Importing required Libraries

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
import plotly.express as px
import plotly graph objs as go
from plotly subplots import make subplots
from plotly.offline import plot, iplot, init notebook mode
init notebook mode(connected=True)
corona dataset csv = pd.read csv('covid19 Confirmed dataset.csv')
corona dataset csv.head(10)
                 Province/State
                                       Country/Region
                                                            Lat
Long
                             NaN
                                          Afghanistan 33.0000
65.0000
                                              Albania 41.1533
                             NaN
20.1683
                             NaN
                                              Algeria 28.0339
1.6596
                             NaN
                                              Andorra 42.5063
1.5218
4
                             NaN
                                               Angola -11.2027
17.8739
                                  Antigua and Barbuda 17.0608 -
5
                             NaN
61.7964
                             NaN
                                            Argentina -38.4161 -
63.6167
7
                             NaN
                                              Armenia 40.0691
45.0382
8 Australian Capital Territory
                                            Australia -35.4735
149.0124
                New South Wales
                                            Australia -33.8688
151.2093
   1/22/20 1/23/20
                    1/24/20 1/25/20
                                       1/26/20
                                                 1/27/20
                                                           . . .
                                                                4/21/20
\
0
         0
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                                              0
                                                        0
                                                                   1092
                                                           . . .
1
                  0
         0
                            0
                                     0
                                              0
                                                        0
                                                                    609
                                                           . . .
2
         0
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                                     0
                                              0
                                                        0
                                                                   2811
                                                           . . .
3
         0
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                                     0
                                              0
                                                        0
                                                          . . .
                                                                    717
```

4	0	0	0	0	0	0		24
5	0	0	0	0	0	0		23
6	0	0	0	0	0	0		3031
7	0	Θ	0	0	0	0		1401
8	0	Θ	Θ	Θ	Θ	Θ		104
9	0	0	0	0	3	4		2969
J	J	· ·	· ·	· ·	J	·		2303
4/ 4/29/	/22/20 /20 \	4/23/20	4/24/20	4/25/20	4/26/20	4/27/20	4/28/20	
0 1939	1176	1279	1351	1463	1531	1703	1828	
1939 1 766	634	663	678	712	726	736	750	
2 3848	2910	3007	3127	3256	3382	3517	3649	
3 3 743	723	723	731	738	738	743	743	
743 4 27	25	25	25	25	26	27	27	
5 24	24	24	24	24	24	24	24	
6 4285	3144	3435	3607	3780	3892	4003	4127	
7 1932	1473	1523	1596	1677	1746	1808	1867	
8 106	104	104	105	106	106	106	106	
9 3016	2971	2976	2982	2994	3002	3004	3016	
4/ 0 1 2 3 4 5 6 7 8 9	730/20 2171 773 4006 745 27 24 4428 2066 106 3025							

[10 rows x 104 columns]
corona_dataset_csv.shape
(266, 104)
corona_dataset_csv.drop(['Lat','Long'],axis=1,inplace=True)
corona_dataset_csv.head(10)

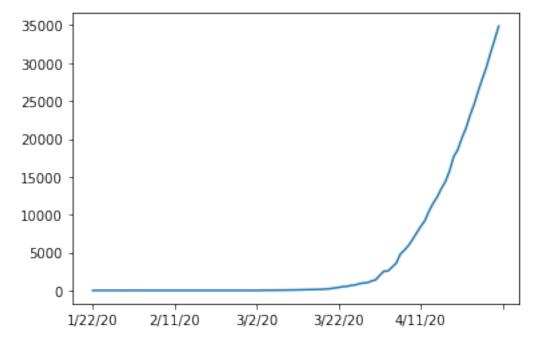
CO	i ona_aaca	366_634.11	caa(10)					
`		Pro	vince/Sta	te	Country/R	egion 1/	22/20	1/23/20
0			Na	϶N	Afghan	istan	Θ	0
1			Na	aN	Al	bania	0	0
2			Na	aN	Al	geria	0	Θ
3			Na	aN	An	dorra	0	Θ
4			Na	aN	А	ngola	0	Θ
5			Na	aN Antig	ua and Ba	rbuda	0	Θ
6			Na	϶N	Arge	ntina	0	Θ
7			Na	϶N	Ar	menia	0	Θ
8	Australian Capital Territory				Aust	ralia	0	0
9		New	South Wale	es	Aust	ralia	0	0
`	1/24/20	1/25/20	1/26/20	1/27/20	1/28/20	1/29/20		4/21/20
0	Θ	Θ	0	0	Θ	0		1092
1	0	0	0	0	0	0		609
2	0	0	0	0	0	0		2811
3	Θ	Θ	Θ	0	Θ	0		717
4	Θ	Θ	Θ	0	Θ	0		24
5	0	0	Θ	0	0	0		23
6	0	0	Θ	0	0	0		3031

```
7
         0
                   0
                             0
                                       0
                                                 0
                                                                       1401
                                                           0
8
                   0
                             0
                                       0
                                                                        104
         0
                                                 0
9
         0
                   0
                             3
                                       4
                                                 4
                                                           4
                                                                       2969
   4/22/20
             4/23/20
                       4/24/20
                                4/25/20
                                          4/26/20
                                                    4/27/20
                                                              4/28/20
4/29/20
      1176
                1279
                          1351
                                    1463
                                              1531
                                                        1703
                                                                  1828
1939
                           678
                                     712
                                               726
                                                         736
                                                                   750
1
       634
                 663
766
      2910
                3007
                          3127
                                    3256
                                              3382
                                                        3517
                                                                  3649
2
3848
       723
                 723
                           731
                                     738
                                               738
                                                         743
                                                                   743
743
        25
                  25
                            25
                                      25
                                                26
                                                          27
                                                                    27
4
27
5
        24
                  24
                            24
                                      24
                                                24
                                                          24
                                                                    24
24
      3144
                3435
                          3607
                                    3780
                                              3892
                                                        4003
                                                                  4127
6
4285
      1473
                1523
                          1596
                                    1677
                                              1746
                                                        1808
                                                                  1867
1932
       104
                 104
                           105
                                     106
                                               106
                                                         106
                                                                   106
106
      2971
                2976
                          2982
                                    2994
                                              3002
                                                        3004
                                                                  3016
3016
   4/30/20
0
      2171
       773
1
2
      4006
3
       745
4
        27
5
        24
6
      4428
7
      2066
8
       106
9
      3025
[10 rows x 102 columns]
# Aggregating the rows with country
corona_dataset_aggregated =
corona_dataset_csv.groupby("Country/Region").sum()
```

```
corona dataset aggregated.head(10).sum(axis=1)
Country/Region
Afghanistan
                        28462
Albania
                        17864
Algeria
                        74325
Andorra
                        21893
Angola
                          649
Antigua and Barbuda
                          678
Argentina
                        84105
Armenia
                        40610
                       224354
Australia
Austria
                       502063
dtype: int64
corona dataset aggregated.shape
(187, 100)
#from matplotlib.pyplot import figure
#countries=list(corona dataset aggregated.index)
#total cases=corona dataset aggregated.sum(axis=1)
\#plt.figure(figsize=(100,40),dpi=380)
#plt.xticks(rotation=90, fontsize=25)
#plt.bar(countries, total cases)
#plt.show()
#figure(num=None, figsize=(8, 6), dpi=80, facecolor='w',
edgecolor='k')
```

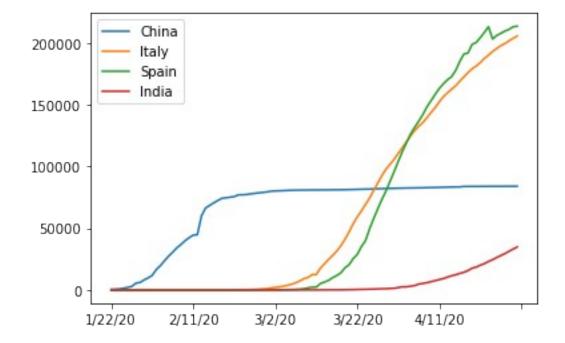
Visualizing data related to a country for example India, China, Spain, Italy

