Project Report

Subject: Python

Subject code: IT606

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Problem Statement:

The patient details from a dataset can't be found directly, such as knowing the growth rate or comparing the cases between the state, district, or nation regarding the active, confirmed, recovered, and deaths cases. The aim is to show the Exploratory Analysis to get insights of covid patients, covid growth on the state, district, national level.

INTRODUCTION:

COVID-19 outbreak was first reported in Wuhan, China, and has spread to more than 50 countries. Since the COVID-19 outbreak in China, the first coronavirus case in India, the second most populated country globally, was reported in Kerala. New confirmed cases were reported in cities such as New Delhi, Mumbai, Bengaluru, Hyderabad, and Patna. COVID-19 is still an unclear infectious disease. We have a lot of datasets on the patients affected by Covid-19. Here we have taken the dataset for India.

So, based on the reports, we have done an analysis.

Starting with importing the libraries

import numpy as np

Numpy: stands for Numerical Python, a library consisting of multidimensional array objects and a collection of routines for processing those arrays. Using NumPy, mathematical and logical operations on arrays can be performed. NumPy is a Python package.

import pandas as pd

Pandas: Pandas is a predominantly used python data analysis library. It provides many functions and methods to expedite the data analysis process.

import matplotlib.pyplot as plt

from matplotlib.pyplot import pie,axis,show

Matplotlib: Matplotlib is a cross-platform, data visualization and graphical plotting library for Python, and its numerical extension is NumPy.

import plotly.offline as pyo

from plotly.offline import iplot

Ploty: Plotly's Python graphing library makes interactive, publication-quality graphs. Examples of how to make line plots, scatter plots, area charts, bar charts, error bars, box plots, histograms, heatmaps, subplots, multiple-axes, polar charts, and bubble charts.

import cufflinks as cf

cf.go_offline()

Cufflinks: Library that connects the Pandas data frame with Plotly enabling users to create visualizations directly from Pandas. The library binds the power of Plotly with the flexibility of Pandas for easy plotting.

import plotly.io as pio

pio.renderers.default='colab'

Plotly.io: Low-level interface for displaying, reading, and writing figures.

import seaborn as sns

Seaborn: An open-source Python library built on top of matplotlib. It is used for data visualization and exploratory data analysis. Seaborn works efficiently with data frames and the Pandas library. The graphs created can also be customized easily.

We have imported from Google drive using the command:

drive.mount('/content/gdrive')

and then we have read four CSV files:

covid_data - contains the longitude and latitude of the state

district_data - contains information about the district

national_data - contains information about nation

state_data - contains information about state

Data Cleaning

Moving Forward to Data Cleaning part:

First, we have done some cleaning on district dataset

 We have renamed our columns using command:

```
district data.rename(columns={'Deceased':'Deaths'},inplace=True)
```

 We Summed up the null values of all the columns using command:

```
district_data.isna().sum()
```

• We dropped the unwanted columns

using command:

```
district_data =
district_data.drop(['District_Notes','Last_Updated'],axis
=1)
```

• We displayed the information about the district data set:

```
Using district data.info() command.
```

It gives an overview of the dataset, such as displaying the total number of rows and columns; for example, it has 800 rows and 14 columns. It also shows the data type of each column of our dataset and represents the total count of not null values of each column. It is generally used in exploratory data analysis.

 We checked all the negative values in the columns using command:

```
check = {}
for columns in district_data:
  if district_data[columns].dtypes == "int64":
    s = district_data[columns].lt(0).sum()
    check[columns]=s
print(check)
```

Convert the negative values to absolute value code:

```
for key in check:
  if check[key] != 0:
    district_data[key] = district_data[key].abs()
```

Verify it again

code:

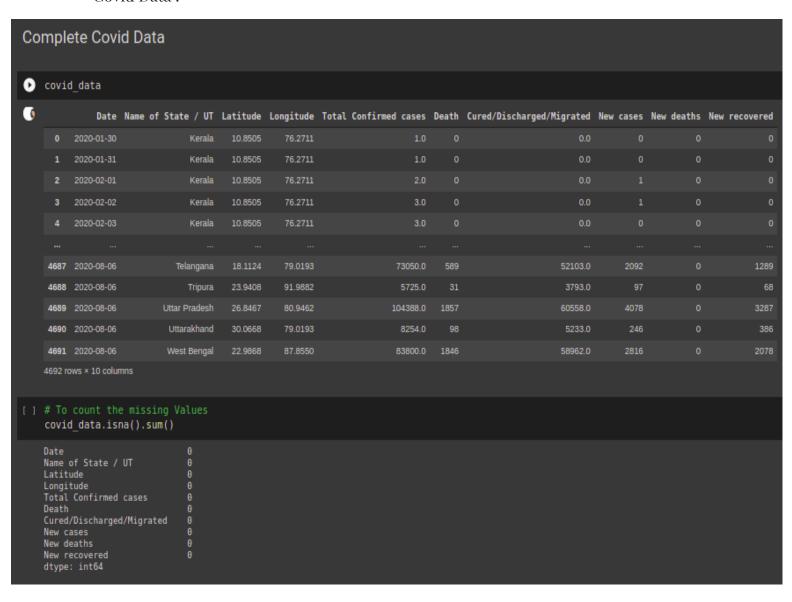
```
verify={}
for columns in district_data:
  if district_data[columns].dtypes == "int64":
    s = district_data[columns].lt(0).sum()
    verify[columns]=s
print(verify)
```

