```
In [1]:
        import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        import seaborn as sns
        pd.options.display.float_format = '{:,.2f}'.format
```

/opt/conda/lib/python3.10/site-packages/scipy/\_\_init\_\_.py:146: UserWar ning: A NumPy version >=1.16.5 and <1.23.0 is required for this versio n of SciPy (detected version 1.23.5 warnings.warn(f"A NumPy version >={np\_minversion} and <{np\_maxversio</pre> n}"

In [2]:

df=pd.read\_csv("/kaggle/input/corona-virus-report/worldometer\_data.csv") df

Out[2]:

	Country/Region	Continent	Population	TotalCases	NewCases	TotalDeaths	New		
0	USA	North America	331,198,130.00	5032179	NaN	162,804.00	NaN		
1	Brazil	South America	212,710,692.00	2917562	NaN	98,644.00	NaN		
2	India	Asia	1,381,344,997.00	2025409	NaN	41,638.00	NaN		
3	Russia	Europe	145,940,924.00	871894	NaN	14,606.00	NaN		
4	South Africa	Africa	59,381,566.00	538184	NaN	9,604.00	NaN		
204	Montserrat	North America	4,992.00	13	NaN	1.00	NaN		
205	Caribbean Netherlands	North America	26,247.00	13	NaN	NaN	NaN		
206	Falkland Islands	South America	3,489.00	13	NaN	NaN	NaN		
207	Vatican City	Europe	801.00	12	NaN	NaN	NaN		
208	Western Sahara	Africa	598,682.00	10	NaN	1.00	NaN		
4				<b>←</b>					

209 rows × 16 columns

In [3]: df.shape Out[3]:

(209, 16)

```
In [4]:
        print(df.info())
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 209 entries, 0 to 208
        Data columns (total 16 columns):
             Column
                                Non-Null Count
                                                Dtype
             Country/Region
                                209 non-null
                                                object
             Continent
         1
                                208 non-null
                                                object
         2
             Population
                                208 non-null
                                                float64
         3
             TotalCases
                                209 non-null
                                                int64
             NewCases
                                4 non-null
         4
                                                float64
                                188 non-null
         5
             TotalDeaths
                                                float64
                                3 non-null
         6
             NewDeaths
                                                float64
         7
             TotalRecovered
                                205 non-null
                                                float64
                                3 non-null
         8
             NewRecovered
                                                float64
         9
                                205 non-null
             ActiveCases
                                                float64
         10
             Serious, Critical 122 non-null
                                                float64
             Tot Cases/1M pop 208 non-null
                                                float64
         11
         12
             Deaths/1M pop
                                187 non-null
                                                float64
             TotalTests
         13
                                191 non-null
                                                float64
         14
             Tests/1M pop
                                191 non-null
                                                float64
            WHO Region
                                184 non-null
         15
                                                object
        dtypes: float64(12), int64(1), object(3)
```

memory usage: 26.2+ KB

None

As you can see there are multiple null values in DataFrame but before that lets run and study some statistical summary

```
In [5]:
        df.describe()
```

Out[5]:

	Population	TotalCases	NewCases	TotalDeaths	NewDeaths	TotalRecovered
count	208.00	209.00	4.00	188.00	3.00	205.00
mean	30,415,486.97	91,718.50	1,980.50	3,792.59	300.00	58,878.98
std	104,766,099.48	432,586.68	3,129.61	15,487.18	451.20	256,698.41
min	801.00	10.00	20.00	1.00	1.00	7.00
25%	966,314.00	712.00	27.50	22.00	40.50	334.00
50%	7,041,972.50	4,491.00	656.00	113.00	80.00	2,178.00
75%	25,756,135.50	36,896.00	2,609.00	786.00	449.50	20,553.00
max	1,381,344,997.00	5,032,179.00	6,590.00	162,804.00	819.00	2,576,668.00
4						<b>&gt;</b>

### lets count null values

```
In [6]:
        df.isnull().sum()
Out[6]:
        Country/Region
                                0
        Continent
        Population
                                1
        TotalCases
                                0
        NewCases
                              205
        TotalDeaths
                               21
        NewDeaths
                              206
        TotalRecovered
                                4
        NewRecovered
                              206
        ActiveCases
                                4
        Serious, Critical
                               87
        Tot Cases/1M pop
                                1
        Deaths/1M pop
                               22
        TotalTests
                               18
        Tests/1M pop
                               18
        WHO Region
                               25
        dtype: int64
```

In such DataFrame if data is null then we have to keep it blank as it might be possible that that's true.