```
corona_dataset_aggregated.loc['India'].plot()
<AxesSubplot:>
```



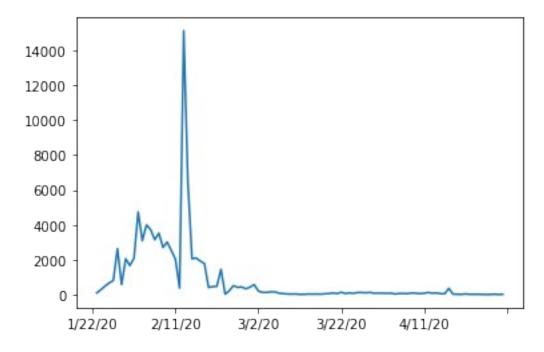
```
corona_dataset_aggregated.loc['China'].plot()
corona_dataset_aggregated.loc['Italy'].plot()
corona_dataset_aggregated.loc['Spain'].plot()
corona_dataset_aggregated.loc['India'].plot()
plt.legend()
```

<matplotlib.legend.Legend at 0x1cfd9066310>

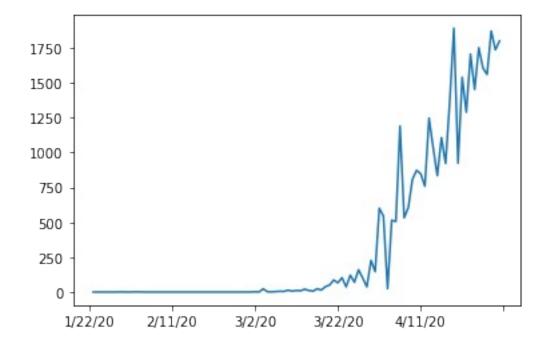


TO GET THE RATE AT WHICH CASES ARE GROWING(DIFFERENTIATING THE CURVE)

corona_dataset_aggregated.loc['China'].diff().plot()
<AxesSubplot:>



corona_dataset_aggregated.loc['India'].diff().plot()
<AxesSubplot:>



```
Find max infection rates for China and India
corona dataset aggregated.loc['China'].diff().max()
15136.0
corona dataset aggregated.loc['India'].diff().max()
1893.0
corona dataset aggregated.loc['India',:]
               0
1/22/20
1/23/20
               0
               0
1/24/20
1/25/20
               0
               0
1/26/20
4/26/20
           27890
4/27/20
           29451
4/28/20
           31324
4/29/20
           33062
4/30/20
           34863
Name: India, Length: 100, dtype: int64
Finding average infection rates
corona dataset aggregated.loc['China'].diff().mean()
842.5050505050505
corona dataset aggregated.loc['India'].diff().mean()
352.1515151515151
Find maximum infection rate for all of the countries.
countries = list(corona dataset aggregated.index)
ave infection rates = []
for country in countries :
ave infection rates.append(corona dataset aggregated.loc[country].diff
().mean())
corona_dataset_aggregated['mean infection rate'] = ave_infection_rates
corona dataset aggregated.head()
                1/22/20 1/23/20 1/24/20 1/25/20 1/26/20 1/27/20
1/28/20 \
Country/Region
Afghanistan
                      0
                               0
                                         0
                                                  0
                                                           0
                                                                     0
```

Albania 0	0	0	0	Θ	0	0
Algeria	0	0	0	0	0	0
0 Andorra	0	Θ	Θ	0	Θ	Θ
0 Angola 0	0	0	0	0	0	0
4/24/20 \	1/29/20	1/30/20	1/31/20		22/20 4/2	23/20
Country/Region						
Afghanistan 1351	0	0	0		1176	1279
Albania 678	0	0	0		634	663
Algeria 3127	Θ	0	Θ		2910	3007
Andorra 731	Θ	0	Θ		723	723
Angola 25	0	0	0		25	25
4/30/20 \ Country/Region	4/25/20	4/26/20	4/27/20	4/28/20	4/29/20	
Afghanistan	1463	1531	1703	1828	1939	2171
Albania	712	726	736	750	766	773
Algeria	3256	3382	3517	3649	3848	4006
Andorra	738	738	743	743	743	745
Angola	25	26	27	27	27	27

mean infection rate

Country/Region
Afghanistan 21.929293
Albania 7.808081
Algeria 40.464646
Andorra 7.525253
Angola 0.272727

[5 rows x 101 columns]

Create a new dataframe with only needed column

corona_data = pd.DataFrame(corona_dataset_aggregated['mean infection
rate'])

corona data.head(10)

	mean	infection rate
Country/Region		
Afghanistan		21.929293
Albania		7.808081
Algeria		40.464646
Andorra		7.525253
Angola		0.272727
Antigua and Barbuda		0.242424
Argentina		44.727273
Armenia		20.868687
Australia		68.343434
Austria		156.080808

Importing the world happiness report

world_happiness_report = pd.read_csv("worldwide_happiness_report.csv")
world_happiness_report.head()

0verall	rank	Country or region	Score	GDP per capita	Social
support \					
0	1	Finland	7.769	1.340	
1.587					
1	2	Denmark	7.600	1.383	
1.573					
2	3	Norway	7.554	1.488	
1.582					
3	4	Iceland	7.494	1.380	
1.624					
4	5	Netherlands	7.488	1.396	
1.522					

Healthy life expectancy Freedom to make life choices Generosity \
0 0.986 0.596

1	0.996	0.592	0.252
2	1.028	0.603	0.271
3	1.026	0.591	0.354
4	0.999	0.557	0.322

0.153

