A Project Report

on

COVID-19 CASE PREDICTION USING ML

Submitted in partial fulfillment of the requirement for the award of the degree of

Bachelor of Technology in Computer Science and Engineering



(Established under Galgotias University Uttar Pradesh Act No. 14 of 2011)

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SCHOOL OF COMPUTING SCIENCE AND ENGINEERING DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING GALGOTIAS UNIVERSITY, GREATER NOIDA INDIA December, 2021



SCHOOL OF COMPUTING SCIENCE AND ENGINEERING GALGOTIAS UNIVERSITY, GREATER NOIDA

CANDIDATE'S DECLARATION

I/We hereby certify that the work which is being presented in the project, entitled "COVID-19 CASE PREDETECTION USING ML: A cross-platform Application " in partial fulfillment of the requirements for the award of the

BACHELOR OF TECHNOLOGY IN COMPUTER SCIENCE AND ENGINEERING submitted in the School of Computing Science and Engineering of Galgotias University, Greater Noida, is an original work carried out during the period of JULY-2021 to DECEMBER-2021, under the supervision of Ms. PRIYANKA SHUKLA, Assistant Professor, Department of Computer Science and Engineering of School of Computing Science and Engineering , Galgotias University, Greater Noida

The matter presented in the project has not been submitted by me/us

for the award of any other degree of this or any other places.

19SCSE1010545 – CHATURVEDI VENUGOPAL 19SCSE1010672 – ANAND KUMAR

This is to certify that the above statement made by the candidates is correct to the best of my knowledge.

Supervisor

(Ms. Priyanka Shukla, Assistant Professor)

CERTIFICATE

The Final Thesis/Project/ Dissertation Viva-Voce examination of 19SCSE1010545 -CHATURVEDI VENUGOPAL, 19SCSE1010672 – ANAND KUMAR has been held on ______and his/her work is recommended for the award of BACHELOR OF TECHNOLOGY IN COMPUTER SCIENCE AND ENGINEERING.

Signature of Examiner(s)

Signature of Supervisor(s)

Signature of Project Coordinator

Signature of Dean

Date:

Place:

ABSTRACT

COVID-19 has already been declared as a pandemic by World Health Organization. Huge number of tests are being carried out across the country and there is a large number COVID positive patients. Everyone wants to know that what will be the numbers in future and when will be the peak and when will the number of cases drop. We will predict the number of cases as per the current and past scenario. Currently, COVID cases are being predicted on the basis of previously infected person and number of patients recovered. But, nowadays vaccination is carried out in different parts of the world including our country India. As people are getting vaccinated, the risk of them getting infected is reducing day by day. So, we need to predict taking all the following variables (including number of people vaccinated and number of people infected after vaccination) accurately in future. Also, we have to check the effectiveness of each and every vaccine and then predict the chances of people getting infected. This will help in getting the very accurate data what will be the scenario in future. Three types of vaccine are available in India (COVAXIN, COVISHIELD, SPUTNIK V). All have different effectiveness. So, after collecting all the data as we can predict future number of cases using machine learning algorithm.

Keywords—pandemic, peak, drop, vaccination, infection, effectiveness

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CHAPTER 1

INTRODUCTION

Coronavirus disease (COVID-19) is an infectious disease caused by the SARS-CoV- 2 virus. It is spreading continuously among the population of a particular locality. This spread of virus has almost stopped the whole world and is also responsible for decrease in economy of our country and many other countries of the world. Lockdown were imposed to contain the spread of this virus. There were a lot of cases increasing on daily basis and fatality rate also increased. The government had taken various measures to protect our population from this pandemic. This has not only affected the whole country economically, but has also caused financial as well as health problems on the individual level. Many lost their jobs and a lot of people have faced losses in business. The overall point is that due to this Corona virus pandemic, people have lost many things. Its affect reduces for sometimes but it again comes back in the form of wave and lefts people losing lives, jobs, business etc. Everyone wants to know the future that when will this pandemic end, when will be the peak phase to get prepared for the next waves to come.