• describe the data

```
district_data.describe()
```

Similarly, we have cleaned all the other datasets like:

Covid Data:

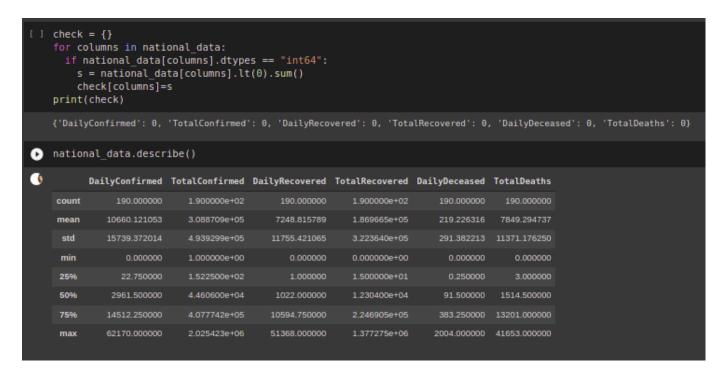


```
check={}
     for columns in covid_data:
       if covid_data[columns].dtypes != "object":
          check[columns]=s
     {'Latitude': 0, 'Longitude': 0, 'Total Confirmed cases': 0, 'Cured/Discharged/Migrated': 0, 'New cases': 0, 'New deaths': 0, 'New recovered': 3}
        if check[key] != 0:
          covid_data[key] = covid_data[key].abs()
     verify={}
     for columns in covid data:
       if covid_data[columns].dtypes != "object":
           verify[columns]=s
     print(verify)
[ ] covid data.info()
     <class 'pandas.core.frame.DataFrame'>
RangeIndex: 4692 entries, 0 to 4691
Data columns (total 10 columns):
# Column
          Date
Name of State / UT
Latitude
Longitude
Total Confirmed cases
                                                                 object
object
float64
float64
float64
           Death
Cured/Discharged/Migrated
New cases
New deaths
                                                                  object
float64
int64
int64
```

```
s = covid data[columns].lt(θ).sum()
           verify[columns]=s
     print(verify)
     {'Latitude': θ, 'Longitude': θ, 'Total Confirmed cases': θ, 'Cured/Discharged/Migrated': θ, 'New cases': θ, 'New deaths': θ, 'New recovered': θ}
    covid_data.info()
    <class 'pandas.core.frame.DataFrame'>
RangeIndex: 4692 entries, 0 to 4691
Data columns (total 10 columns):
# Column Non-N
                                             Non-Null Count Dtype
          Date
Name of State / UT
Latitude
                                                                 object
object
float64
                                             4692 non-null
4692 non-null
                                             4692 non-null
4692 non-null
           Total Confirmed cases
                                                                 float64
                                             4692 non-null
4692 non-null
                                                                 float64
                                             4692 non-null
4692 non-null
           New deaths
New recovered
                                                                 int64
int64
    dtypes: float64(4), int64(3), object(3)
memory usage: 366.7+ KB
[ ] covid data.describe()
                Latitude Longitude Total Confirmed cases Cured/Discharged/Migrated New cases New deaths New recovered
      count 4692.000000 4692.000000
       std
                                                        37208.600846
                                                                                         23390.671258 1259.748923
                                                                                                                                            947.925429
                                                                                                                                            0.000000
       min
                23.940800
                              79.019300
                                                          619.000000
                                                                                           197.500000
                                                                                                            26.000000
                                                                                                                                              8.000000
                                                        5233.000000
                                                                                          2736.000000
```