```
Perceptions of corruption
0
                        0.393
1
                        0.410
2
                        0.341
3
                        0.118
4
                       0.298
world happiness report.dtypes
Overall rank
                                   int64
Country or region
                                  obiect
Score
                                 float64
GDP per capita
                                 float64
Social support
                                 float64
Healthy life expectancy
                                 float64
Freedom to make life choices
                                 float64
Generosity
                                 float64
Perceptions of corruption
                                 float64
dtype: object
world happiness report.shape
(156, 9)
Dropping useless Columns
columns to dropped = ['Overall rank', 'Score', 'Generosity', 'Perceptions
of corruption']
world happiness report.drop(columns to dropped,axis=1 , inplace=True)
world happiness report.head()
  Country or region GDP per capita Social support Healthy life
expectancy \
            Finland
                               1.340
                                                1.587
0.986
            Denmark
                               1.383
                                                1.573
1
0.996
             Norway
                               1.488
                                                1.582
1.028
3
            Iceland
                               1.380
                                               1.624
1.026
        Netherlands
                               1.396
                                                1.522
4
0.999
   Freedom to make life choices
0
                           0.596
                           0.592
1
2
                           0.603
3
                           0.591
4
                           0.557
```

Changing the indices of the dataframe

world_happiness_report.set_index(['Country or region'],inplace=True)
world_happiness_report.head()

	GDP per capita	Social support	Healthy life
expectancy \ Country or region			

Finland 0.986	1.340	1.587
Denmark	1.383	1.573
0.996 Norway	1.488	1.582
1.028 Iceland	1.380	1.624
1.026 Netherlands	1.396	1.522
0.999		_

Freedom to make life choices

Country or region	
Finland	0.596
Denmark	0.592
Norway	0.603
Iceland	0.591
Netherlands	0.557

Joining the two datasets

data = world_happiness_report.join(corona_data,how="inner")
data.head()

ovnostansv	GDP per capita	Social support	Healthy life
expectancy Finland	1.340	1.587	0.986
Denmark	1.383	1.573	0.996
Norway	1.488	1.582	1.028
Iceland	1.380	1.624	1.026
Netherlands	1.396	1.522	0.999

	Freedom	to	make	life	choices	mean	infection rate
Finland					0.596		50.454545
Denmark					0.592		94.505051
Norway					0.603		78.161616

Iceland	0.591	18.151515
Netherlands	0.557	399.111111

Correlation Matrix

data.corr()

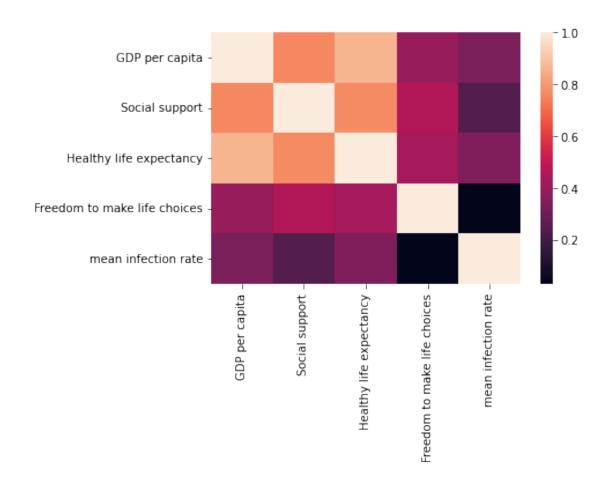
it is representing the correlation between every two columns of our dataset

dataset	
GDP per capita Social support Healthy life expectancy Freedom to make life choices mean infection rate	GDP per capita Social support \ 1.000000
GDP per capita Social support Healthy life expectancy Freedom to make life choices mean infection rate	Healthy life expectancy \
GDP per capita Social support Healthy life expectancy Freedom to make life choices mean infection rate	Freedom to make life choices \
GDP per capita Social support	mean infection rate 0.329073 0.241618

GDP per capita 0.329073
Social support 0.241618
Healthy life expectancy 0.341890
Freedom to make life choices 0.029595
mean infection rate 1.000000

sns.heatmap(data.corr())

<AxesSubplot:>



Visualization of the results

Plotting GDP vs maximum Infection rate

x = data['GDP per capita']

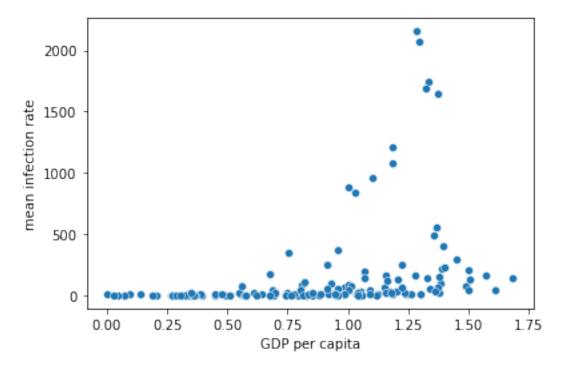
y = data['mean infection rate']

sns.scatterplot(x,y) #VISUALISATION IS DIFFICULT BECAUSE OF DIFFERENCE Y SCALE AND X-SCALE

D:\Users\dell\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning:

Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

<AxesSubplot:xlabel='GDP per capita', ylabel='mean infection rate'>

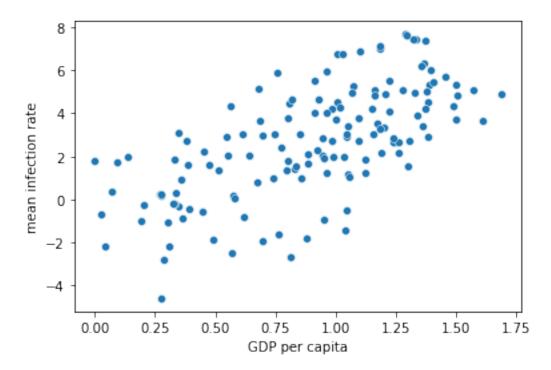


```
x = data['GDP per capita']
y = data['mean infection rate']
sns.scatterplot(x,np.log(y))
```

D:\Users\dell\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning:

Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

<AxesSubplot:xlabel='GDP per capita', ylabel='mean infection rate'>

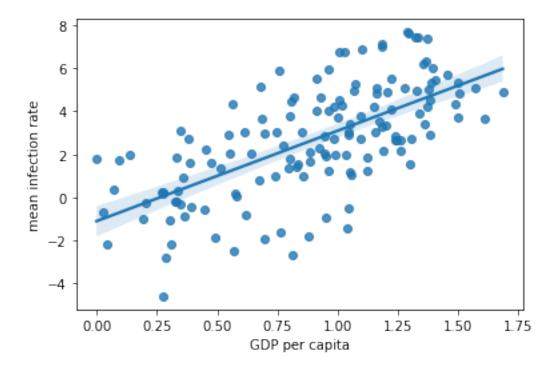


sns.regplot(x,np.log(y))

D:\Users\dell\anaconda3\lib\site-packages\seaborn_decorators.py:36:
FutureWarning:

Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

<AxesSubplot:xlabel='GDP per capita', ylabel='mean infection rate'>



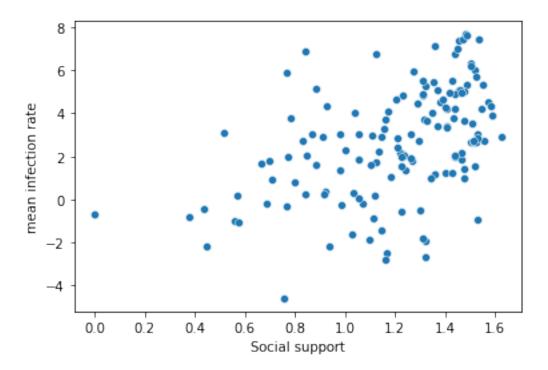
Plotting Social support vs maximum Infection rate

