

Covid-19 Data Analysis

1

Download the data from the given link

2

Create a DataFrame from the CSV file.

In [104]:

```
1 import pandas as pd
2 data = pd.read_csv("time-series-19-covid-combined.csv")
3 df=pd.DataFrame(data)
```

In [105]:

```
1 df
```

Out[105]:

| | Date | Country/Region | Province/State | Confirmed | Recovered | Deaths |
|--------|------------|----------------|----------------|-----------|-----------|--------|
| 0 | 2020-01-22 | Afghanistan | NaN | 0 | 0.0 | 0 |
| 1 | 2020-01-23 | Afghanistan | NaN | 0 | 0.0 | 0 |
| 2 | 2020-01-24 | Afghanistan | NaN | 0 | 0.0 | 0 |
| 3 | 2020-01-25 | Afghanistan | NaN | 0 | 0.0 | 0 |
| 4 | 2020-01-26 | Afghanistan | NaN | 0 | 0.0 | 0 |
| ... | ... | ... | ... | ... | ... | ... |
| 154561 | 2021-07-24 | Zimbabwe | NaN | 97277 | 64628.0 | 3050 |
| 154562 | 2021-07-25 | Zimbabwe | NaN | 97894 | 65913.0 | 3094 |
| 154563 | 2021-07-26 | Zimbabwe | NaN | 99944 | 67827.0 | 3173 |
| 154564 | 2021-07-27 | Zimbabwe | NaN | 101711 | 70496.0 | 3280 |
| 154565 | 2021-07-28 | Zimbabwe | NaN | 103567 | 71383.0 | 3340 |

154566 rows × 6 columns

3

Merge the data for countries with multiple regions in order to provide a single time-series for each country.

In [89]:

```
1 df_new = df.groupby(['Date', 'Country/Region'])["Confirmed", "Recovered", 'Deaths'].app
```

In [90]:

```
1 df_new
```

Out[90]:

| | | Confirmed | Recovered | Deaths |
|------------|--------------------|-----------|-----------|--------|
| Date | Country/Region | | | |
| 2020-01-22 | Afghanistan | 0.0 | 0.0 | 0.0 |
| | Albania | 0.0 | 0.0 | 0.0 |
| | Algeria | 0.0 | 0.0 | 0.0 |
| | Andorra | 0.0 | 0.0 | 0.0 |
| | Angola | 0.0 | 0.0 | 0.0 |
| ... | ... | ... | ... | ... |
| 2021-07-28 | Vietnam | 123640.0 | 27457.0 | 630.0 |
| | West Bank and Gaza | 316328.0 | 311622.0 | 3601.0 |
| | Yemen | 7027.0 | 4168.0 | 1374.0 |
| | Zambia | 193432.0 | 183957.0 | 3338.0 |
| | Zimbabwe | 103567.0 | 71383.0 | 3340.0 |

108030 rows × 5 columns

4

Print the total number of confirmed cases and number of deaths in each country in the last reported day. What are the 10 countries with the highest number of confirmed COVID-19 cases? What are the 10 countries with the highest number of deaths?

In [37]:

```
1 df_new.reset_index(level=0, inplace=True)
2 latest=df_new['Date'].max()
3 df_1 = df_new[df_new['Date'] == latest]
```

In [38]:



```
1 df_1
```

Out[38]:

| | Date | Confirmed | Recovered | Deaths |
|---------------------------|------------|-----------|-----------|--------|
| Country/Region | | | | |
| Afghanistan | 2021-07-28 | 145552.0 | 82586.0 | 6577.0 |
| Albania | 2021-07-28 | 132952.0 | 130174.0 | 2457.0 |
| Algeria | 2021-07-28 | 167131.0 | 112900.0 | 4161.0 |
| Andorra | 2021-07-28 | 14586.0 | 14113.0 | 127.0 |
| Angola | 2021-07-28 | 42288.0 | 35964.0 | 1000.0 |
| ... | ... | ... | ... | ... |
| Vietnam | 2021-07-28 | 123640.0 | 27457.0 | 630.0 |
| West Bank and Gaza | 2021-07-28 | 316328.0 | 311622.0 | 3601.0 |
| Yemen | 2021-07-28 | 7027.0 | 4168.0 | 1374.0 |
| Zambia | 2021-07-28 | 193432.0 | 183957.0 | 3338.0 |
| Zimbabwe | 2021-07-28 | 103567.0 | 71383.0 | 3340.0 |

195 rows × 4 columns

In [21]:



```
1 df_1['Confirmed'].nlargest(10)
```

Out[21]:

```
Country/Region
US                34672690.0
India             31484605.0
Brazil            19797086.0
France            6116853.0
Russia            6116249.0
United Kingdom    5797445.0
Turkey            5660469.0
Argentina         4891810.0
Colombia          4757139.0
Spain             4395602.0
Name: Confirmed, dtype: float64
```

In [22]:



```
1 df_1['Deaths'].nlargest(10)
```

Out[22]:

| Country/Region | |
|----------------|----------|
| US | 611801.0 |
| Brazil | 553179.0 |
| India | 422022.0 |
| Mexico | 239616.0 |
| Peru | 196138.0 |
| Russia | 153620.0 |
| United Kingdom | 129718.0 |
| Italy | 128010.0 |
| Colombia | 119801.0 |
| France | 111923.0 |

Name: Deaths, dtype: float64

5

Plot a graph of the number of confirmed cases over time for each country. Which countries present exponential growth in the number of cases and which countries are already leaving exponential growth?

