0 | 2131571.0 | |
| 150 | PER | Peru | 0.599490 | 74882695.0 | 3020038.0 | |
| 125 | MEX | Mexico | 0.774000 | 74347548.0 | 7295850.0 | |
| 178 | ESP | Spain | 0.887969 | 73717676.0 | 5510624.0 | |
| 175 | ZAF | South Africa | 0.608653 | 63027659.0 | 1357682.0 | |
| 42 | COL | Colombia | 0.581847 | 60543682.0 | 1936134.0 | |
| 199 | GBR | United Kingdom | 0.922000 | 59475032.0 | 7249573.0 | |

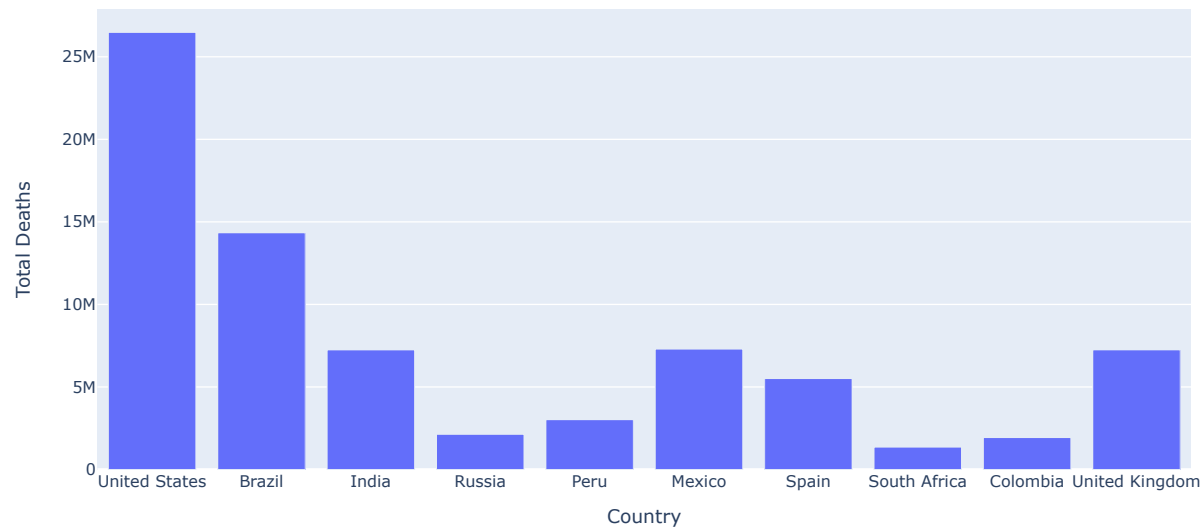
| | Stringency Index | Population | GDP Before Covid | GDP During Covid |
|-----|------------------|------------|------------------|------------------|
| 200 | 3.350949 | 19.617637 | 65279.53 | 63543.58 |
| 27 | 3.136028 | 19.174732 | 8897.49 | 6796.84 |
| 90 | 3.610552 | 21.045353 | 2100.75 | 1900.71 |
| 157 | 3.380088 | 18.798668 | 11497.65 | 10126.72 |
| 150 | 3.430126 | 17.311165 | 7027.61 | 6126.87 |
| 125 | 3.019289 | 18.674802 | 9946.03 | 8346.70 |
| 178 | 3.393922 | 17.660427 | 29564.74 | 27057.16 |
| 175 | 3.364333 | 17.898266 | 6001.40 | 5090.72 |
| 42 | 3.357923 | 17.745037 | 6424.98 | 5332.77 |
| 199 | 3.353883 | 18.033340 | 42354.41 | 40284.64 |


```
figure = px.bar(data, y='Total Cases', x='Country',  
                title="Countries with Highest Covid Cases")  
figure.show()
```