```
x = data['Social support']
y = data['mean infection rate']
sns.scatterplot(x,np.log(y))
```

D:\Users\dell\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning:

Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

<AxesSubplot:xlabel='Social support', ylabel='mean infection rate'>

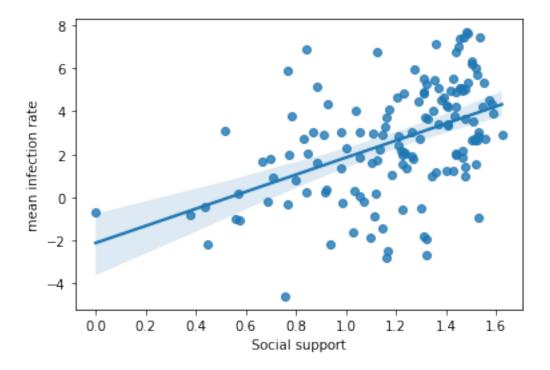


sns.regplot(x,np.log(y))

D:\Users\dell\anaconda3\lib\site-packages\seaborn_decorators.py:36:
FutureWarning:

Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

<AxesSubplot:xlabel='Social support', ylabel='mean infection rate'>



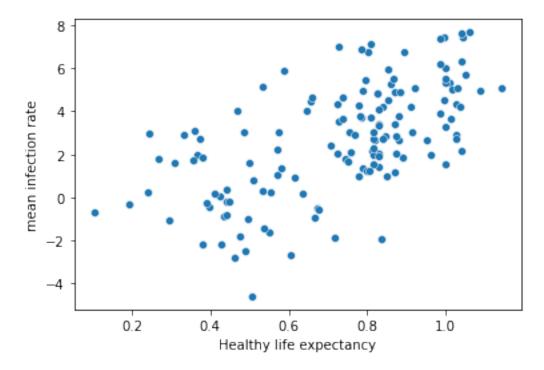
Plotting Healthy life expectancy vs maximum Infection rate

x = data['Healthy life expectancy']
y = data['mean infection rate']
sns.scatterplot(x,np.log(y))

 $\label{lem:conda} $$D:\Users\dell\anaconda3\lib\site-packages\seaborn_decorators.py:36: Future\Warning:$

Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

<AxesSubplot:xlabel='Healthy life expectancy', ylabel='mean infection
rate'>

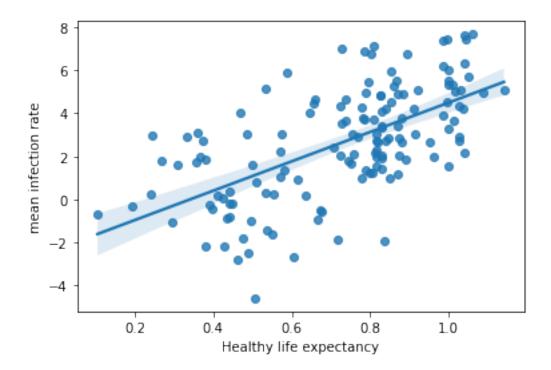


sns.regplot(x,np.log(y))

D:\Users\dell\anaconda3\lib\site-packages\seaborn_decorators.py:36:
FutureWarning:

Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

<AxesSubplot:xlabel='Healthy life expectancy', ylabel='mean infection
rate'>



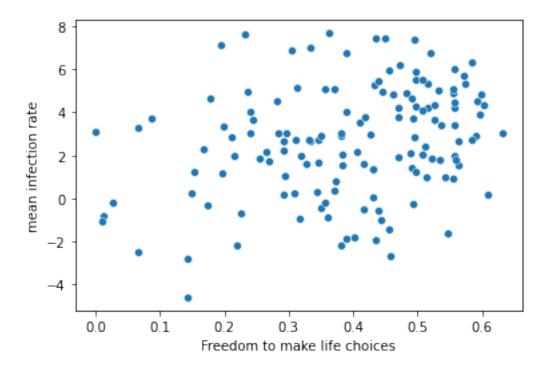
Plotting Freedom to make life choices vs maximum Infection rate

x = data['Freedom to make life choices']
y = data['mean infection rate']
sns.scatterplot(x,np.log(y))

D:\Users\dell\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning:

Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

<AxesSubplot:xlabel='Freedom to make life choices', ylabel='mean
infection rate'>

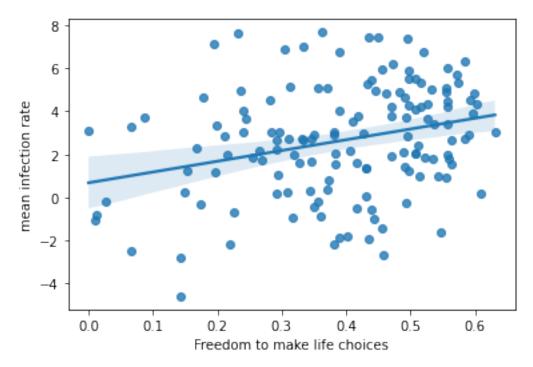


sns.regplot(x,np.log(y))

D:\Users\dell\anaconda3\lib\site-packages\seaborn_decorators.py:36:
FutureWarning:

Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

<AxesSubplot:xlabel='Freedom to make life choices', ylabel='mean
infection rate'>



```
from covid import Covid
covid = Covid()
covid.get data()
len(covid.get data())
196
countries = covid.list countries()
italy cases = covid.get status by country id(115)
india_cases = covid.get_status_by_country_name("india")
india_cases
{'id': '80',
 'country': 'India',
 'confirmed': 34674643,
 'active': None,
 'deaths': 474479,
 'recovered': None,
 'latitude': 20.593684,
 'longitude': 78.96288,
 'last_update': 1639146106000}
df=pd.DataFrame(data=covid.get data())
df.head()
    id
               country confirmed active deaths recovered
                                                               latitude
\
```

```
0
   183
                     US
                          49666654
                                      None
                                            794656
                                                         None
                                                               40.000000
1
    80
                  India
                          34674643
                                      None
                                            474479
                                                         None
                                                               20.593684
2
    24
                 Brazil
                          22177059
                                            616457
                                                         None -14.235000
                                      None
3
   187
        United Kingdom
                          10722083
                                      None
                                            146592
                                                         None
                                                               55.000000
4
   145
                 Russia
                           9782723
                                      None
                                            281571
                                                         None
                                                               61.524000
   longitude
                 last update
0 -100.00000
               1639146106000
1
    78.96288
              1639146106000
2
   -51.92530
              1639146106000
3
   -3.00000
              1639146106000
   105.31880
              1639146106000
def plot map(df, col, pal):
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Vincent and the Grenadines", "Samoa", "San Marino", "Sao Tome and
Principe", "Saudi Arabia", "Senegal", "Serbia", "Seychelles", "Sierra
Leone", "Singapore", "Slovakia", "Slovenia", "Solomon
Islands", "Somalia", "South Africa", "South Sudan", "Spain", "Sri
Lanka", "Sudan", "Summer Olympics
2020", "Suriname", "Sweden", "Switzerland", "Syria", "Taiwan*", "Tajikistan"
```

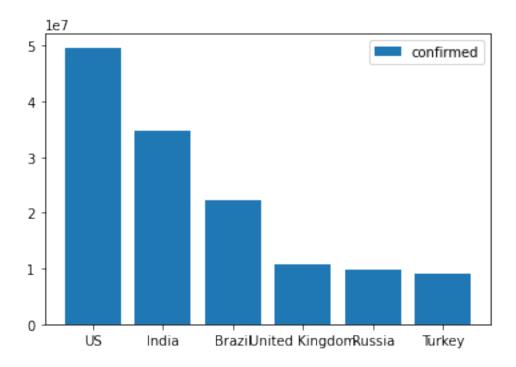
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df.drop(['latitude','longitude','id'],axis=1,inplace=True)
df.head()
          country
                   confirmed active
                                     deaths recovered
                                                         last update
0
               US
                    49666654
                               None
                                     794656
                                                 None
                                                       1639146106000
1
            India
                    34674643
                               None 474479
                                                 None
                                                       1639146106000
2
           Brazil
                    22177059
                               None 616457
                                                 None
                                                       1639146106000
3
  United Kingdom
                    10722083
                               None 146592
                                                 None
                                                       1639146106000
                               None 281571
           Russia
                    9782723
                                                 None
                                                       1639146106000
df.set index('country',inplace=True)
df.head()
```

```
confirmed active deaths recovered
                                                       last update
country
                 49666654
US
                            None
                                  794656
                                              None
                                                    1639146106000
India
                 34674643
                            None
                                  474479
                                              None
                                                    1639146106000
                                  616457
Brazil
                 22177059
                            None
                                              None
                                                    1639146106000
United Kingdom
                 10722083
                            None
                                  146592
                                              None
                                                    1639146106000
Russia
                  9782723
                            None 281571
                                              None
                                                    1639146106000
countries=df.index[:6]
confirm=df['confirmed'][:6].to_numpy()
plt.bar(countries,confirm,label='confirmed')
plt.legend()
```

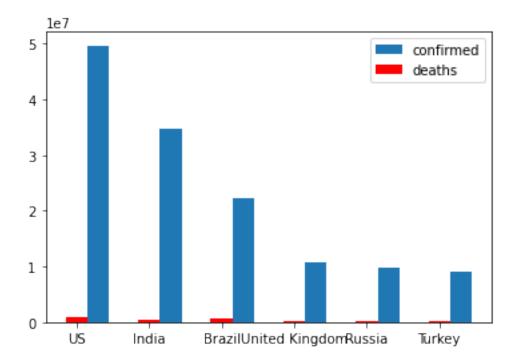
<matplotlib.legend.Legend at 0x1cfdba9dee0>



```
countries=df.index[:6]
confirm=df['confirmed'][:6].to_numpy()
total_deaths=df['deaths'][:6].to_numpy()

xpos=np.arange(6)

plt.bar(xpos+0.3,confirm,label='confirmed',width=0.3)
plt.bar(xpos,total_deaths,label='deaths',width=0.3,color='red')
plt.xticks(xpos,countries)
plt.legend()
<matplotlib.legend.Legend at 0x1cfdbb19df0>
```



final=df.join(corona_data,how='inner')
len(final)