In []:



```
1 # dividing the column into 16 equal arrays for subplotting purposes
```

In [91]:



```
1 df_new.reset_index(level=0, inplace=True)
```

In [93]:



```
1 import numpy as np
2 df1, df2, df3, df4, df5, df6, df7, df8 = np.array_split(df_new[['Date', 'Confirmed']], 8)
```

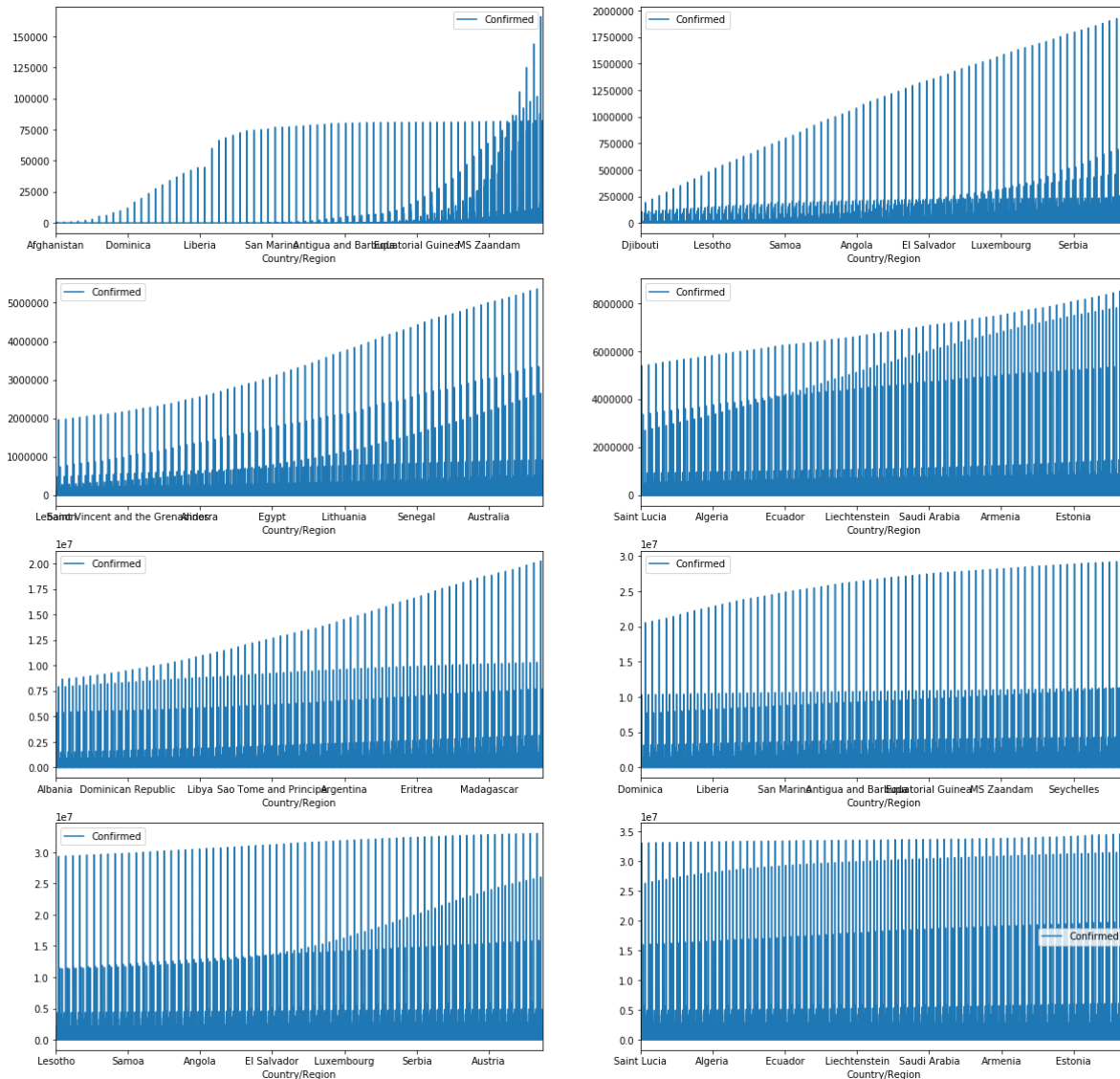
In [94]:



```

1 # plotting the graphs together by using subplot method
2 df_list = [df1 ,df2, df3, df4, df5, df6, df7, df8]
3 fig, axes = plt.subplots(4, 2)
4
5 #counter
6 count=0
7 for i in range(4):
8     for j in range(2):
9         df_list[count].plot(ax=axes[i,j], figsize=(20,20))
10        count+=1

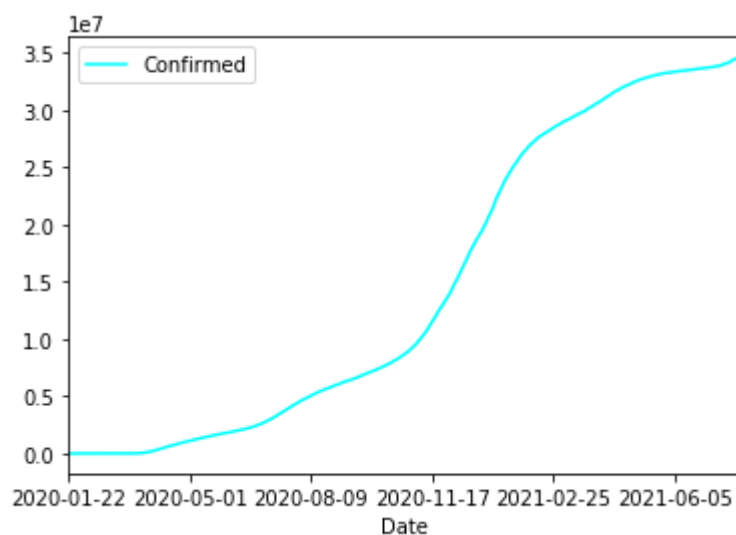
```



In [106]:



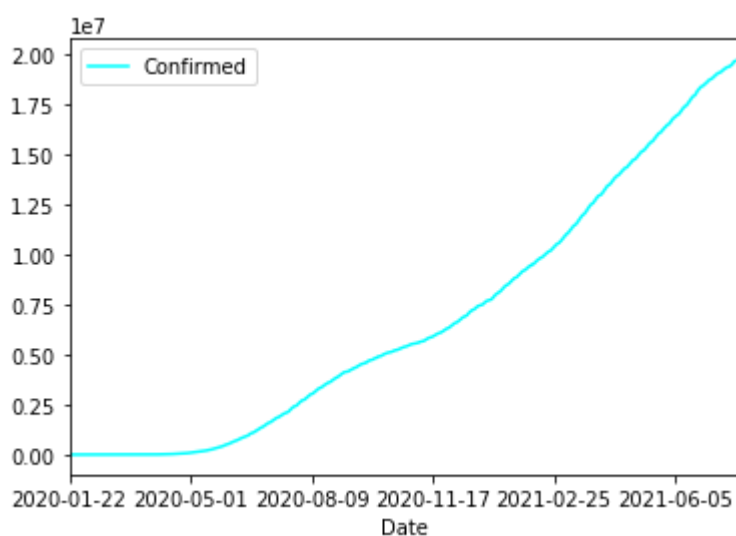
```
1 # to make plots look comprehensible, i tried plotting them one at a time for one country
2
3 import matplotlib.pyplot as plt
4 df_US=df[df['Country/Region'] == 'US']
5 df_US.plot( 'Date' , 'Confirmed', color='cyan' )
6 plt.show()
```



In [108]:



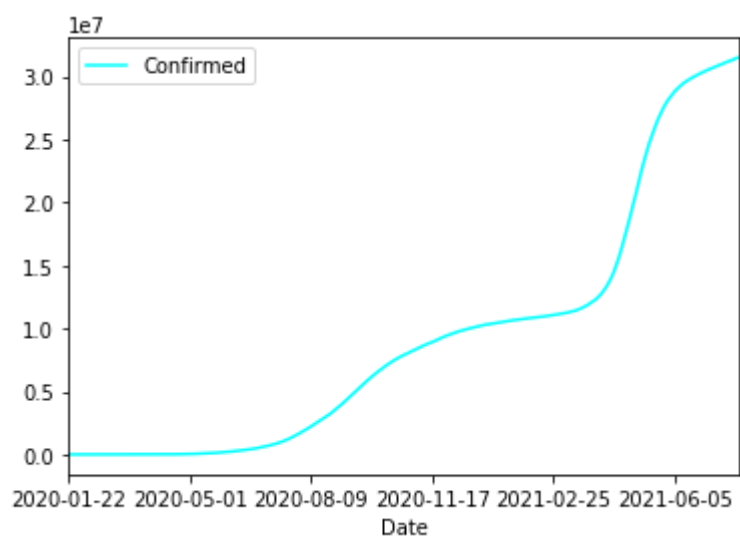
```
1 df_Brazil=df[df['Country/Region'] == 'Brazil']
2 df_Brazil.plot( 'Date' , 'Confirmed', color='cyan' )
3 plt.show()
```



In [109]:



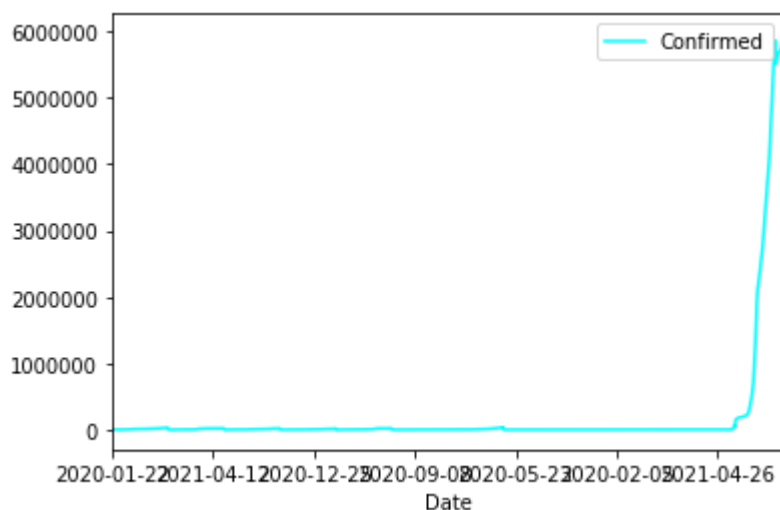
```
1 df_India=df[df['Country/Region'] == 'India']  
2 df_India.plot( 'Date' , 'Confirmed', color='cyan' )  
3 plt.show()
```



In [110]:

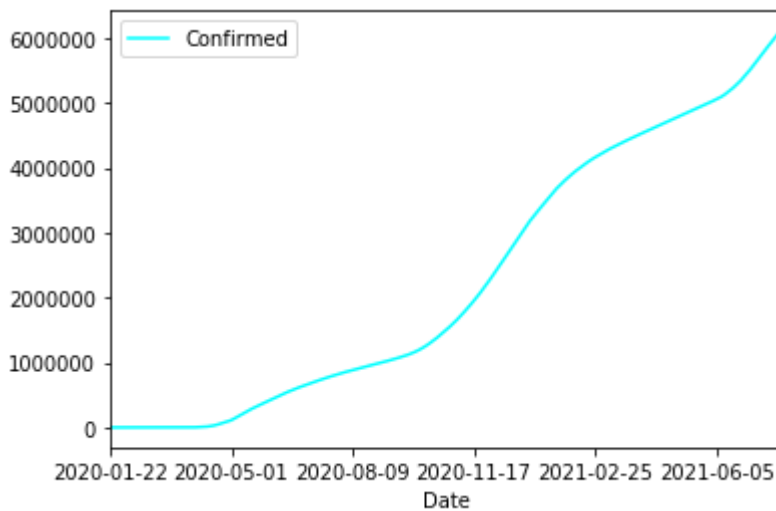


```
1 df_France=df[df['Country/Region'] == 'France']  
2 df_France.plot( 'Date' , 'Confirmed', color='cyan' )  
3 plt.show()
```



In [111]:

```
1 df_Russia=df[df['Country/Region'] == 'Russia']
2 df_Russia.plot( 'Date' , 'Confirmed', color='cyan' )
3 plt.show()
```



Through these plots we found out that, while countries like US, Brazil, India, Russia had a similar increase in the covid confirmed cases throughout the duration, there were countries like France which encountered a sudden upheaval in the cases around April, 2021.

6

Create a bar plot that shows the number of deaths per 100 confirmed cases (observed case-fatality ratio) for the 20 most affected countries. See Figure 1 for an example.

