

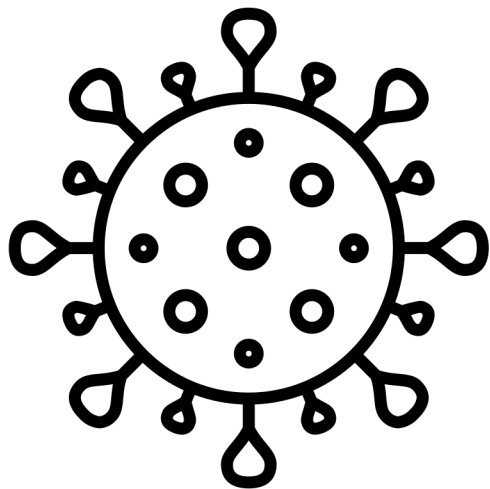
# DATA SCIENCE MINI PROJECT

TEAM 11

AGGARWAL ANUSHA

SARAF ISHITA





*Dataset used:  
Covid Tracking  
Project*



Aim:

- To help analyze the current COVID situation in different states by means by relevant visualizations
- To predict useful variables such as new cases, active cases, deaths, hospitalizations, probability of an individual getting COVID, etc. to help the state governments, testing agencies, hospitals, and individuals during the pandemic

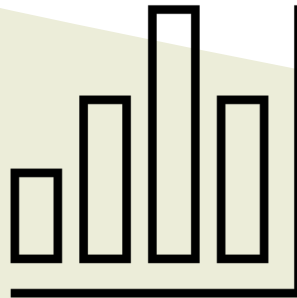


## Importing US Statewise population

```
StateWisePopulation = pd.read_csv('US Statewise Population.csv')  
StateWisePopulation.head()
```

	state	Current Population
0	AK	731545
1	AL	4903185
2	AR	3017825
3	AS	55641
4	AZ	7278717

# Data Collection



# Data Cleaning and Data Merging



## Data Frame only containing relevant variables

```
CovidUseful= pd.DataFrame(CovidData[["date","state","positive","recovered","death",  
                                     "positiveTestsPeopleAntibody","positiveIncrease","deathIncrease","hospitalizedIncrease",  
                                     "totalTestResultsIncrease","hospitalized", "totalTestResults","positiveCasesViral"]])
```

## Setting the NaN values in the data to zero

```
CovidUseful= CovidUseful.replace(np.nan, 0)
```

## Converting "date" from string to a datetime object

```
CovidUseful['date'] = pd.to_datetime(CovidUseful['date'], format='%Y%m%d')
```

## Grouping Data according to states

```
StatewiseData = CovidUseful.groupby(["state"])
```

## Function to obtain data frame for each state

```
def StateDataFrame(StatewiseData, state):  
    return StatewiseData.get_group(state).reset_index(drop=True)
```



# Data Curation

CALCULATE NUMBER OF PEOPLE  
CURRENTLY INFECTED WITH COVID  
(ACTIVE CASES)



Function to calculate number of people currently infected with covid (active cases)

*function that creates a new column in a state's data frame of the people who have covid currently*

```
In [19]: def addCurrentlyInfected (stateDF, state):  
stateDF['activeCases'] = stateDF['positive'] - stateDF['recovered'] - stateDF['death']  
return stateDF
```





# *Data Curation*

## CALCULATE DAILY POPULATION

Function to calculate daily population

*Calculating daily growth in population*

considering daily births, deaths and migration

```
In [14]: DailyGrowth = 1*24*60*60/40
```

*function that creates a new column in a state's data frame of the daily population*

```
In [16]: def addPopulation(StateWisePopulation, StatewiseData, state):  
         stateDF=StateDataFrame(StatewiseData, state)  
         stateDF['DailyPopulation']=StateWisePopulation[StateWisePopulation['state'] == state]['Current Population'].values[0]  
         stateDF.at[0, 'DailyPopulation']=stateDF.at[0, 'DailyPopulation']-stateDF.at[0, 'deathIncrease']  
         for i in range(1, len(stateDF['date'])):  
             stateDF.at[i, 'DailyPopulation']= stateDF.at[i-1, 'DailyPopulation']+DailyGrowth-stateDF.at[i, 'deathIncrease']  
         return stateDF
```



# Data Curation

CALCULATE THE PROBABILITY OF A  
PERSON GETTING COVID ON A  
PARTICULAR DAY

## Function to calculate the probability of a person getting covid on a particular day

*function that creates a new column in a state's data frame of the probability of a person getting COVID on that day*