186

final.head()

	confirmed	active	deaths	recovered	last_update	\
US	49666654	None	794656	None	$163914\overline{6}106000$	
India	34674643	None	474479	None	1639146106000	
Brazil	22177059	None	616457	None	1639146106000	
United Kingdom	10722083	None	146592	None	1639146106000	
Russia	9782723	None	281571	None	1639146106000	

mean infection rate
US 10802.252525
India 352.151515
Brazil 880.676768
United Kingdom 1742.232323
Russia 1075.737374

Joining Data from covid library and World Happiness Index

Finland	1.340	1.587	7		0.986
Denmark	1.383	1.573	3		0.996
Norway	1.488	1.582	2		1.028
Iceland	1.380	1.624	4		1.026
Netherlands	1.396	1.522	2		0.999
Switzerland	1.452	1.526	6		1.052
Sweden	1.387	1.48	7		1.009
New Zealand	1.303	1.55	7		1.026
Canada	1.365	1.50	5		1.039
Austria	1.376	1.47	5		1.016
recovered \ Finland None Denmark None Norway None Iceland None Netherlands None Switzerland None Sweden None New Zealand None Canada None Austria None	Freedom to make life	0.596 0.592 0.603 0.591 0.557 0.572 0.574 0.585 0.584 0.532	201051 542181 305368 19159 2870459 1111923 1226055 12623 1831341 1221095	None None None None	deaths 1421 2996 1135 35 20372 11739 15185 44 29930 13076
Finland Denmark Norway Iceland	last_update 1639146106000 1639146106000 1639146106000 1639146106000				

1639146106000
1639146106000
1639146106000
1639146106000
1639146106000
1639146106000

Scaling data

data1.drop(['active','recovered','last_update'],axis=1 ,inplace =
True)

data1.head()

ovnostansv	GDP per capita	Social support	Healthy life	
expectancy Finland	1.340	1.587	0.98	6
Denmark	1.383	1.573	0.99	6
Norway	1.488	1.582	1.02	8
Iceland	1.380	1.624	1.02	6
Netherlands	1.396	1.522	0.99	9

	Freedom	to	make	life	choices	confirmed	deaths
Finland					0.596	201051	1421
Denmark					0.592	542181	2996
Norway					0.603	305368	1135
Iceland					0.591	19159	35
Netherlands					0.557	2870459	20372

from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
scaled = scaler.fit_transform(data1)

from pandas import DataFrame
datax = DataFrame(scaled)
datax.head()

	0	1	2	3	4	5
0	0.795724	0.977217	0.850386	0.944532	0.005722	0.002300
1	0.821259	0.968596	0.860039	0.938193	0.015561	0.004855
2	0.883610	0.974138	0.890927	0.955626	0.008731	0.001836
3	0.819477	1.000000	0.888996	0.936609	0.000476	0.000052
4	0.828979	0.937192	0.862934	0.882726	0.082713	0.033042

Performing linear regression

```
from sklearn.linear model import LinearRegression
regression model = LinearRegression()
x = datax.iloc[:, -6].values #GDP
y = datax.iloc[:, -1].values #deaths
x = x.reshape(-1,1)
regression model.fit(x,y) #GDP vs Deaths
LinearRegression()
y predicted = regression model.predict(x)
from sklearn.metrics import mean squared error ,r2 score
mse = mean squared error(y,y predicted)
mse
0.015631230121152265
rmse = np.sqrt(mean squared error(y,y predicted))
r2 = r2 score(y,y predicted)
rmse
0.12502491800098198
r2
0.021593705847284927
```

GDP alone cannot explain variance of deaths

Gradient descent approach

```
# Gradient descent_Build the model
m = 0
c = 0
L = 0.00001
epochs = 10000 #the number of iterations to perform gradiation
X = datax.iloc[:, -6].values #attributes to determine dependant
variable/ Class
Y = datax.iloc[:, -1].values #dependant variable/ Class
n = float(len(X))
n

144.0

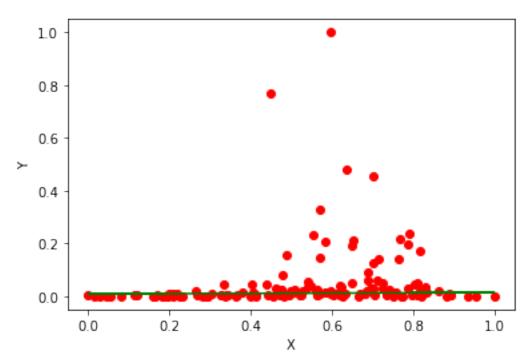
for i in range(epochs):
    Y_pred = m*X + c #The current predicted value of Y
    D_m = (-2/n) * sum(X* (Y - Y_pred)) # Derivative wrt m
    D_c = (-2/n) * sum(Y - Y_pred) # Derivative of c
```

```
m = m - L* D_m # Update m
    c = c - L*D_c # Update c
print (m,c)

0.005452208077273036  0.008711837822156551

Y_pred = m*X + c
plt.scatter(X,Y,color = "red")
plt.plot(X,Y_pred,color = "green")
plt.xlabel("X")
plt.ylabel("Y")

Text(0, 0.5, 'Y')
```



Outliers of the Data Frame

Social support index vs Deaths due to covid

```
x = datax.iloc[:, -5].values #attributes to determine dependant
variable/ Class
y = datax.iloc[:, -1].values #dependant variable/ Class
x= x.reshape(-1,1)
regression_model.fit(x,y)
LinearRegression()
y_predicted = regression_model.predict(x)
mse = mean_squared_error(y,y_predicted)
mse
```

```
0.015785897837850917
rmse = np.sqrt(mean squared error(y,y predicted))
r2 = r2 score(y,y predicted)
r2
0.011912582458575072
Healthy Life Expectancy Vs Deaths due to covid
x = datax.iloc[:, -4].values #attributes to determine dependant
variable/ Class
y = datax.iloc[:, -1].values #dependant variable/ Class
x = x.reshape(-1,1)
regression model.fit(x,y)
y predicted = regression model.predict(x)
mse = mean squared_error(y,y_predicted)
0.0156979693574256
rmse = np.sqrt(mean squared error(y,y predicted))
r2 = r2 score(y,y predicted)
r2
0.017416293811848615
Vaccinations
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import plotly.express as px
import plotly.graph objects as go
import matplotlib.patches as mpatches
from plotly.subplots import make subplots
from wordcloud import WordCloud
import seaborn as sns
sns.set(color codes = True)
sns.set(style="whitegrid")
import plotly.figure factory as ff
from plotly.colors import n colors
df = pd.read csv('country vaccinations.csv')
df.head()
       country iso_code
                               date total vaccinations
people_vaccinated \
                  AFG 2021-02-22
                                                    0.0
0 Afghanistan
0.0
1 Afghanistan AFG 2021-02-23
                                                    NaN
```

```
NaN
2 Afghanistan
                    AFG 2021-02-24
                                                     NaN
NaN
3 Afghanistan
                    AFG 2021-02-25
                                                     NaN
NaN
4 Afghanistan
                    AFG 2021-02-26
                                                     NaN
NaN
   people fully vaccinated daily vaccinations raw daily vaccinations
0
                       NaN
                                                NaN
                                                                     NaN
1
                                                NaN
                                                                  1367.0
                       NaN
2
                                                NaN
                       NaN
                                                                  1367.0
3
                       NaN
                                                NaN
                                                                  1367.0
4
                                                NaN
                                                                  1367.0
                       NaN
   total vaccinations per hundred people vaccinated per hundred
0
                               0.0
                                                               0.0
1
                               NaN
                                                               NaN
2
                               NaN
                                                               NaN
3
                               NaN
                                                               NaN
4
                               NaN
                                                               NaN
   people fully vaccinated per hundred daily vaccinations per million
\
0
                                    NaN
                                                                     NaN
1
                                    NaN
                                                                    34.0
2
                                    NaN
                                                                    34.0
3
                                    NaN
                                                                    34.0
4
                                                                    34.0
                                    NaN
                                             vaccines \
  Johnson&Johnson, Oxford/AstraZeneca, Pfizer/Bi...
  Johnson&Johnson, Oxford/AstraZeneca, Pfizer/Bi...
  Johnson&Johnson, Oxford/AstraZeneca, Pfizer/Bi...
  Johnson&Johnson, Oxford/AstraZeneca, Pfizer/Bi...
  Johnson&Johnson, Oxford/AstraZeneca, Pfizer/Bi...
```