In [39]:

```
1 df_6=df_1
2 df_6.fillna(0)
3 df_6['Death_percent_confirmed']=df_6['Deaths']*100/df_6['Confirmed']
4 del df_6['Date']
```


In [46]:

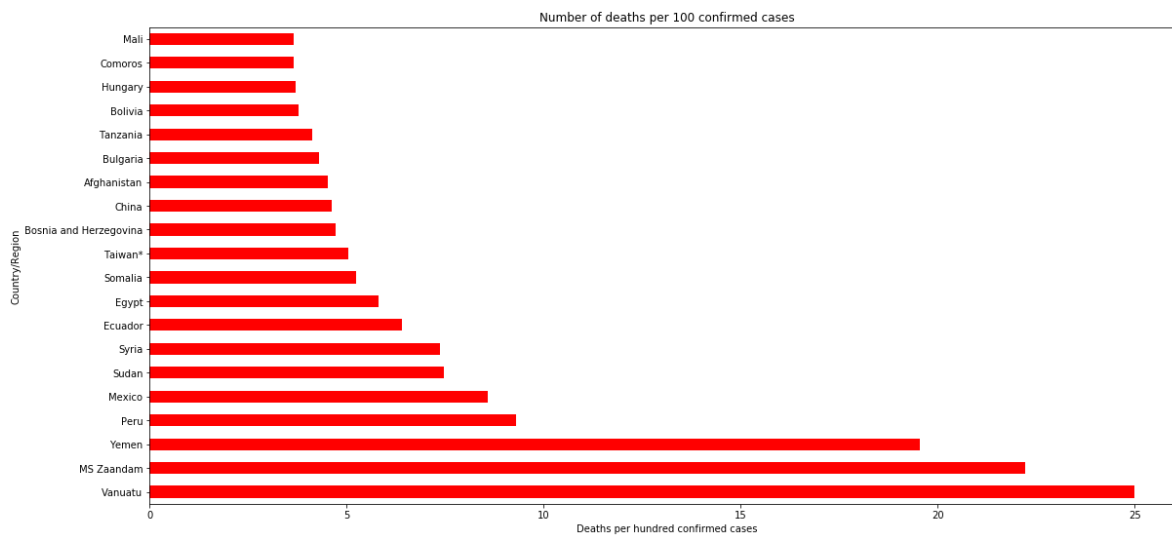
```

1 # plotting a horizontal bar graph
2 import matplotlib.pyplot as plt
3
4 plt.title('Number of deaths per 100 confirmed cases')
5 plt.xlabel('Deaths per hundred confirmed cases')
6 plt.ylabel('Country/Region')
7 df_6['Death_percent_confirmed'].nlargest(20).plot.barh(figsize=(19,9), color='red')

```

Out[46]:

<matplotlib.axes._subplots.AxesSubplot at 0x14764aa1248>



7

Compute the ratio between the total number of confirmed cases and the population size of each country. The file `worldpopulation.json` contains data on the population size of each country.

Data Cleaning

In [49]:

```

1 population=pd.read_json('worldpopulation.json')
2 df_6.reset_index(level=0, inplace=True)
3 df_7=df_6

```

In [50]:

```

1 # Finding out the non-matching countries in the two datasets
2 x=pd.DataFrame()
3 x=population['country'].sort_values()
4 x=x.reset_index(drop=True)
5 data=[df_7['Country/Region'], x]
6 names = ["country_1", "country_2"]
7 df_check= pd.concat(data, axis=1, keys=names)
8 print(list(set(df_check.country_1)-set(df_check.country_2)))

```

['West Bank and Gaza', 'Taiwan*', 'US', 'Burma', 'Summer Olympics 2020', 'Saint Vincent and the Grenadines', 'Vietnam', 'Kosovo', 'United Kingdom', 'Cote d'Ivoire', 'Congo (Brazzaville)', 'Eswatini', 'Korea, South', 'North Macedonia', 'Diamond Princess', 'Congo (Kinshasa)', 'MS Zaandam', 'Czechia']

In [52]:

```

1 # preprocessing the data
2 # changing the names of countries so that they match
3 update_df=pd.DataFrame()
4 update_df['Country/Region']=df_7['Country/Region'].replace("Czechia", "Czech Republic")
5 update_df['Country/Region']=df_7['Country/Region'].replace("North Macedonia", "North Korea")
6 update_df['Country/Region']=df_7['Country/Region'].replace("Taiwan*", "TFYR Macedonia")
7 update_df['Country/Region']=df_7['Country/Region'].replace("US", "U.S.", inplace=True)
8 update_df['Country/Region']=df_7['Country/Region'].replace("United Kingdom", "U.K.", inplace=True)
9 update_df['Country/Region']=df_7['Country/Region'].replace("Vietnam", "Viet Nam", inplace=True)
10
11 # removing the rows with countries that are only present in one dataset
12 update_df = df_7.drop(df_7.index[[40, 42, 48, 58, 92, 93, 106, 148, 167]])
13 update_pop= population.drop(population.index[[54, 45, 15, 24, 118, 25, 150, 178, 180, 181]])
14
15 update_df = update_df.reset_index(drop=True)
16 update_pop = update_pop.reset_index(drop=True)
17 update_df = update_df.set_index('Country/Region')
18 update_df = update_df.reindex(index=population['country'])
19 update_df = update_df.reset_index()
20 update_df= update_df.dropna()
21 update_df = update_df.reset_index(drop=True)
22 update_pop= update_pop.dropna()
23 update_pop = update_pop.reset_index(drop=True)
24
25 # creating a new dataframe to check whether all the country names are same or not
26 data2=[update_df['country'], update_pop['country']]
27 names2= ["country_1", "country_2"]
28 df_check2= pd.concat(data2, axis=1, keys=names2)

```

In [55]:



```

1 # displaying the dataframe containing both the columns of countries to compare them
2 with pd.option_context('display.max_rows', None,
3                          'display.max_columns', None,
4                          'display.precision', 3,
5                          ):
6     print(df_check2)