Here is the list of number of infected people in different states of

India.	Top 5	most	affected	states	by	coronav	irus	disease	201	19:
--------	-------	------	----------	--------	----	---------	------	---------	-----	-----

State	Case Reporte d	Death	Recovere d
Maharashtr	66.3L	1.41L	63.5L
а			

Kerala	51L	37,495	50.4L
Karnataka	29.9L	38,175	29.5L
Tamil Nadu	27.2L	36,375	26.7L
Andhra Prades h	20.7L	14,426	20.5L

There are a lot of factors on which the spread of Covid 19 (SARS-CoV-2 virus) depends. We will have to find all the factors and then find there dependencies on each other. After finding the dependency, we can classify the factors as dependent and independent variables, which we will further use for predicting the impact of covid 19 on our population and we will be able to predict the cases with accuracy. Some of the factors on the which the future number of new infected people depends are percentage of people infected in a particular area, number of people recovered, number of people who have got both the doses of vaccination etc. With all the details, we will make a machine learning model to predict the future number of cases. This can be helpful for the government as well as people to know when the peak will come and to get alert and take all the safety measures to prevent the spread of Coronavirus disease (SARS-CoV-2 virus).

CHAPTER 2

LITERATURE SURVEY

Coronaviruses are among the main pathogens that predominantly affect the human respiratory system. The focus of the literature review was, therefore, to outline the predominant variables and methodology used in studies related to the spread of the virus. People with prevalent illnesses such as diabetes, hypertension, diabetes, stroke, heart, or kidney failure, as well as elderly people with impaired immune systems, are at an increased risk of infection. Closed areas with low ventilation and airflow may increase the risk of infection. The spread of the virus is believed to occur through respiratory droplets from coughing and sneezing, as with other respiratory viruses, including influenza virus and rhinoviruses. Aerosol transmission is also possible in case of protracted exposure to elevated aerosol concentrations in closed spaces. In the context of the COVID- 19 pandemic, people across the world are using various methods to explore prediction models with the goal of addressing the problems caused by the pandemic. The motivation for this SLR was to help researchers across the world study the various prediction models that have been created by numerous authors from multiple countries by providing information on a comprehensive range of models in one place. A systematic review is a compilation of various studies related to a single topic. It aims to provide a comprehensive and unbiased review of all the relevant studies in a given field. Our SLR was conducted to determine which prediction models are currently available, and the objective of the study was to identify the various methods used to develop different types of prediction models and to conduct an effectiveness or quality assessment of the models, which helps to evaluate their accuracy. It is hoped that this SLR will help healthcare workers and researchers wisely and confidently choose accurate prediction models to facilitate healthcare management by arranging medical

facilities and equipment. Researchers or scholars can enhance their research program by using this SLR to obtain up-to date information on the various techniques used in prediction models, as well as their efficiency and accuracy. All currently available prediction models for COVID-19 were systematically reviewed and critically appraised. There are presently a number of individual and prognostic models for COVID-19, all of which show moderate to excellent demarcation. To explore the different vaticination models and find the best-suited model in terms of furnishing high delicacy while minimizing the burden on the healthcare system and perfecting care for cases, both the opinion and prognostic evaluation of conditions need to be bettered. This study will impact decision-makers in colorful aspects.

CHAPTER 3

SYSTEM ANALYSIS

3.1 REQUIRED TOOLS

• 3.1.1 Hardware Requirements

- Processor i3
- **O** Hard Disk -5 GB
- $\bullet \quad Memory-1GB \ RAM$

• 3.1.2 Software Requirements

- **O** Windows Xp,
- Windows 7(ultimate, enterprise) Visual studio 2010

ABOUT WINDOWS XP - Windows XP is a version of the Microsoft Windows operating system for personal computers. The letters "XP" stand for eXPerience. Microsoft released Windows XP on October 25, 2001. Windows XP replaced Windows 2000 and Windows ME, which helped bring the NT and 9x versions of Windows together.

ABOUT WINDOWS 7 - Windows 7 is a major release of the Windows NT operating system developed by Microsoft. It was released to manufacturing on July 22, 2009, and became generally available on October 22, 2009.

Required tools & Libraries

- Jupiter Notebook
- Pandas 1.2.4
- Scikit-learn 0.24.1
- NLTK 3.6.2

To install above modules please run the following command: pip install pandas scikit-learn nltk

CHAPTER 4

Problem Formulation

My goal was to develop public models that would reasonably predict the growth of the coronavirus for the next day. This goal can be broken down and formalized into two parts:

- 1. To build a model that will predict the next day's number of coronavirus cases based on the last n day's number of coronavirus cases and deaths.
- 2. To build a model that will predict the next day's number of coronavirus deaths based on the last k day's number of coronavirus cases and deaths.

While I learned that n and k should be the same value after rigorous testing of my models, when I first defined the problem, I assumed that n and k may be different numbers as to not limit the possibilities of improving my individual models.