source name \

```
World Health Organization
  World Health Organization
1
2 World Health Organization
3 World Health Organization
4 World Health Organization
                                      source website
  https://reliefweb.int/sites/reliefweb.int/file...
  https://reliefweb.int/sites/reliefweb.int/file...
  https://reliefweb.int/sites/reliefweb.int/file...
3
  https://reliefweb.int/sites/reliefweb.int/file...
4 https://reliefweb.int/sites/reliefweb.int/file...
df.isnull().sum()
                                           0
country
                                            0
iso code
                                            0
date
                                       25685
total vaccinations
people_vaccinated
                                       27190
people fully vaccinated
                                       30037
daily_vaccinations raw
                                       31124
daily_vaccinations
                                         253
total vaccinations per hundred
                                       25685
people vaccinated per hundred
                                       27190
people fully vaccinated per hundred
                                       30037
                                         253
daily_vaccinations_per_million
vaccines
                                           0
                                           0
source name
                                            0
source website
dtype: int64
df.fillna(value = 0, inplace = True)
df.total vaccinations = df.total vaccinations.astype(int)
df.people_vaccinated = df.people_vaccinated.astype(int)
df.people fully vaccinated = df.people fully vaccinated.astype(int)
df.daily_vaccinations_raw = df.daily_vaccinations_raw.astype(int)
df.daily vaccinations = df.daily vaccinations.astype(int)
df.total vaccinations per hundred =
df.total_vaccinations_per_hundred.astype(int)
df.people fully vaccinated per hundred =
df.people fully vaccinated per hundred.astype(int)
df.daily vaccinations per million =
df.daily vaccinations per million.astype(int)
df.people vaccinated per hundred =
df.people vaccinated per hundred.astype(int)
date = df.date.str.split('-', expand =True)
date
              1
                  2
0
       2021
             02
                 22
```

```
2021
             02
                 23
1
2
       2021
             02
                 24
3
       2021
             02
                 25
4
       2021
             02
                 26
        . . .
55819
       2021
             11
                 01
                 02
55820
       2021
             11
55821
       2021
             11
                 03
55822
       2021
             11
                 04
55823
      2021
            11
                 05
[55824 rows x 3 columns]
df['year'] = date[0]
df['month'] = date[1]
df['day'] = date[2]
df.year = pd.to numeric(df.year)
df.month = pd.to numeric(df.month)
df.day = pd.to numeric(df.day)
df.date = pd.to_datetime(df.date)
df.head()
       country iso_code
                               date total vaccinations
people vaccinated \
  Afghanistan
                   AFG 2021-02-22
                                                       0
1
  Afghanistan
                    AFG 2021-02-23
                                                       0
0
2
                    AFG 2021-02-24
                                                       0
  Afghanistan
0
3
  Afghanistan
                    AFG 2021-02-25
                                                       0
0
4
  Afghanistan
                    AFG 2021-02-26
                                                       0
   people fully vaccinated daily vaccinations raw daily vaccinations
\
0
                          0
                                                   0
                                                                        0
1
                          0
                                                   0
                                                                    1367
2
                          0
                                                   0
                                                                    1367
3
                          0
                                                   0
                                                                    1367
4
                          0
                                                   0
                                                                    1367
   total_vaccinations_per_hundred people_vaccinated_per_hundred
0
```

```
0
                                                               0
1
2
                                0
                                                               0
3
                                0
                                                               0
4
                                0
                                                               0
   people fully vaccinated per hundred daily vaccinations per million
\
0
                                     0
                                                                     0
1
                                     0
                                                                    34
2
                                     0
                                                                    34
3
                                     0
                                                                    34
4
                                                                    34
                                     0
                                            vaccines
  Johnson&Johnson, Oxford/AstraZeneca, Pfizer/Bi...
   Johnson&Johnson, Oxford/AstraZeneca, Pfizer/Bi...
  Johnson&Johnson, Oxford/AstraZeneca, Pfizer/Bi...
   Johnson&Johnson, Oxford/AstraZeneca, Pfizer/Bi...
  Johnson&Johnson, Oxford/AstraZeneca, Pfizer/Bi...
                 source name
  World Health Organization
  World Health Organization
  World Health Organization
  World Health Organization
4 World Health Organization
                                      source website year month day
  https://reliefweb.int/sites/reliefweb.int/file...
                                                      2021
                                                                2
                                                                    22
  https://reliefweb.int/sites/reliefweb.int/file...
                                                                    23
1
                                                      2021
                                                                2
                                                                    24
2 https://reliefweb.int/sites/reliefweb.int/file... 2021
                                                                2
                                                                    25
  https://reliefweb.int/sites/reliefweb.int/file... 2021
                                                                2
                                                                2
                                                                    26
   https://reliefweb.int/sites/reliefweb.int/file... 2021
df.describe()
       total_vaccinations people_vaccinated people_fully_vaccinated
                                                         5.582400e+04
count
             5.582400e+04
                                5.582400e+04
```

mean	9.954941e+06	5.624392e+06		3.438866e+06
std	1.057154e+08	3.158674e+07		1.941963e+07
min	-2.147484e+09	0.000000e+00		0.000000e+00
25%	0.000000e+00	0.000000e+00		0.000000e+00
50%	3.108000e+04	4.969000e+03		0.000000e+00
75%	2.477739e+06	1.369812e+06		6.764480e+05
max	2.142580e+09	1.100842e+09		1.070386e+09
count mean std min 25% 50% 75% max total	_vaccinations_raw 5.582400e+04 1.134868e+05 8.412834e+05 0.0000000e+00 0.000000e+00 1.709025e+04 2.474100e+07 _vaccinations_per_l nated_per_hundred	5.582400e+04 1.297107e+05 8.228933e+05 0.000000e+00 9.220000e+02 7.007000e+03 4.154275e+04 2.242429e+07	\	
count		. 000000		55824.000000
mean	28	. 357785		15.636644
std	45	. 498547		24.255363
min	0	. 000000		0.000000
25%	Θ	. 000000		0.000000
50%	0	. 000000		0.000000
75%	42	. 000000		26.000000
max	269	. 000000		120.000000