```

| | country_1 | country_2 |
|----|--------------|--------------|
| 0 | China | China |
| 1 | India | India |
| 2 | U.S. | U.S. |
| 3 | Indonesia | Indonesia |
| 4 | Brazil | Brazil |
| 5 | Pakistan | Pakistan |
| 6 | Nigeria | Nigeria |
| 7 | Bangladesh | Bangladesh |
| 8 | Russia | Russia |
| 9 | Mexico | Mexico |
| 10 | Japan | Japan |
| 11 | Ethiopia | Ethiopia |
| 12 | Philippines | Philippines |
| 13 | Viet Nam | Viet Nam |
| 14 | Egypt | Egypt |
| 15 | Iran | Iran |
| 16 | Germany | Germany |
| 17 | Turkey | Turkey |
| 18 | Thailand | Thailand |
| 19 | U.K. | U.K. |
| 20 | France | France |
| 21 | Italy | Italy |
| 22 | Tanzania | Tanzania |
| 23 | South Africa | South Korea |
| 24 | Colombia | Colombia |
| 25 | Kenya | Kenya |
| 26 | Spain | Spain |
| 27 | Ukraine | Ukraine |
| 28 | Argentina | Argentina |
| 29 | Sudan | Sudan |
| 30 | Uganda | Uganda |
| 31 | Algeria | Algeria |
| 32 | Iraq | Iraq |
| 33 | Poland | Poland |
| 34 | Canada | Canada |
| 35 | Morocco | Morocco |
| 36 | Afghanistan | Afghanistan |
| 37 | Saudi Arabia | Saudi Arabia |
| 38 | Peru | Peru |
| 39 | Venezuela | Venezuela |
| 40 | Malaysia | Malaysia |
| 41 | Uzbekistan | Uzbekistan |
| 42 | Mozambique | Nepal |
| 43 | Nepal | Ghana |
| 44 | Ghana | Yemen |
| 45 | Yemen | Angola |
| 46 | Angola | Madagascar |
| 47 | North Korea | North Korea |
| 48 | Australia | Australia |
| 49 | Cameroon | Cameroon |

| | | |
|-----|----------------------|----------------------|
| 50 | Niger | Niger |
| 51 | Sri Lanka | Sri Lanka |
| 52 | Romania | Romania |
| 53 | Burkina Faso | Burkina Faso |
| 54 | Syria | Syria |
| 55 | Mali | Mali |
| 56 | Chile | Chile |
| 57 | Malawi | Malawi |
| 58 | Kazakhstan | Kazakhstan |
| 59 | Zambia | Zambia |
| 60 | Netherlands | Netherlands |
| 61 | Guatemala | Guatemala |
| 62 | Ecuador | Ecuador |
| 63 | Zimbabwe | Zimbabwe |
| 64 | Cambodia | Cambodia |
| 65 | Senegal | Senegal |
| 66 | Chad | Chad |
| 67 | Guinea | Guinea |
| 68 | South Sudan | South Sudan |
| 69 | Rwanda | Rwanda |
| 70 | Burundi | Burundi |
| 71 | Tunisia | Tunisia |
| 72 | Benin | Benin |
| 73 | Belgium | Belgium |
| 74 | Somalia | Somalia |
| 75 | Cuba | Cuba |
| 76 | Bolivia | Bolivia |
| 77 | Haiti | Haiti |
| 78 | Greece | Greece |
| 79 | Dominican Republic | Dominican Republic |
| 80 | Czech Republic | Czech Republic |
| 81 | Portugal | Portugal |
| 82 | Azerbaijan | Azerbaijan |
| 83 | Sweden | Sweden |
| 84 | Hungary | Hungary |
| 85 | Belarus | Belarus |
| 86 | United Arab Emirates | United Arab Emirates |
| 87 | Tajikistan | Tajikistan |
| 88 | Serbia | Serbia |
| 89 | Austria | Austria |
| 90 | Switzerland | Switzerland |
| 91 | Israel | Israel |
| 92 | Honduras | Honduras |
| 93 | Papua New Guinea | Papua New Guinea |
| 94 | Jordan | Jordan |
| 95 | Togo | Togo |
| 96 | Bulgaria | Bulgaria |
| 97 | Laos | Laos |
| 98 | Paraguay | Paraguay |
| 99 | Sierra Leone | Sierra Leone |
| 100 | Libya | Libya |
| 101 | Nicaragua | Nicaragua |
| 102 | El Salvador | El Salvador |
| 103 | Kyrgyzstan | Lebanon |
| 104 | Lebanon | Singapore |
| 105 | Singapore | Denmark |
| 106 | Denmark | Finland |
| 107 | Finland | Turkmenistan |
| 108 | Eritrea | Eritrea |
| 109 | Slovakia | Slovakia |
| 110 | Norway | Norway |