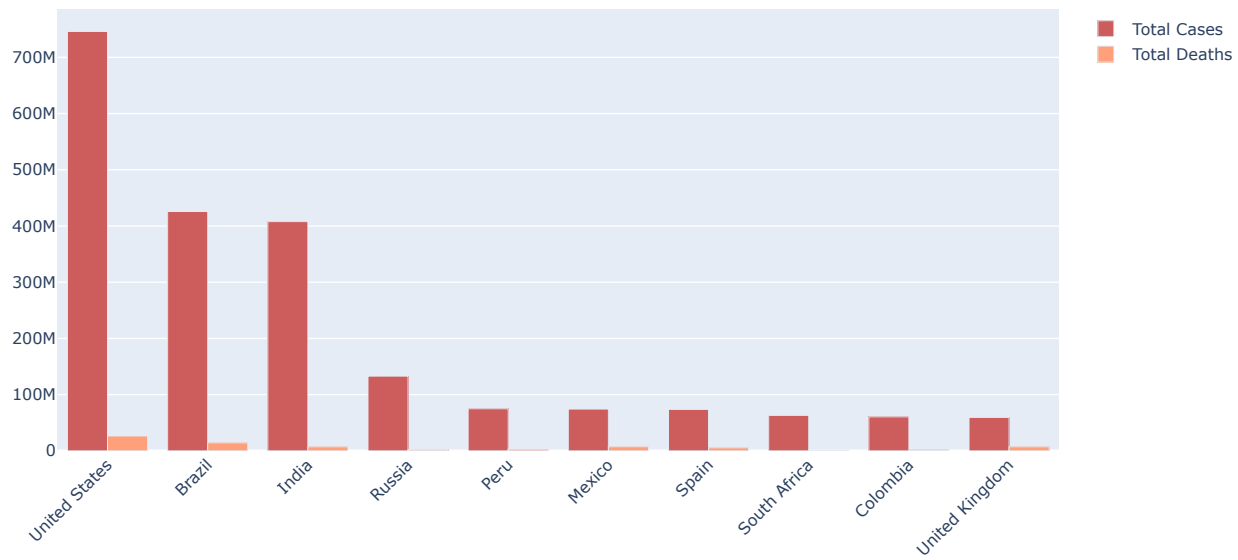


```
figure = px.bar(data, y='Total Deaths', x='Country',  
                title="Countries with Highest Deaths")  
figure.show()
```

Countries with Highest Deaths



```
fig = go.Figure()
fig.add_trace(go.Bar(
    x=data["Country"],
    y=data["Total Cases"],
    name='Total Cases',
    marker_color='indianred'
))
fig.add_trace(go.Bar(
    x=data["Country"],
    y=data["Total Deaths"],
    name='Total Deaths',
    marker_color='lightsalmon'
))
fig.update_layout(barmode='group', xaxis_tickangle=-45)
fig.show()
```

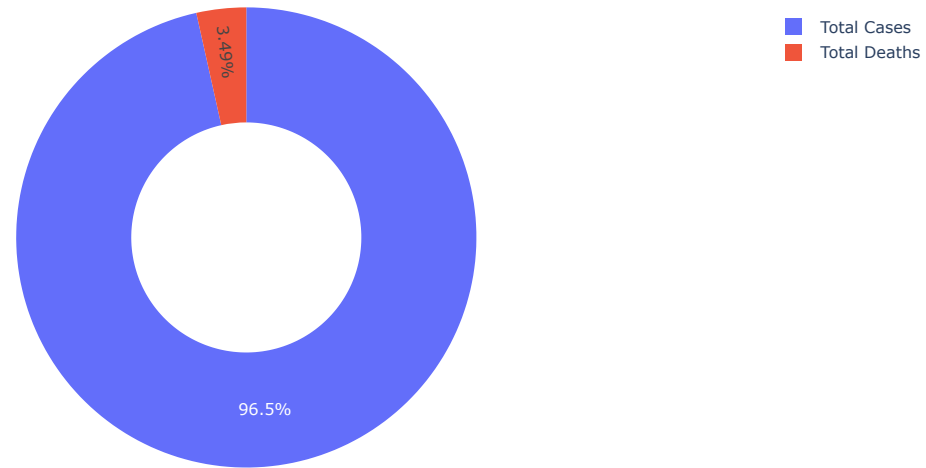


```
# Percentage of Total Cases and Deaths
cases = data["Total Cases"].sum()
deceased = data["Total Deaths"].sum()

labels = ["Total Cases", "Total Deaths"]
values = [cases, deceased]

fig = px.pie(data, values=values, names=labels,
             title='Percentage of Total Cases and Deaths', hole=0.5)
fig.show()
```