Probability of a person getting covid on a particular day= No. of positive cases on that day/No. of people who can have COVID on that day  
Probability of a person getting covid on a particular day= No. of positive cases on that day/(Population on that day - People who have antibodies on that day - Active cases on that day)

```
In [21]: def findProbability(StatewisePopulation,StatewiseData,state):
         stateDF = addPopulation( StatewisePopulation, StatewiseData, state)
         stateDF= addAntibodies(stateDF, state)
         stateDF= addCurrentlyInfected(stateDF, state)
         stateDF['peopleWhoCanGetCOVID']=stateDF['DailyPopulation']-stateDF['immunisedPopulation']-stateDF['activeCases']
         stateDF['probabilityOfGettingCOVID']=stateDF['positiveIncrease']/stateDF['peopleWhoCanGetCOVID']
         return stateDF

         AlaskaDF=findProbability(StatewisePopulation,StatewiseData, 'AK')
         AlaskaDF.head()
```

```
Out[21]:
```

stResults	positiveCasesViral	DailyPopulation	positiveAntibodiesIncrease	immunisedPopulation	activeCases	peopleWhoCanGetCOVID	probabilityOfGettingCOVID
1731628.0	0.0	731545	0.0	0.0	56581.0	674964.0	0.000000
1731628.0	0.0	733705	0.0	0.0	56581.0	677124.0	0.000000
1731628.0	0.0	735863	0.0	0.0	56581.0	679282.0	0.000208





# Data Curation

CALCULATE NUMBER OF PEOPLE  
WHO HAVE ANTIBODIES AND  
IMMUNITY AGAINST COVID

Function to calculate number of people who have antibodies and immunity against covid currently

*function that creates a new column in a state's data frame of the people who have antibodies currently*

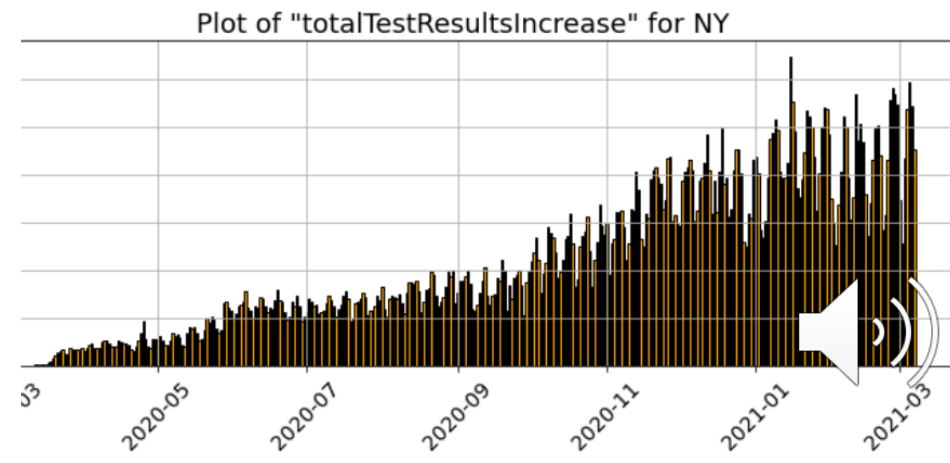
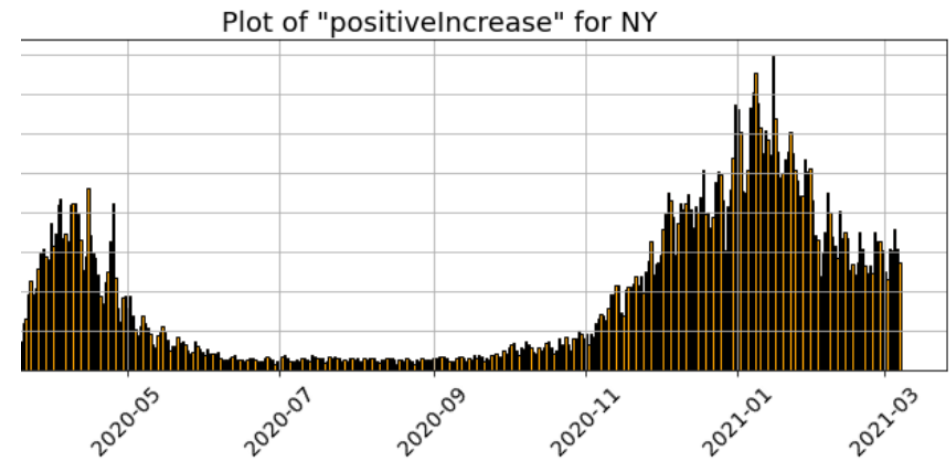
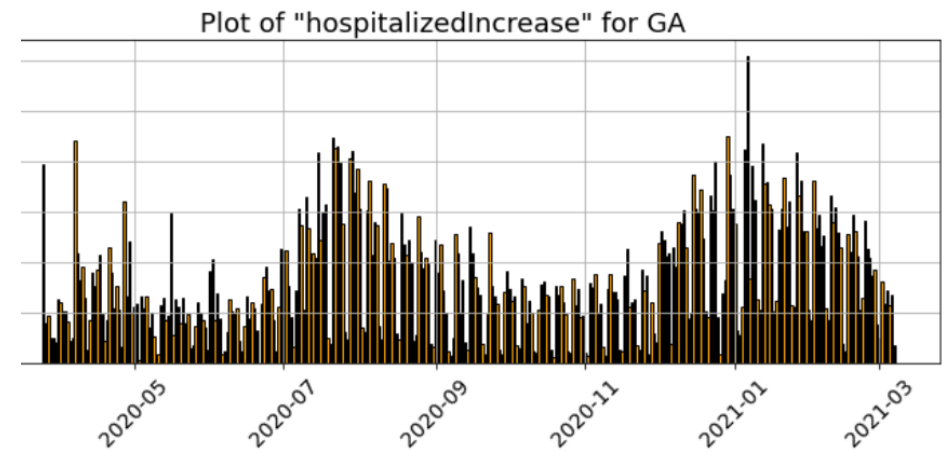
according to CDC, antibodies and immunity against COVID officially lasts for 3 months (90 days) on average

```
In [17]: def addAntibodies(stateDF, state):
stateDF['positiveAntibodiesIncrease'] = stateDF['positiveTestsPeopleAntibody']
for k in range(len(stateDF['date'])-1, 0, -1):
    stateDF.at[k, 'positiveAntibodiesIncrease'] = stateDF.at[k, 'positiveTestsPeopleAntibody'] - stateDF.at[k-1, 'positiveTestsPeopleAntibody']
stateDF['immunisedPopulation'] = stateDF['positiveAntibodiesIncrease']
for i in range(0, len(stateDF['date'])):
    for j in range(1, 90):
        if (i-j) >= 0:
            temp = stateDF.at[i-j, 'positiveAntibodiesIncrease']
            stateDF.at[i, 'immunisedPopulation'] = stateDF.at[i, 'immunisedPopulation'] + temp
        else:
            break
    return stateDF
```