| | | |
|-----|--------------------------|--------------------------|
| 111 | Central African Republic | Central African Republic |
| 112 | Costa Rica | Costa Rica |
| 113 | Ireland | Congo |
| 114 | Oman | Ireland |
| 115 | Liberia | Oman |
| 116 | New Zealand | Liberia |
| 117 | Mauritania | New Zealand |
| 118 | Croatia | Mauritania |
| 119 | Kuwait | Croatia |
| 120 | Moldova | Kuwait |
| 121 | Panama | Moldova |
| 122 | Georgia | Panama |
| 123 | Bosnia and Herzegovina | Georgia |
| 124 | Uruguay | Bosnia and Herzegovina |
| 125 | Mongolia | Uruguay |
| 126 | Armenia | Mongolia |
| 127 | Albania | Armenia |
| 128 | Lithuania | Albania |
| 129 | Jamaica | Lithuania |
| 130 | Namibia | Jamaica |
| 131 | Botswana | Namibia |
| 132 | Qatar | Botswana |
| 133 | Lesotho | Qatar |
| 134 | Gambia | Lesotho |
| 135 | TFYR Macedonia | Gambia |
| 136 | Slovenia | TFYR Macedonia |
| 137 | Latvia | Slovenia |
| 138 | Guinea-Bissau | Latvia |
| 139 | Gabon | Guinea-Bissau |
| 140 | Bahrain | Bahrain |
| 141 | Trinidad and Tobago | Trinidad and Tobago |
| 142 | Estonia | Estonia |
| 143 | Mauritius | Mauritius |
| 144 | Timor-Leste | Timor-Leste |
| 145 | Cyprus | Cyprus |
| 146 | Djibouti | Djibouti |
| 147 | Fiji | Fiji |
| 148 | Equatorial Guinea | Equatorial Guinea |
| 149 | Comoros | Comoros |
| 150 | Bhutan | Bhutan |
| 151 | Guyana | Guyana |
| 152 | Montenegro | Montenegro |
| 153 | Solomon Islands | Solomon Islands |
| 154 | Luxembourg | Luxembourg |
| 155 | Suriname | Suriname |
| 156 | Cabo Verde | Cabo Verde |
| 157 | Brunei | Brunei |
| 158 | Malta | Malta |
| 159 | Bahamas | Bahamas |
| 160 | Maldives | Maldives |
| 161 | Belize | Belize |
| 162 | Iceland | Iceland |
| 163 | Barbados | Barbados |
| 164 | Vanuatu | Vanuatu |
| 165 | Sao Tome and Principe | Sao Tome and Principe |
| 166 | Samoa | Samoa |
| 167 | Saint Lucia | Saint Lucia |
| 168 | Kiribati | Kiribati |
| 169 | Grenada | Grenada |
| 170 | Micronesia | Micronesia |
| 171 | Seychelles | Seychelles |

| | | |
|-----|-----------------------|-----------------------|
| 172 | Antigua and Barbuda | Antigua and Barbuda |
| 173 | Dominica | Dominica |
| 174 | Andorra | Andorra |
| 175 | Saint Kitts and Nevis | Saint Kitts and Nevis |
| 176 | Marshall Islands | Marshall Islands |
| 177 | Liechtenstein | Liechtenstein |
| 178 | Monaco | Monaco |
| 179 | San Marino | San Marino |
| 180 | Holy See | Palau |
| 181 | NaN | Holy See |

Finding the Ratio

In [59]:

```
1 df_ratio=pd.DataFrame()
2 df_ratio['Cases_per_capita']=update_df['Confirmed']/update_pop['population']
3 df_ratio['Country']=update_pop['country']
```

In [60]:

```
1 df_ratio
```

Out[60]:

| | Cases_per_capita | Country |
|-----|------------------|---------------|
| 0 | 0.000076 | China |
| 1 | 0.023452 | India |
| 2 | 0.106204 | U.S. |
| 3 | 0.012477 | Indonesia |
| 4 | 0.093717 | Brazil |
| ... | ... | ... |
| 177 | 0.081085 | Liechtenstein |
| 178 | 0.074533 | Monaco |
| 179 | 0.159793 | San Marino |
| 180 | 0.001243 | Palau |
| 181 | NaN | Holy See |

182 rows × 2 columns

What are the 10 countries with the highest number of confirmed COVID-19 cases per capita?

In [57]:

```
1 df_per_capita=pd.DataFrame()
2 df_per_capita['Cases_per_capita']=update_df['Confirmed']*100/update_pop['population']
3 df_per_capita['Country']=update_pop['country']
```

In [58]:

```
1 df_per_capita.nlargest(10, 'Cases_per_capita')
```

Out[58]:

| | Cases_per_capita | Country |
|-----|------------------|----------------|
| 174 | 21.222791 | Andorra |
| 160 | 20.510979 | Maldives |
| 140 | 18.947561 | Bahrain |
| 171 | 18.517721 | Seychelles |
| 152 | 16.206946 | Montenegro |
| 179 | 15.979317 | San Marino |
| 80 | 15.850274 | Czech Republic |
| 154 | 12.625171 | Luxembourg |
| 136 | 12.425623 | TFYR Macedonia |
| 83 | 11.075866 | Sweden |

8

In this part we would like to test the hypothesis that the spread of the virus is slowed down by warm weather. Plot a graph of the monthly number of confirmed cases vs. the average monthly temperature for a few selected countries, and analyze the correlation between these two factors.

Data Processing

In [64]:

```
1 import json
2 from pandas.io.json import json_normalize
3 with open('climate.json') as file:
4     data=json.load(file)
5 climate=json_normalize(data, 'monthlyAvg', ['country'], errors='ignore')
6 climate.columns=['high', 'low', "dryDays", "snowDays", 'rainfall', 'Country/Region']
7 cl=pd.DataFrame(climate)
```

In [65]:

```
1 # taking the average of all the temperatures of each country
2 average_high_temp=[]
3 for j in cl['Country/Region']:
4     cl_new= cl[cl['Country/Region'] == j]
5     cl_new=cl.groupby(['Country/Region'])['high'].mean().sort_values().reset_index(name='high')
6     cl_new2=cl.groupby(['Country/Region'])['low'].mean().sort_values().reset_index(name='low')
7     cl_merge=pd.merge(cl_new, cl_new2, on= 'Country/Region')
```

In [66]:



```
1 # finding the confirmed cases per month and averaging all the confirmed cases per month
2 monthly_confirmed=[]
3 for i in df_1['Country/Region']:
4     df_us = df[df['Country/Region'] == i]
5     df_us['Date'] = pd.to_datetime(df['Date'])
6     df_us=df_us.groupby(df_us['Date'].dt.strftime('%B'))['Confirmed'].sum().sort_values
7     x=df_us.mean()
8     monthly_confirmed.append(x)
```

...