CHAPTER 5

Proposed System

The proposed system is grounded on machine literacy frame. Logistic Retrogression algorithm has been used for prognosticating the COVID-19 in someone's





Implementation

import numpy as np import pandas as pd import os for dirname, filenames in , os.walk('/kaggle/input'): for filename in filenames: print(os.path.join(dirname, filename)) df=pd.read_csv("downloads/covid_19_india.c sv")

```
state_testing = pd.read_csv('downloads/StatewiseTestingDetails.csv')
```

 $df_vc =$

pd.read_csv('downloads/covid_vaccine_statewise.csv/covid_vaccine_statewise.csv'

Out[28]:

[20].										
		Sno	Date	Time	State/UnionTerritory	ConfirmedIndianNational	ConfirmedForeignNational	Cured	Deaths	Confirmed
	0	1	2020-01-30	6:00 PM	Kerala	1	0	0	0	1
	1	2	2020-01-31	6:00 PM	Kerala	1	0	0	0	1
	2	3	2020-02-01	6:00 PM	Kerala	2	0	0	0	2
	3	4	2020-02-02	6:00 PM	Kerala	3	0	0	0	3
	4	5	2020-02-03	6:00 PM	Kerala	3	0	0	0	3
df.h	e	ас	1()							

Fig. head of dataset *df.info()*

In [29]: ▶ df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 18110 entries, 0 to 18109
Data columns (total 9 columns):
# Column
                            Non-Null Count Dtype
--- -----
                             ----- -----
0 Sno
                           18110 non-null int64
1 Date
                           18110 non-null object
2Time18110 non-nullobject3State/UnionTerritory18110 non-nullobject
 4 ConfirmedIndianNational 18110 non-null object
 5 ConfirmedForeignNational 18110 non-null object
 6 Cured
                           18110 non-null int64
                            18110 non-null int64
    Deaths
 7
8
    Confirmed
                            18110 non-null int64
dtypes: int64(4), object(5)
memory usage: 1.2+ MB
```

df.isnull().sum()

Out[30]:	Sno					
	Date	0				
	Time					
	State/UnionTerritory					
	ConfirmedIndianNational					
	ConfirmedForeignNational					
	Cured	0				
	Deaths	0				
	Confirmed	0				
	dtype: int64					

df.drop(["Sno", "Time", "ConfirmedIndianNational", "ConfirmedForeignNational"], inplace=True, axis=1)

df['Active_cases']=df['Confirmed']-(df['Cured']+df['Deaths'])

df.head()

Out[32]:

	Date	State/UnionTerritory	Cured	Deaths	Confirmed	Active_cases
0	2020-01-30	Kerala	0	0	1	1
1	2020-01-31	Kerala	0	0	1	1
2	2020-02-01	Kerala	0	0	2	2
3	2020-02-02	Kerala	0	0	3	3
4	2020-02-03	Kerala	0	0	3	3

df['Confirmed'] = pd.to_numeric(df['Confirmed'],
errors='coerce') df['Confirmed']=df['Confirmed'].fillna(0)
df['Confirmed']=df['Confirmed'].astype('int')

df['Deaths'] = pd.to_numeric(df['Deaths'],
errors='coerce') df['Deaths']=df['Deaths'].fillna(0)
df['Deaths']=df['Deaths'].astype('int')

```
df['Cured'] = pd.to_numeric(df['Cured'],
errors='coerce') df['Cured']=df['Cured'].fillna(0)
df['Cured']=df['Cured'].astype('int')
df['Deaths'] = pd.to_numeric(df['Deaths'],
errors='coerce') df['Deaths']=df['Deaths'].fillna(0)
df['Deaths']=df['Deaths'].astype('int')
```