Let's talk about the "TotalRecovered" column. we can assume that 4 null values mean no recovery at that point in time

formatting column name for ease of use

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```
In [7]:
        lowercase_columns = [column.lower() for column in df.columns]
        df.columns = lowercase_columns
        df.head()
Out[7]:
```

	country/region	continent	population	totalcases	newcases	totaldeaths	newdeath
0	USA	North America	331,198,130.00	5032179	NaN	162,804.00	NaN
1	Brazil	South America	212,710,692.00	2917562	NaN	98,644.00	NaN
2	India	Asia	1,381,344,997.00	2025409	NaN	41,638.00	NaN
3	Russia	Europe	145,940,924.00	871894	NaN	14,606.00	NaN
4	South Africa	Africa	59,381,566.00	538184	NaN	9,604.00	NaN
4							<b>&gt;</b>

# What is the total population covered by the dataset, and which country/region and continent has the highest population?

```
In [8]:
        df['population'].sum()
Out[8]:
         6326421290.0
```

```
In [9]:
        df[['country/region','population']].sort_values(by='population',ascending
        =False)
```

Out[9]:

	country/region	population
2	India	1,381,344,997.00
0	USA	331,198,130.00
22	Indonesia	273,808,365.00
13	Pakistan	221,295,851.00
1	Brazil	212,710,692.00
•••		
205	Caribbean Netherlands	26,247.00
204	Montserrat	4,992.00
206	Falkland Islands	3,489.00
207	Vatican City	801.00
156	Diamond Princess	NaN

209 rows × 2 columns

```
In [10]:
         df.groupby('continent')['population'].sum().sort_values(ascending=False)
Out[10]:
         continent
                             3,173,656,415.00
         Asia
         Africa
                             1,343,515,489.00
                               747,677,546.00
         Europe
                               589,503,467.00
         North America
         South America
                               431,110,464.00
         Australia/Oceania
                                40,957,909.00
         Name: population, dtype: float64
```

## How many countries/regions are missing data for the "Continent" column, and can you identify them?

```
9/19/23, 7:52 PM
                                                        notebook
   In [11]:
              df[df['continent'].isna()]['country/region'].count()
    Out[11]:
              1
```

### Calculate the total number of active cases, and what is the average number of active cases per country/region?

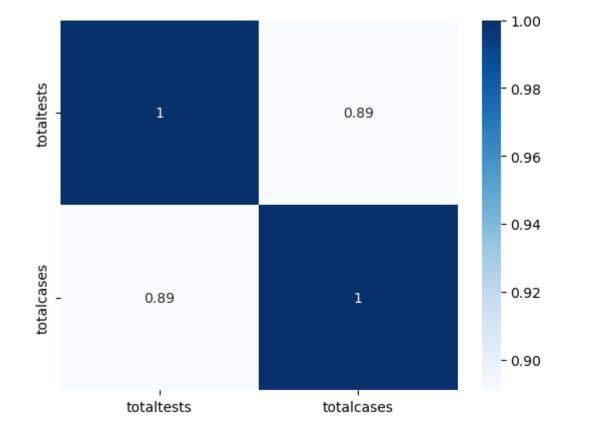
```
In [12]:
         df['activecases'].sum()
Out[12]:
         5671187.0
In [13]:
         df.groupby('country/region')['activecases'].mean().sort_values(ascending=
         False)
Out[13]:
         country/region
         USA
                             2,292,707.00
         Brazil
                                771,258.00
         India
                                606,387.00
         Russia
                                180,931.00
                                153,416.00
         Colombia
         Falkland Islands
                                      0.00
         Netherlands
                                       NaN
         Spain
                                       NaN
         Sweden
                                       NaN
         UK
                                       NaN
         Name: activecases, Length: 209, dtype: float64
```

How does the number of total tests conducted correlate with the number of total cases? Visualize the relationship.