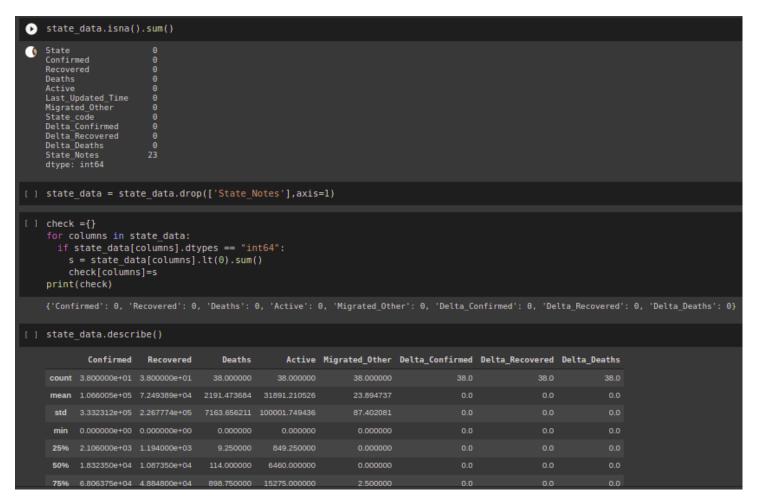
National Data:

```
National Level Data
    national data.columns = national data.columns.str.replace(" ","")
[ ] national data.columns
    [ ] national data.rename(columns={'TotalDeceased':'TotalDeaths'},inplace=True)
   national_data.isna().sum()
   Date
DailyConfirmed
TotalConfirmed
DailyRecovered
TotalRecovered
    DailyDeceased
TotalDeaths
    dtype: int64
[ ] national_data.dtypes
    Date
                       int64
    TotalConfirmed
DailyRecovered
TotalRecovered
                       int64
                       int64
                       int64
    DailyDeceased
TotalDeaths
                       int64
                       int64
    dtype: object
```



State Data:

ate Level Late	st data											
] state_data												
	State	Confirmed	Recovered	Deaths	Active	Last_Updated_Time	Migrated_Other	State_code	Delta_Confirmed	Delta_Recovered	Delta_Deaths	State_Notes
												NaN
	Maharashtra					06/08/2020 20:42:51						307 cases are marked as non-covid deaths in MH
												[July 22]: 444 backdated deceased entries adde
	Delhi				10348	06/08/2020 18:39:45						[July 14]: Value for the total tests conducted
	Karnataka											
	Andhra Pradesh	196789	112870			06/08/2020 19:45:03						Total includes patients from other states and \dots
	Uttar Pradesh											
	Gujarat					06/08/2020 21:28:40						NaN
	West Bengal											
	Telangana					06/08/2020 11:26:06						[July 27]\nTelangana bulletin for the previous
	Rajasthan											
	Bihar	68148	43820			06/08/2020 17:29:43						NaN
												[Aug 2]: 21 Foreign Evacuees have been merged
					15464	06/08/2020 23:00:49						Includes 1 case from Nagaland
	Madhya Pradesh											
15	Odisha		26888	280	13549	06/08/2020 20:44:03		OR				[July 12th] 20 non-covid deaths reported in st
	Kerala	30449			11983	06/08/2020 20:08:58						Mahe native who expired in Kannur included in
	Punjab											
19	Jharkhand		6594			06/08/2020 23:00:52						NaN
	Chhattisgarh											
21	Uttarakhand	8623	5427		3056	06/08/2020 23:46:40			0		0	NaN



DATA ANALYSIS

Starting with the analysis part:

Firstly, we extracted the date on which the nation had the highest number of recovered cases.

```
national_data ['Date'][national_data.TotalRecovered ==
national_data. TotalRecovered.max()]
```

Then using the District Data, we have extracted the name of the state which has the highest number of confirmed cases and the details of that State.

```
sc = district_data ['State'][district_data .Confirmed == district_data
.Confirmed.max()]
```

```
data = district_data .loc[district_data ['Confirmed'].idxmax()]
print("Detailed Information about the state having highest confirmed
cases \n",data)
```

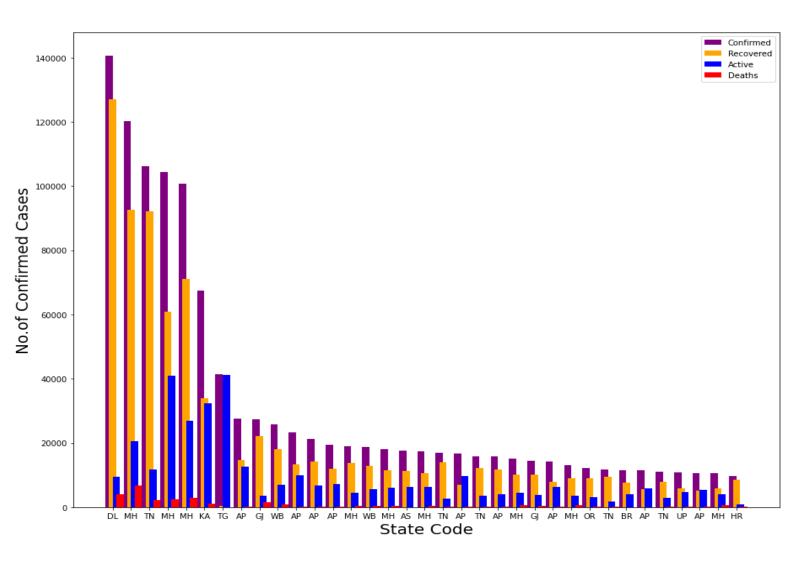
Next, we have created a new variable containing the top 35 places with the highest number of confirmed cases.

```
top35 = district_data .nlargest(35,'Confirmed')
```

Bar Plot:

Now using the top 35 data, we have plotted the bar graph, which displays the Confirmed, Recovered, Active, and Death cases according to the State Code. Here we can see of each State Code the highest number of the particular cases and how it varies as we move from left to right.