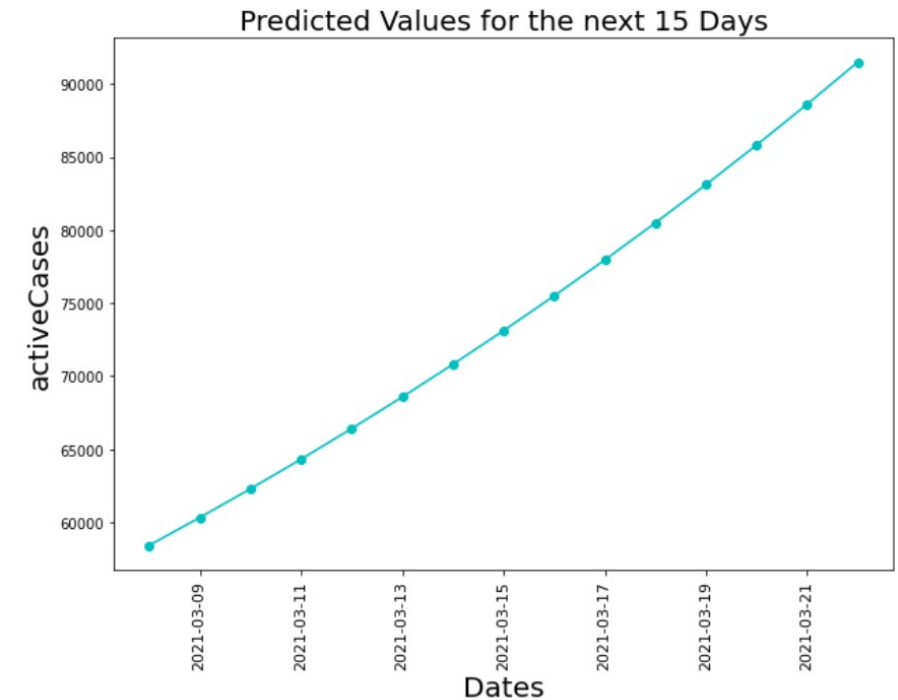
# Exploratory Data Analysis ( EDA )

## BAR CHARTS



# Method 1: Using average growth factor

Average growth factor 1.0325617183776112



# Method 2: ARIMA Model

## STEP 1: CONVERTING DATA TO TIME SERIES

### i. Converting data to a time series

*function to convert the generated date-probability dataframe into a series*

```
def series(stateDF):  
    stateDF.drop(['state'], axis=1)  
    stateDF= stateDF.set_index('date')  
    TS=stateDF[['probabilityOfGettingCOVID']].squeeze()  
    return TS
```

```
AlaskaTS= series(AlaskaDF)  
print(AlaskaTS)  
print(type(AlaskaTS))
```