Percentage of Total Cases and Deaths

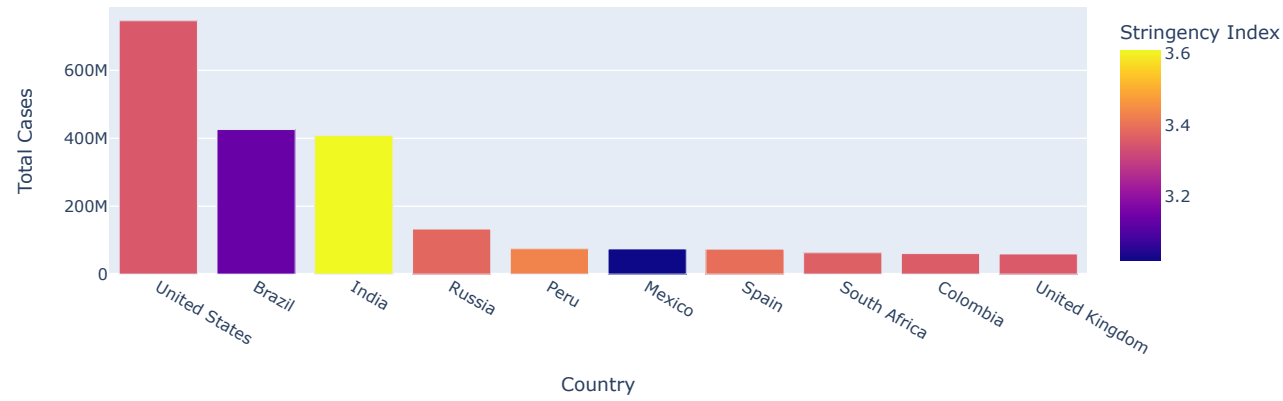


```
death_rate = (data["Total Deaths"].sum() / data["Total Cases"].sum()) * 100  
print("Death Rate = ", death_rate)
```

```
Death Rate = 3.6144212045653767
```

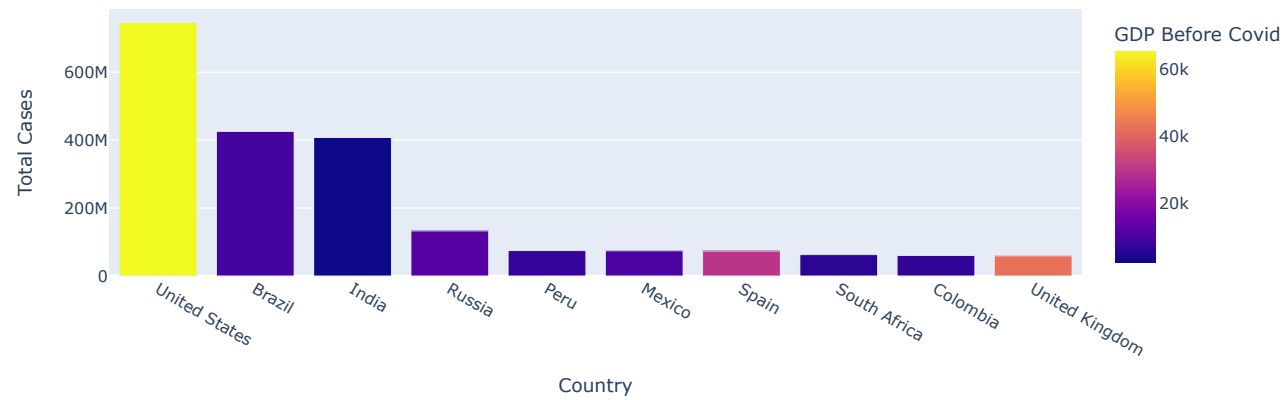
```
fig = px.bar(data, x='Country', y='Total Cases',  
             hover_data=['Population', 'Total Deaths'],  
             color='Stringency Index', height=400,  
             title= "Stringency Index during Covid-19")  
fig.show()
```

Stringency Index during Covid-19

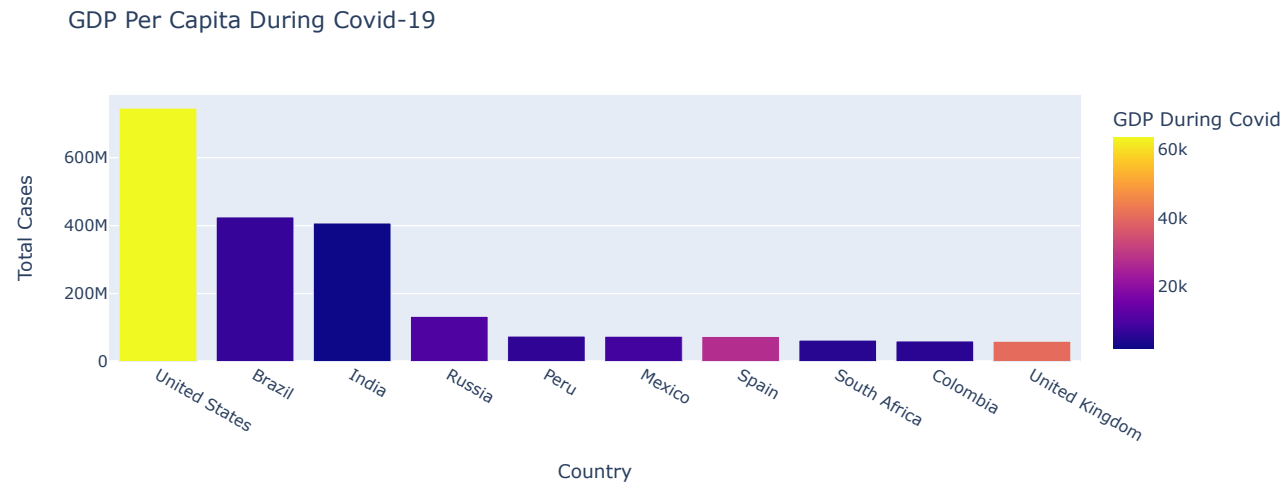


```
fig = px.bar(data, x='Country', y='Total Cases',
             hover_data=['Population', 'Total Deaths'],
             color='GDP Before Covid', height=400,
             title="GDP Per Capita Before Covid-19")
fig.show()
```