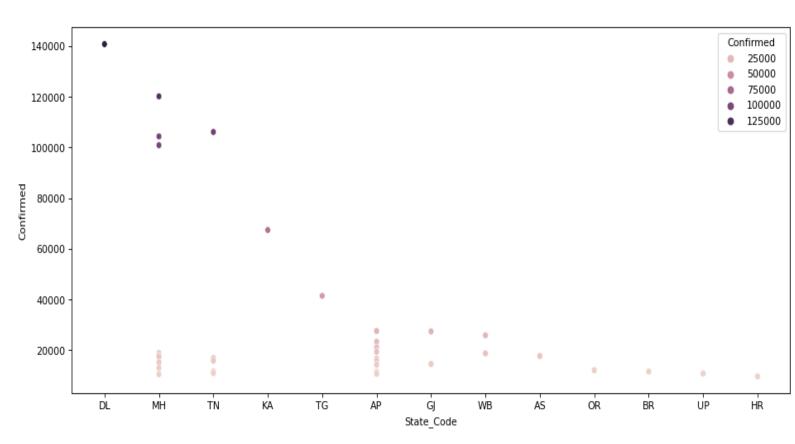
```
ind = np.arange(35)
width = 0.4
plt.figure(figsize=(15,12))
x = top35['State_Code']
y = top35['Confirmed']
plt.bar(ind+width/2,y,align='edge',width=width,label="Confirmed",
color='Purple')
y = top35['Recovered']
plt.bar(ind+width,y,align='edge',width=width,label="Recovered",
color='orange')
y = top35['Active']
plt.bar(ind+3*width/2,y,align='edge',width=width,label="Active",color='
Blue')
y = top35['Deaths']
plt.bar(ind+2*width,y,align='edge',width=width,label="Deaths",
color='Red')
plt.xticks(ind + 3*width/2, x)
plt.xlabel('State Code',fontsize=20)
plt.ylabel('No.of Confirmed Cases',fontsize=20)
plt.legend();
```



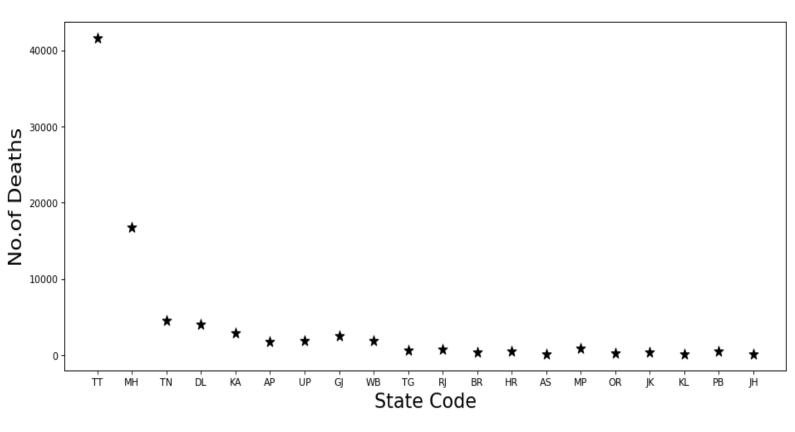
Scatter Plots:

Here we have a scatter plot that displays how the confirmed cases are spread in each state. Here the dark dots represent a higher no. of cases, and the light dots represent a lower no. of cases. Here we can see that states like Delhi, Maharashtra, and Tamil Nadu have more cases than others. In Maharashtra, we can say it had fewer cases in some time frame while it had more for a certain period of time.

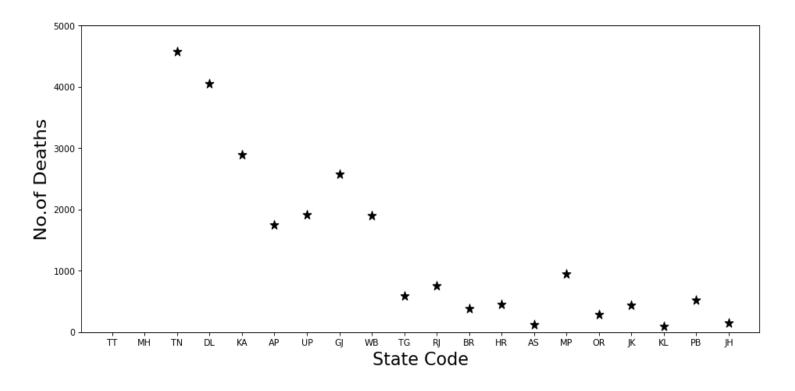
```
sc = sns.scatterplot(x='State_Code', y ='Confirmed',data = top35, hue =
'Confirmed')
```



In the next scatter plot, we can see how the death cases are spread. Here TT represents the total, and we can see it is more than 40000 and here, Maharashtra has the highest no. of death cases, and as we move to Jharkhand, it decreases. It shows that states between Tamil Nadu and Jharkhand have less than 10000 cases but a clear picture is not seen.