```
df['Cured'] = pd.to_numeric(df['Cured'],
errors='coerce') df['Cured']=df['Cured'].fillna(0)
df['Cured']=df['Cured'].astype('int')
```

#statewise analysis

statewise = pd.pivot_table(df, values=['Confirmed', 'Deaths', 'Cured'], index='State/UnionTerritory', aggfunc='max') statewise['Recovery Rate'] = statewise['Cured']*100 / statewise['Confirmed'] statewise['Mortality Rate'] = statewise['Deaths']*100 /statewise['Confirmed'] statewise = statewise.sort_values(by='Confirmed', ascending= False) statewise.style.background_gradient(cmap='YIOrRd')

Confirmed Cured Deaths Recovery Rate

Maharashtra 6363442 6159676 134201 96.797865 Maharashtra*** 6229596 6000911 130753 96.329056 3586693 3396184 Kerala 18004 94.688450 2921049 2861499 97.961349 Karnataka 36848 2821491 97.790581 Karanataka 2885238 36197 Tamil Nadu 2579130 2524400 34367 97.877967 Andhra Pradesh 1985182 1952736 13564 98.365591 Uttar Pradesh 98.635309 1708812 1685492 22775 West Bengal 1534999 1506532 18252 Delhi 1436852 1411280 25068 98.220276

State/UnionTerritory

Chhattisgarh	1003356	988189	13544	98.488373
Odisha	988997	972710	6565	98.353180
Rajasthan	953851	944700	8954	99.040626
Gujarat	825085	814802	10077	98.753704
Madhya Pradesh	791980	781330	10514	98.655269
Madhya Pradesh***	791656	780735	10506	98.620487
Haryana	770114	759790	9652	98.659419
Bihar	725279	715352	9646	98.631285
Bihar****	715730	701234	9452	97.974655
Telangana	650353	638410	3831	98.163613
Punjab	599573	582791	16322	97.201008
Assam	576149	559684	5420	97.142232
Telengana	443360	362160	2312	81.685312
Jharkhand	347440	342102	5130	98.463620

State/UnionTerritory

State/UnionTerritory

Uttarakhand	342462	334650	7368	97.718871
Jammu and Kashmir	322771	317081	4392	98.237140
Himachal Pradesh	208616	202761	3537	97.193408
Himanchal Pradesh	204516	200040	3507	97.811418
Goa	172085	167978	3164	97.613389
Puducherry	121766	119115	1800	97.822873
Manipur	105424	96776	1664	91.796934
Tripura	80660	77811	773	96.467890
Meghalaya	69769	64157	1185	91.956313
Chandigarh	61992	61150	811	98.641760
Arunachal Pradesh	50605	47821	248	94.498567
Mizoram	46320	33722	171	72.802245
Nagaland	28811	26852	585	93.200514

Cor

Confirmed Cured

Sikkim 28018 25095 356 89.567421 I adakh 20411 20130 207 98.623291 Dadra and Nagar Haveli and Daman and 10646 99.924911 10654 4 Diu 98.882143 Dadra and Nagar Haveli 10377 10261 4 Lakshadween 10263 10165 51 99 045114 Cases being reassigned to states 9265 0 0.000000 0 Andaman and Nicobar Islands 7548 7412 129 98.198198 Unassigned 77 0 0 0.000000 Daman & Diu 2 0 0 0.000000

State/UnionTerritory

pip install plotly

Collecting plotly
 Downloading plotly-5.5.0-py2.py3-none-any.whl (26.5 MB)
Collecting tenacity>=6.2.0
 Downloading tenacity-8.0.1-py3-none-any.whl (24 kB)
Requirement already satisfied: six in c:\programdata\anaconda3\lib\site-p
ackages (from plotly) (1.15.0)
Installing collected packages: tenacity, plotly
Successfully installed plotly-5.5.0 tenacity-8.0.1
Note: you may need to restart the kernel to use updated packages.

import seaborn as sns

import matplotlib.pyplot as plt

%matplotlib inline

import plotly.express as px

import plotly.offline as pyo

import plotly.graph_objs as

go

from plotly.subplots import make_subplots

from IPython.display import display,

Markdown

plt.figure(figsize = (18, 10))
figure = px.line(df, x='Date', y='Confirmed',
 color='State/UnionTerritory')
figure.update_xaxes(rangeslider_visible=True)





"Kerala" : [10.8505,76.2711], "Maharashtra" : [19.7515,75.7139], "Karnataka": [15.3173,75.7139], "Telangana": [18.1124,79.0193], "Uttar Pradesh": [26.8467,80.9462], "Rajasthan": [27.0238,74.2179], "Gujarat":[22.2587,71.1924], "Delhi" : [28.7041,77.1025], "Punjab":[31.1471,75.3412], "Tamil Nadu": [11.1271,78.6569], "Haryana": [29.0588,76.0856], "Madhya Pradesh":[22.9734,78.6569], "Jammu and Kashmir":[33.7782,76.5762], "Ladakh": [34.1526,77.5770], "Andhra Pradesh":[15.9129,79.7400], "West Bengal": [22.9868,87.8550], "Bihar": [25.0961,85.3131], "Chhattisgarh":[21.2787,81.866 1], "Chandigarh":[30.7333,76.7794], "Uttarakhand":[30.0668,79.0193], "Himachal Pradesh":[31.1048,77.1734], "Goa": [15.2993,74.1240], "Odisha":[20.9517,85.0985], "Andaman and Nicobar Islands": [11.7401,92.6586], "Puducherry":[11.9416,79.8083],

"Manipur":[24.6637,93.9063], "Mizoram":[23.1645,92.9376], "Assam":[26.2006,92.9376], "Meghalaya":[25.4670,91.366 2],