people_fully_vaccinated_per_hundred
daily_vaccinations_per_million \
count 55824.000000

```
55824.000000
                                        11.008222
mean
3480.269669
                                        20.232838
std
4241.817117
min
                                         0.000000
0.000000
25%
                                         0.000000
586.000000
                                         0.000000
50%
2205.000000
75%
                                        12.000000
5095.000000
                                      118.000000
117497.000000
                  vear
                                 month
                                                     day
        55824.000000
                         55824.000000
                                          55824.000000
count
         2020.993730
                              6.266946
                                              15.846106
mean
             0.078934
                              2.637807
                                               8.869602
std
min
         2020.000000
                              1.000000
                                               1.000000
25%
         2021.000000
                              4.000000
                                               8.000000
50%
         2021.000000
                              6.000000
                                              16.000000
75%
         2021.000000
                              8.000000
                                              24.000000
         2021.000000
                             12,000000
                                              31,000000
max
df.country.unique()
array(['Afghanistan', 'Albania', 'Algeria', 'Andorra', 'Angola',
         'Anguilla', 'Antigua and Barbuda', 'Argentina', 'Armenia',
'Aruba'
         Australia', 'Austria', 'Azerbaijan', 'Bahamas', 'Bahrain',
         'Bangladesh', 'Barbados', 'Belarus', 'Belgium', 'Belize',
'Benin',
         'Bermuda', 'Bhutan', 'Bolivia', 'Bonaire Sint Eustatius and
Saba',
        'Bosnia and Herzegovina', 'Botswana', 'Brazil',
'British Virgin Islands', 'Brunei', 'Bulgaria', 'Burkina Faso',
'Burundi', 'Cambodia', 'Cameroon', 'Canada', 'Cape Verde',
'Cayman Islands', 'Central African Republic', 'Chad', 'Chile',
        'China', 'Colombia', 'Comoros', 'Congo', 'Cook Islands', 'Costa Rica', "Cote d'Ivoire", 'Croatia', 'Cuba', 'Curacao',
         'Cyprus', 'Czechia', 'Democratic Republic of Congo', 'Denmark',
        'Djibouti', 'Dominica', 'Dominican Republic', 'Ecuador',
'Egypt',
         'El Salvador', 'England', 'Equatorial Guinea', 'Estonia',
        'Eswatini', 'Ethiopia', 'Faeroe Islands', 'Falkland Islands',
         'Fiji', 'Finland', 'France', 'French Polynesia', 'Gabon',
'Gambia',
         'Georgia', 'Germany', 'Ghana', 'Gibraltar', 'Greece',
```

```
'Greenland',
        'Grenada', 'Guatemala', 'Guernsey', 'Guinea', 'Guinea-Bissau',
        'Guyana', 'Haiti', 'Honduras', 'Hong Kong', 'Hungary',
'Iceland',
        'India', 'Indonesia', 'Iran', 'Iraq', 'Ireland', 'Isle of Man',
        'Israel', 'Italy', 'Jamaica', 'Japan', 'Jersey', 'Jordan',
        'Kazakhstan', 'Kenya', 'Kiribati', 'Kosovo', 'Kuwait',
        'Kyrgyzstan', 'Laos', 'Latvia', 'Lebanon', 'Lesotho',
'Liberia',
        'Libya', 'Liechtenstein', 'Lithuania', 'Luxembourg', 'Macao',
        'Madagascar', 'Malawi', 'Malaysia', 'Maldives', 'Mali',
'Malta',
        'Mauritania', 'Mauritius', 'Mexico', 'Moldova', 'Monaco',
        'Mongolia', 'Montenegro', 'Montserrat', 'Morocco',
'Mozambique',
        'Myanmar', 'Namibia', 'Nauru', 'Nepal', 'Netherlands',
        'New Caledonia', 'New Zealand', 'Nicaragua', 'Niger',
'Nigeria',
        'Niue', 'North Macedonia', 'Northern Cyprus', 'Northern
Ireland',
        'Norway', 'Oman', 'Pakistan', 'Palestine', 'Panama',
        'Papua New Guinea', 'Paraguay', 'Peru', 'Philippines',
'Pitcairn',
        'Poland', 'Portugal', 'Qatar', 'Romania', 'Russia', 'Rwanda',
        'Saint Helena', 'Saint Kitts and Nevis', 'Saint Lucia',
        'Saint Vincent and the Grenadines', 'Samoa', 'San Marino',
       'San Tome and Principe', 'Saudi Arabia', 'Scotland', 'Senegal', 'Serbia', 'Seychelles', 'Sierra Leone', 'Singapore', 'Sint Maarten (Dutch part)', 'Slovakia', 'Slovenia', 'Solomon Islands', 'Somalia', 'South Africa', 'South Korea', 'South Sudan', 'Spain', 'Sri Lanka', 'Sudan', 'Suriname',
'Sweden',
        'Switzerland', 'Syria', 'Taiwan', 'Tajikistan', 'Tanzania',
        'Thailand', 'Timor', 'Togo', 'Tokelau', 'Tonga',
        'Trinidad and Tobago', 'Tunisia', 'Turkey', 'Turkmenistan',
        'Turks and Caicos Islands', 'Tuvalu', 'Uganda', 'Ukraine',
        'United Arab Emirates', 'United Kingdom', 'United States',
        'Uruguay', 'Uzbekistan', 'Vanuatu', 'Venezuela', 'Vietnam'
        'Wales', 'Wallis and Futuna', 'Yemen', 'Zambia', 'Zimbabwe'],
       dtype=object)
def size(m,n):
    fig = plt.gcf();
    fig.set_size_inches(m,n);
wordCloud = WordCloud(
    background color='white',
    max_font_size = 50).generate(' '.join(df.country))
plt.figure(figsize=(15,7))
plt.axis('off')
```

```
plt.imshow(wordCloud)
plt.show()
```

```
Sweden Sweden
                                                        Costa Rica
country_wise_total_vaccinated = {}
for country in df.country.unique() :
    vaccinated = 0
    for i in range(len(df)) :
        if df.country[i] == country :
            vaccinated += df.daily_vaccinations[i]
    country wise total vaccinated[country] = vaccinated
      made a seperate dict from the df
#
    country_wise_total_vaccinated_df =
pd.DataFrame.from_dict(country_wise_total vaccinated,
orient='index',
                                                          columns =
['total_vaccinted_till_date'])
      converted dict to df
country wise total vaccinated df.sort values(by =
'total_vaccinted_till_date', ascending = False, inplace = True)
country_wise_total_vaccinated_df
C:\Users\dell\AppData\Local\Temp/ipykernel 8784/469794026.py:6:
RuntimeWarning:
overflow encountered in long scalars
               total vaccinted till date
India
                               1070252854
United States
                                426856166
Brazil
                                277092585
Indonesia
                                196518393
```

Japan	189787850	
Niue Tokelau Burundi Pitcairn China	2484 1966 610 43 -2007914084	
[223 rows x 1 d	columns]	
datar = country	/_wise_total_vaccinated_df.j	oin(datal,how="inner")
datar.head(10)		
support \	total_vaccinted_till_date	GDP per capita Social
India 0.765	1070252854	0.755
Brazil 1.439	277092585	1.004
Indonesia 1.203	196518393	0.931
Japan 1.419	189787850	1.327
Mexico 1.323	125372969	1.070
Turkey 1.360	117406602	1.183
Germany 1.454	112306162	1.373
Pakistan 0.886	107371250	0.677
Russia 1.452	105282120	1.183
United Kingdom 1.538	102153121	1.333
,	Healthy life expectancy F	reedom to make life choices
\ India	0.588	0.498
Brazil	0.802	0.390
Indonesia	0.660	0.491
Japan	1.088	0.445

Mexico	0.861	0.433
Turkey	0.808	0.195
Germany	0.987	0.495
Pakistan	0.535	0.313
Russia	0.726	0.334
United Kingdom	0.996	0.450

	confirmed	deaths
India	34674643	474479
Brazil	22177059	616457
Indonesia	4258560	143918
Japan	1727654	18367
Mexico	3911714	296188
Turkey	8986377	78602
Germany	6427417	105000
Pakistan	1288366	28812
Russia	9782723	281571
United Kingdom	10722083	146592

from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
scaled = scaler.fit_transform(datar)

from pandas import DataFrame
data_scaled = DataFrame(scaled)
data_scaled.head()

0	1	2	3	4	5
6					
0 1.000000	0.448337	0.471059	0.466216	0.789223	1.000000
0.769686					
1 0.742327	0.596200	0.886084	0.672780	0.618067	0.639548
1.000000					
2 0.716151	0.552850	0.740764	0.535714	0.778130	0.122748
0.233456					
3 0.713965	0.788005	0.873768	0.948842	0.705230	0.049752
0.029790					
4 0.693038	0.635392	0.814655	0.729730	0.686212	0.112744
0.480466					

from sklearn.linear_model import LinearRegression
regression_model = LinearRegression()
from sklearn.metrics import mean_squared_error ,r2_score

```
x = data \ scaled.iloc[:, -7].values #attributes to determine dependent
variable/ Class
y = data scaled.iloc[:, -1].values #dependant variable/ Class
x=x.reshape(-1,1)
regression model.fit(x,y)
y predicted = regression model.predict(x)
mse = mean squared error(y,y predicted)
mse
0.013823870773761234
rmse = np.sqrt(mean squared error(y,y predicted))
r2 = r2 \text{ score}(y, y \text{ predicted})
r2
0.1449725320537083
x = data \ scaled.iloc[:, -7].values #attributes to determine dependent
variable/ Class
y = data scaled.iloc[:, -2].values #dependant variable/ Class
x=x.reshape(-1,1)
regression model.fit(x,y)
y predicte\overline{d} = regression model.predict(x)
mse = mean squared error(y,y predicted)
mse
0.009501470187026878
rmse = np.sqrt(mean squared error(y,y predicted))
r2 = r2 score(y,y predicted)
r2
0.22022315976696827
```