So using the following graph, we can see that there are states with death cases less than 1000 also.



HeatMap:

Here we have a heat-map that shows the correlation between each column. As we know, the correlation of anything with itself is one; so we can see that the diagonal values are one which saws the correlation of each column with itself. The value of the correlation between Confirmed and Recovered is 0.946, so we can say they are highly correlated i.e., as the number of Confirmed cases increases, the number of Recovered cases is also increasing.

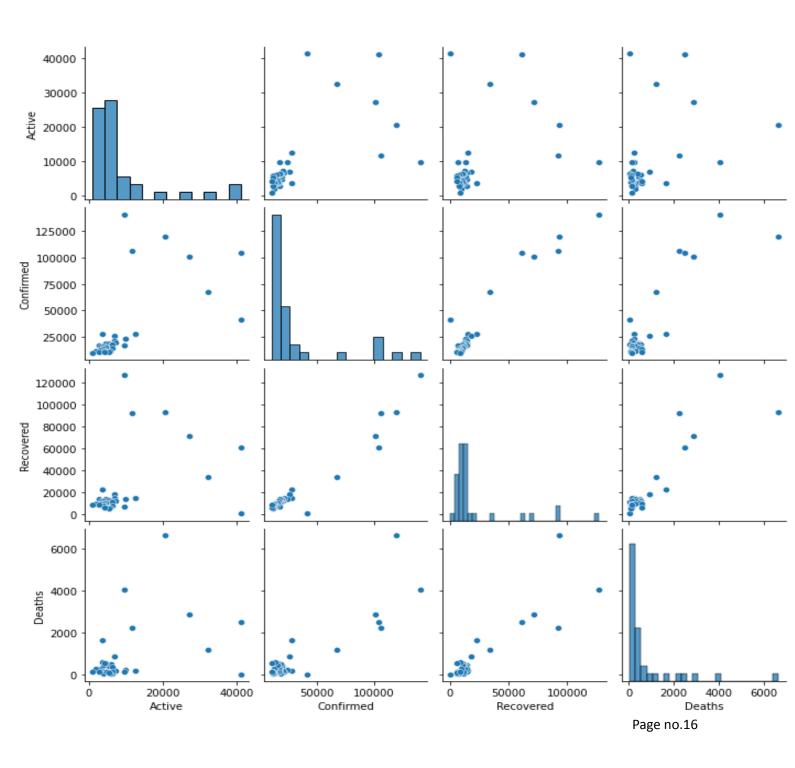
Similarly, the value between Migrated Other and Delta Active is 0.024, so we can say that they are the least related. If we had any negative value, we could say that they are negatively correlated i.e as one increases, the other decreases.

	Confirmed	Active	Recovered	Deaths	Migrated_Other	Delta_Confirmed	Delta_Active	Delta_Recovered	Delta_Deceased
Confirmed	1.000	0.661	0.946	0.900	0.417	0.688	0.554	0.665	0.744
Active	0.661	1.000	0.631	0.530	0.205	0.710	0.567	0.589	0.603
Recovered	0.946	0.631	1.000	0.900	0.412	0.542	0.437	0.565	0.650
Deaths	0.900	0.530	0.900	1.000	0.704	0.483	0.403	0.514	0.742
Migrated_Other	0.417	0.205	0.412	0.704	1.000	0.146	0.024	0.140	0.459
Delta_Confirmed	0.688	0.710	0.542	0.483	0.146	1.000	0.651	0.766	0.732
Delta_Active	0.554	0.567	0.437	0.403	0.024	0.651	1.000	0.775	0.691
Delta_Recovered	0.665	0.589	0.565	0.514	0.140	0.766	0.775	1.000	0.704
Delta_Deceased	0.744	0.603	0.650	0.742	0.459	0.732	0.691	0.704	1.000

Pairplot:

Pairplot is used to show the distribution of each variable and the relationship between them. Here for the relation between each column with itself, we have represented it using the histogram while the other relationships we have shown it using scatter plot.