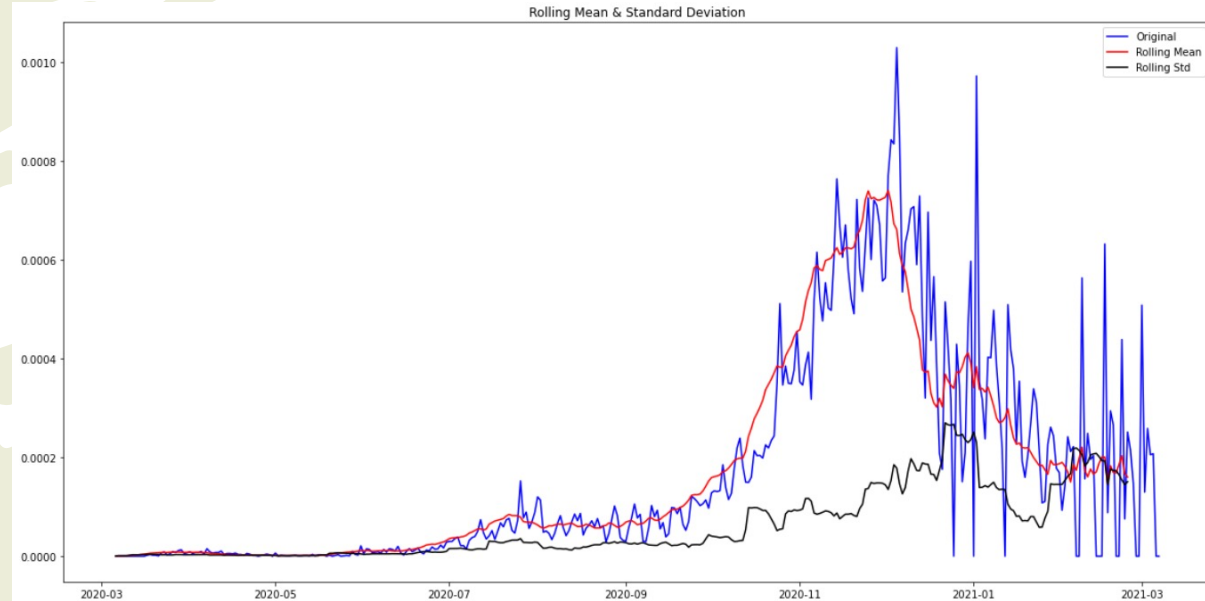
```
date  
2021-03-07    0.000000  
2021-03-06    0.000000  
2021-03-05    0.000208  
2021-03-04    0.000205  
2021-03-03    0.000259
```

```
...  
2020-03-10    0.000000  
2020-03-09    0.000000  
2020-03-08    0.000000  
2020-03-07    0.000000  
2020-03-06    0.000000
```

```
Name: probabilityOfGettingCOVID, Length: 367, dtype: float64  
<class 'pandas.core.series.Series'>
```



- STEP 2: CHECKING STATIONARITY OF TIME SERIES



Results of Dickey-Fuller Test:

Test Statistic	-1.254630
p-value	0.649665
#Lags Used	12.000000
Number of Observations Used	354.000000
Critical Value (1%)	-3.448958
Critical Value (5%)	-2.869739
Critical Value (10%)	-2.571138

dtype: float64

Results of KPSS Test:

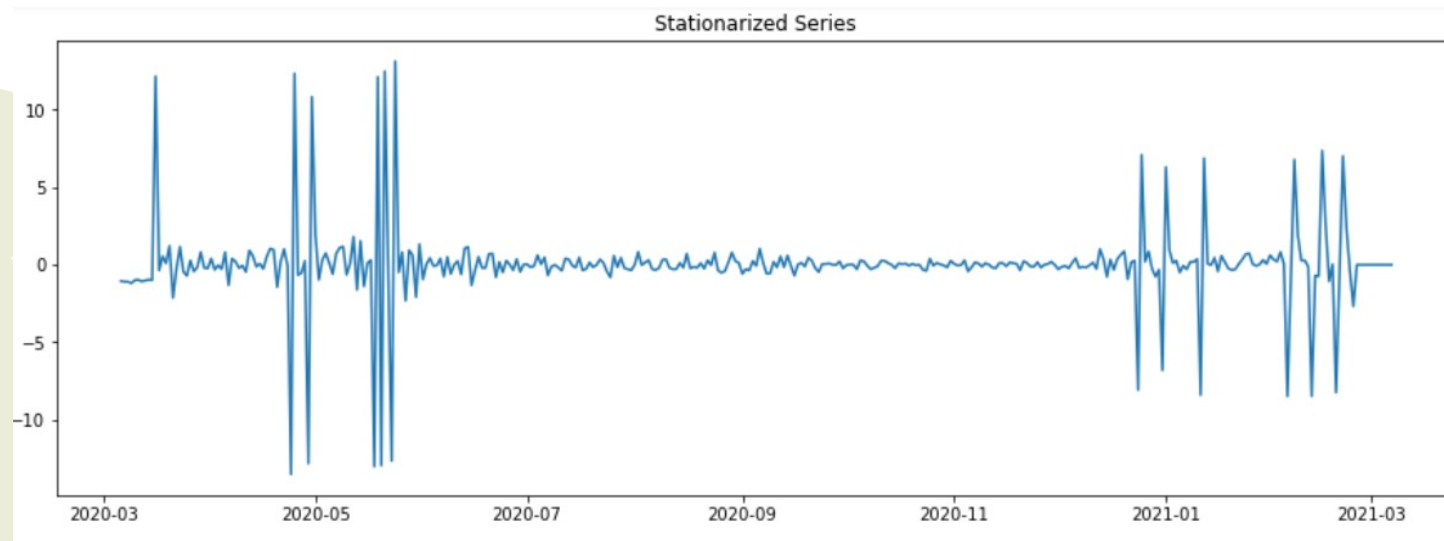
Test Statistic	1.167364
p-value	0.010000
Lags Used	17.000000
Critical Value (10%)	0.347000
Critical Value (5%)	0.463000
Critical Value (2.5%)	0.574000
Critical Value (1%)	0.739000