```
"Tripura":[23.9408,91.9882],

"Arunachal

Pradesh":[28.2180,94.7278],

"Jharkhand" : [23.6102,85.2799],

"Nagaland": [26.1584,94.5624],

"Sikkim": [27.5330,88.5122],

"Dadra and Nagar

Haveli":[20.1809,73.0169],

"Lakshadweep":[10.5667,72.6417],

"Daman and Diu":[20.4283,72.8397]

}

statewise["Lat"] = ""
```

```
....
```

```
for index in
```

statewise["Long"] =

```
statewise.index : if
index in locations:
    statewise.loc[statewise.index == index,"Lat"] =
    locations[index][0] statewise.loc[statewise.index ==
    index,"Long"] = locations[index][1]
    else:
    statewise.drop([index],inplace=True)

import folium
from IPuthon display import HTMI
```

from IPython.display import HTML,
display india =
folium.Map(location=[23,80],
zoom_start=4,max_zoom=6,min_zoom=4,height=500,width="
80%") for i in
range(0,len(statewise[statewise['Confirmed']>0].index)):
folium.Circle(
 location=[statewise.iloc[i]['Lat'], statewise.iloc[i]['Long']],

tooltip = "<h5 style='text-align:center;font-weight:

bold'>"+statewise.iloc[i].name+"</h5>"+ "<hr style='margin:10px;'>"+

"
 style='color: #444;list-style-type:circle;align-item:left;paddingleft:20px;padding-right:20px'>"+

```
"Confirmed:
"+str(statewise.iloc[i]['Confirmed'])+""+"Cured:
"+str(statewise.iloc[i]['Cured'])+""+
```

"Deaths: "+str(statewise.iloc[i]['Deaths'])+""+
"Recovery Rate: "+str(np.round(statewise.iloc[i]['Recovery
Rate'],2))+""+ "Mortality Rate'],2))+"",
Rate: "+str(np.round(statewise.iloc[i]['Mortality Rate'],2))+"",

```
radius=(int(np.log2(statewise.iloc[i]['Confirmed']+1)))*15
000, color='#ff6600',
fill_color='#ff8533',
fill=True).add_to(indi
a)
```

india



df1=df.groupby('Date')[['Cured', 'Deaths', 'Confirmed']].s um() df1.tail()

Out[19]:

	Cured	Deaths	Confirmed
Date			
2021-08-07	31055861	427371	31895385
2021-08-08	31099771	427862	31934455
2021-08-09	31139457	428309	31969954
2021-08-10	31180968	428682	31998158
2021-08-11	31220981	429179	32036511

fig=go.Figure()

fig.add_trace(go.Scatter(x=df1.index,

y=df1["Confirmed"], mode='lines+markers',

name='Confirmed Cases'))

fig.add_trace(go.Scatter(x=df1.index,

y=df1["Cured"],

mode='lines+marker

s', name='Cured

Cases'))

fig.add_trace(go.Scatter(x=df1.index,

y=df1["Deaths"], mode='lines+markers',

name='Death Cases'))

fig.update_layout(title="Growth of different types of cases in India",

xaxis_title="Date",yaxis_title="Number of

```
Cases", legend=dict(x=0, y=1, traceorder="normal"))
```

fig.show()