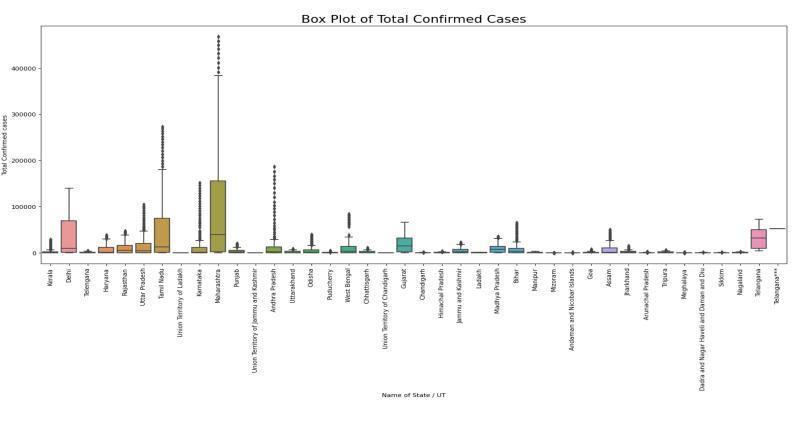
As seen in the above heatmap, the correlation between Confirmed and Recovered is highly positive and in the scatter plot, we can see that the values approximately lie in a straight line and for the value is scattering more the correlation between Active and Deaths.



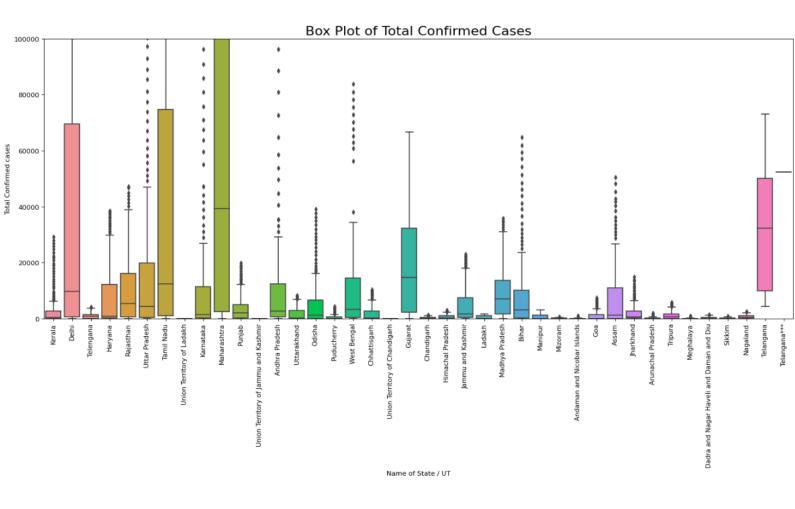
Box Plot:

Box plot is used to describe the distribution across different variables. The first line indicates the minimum value; the last line indicates the maximum value, and the middle value indicates the median value. The box represents the range of the maximum value lying for the particular variable. And the dots seen are the outliers.

```
plt.figure(figsize=(20,8), dpi= 80)
sns.boxplot(x='Name of State / UT', y='Total Confirmed cases',
data=covid_data, notch=False)
plt.xticks(rotation=90)
plt.title('Box Plot of Total Confirmed Cases', fontsize=20)
plt.show()
```



Taking the zoomed picture, we can see Telangana in the first line, which lies near 5500 is the minimum no. of confirmed cases, the second line, which lies near 70000, and the median line near 38000. The box in the middle ranges from about 12000 to 49000, indicating that the maximum number of cases lies between them.

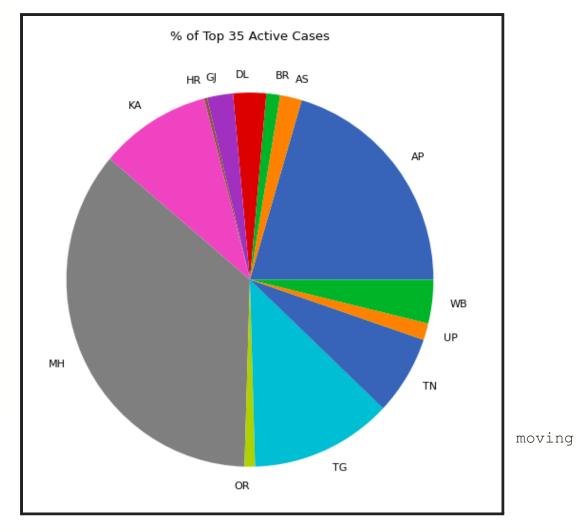


Pie Charts:

We have represented the pie charts to display the percentage share of cases present in our dataset.