```
In [14]:
         df_corr=df[['totaltests','totalcases']].corr()
         df_corr
Out[14]:
```

	totaltests	totalcases
totaltests	1.00	0.89
totalcases	0.89	1.00

```
In [15]:
         sns.heatmap(data=df_corr,annot=True,cmap='Blues')
Out[15]:
         <Axes: >
```



What is the distribution of "Serious, Critical" cases by continent, and which continent has the highest average?

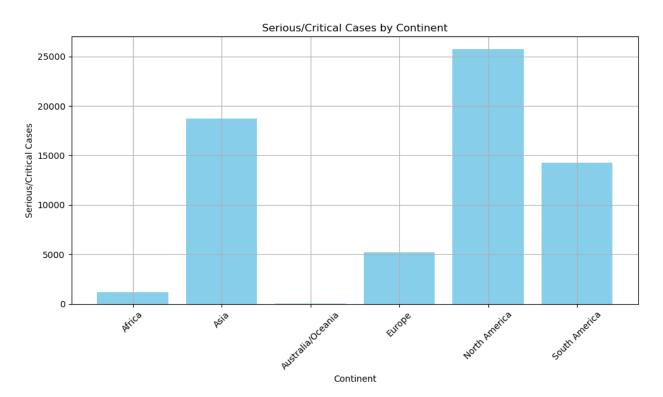
```
In [16]:
```

cases\_dis=df.groupby('continent')['serious,critical'].sum().reset\_index() cases\_dis

#### Out[16]:

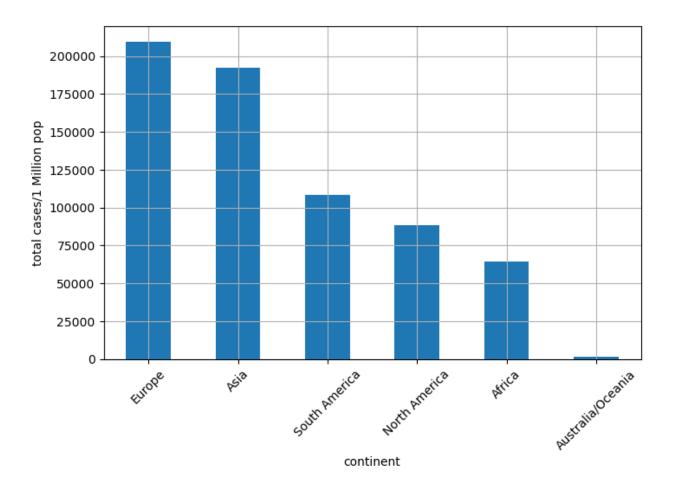
	continent	serious,critical
0	Africa	1,187.00
1	Asia	18,749.00
2	Australia/Oceania	52.00
3	Europe	5,200.00
4	North America	25,709.00
5	South America	14,295.00

```
In [17]:
         plt.figure(figsize=(10, 6))
         plt.bar(cases_dis['continent'], cases_dis['serious,critical'], color='sky
         blue')
         plt.xlabel('Continent')
         plt.ylabel('Serious/Critical Cases')
         plt.title('Serious/Critical Cases by Continent')
         plt.xticks(rotation=45)
         plt.grid()
         plt.tight_layout()
         plt.show()
```



#### Continents with the Highest Cases per Million Population:

```
In [18]:
         plt.figure(figsize=(8,5))
         df.groupby('continent')['tot cases/1m pop'].sum().sort_values(ascending=F
         alse).plot(kind='bar')
         plt.ylabel("total cases/1 Million pop")
         plt.xticks(rotation=45)
         plt.grid()
```



Identify the continents with the highest and lowest average cases per million population.

```
In [19]:
         df.groupby('continent')['tot cases/1m pop'].mean().sort_values(ascending=
         False)
Out[19]:
         continent
                             7,745.79
         South America
                             4,363.62
         Europe
                             4,008.94
         Asia
                             2,529.91
         North America
                             1,130.81
         Africa
                               241.00
         Australia/Oceania
         Name: tot cases/1m pop, dtype: float64
```

Calculate the percentage of serious/critical cases relative to the total cases for each country/region. and Identify countries/regions where a high percentage of cases are serious or critical.

```
In [20]:
```

```
df['serious % of total']=(df['serious,critical']/df['totalcases'])*100
x=df[['country/region','totalcases','serious % of total']].dropna()
x.sort_values(by='serious % of total',ascending=False)
```

#### Out[20]:

	country/region	totalcases	serious % of total
68	El Salvador	19126	2.66
189	Belize	86	2.33
183	Turks and Caicos	129	2.33
23	Canada	118561	1.91
192	Saint Martin	53	1.89
•••			
50	Ghana	39642	0.02
15	Italy	249204	0.02
45	Nigeria	45244	0.02
73	Denmark	14306	0.01
42	Guatemala	54339	0.01

122 rows × 3 columns

In [ ]:			