dtype: float64

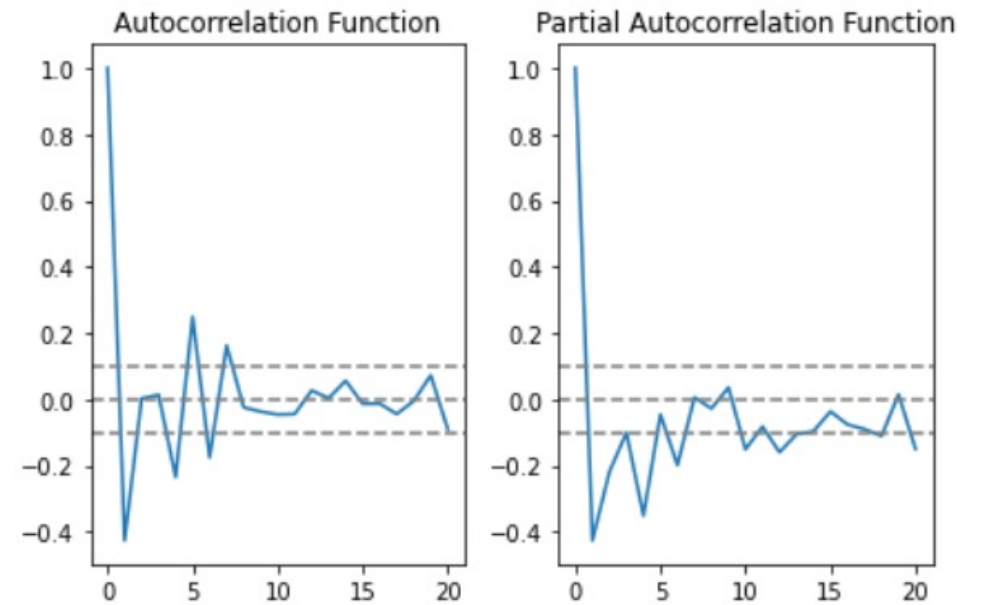
Since the Test Statistic of Dickey-Fuller Test is bigger than all the critical values we cannot say that the time series is stationary



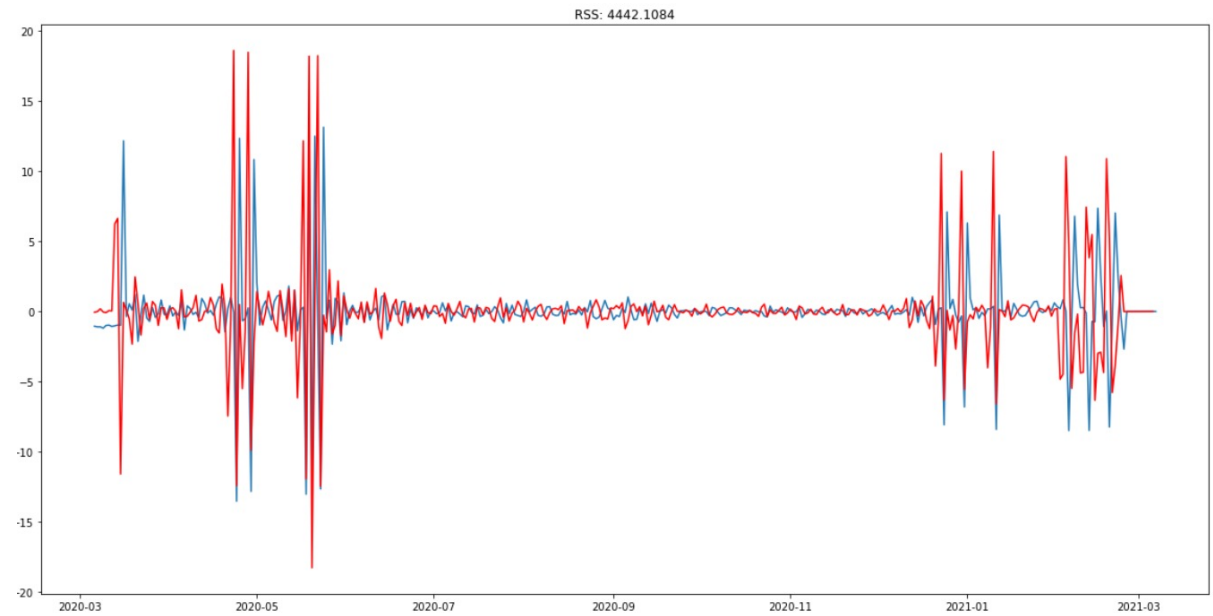
- STEP 3: MAKING THE TIME SERIES STATIONARY