So, we have displayed the percentage share of Top 35 active cases using the pie chart. Here we have grouped up the states. So what we are doing is summing up the top 35 cases according to their state and then storing its value in a variable called sums and then, with the help of the pie chart, representing the % share.

```
plt.figure(figsize=(15,8), dpi= 80)
sums = top35.groupby(top35['State_Code'])['Active'].sum()
axis('equal');
pie(sums, labels=sums.index);
plt.title("% of Top 35 Active Cases")
show()
```



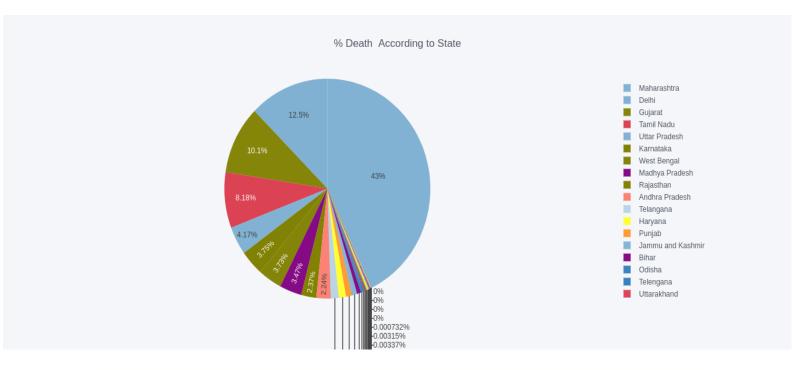
forward to our 2nd pie chart:

```
plt.figure(figsize=(20,20), dpi= 80)

data_plot=pd.DataFrame({'labels': covid_data['Name of State /
UT'],'counts': covid_data['Death']})
```

```
data_plot.iplot(kind='pie',labels='labels',values='counts',tit
le="% Death According to State")
plt.show()
```

here we declare a variable called data_plot and store the percentage count of Deaths according to their states and then display the pie chart using **data plot.iplot.**



As we can see, Maharashtra has the highest number of death cases at 43% followed by Delhi at 12.5%, Gujarat at 10.1%, and Tamil Nadu at 8.18%.

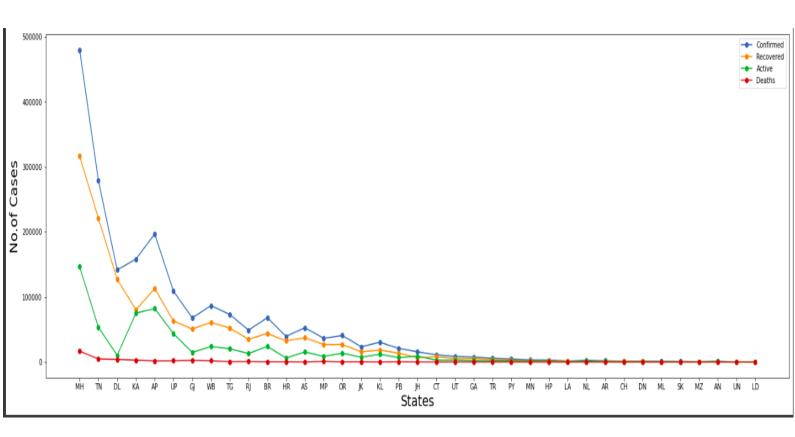
Line Graphs:

Line charts are used to represent the relation between two data X and Y on a different axis.

so, below we have shown the code of our first line graph, which shows the relationship between the no of cases and states.

```
plt.figure(figsize=(25,8))
x = state_data ['State_code']
y = state_data ['Confirmed']
plt.plot(x,y,marker='o',label="Confirmed")
y = state_data ['Recovered']
plt.plot(x,y,marker='o',label="Recovered")
y = state_data ['Active']
plt.plot(x,y,marker='o',label="Active")
y = state_data ['Deaths']
plt.plot(x,y,marker='o',label="Deaths")
plt.xlabel('States',fontsize=20)
plt.ylabel('No.of Cases',fontsize=20)
plt.legend();
```

Now, here we are going to show the Number of confirmed cases, recovered cases, active cases, and deaths cases in one frame with respect to their states.



As we can see that the blue line represents the number of Confirmed cases. Similarly, the orange line shows the no. of Recovered cases, the green line represents the number of Active cases and the red line represents the no. of Death cases.

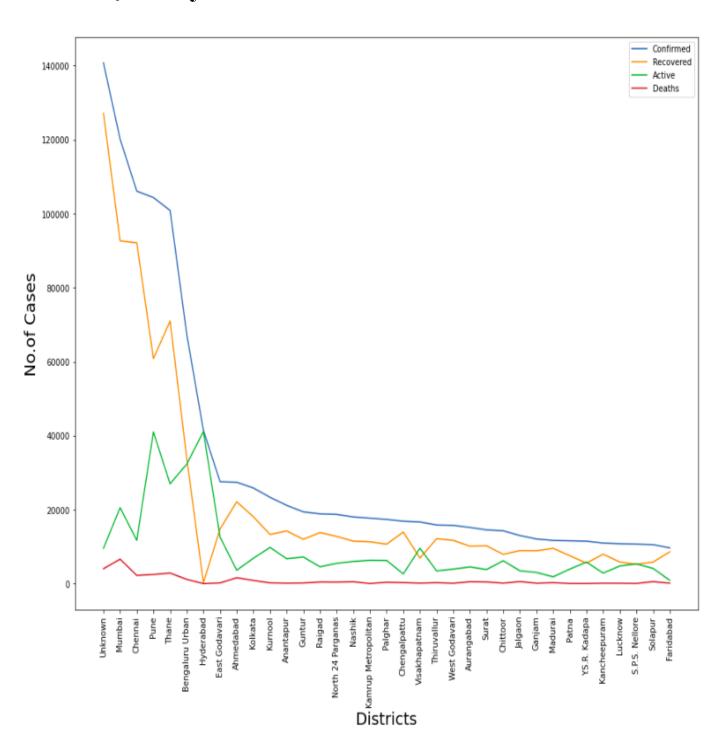
If we just focus on the line that shows the no. of confirmed cases, we can see that the highest number of confirmed cases are present in Maharashtra state and are decreasing as we move towards Lakshadweep.