Multinomial regression

Performing multinomial regreesion on (Vaccination, GDP) vs Deaths due to Covid data_scaled

```
0
                                                                  5
                       1
                                  2
                                            3
                                                       4
6
     1.000000 0.448337 0.471059 0.466216 0.789223 1.000000
0
0.769686
     0.742327 0.596200 0.886084 0.672780 0.618067 0.639548
1.000000
     0.716151 \quad 0.552850 \quad 0.740764 \quad 0.535714 \quad 0.778130 \quad 0.122748
0.233456
3
     0.713965 0.788005 0.873768 0.948842 0.705230 0.049752
```

```
0.029790
     0.693038 0.635392 0.814655 0.729730 0.686212 0.112744
0.480466
. .
          . . .
                    . . .
                              . . .
                                        . . .
                                                  . . .
                                                            . . .
137 0.652380 0.207838 0.471675
                                   0.083977 0.275753 0.000088
0.000289
138 0.652353 0.181710 0.354064 0.183398 0.015848 0.000294
0.000211
139 0.652349 0.191805 0.423645 0.332046 0.041204 0.000666
0.001225
140 0.652309 0.027316 0.275246 0.265444 0.348653 0.000518
0.000057
141 0.000000 0.611045 0.692734 0.760618 0.825674 0.003156
0.007861
[142 rows x 7 columns]
from sklearn import linear model
import statsmodels.api as sm
from sklearn.metrics import r2 score
X = data scaled[[0,1]] # here we have 2 variables for multiple
regression. If you just want to use one variable for simple linear
regression, then use X = df['Interest Rate'] for
example. Alternatively, you may add additional variables within the
brackets
Y = data scaled[6]
# with sklearn
regr = linear model.LinearRegression()
regr.fit(X, Y)
print('Intercept: \n', regr.intercept )
print('Coefficients: \n', regr.coef )
print('R sqaure: \n')
regr.score(X, Y)
Intercept:
 -0.4866183114101143
Coefficients:
 [0.75670296 0.07414227]
R sqaure:
```

0.16382754007909928

Performing multinomial regreesion on (Vaccination, GDP) vs Deaths due to Covid using stastical models

```
# with statsmodels
X = sm.add_constant(X) # adding a constant
```

```
model = sm.OLS(Y, X).fit()
predictions = model.predict(X)

print_model = model.summary()
print(print_model)
```

D:\Users\dell\anaconda3\lib\site-packages\statsmodels\tsa\tsatools.py:142: FutureWarning:

In a future version of pandas all arguments of concat except for the argument 'objs' will be keyword-only

OLS Regression Results

Dep. Variable: 6 R-squared: 0.164

Model: OLS Adj. R-squared:

0.152

Method: Least Squares F-statistic:

13.62

Date: Fri, 10 Dec 2021 Prob (F-statistic):

3.98e-06

Time: 20:38:27 Log-Likelihood:

104.07

No. Observations: 142 AIC:

-202.1

Df Residuals: 139 BIC:

-193.3

Df Model: 2

Covariance Type: nonrobust

=========					
======	coef	std err	t	P> t	[0.025
0.975]					
const -0.280	-0.4866	0.104	-4.660	0.000	-0.693
0	0.7567	0.155	4.885	0.000	0.450
1.063 1 0.157	0.0741	0.042	1.770	0.079	-0.009

======

```
Omnibus:
                              168.792
                                        Durbin-Watson:
1.099
Prob(Omnibus):
                                0.000
                                        Jarque-Bera (JB):
4361.505
Skew:
                                4.557
                                        Prob(JB):
0.00
                               28.575
                                        Cond. No.
Kurtosis:
24.7
```

=======

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

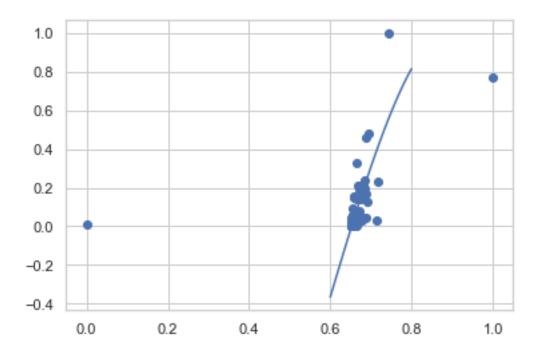
Polonomial Regression

```
X = data_scaled[0] # Vaccinated list of different countries
Y = data_scaled[6] # Covid Confirmed cases
```

Polynomial regression (Vaccinations vs Confirmed cases)

```
import numpy as np
mymodel = np.poly1d(np.polyfit(X, Y, 3))
myline = np.linspace(0.6, 0.8, 100)

plt.scatter(X, Y)
plt.plot(myline, mymodel(myline))
plt.show()
```



Rsquare value of Vaccinations vs Confirmed cases

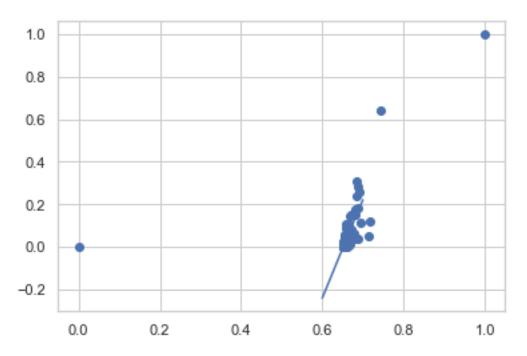
print(r2_score(Y, mymodel(X)))

0.6847344781455831

Polynomial regression (Vaccinations vs Deaths)

```
X = data_scaled[0] #Vaccinated list
Y =data_scaled[5] # Deaths due to covid
import numpy as np
mymodel = np.poly1d(np.polyfit(X, Y, 3))
myline = np.linspace(0.6, 0.7, 100)

plt.scatter(X, Y)
plt.plot(myline, mymodel(myline))
plt.show()
print("R2 value of Vaccinations vs Deaths:")
print(r2_score(Y, mymodel(X)))
```

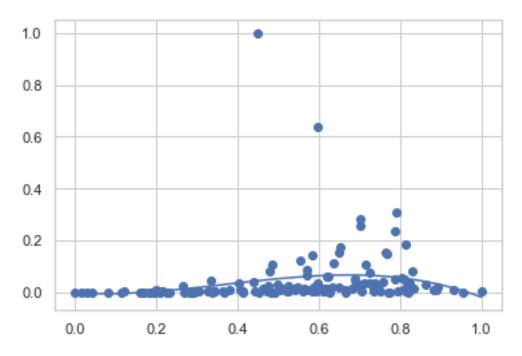


R2 value of Vaccinations vs Deaths: 0.8397867607203168

Polynomial Regression (GDP vs Deaths)

```
X = data_scaled[1]
Y =data_scaled[5]
import numpy as np
mymodel = np.polyld(np.polyfit(X, Y, 3))
myline = np.linspace(0, 1, 100)

plt.scatter(X, Y)
plt.plot(myline, mymodel(myline))
plt.show()
print("R2 value of Gdp vs Deaths:")
print(r2_score(Y, mymodel(X)))
```



R2 value of Gdp vs Deaths: 0.049696032274289315