- STEP 4: FINDING THE BEST FIT PARAMETERS FOR THE ARIMA MODEL



- STEP 5: FITTING THE ARIMA MODEL





- STEP 6: ACCURACY

#### vi. Testing Accuracy of fitted model

*function to compute accuracy measures for the fitted model*

Mean Error (ME)

Mean Absolute Error (MAE)

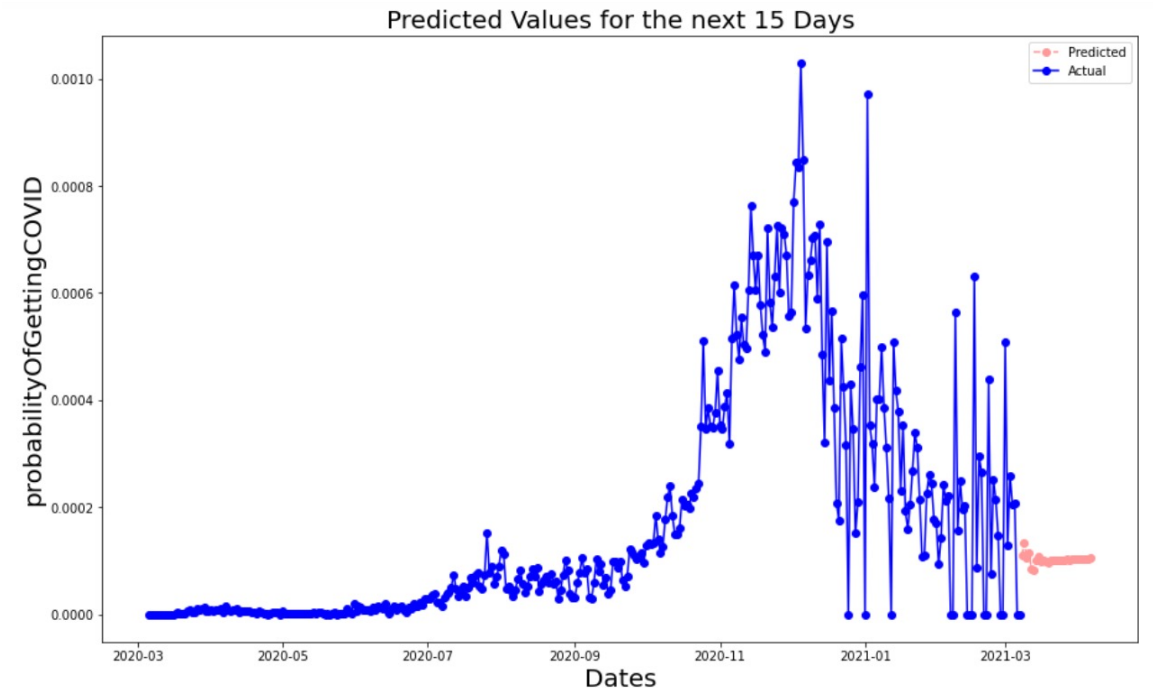
Root Mean Squared Error (RMSE)

Correlation between the Actual and the Forecast (corr)

```
def forecast_accuracy(forecast, actual):  
  
    me = np.mean(forecast - actual)           # ME  
    mae = np.mean(np.abs(forecast - actual))   # MAE  
    rmse = np.mean((forecast - actual)**2)**.5 # RMSE  
    corr = np.corrcoef(forecast, actual)[0,1] # corr  
  
    return({'me':me, 'mae': mae,  
           'rmse':rmse,'corr':corr})
```