Moving forward to another line graph, we show the number of confirmed, recovered, active, and deaths cases in one frame with respect to their districts.

below is the code:

```
ind = np.arange(35)
width = 0.4
plt.figure(figsize=(15,12))
x = top35['District']
y = top35['Confirmed']
plt.plot(x,y,label="Confirmed")
y = top35['Recovered']
plt.plot(x,y,label="Recovered")
y = top35['Active']
plt.plot(x,y,label="Active")
y = top35['Deaths']
plt.plot(x,y,label="Deaths")
plt.xticks(rotation=90)
plt.xlabel('Districts',fontsize=20)
plt.ylabel('No.of Cases',fontsize=20)
plt.legend();
```

and now here we are displaying a plot of the line graph of the top 35 cases according to their districts.



we can see that here also, the blue line represents the no. of Confirmed cases, orange represents the no. of Recovered cases, green represents no. of Active cases, red represents no. of Death cases.

As we can see in the above graph that there are different peak points for different types of cases, for example, if see the graph of no. of death cases, Mumbai has the highest number of death cases and similarly, if we see the graph of no. of active cases, Pune and Hyderabad are pretty close.

Now Moving forward to our 3rd line graph that represents the no. of Active cases VS no. of Recovered cases.

below is the following code:

```
fig = plt.figure(figsize = (18,8))

x=district_data ['Active']

y=district_data ['Recovered']

plt.plot(x, label='Active')

plt.plot(y, label='Recovered')

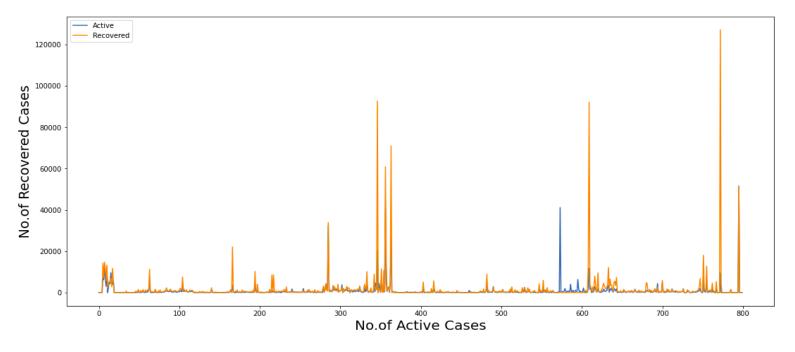
plt.legend()

plt.xlabel('No.of Active Cases', fontsize=20)

plt.ylabel('No.of Recovered Cases', fontsize=20)

plt.show()
```

We are displaying the line graph showcasing the no. of active cases with respect to no. of recovered cases, the blue line shows the no. of active cases and the orange line shows the no. of recovered cases.



here the x-axis shows the no. of Active cases and the y-axis shows the no. of Recovered cases. we can see that the no. of Active cases, in the range of 550 to 600 are more than the no. of recovered cases. while other peak points show the no. of Recovered cases are more than the no. of Active cases.

Conclusion

Through this program, we tried to perform a detailed data analysis on the Covid-19 outbreak in India. We have used different python libraries useful in performing Big-data analysis like NumPy, Pandas, Matplotlib, Plotly, Seaborn, etc. We performed various diagrammatic and statistical analyses on data generated from different sources using bar graphs, Scattered plots, pie charts, line graphs, Box plot and many more. Thus, providing an analytical view of the outbreak, which can further be used in planning and estimation.

Python provides a wide range of features for not just developing interfaces but also for handling & analyzing data. In the near future, we can even perform statistical tool of hypothesis testing and modelling using python libraries such as pandas, numpy, matplotlib etc, and can provide a conclusive remark to the analysis performed above in a computerized form.

Bibliography

Link: https://pandas.pydata.org/

Link: https://matplotlib.org/

Link: https://numpy.org/

Link :https://plotly.com/

Link: https://seaborn.pydata.org/

Link:

 $\underline{https://www.kaggle.com/saurabhprakashgiri/covid-19-india-mar-aug-analysis-visu\ alization/}$

Link: https://www.geeksforgeeks.org/