In [67]:



```
1 # merging all the processed data into a single DataFrame
2 df_climate=pd.DataFrame()
3 df_climate['Country/Region']=df_1['Country/Region']
4 df_climate['average_monthly_confirmed']=monthly_confirmed
5
6 final_data=pd.merge(df_climate, cl_merge, on='Country/Region')
7 final_data['avg_temp']=(final_data['high']+final_data['low'])/2
```


In [68]:



| | |
|---|------------|
| 1 | final_data |
|---|------------|

Out[68]:

| | Country/Region | average_monthly_confirmed | high | low | avg_temp |
|----|----------------|---------------------------|-----------|-----------|-----------|
| 0 | Argentina | 6.408577e+07 | 22.166667 | 14.500000 | 18.333333 |
| 1 | Australia | 9.423037e+05 | 22.041667 | 11.104167 | 16.572917 |
| 2 | Austria | 1.141650e+07 | 14.916667 | 5.583333 | 10.250000 |
| 3 | Belgium | 2.035460e+07 | 14.416667 | 7.000000 | 10.708333 |
| 4 | Brazil | 3.127759e+08 | 29.055556 | 18.916667 | 23.986111 |
| 5 | Bulgaria | 6.818600e+06 | 16.416667 | 4.833333 | 10.625000 |
| 6 | Canada | 2.324708e+07 | 12.130952 | 1.797619 | 6.964286 |
| 7 | Chile | 2.708727e+07 | 23.666667 | 7.333333 | 15.500000 |
| 8 | China | 4.143917e+06 | 19.625000 | 10.375000 | 15.000000 |
| 9 | Czech Republic | 0.000000e+00 | 13.833333 | 5.083333 | 9.458333 |
| 10 | Denmark | 5.041026e+06 | 12.583333 | 5.750000 | 9.166667 |
| 11 | France | 1.005386e+08 | 18.041667 | 8.791667 | 13.416667 |
| 12 | Germany | 6.234787e+07 | 13.250000 | 6.000000 | 9.625000 |
| 13 | Greece | 5.864084e+06 | 19.666667 | 13.666667 | 16.666667 |
| 14 | Hungary | 1.228024e+07 | 16.000000 | 6.750000 | 11.375000 |
| 15 | Iceland | 1.776009e+05 | 8.000000 | 2.416667 | 5.208333 |
| 16 | India | 4.198417e+08 | 30.472222 | 19.333333 | 24.902778 |
| 17 | Indonesia | 3.362925e+07 | 31.666667 | 23.000000 | 27.333333 |
| 18 | Ireland | 4.956613e+06 | 12.416667 | 5.916667 | 9.166667 |
| 19 | Israel | 1.699681e+07 | 23.458333 | 14.333333 | 18.895833 |
| 20 | Italy | 7.437087e+07 | 20.250000 | 10.083333 | 15.166667 |
| 21 | Japan | 1.146843e+07 | 20.875000 | 11.583333 | 16.229167 |
| 22 | Malaysia | 8.298775e+06 | 32.833333 | 24.000000 | 28.416667 |
| 23 | Mexico | 5.108411e+07 | 24.083333 | 10.500000 | 17.291667 |
| 24 | Morocco | 1.140396e+07 | 26.250000 | 12.583333 | 19.416667 |
| 25 | Netherlands | 2.829252e+07 | 12.833333 | 7.583333 | 10.208333 |
| 26 | New Zealand | 8.345533e+04 | 16.888889 | 9.888889 | 13.388889 |
| 27 | Norway | 2.094937e+06 | 10.666667 | 3.666667 | 7.166667 |
| 28 | Poland | 4.611836e+07 | 13.000000 | 4.166667 | 8.583333 |
| 29 | Portugal | 1.627157e+07 | 18.166667 | 11.875000 | 15.020833 |
| 30 | Russia | 1.049129e+08 | 9.333333 | 1.041667 | 5.187500 |
| 31 | Singapore | 2.103296e+06 | 31.416667 | 24.666667 | 28.041667 |
| 32 | South Africa | 3.877769e+07 | 23.500000 | 10.583333 | 17.041667 |
| 33 | Spain | 7.354313e+07 | 21.000000 | 10.833333 | 15.916667 |

| | Country/Region | average_monthly_confirmed | high | low | avg_temp |
|----|----------------------|---------------------------|-----------|-----------|-----------|
| 34 | Sweden | 1.728602e+07 | 10.583333 | 2.666667 | 6.625000 |
| 35 | Switzerland | 1.317681e+07 | 14.666667 | 5.000000 | 9.833333 |
| 36 | Thailand | 2.023818e+06 | 32.722222 | 22.722222 | 27.722222 |
| 37 | Turkey | 7.868205e+07 | 20.416667 | 9.083333 | 14.750000 |
| 38 | United Arab Emirates | 1.011001e+07 | 33.500000 | 22.750000 | 28.125000 |

Visualization

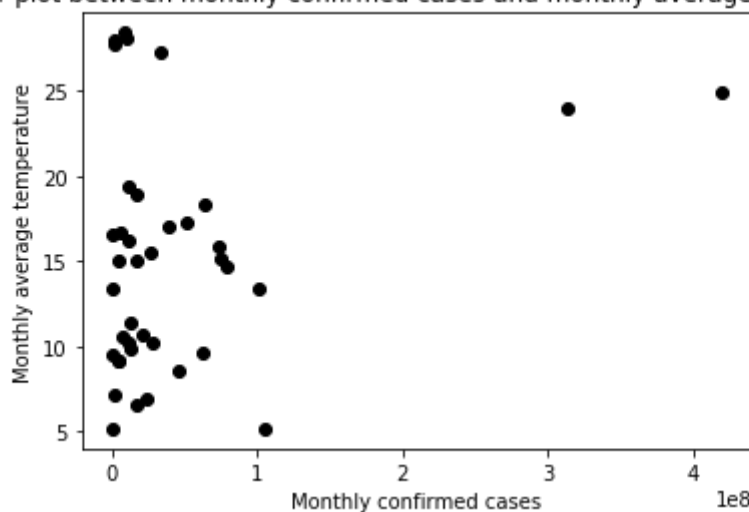
In [83]:

```
1 import matplotlib.pyplot as plt
2 plt.title('Scatter plot between monthly confirmed cases and monthly average temperature')
3 plt.xlabel('Monthly confirmed cases')
4 plt.ylabel('Monthly average temperature')
5 plt.scatter(final_data['average_monthly_confirmed'], final_data['avg_temp'], color='black')
```

Out[83]:

<matplotlib.collections.PathCollection at 0x1475bd9e448>

Scatter plot between monthly confirmed cases and monthly average temperature



In [85]:

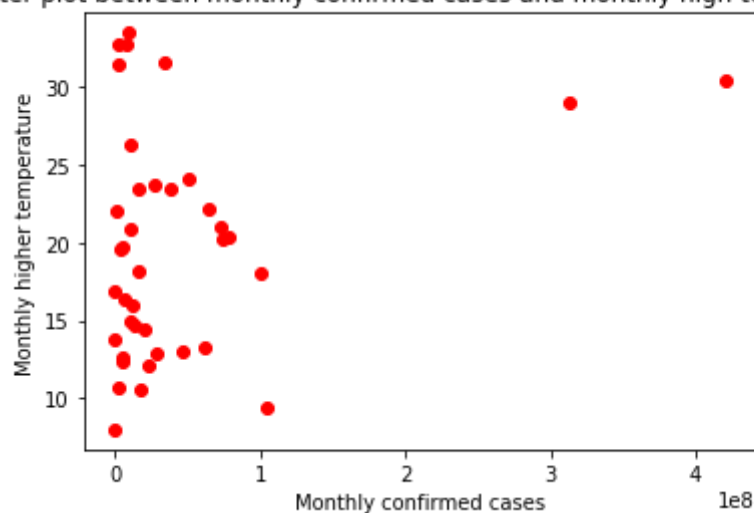


```
1 plt.title('Scatter plot between monthly confirmed cases and monthly high temperature')
2 plt.xlabel('Monthly confirmed cases')
3 plt.ylabel('Monthly higher temperature')
4 plt.scatter(final_data['average_monthly_confirmed'],final_data['high'], color='red')
```

Out[85]:

<matplotlib.collections.PathCollection at 0x1475cbfa288>

Scatter plot between monthly confirmed cases and monthly high temperature



In [86]:

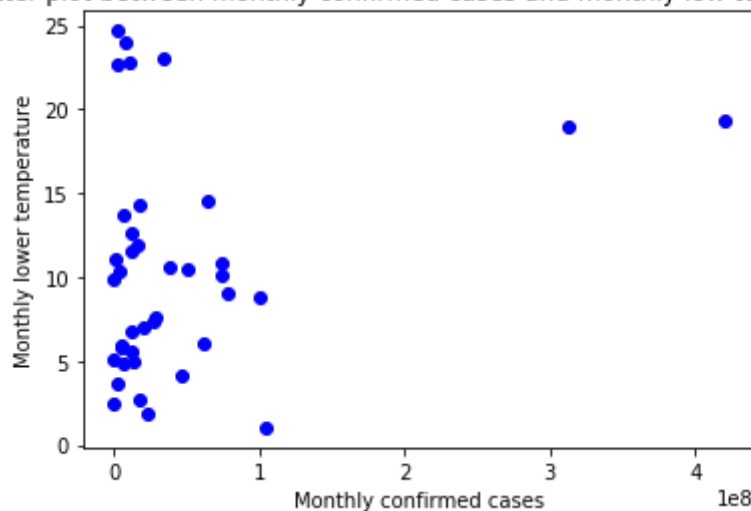


```
1 plt.title('Scatter plot between monthly confirmed cases and monthly low temperature')
2 plt.xlabel('Monthly confirmed cases')
3 plt.ylabel('Monthly lower temperature')
4 plt.scatter(final_data['average_monthly_confirmed'],final_data['low'], color='blue')
```

Out[86]:

<matplotlib.collections.PathCollection at 0x1475bdd05c8>

Scatter plot between monthly confirmed cases and monthly low temperature



Through this plot we can say that the covid cases does not seems to be decreasing with the increase in temperature.

9

Articulate your own research question related to COVID-19 and try to provide an answer to it using the given data set.

The effect of population level of a country on the amount of confirmed cases

In [115]:



```
1 data3=[update_df['Confirmed'], update_pop['population']]
2 names3= ["Confirmed", "population"]
3 df_population= pd.concat(data3, axis=1, keys=names3)
```

In [116]:



```
1 df_population
```

Out[116]:

| | Confirmed | population |
|-----|------------|------------|
| 0 | 104850.0 | 1388232693 |
| 1 | 31484605.0 | 1342512706 |
| 2 | 34672690.0 | 326474013 |
| 3 | 3287727.0 | 263510146 |
| 4 | 19797086.0 | 211243220 |
| ... | ... | ... |
| 177 | 3083.0 | 38022 |
| 178 | 2833.0 | 38010 |
| 179 | 5130.0 | 32104 |
| 180 | 27.0 | 21726 |
| 181 | NaN | 801 |

182 rows × 2 columns

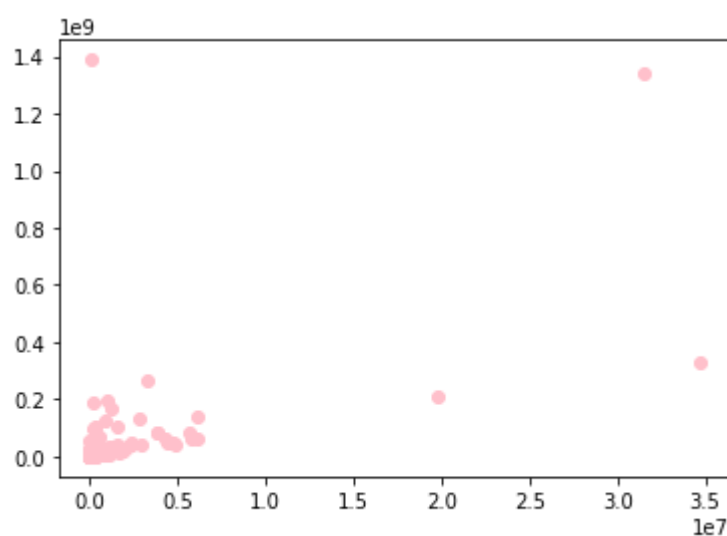
In [121]:



```
1 plt.scatter(df_population['Confirmed'],df_population['population'], color='pink')
```

Out[121]:

<matplotlib.collections.PathCollection at 0x1476242aa88>



We can see a slight correlation between the population and the confirmed cases. It shows that the number of confirmed cases in a country does depend on the population of the particular country.

| |
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