- STEP 7: PREDICTING  
VALUES FOR NEXT 15 DAYS  
S



# *User Interface*

```
51 : VI  
52 : VT  
53 : WA  
54 : WI  
55 : WV  
56 : WY
```

Enter the state you wish to get predictions for: NY

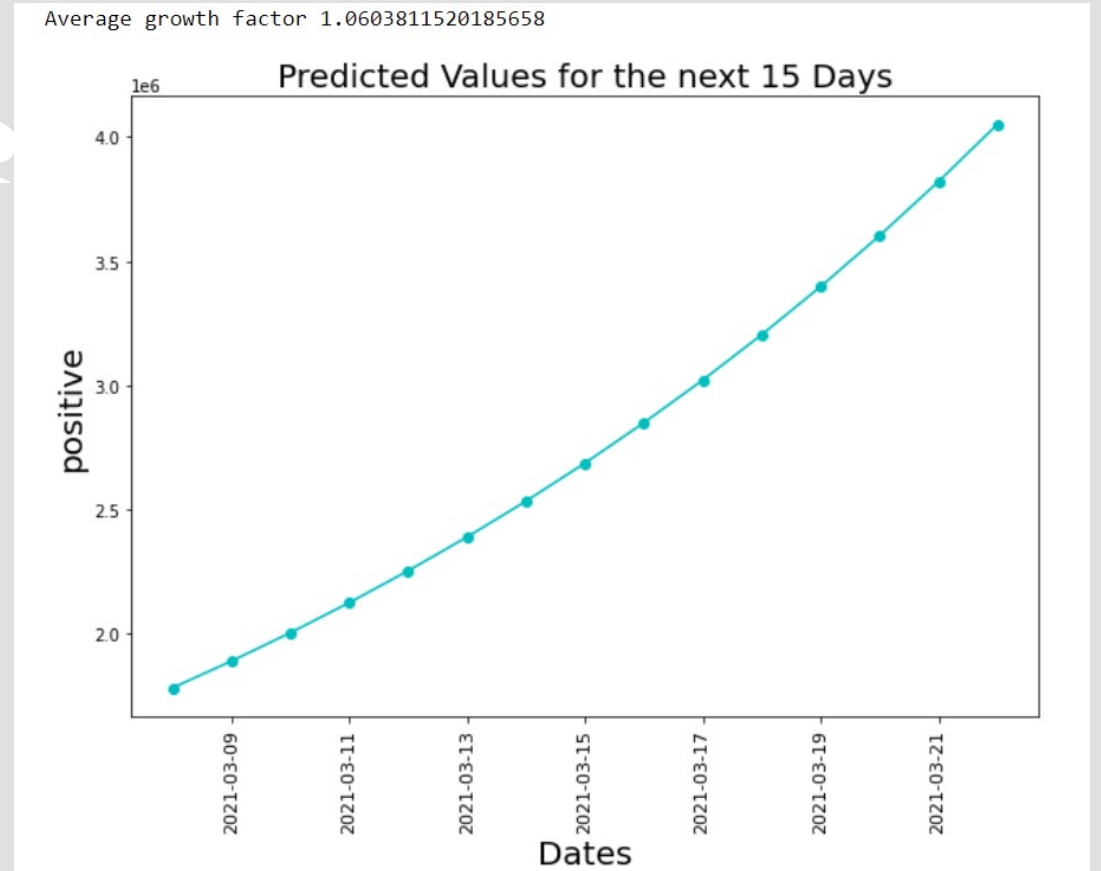
Possible values to predict for next 15 days:

1. positive: Total confirmed and probable cases in the state
2. recovered: Total number of people who have recovered from COVID in the state
3. death: Total number of confirmed and probable deaths due to COVID in the state
4. hospitalizedCumulative: Total number of people who have been hospitalized due to COVID in the state
5. inIcuCumulative: Total number of people who have been in ICU due to COVID in the state
6. onVentilatorCumulative: Total number of people who have been on ventilator due to COVID in the state
7. immunisedPopulation: Total number of people who who have antibodies currently i.e. immune to infection
8. activeCases: Total number of people who currently have COVID in the state
9. probabilityOfGettingCOVID: Probability of a person getting COVID on a particular day in the state