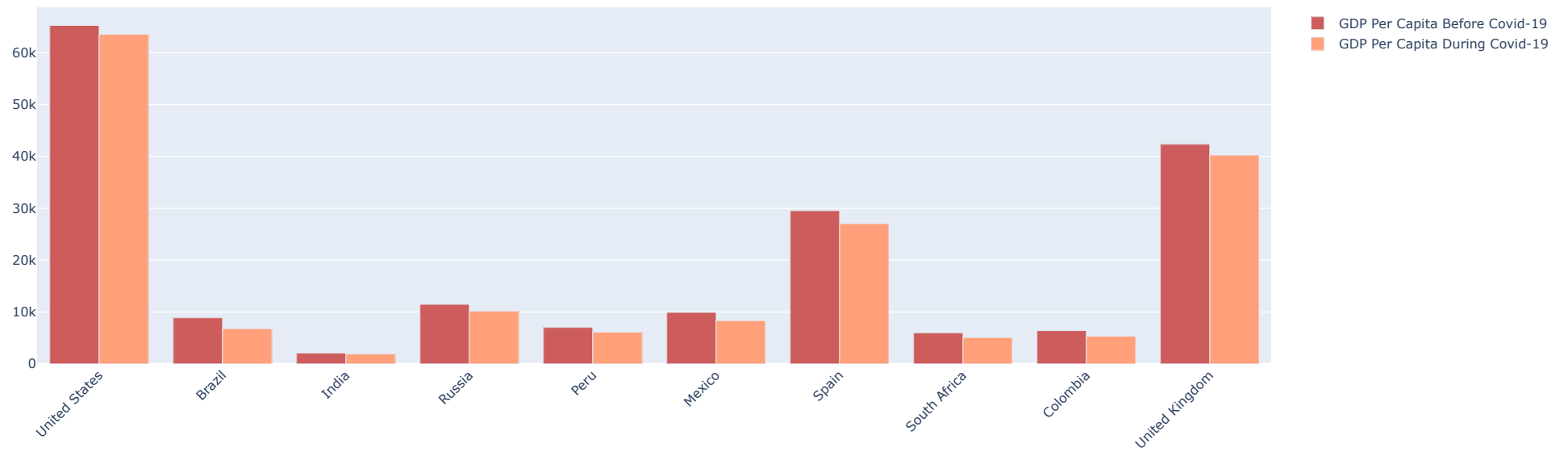
GDP Per Capita Before Covid-19



```
fig = px.bar(data, x='Country', y='Total Cases',
             hover_data=['Population', 'Total Deaths'],
             color='GDP During Covid', height=400,
             title="GDP Per Capita During Covid-19")
fig.show()
```



```
fig = go.Figure()
fig.add_trace(go.Bar(
    x=data["Country"],
    y=data["GDP Before Covid"],
    name='GDP Per Capita Before Covid-19',
    marker_color='indianred'
))
fig.add_trace(go.Bar(
    x=data["Country"],
    y=data["GDP During Covid"],
    name='GDP Per Capita During Covid-19',
    marker_color='lightsalmon'
))
fig.update_layout(barmode='group', xaxis_tickangle=-45)
fig.show()
```



```
fig = px.bar(data, x='Country', y='Total Cases',  
             hover_data=['Population', 'Total Deaths'],  
             color='HDI', height=400,  
             title="Human Development Index during Covid-19")  
fig.show()
```

```
import pandas as pd
import json

# Load the COVID-19 data
covid_data = pd.read_csv("owid-covid-data.csv")

# Assuming your data has columns named 'Date' and 'Cases'
dates = covid_data['Date']
cases = covid_data['Cases']

# Convert dates to a format suitable for d3.js
dates = pd.to_datetime(dates).dt.strftime('%Y-%m-%d')

# Create a JSON-like data structure for d3.js
data = [{'date': date, 'cases': case} for date, case in zip(dates, cases)]

# Write the data to a JSON file
with open('covid_data.json', 'w') as f:
```