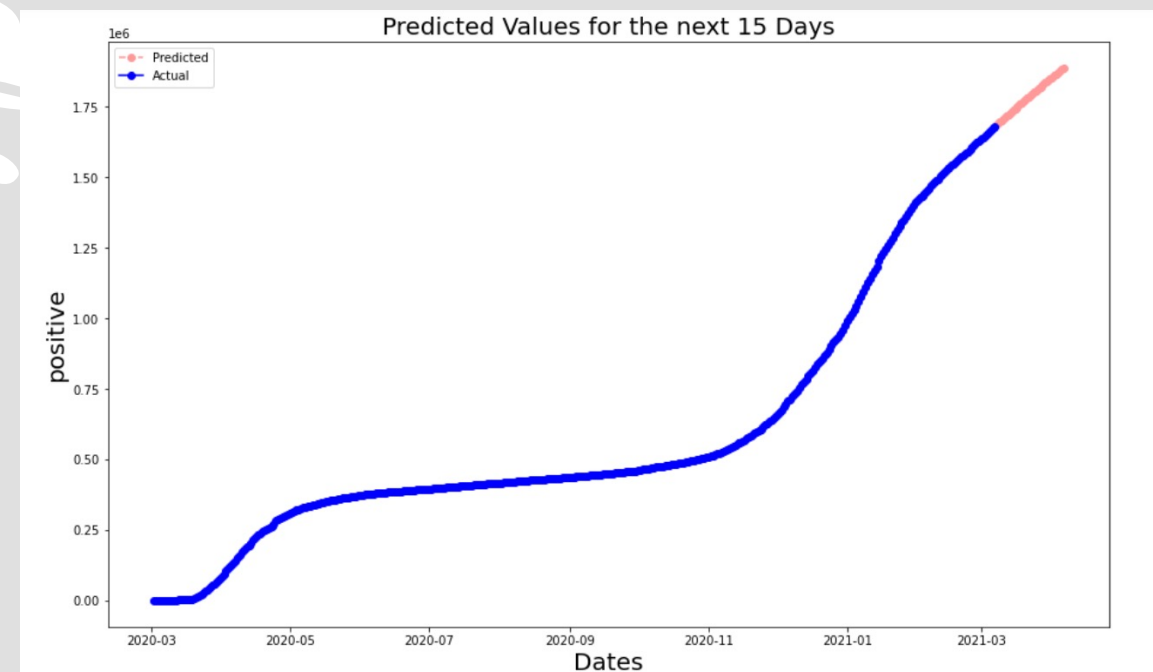
Enter the value you want to predict for the next 15 days: positive



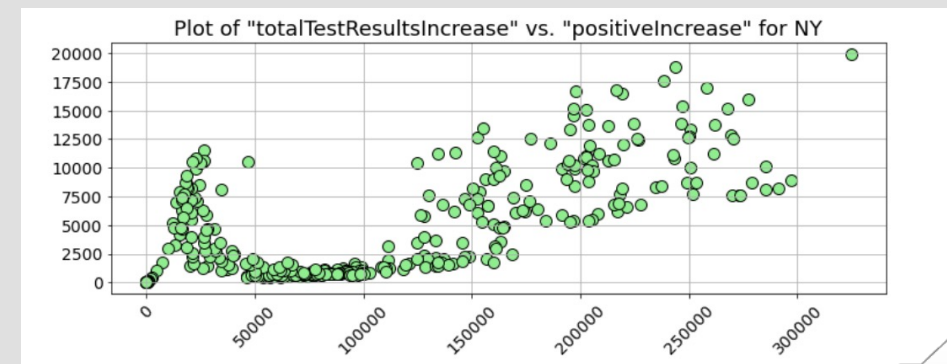
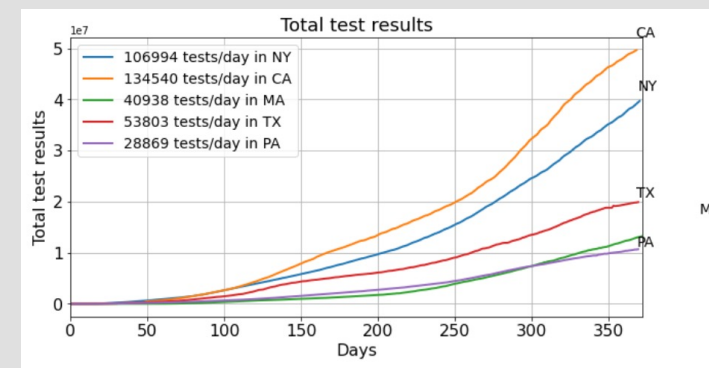
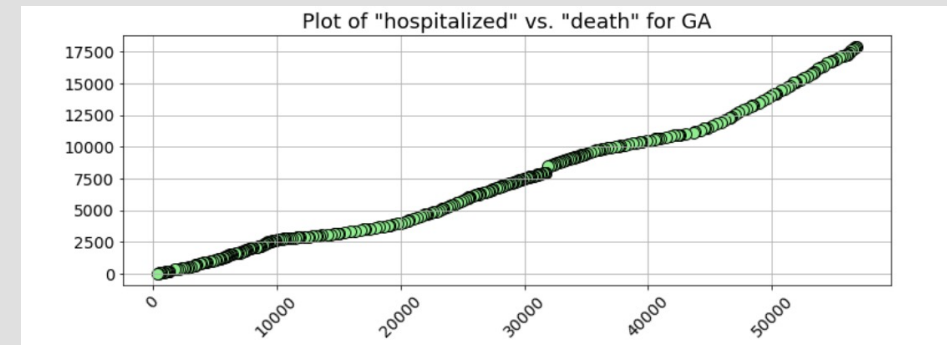
# Result 1:



# Result 2:



# Data driven insights



# Conclusion



<https://www.analyticsvidhya.com/blog/2016/02/time-series-forecasting-codes-python/>

<https://www.machinelearningplus.com/time-series/arma-model-time-series-forecasting-python/>

# Citation