```
fig = make_subplots(rows=2, cols=1,
```

```
subplot_titles=("Recovery Rate", "Mortatlity
```

```
Rate"))
```

fig.add_trace(

```
go.Scatter(x=df1.index,
```

```
y=(df1["Cured"]/df1["Confirmed"])*100,
```

```
name="Recovery Rate"),
```

```
row=1, col=1
```

```
)
```

```
fig.add_trace(
```

```
go.Scatter(x=df1.index,
```

```
y=(df1["Deaths"]/df1["Confirmed"])*100,
```

```
name="Mortality Rate"),
```

```
row=2, col=1
```

)

fig.update_layout(height=1000,legend=dict(x=-0.1,y=1.2,traceorder="normal")) fig.update_xaxes(title_text="Date", row=1, col=1) fig.update_yaxes(title_text="Recovery Rate", row=1, col=1) fig.update_xaxes(title_text="Date", row=1, col=2) fig.update_yaxes(title_text="Mortality Rate", row=1, col=2) fig.show()







Mortality rate = (Number of Death Cases / Number of Confirmed Cases) x 100

Recovery Rate= (Number of Recoverd Cases / Number of Confirmed Cases) x 100

Recovery Rate was initially very high when the number of positive (Confirmed) cases were low and showed a drastic drop with increasing number of cases. Increasing Mortality rate and dropped Recovery Rate is worrying sign for India.

Increasing Mortality Rate and very slowly increasing Recovery Rate is conclusive evidence for increase in number of Closed Cases

Recovery Rate is showing an upward trend which is a really good sign. Mortality Rate is showing a slight dips but with occasional upward trends.

growth_diff = []

for i in range(1,len(df1['Confirmed'])):

growth_diff.append(df1['Confirmed'][i] / df1['Confirmed'][i-

1])

growth_factor =

sum(growth_diff)/len(growth_diff)

print('Average growth factor',growth_factor)

Average growth factor 1.0370264842594674

df1.reset_index(inplace=Tr ue) df1.head()

Date	Cured	Deaths	Confirmed	
0	2020-01-30	0	0	1
1	2020-01-31	0	0	1
2	2020-02-01	0	0	2
3	2020-02-02	0	0	3

Date	Cured	Deaths	Confirmed	
4	2020-02-03	0	0	3

prediction_dates = []

```
from datetime import datetime,timedelta
df1['Date']=pd.to_datetime(df1['Date'], format='%Y-%m-
%d') start_date = df1['Date'][len(df1['Date']) - 1]
for i in range(300):
    date = start_date
    +timedelta(days=1)
    prediction_dates.append(date)
    start_date = date
previous_day_cases =
    df1['Confirmed'][len(df1['Date']) - 1] predicted_cases
    _____
```

```
=[]
```

```
for i in range(300):
    predicted_value = previous_day_cases *
    growth_factor
    predicted_cases.append(predicted_value)
    previous_day_cases = predicted_value
```

```
plt.figure(figsize= (15, 10))
plt.xticks(rotation = 90 ,fontsize = 11)
plt.yticks(fontsize = 10)
plt.xlabel("Dates",fontsize = 20)
plt.ylabel('Total cases',fontsize =
20)
plt.title("Predicted Values for the next 10 months" , fontsize = 20)
```

ax1 = plt.plot_date(y= predicted_cases,x= prediction_dates,linestyle ='-',color = 'c')



popt, pcov = curve_fit(sigmoid, (predIndiaDF.index+1).astype(float), predIndiaDF['Confirmed'], p0=(0,0,0),maxfev=5000)

x0 = int(popt[2]/10) print('\033[1mx0 (point/day of inflexion):\033[0m',int(popt[2]/10)) print('\033[1mL (Maximum no.of cases):\033[0m',int(popt[0]/10000)) print('\033[1mk (Growth Rate):\033[0m',round(float(popt[1]),2)) print('\033[1mPCOV: \033[0m\n',pcov)

```
fig.add_trace(go.Scatter(x =
```

```
predIndiaDF['Date'], y =
predIndiaDF['Confirmed'], mode
= 'lines',
name = 'Observed'))
```

fig.add_trace(go.Scatter(x =

dateRange[:600],

y = sigmoid([x for x in
range(extended_period)], *popt), mode = 'lines',
name = 'Predicted'))
fig.add_trace(go.Scatter(x = [dateRange[x0], dateRange[x0]],
y = [0, sigmoid([x for x in

range(extended_period)],*popt)[x0]], name = 'X0 -

```
Inflexion point', mode = 'lines'))
```

fig.update_layout(template='plotly_dark',title='Projected Confirmed

Cases') fig.show()

x0 (point/day of inflexion): 59 L (Maximum no of cases): 8151						
k (Growth Rate): 0.01						
PCOV:						
[[9.65384796e+13 -	2 72822826e+03	2_43080876e+08				
[-2.72822826e+03 8	3.91315475e-08	-7.05572390e-03]				
[2.43080876e+08 -7	_05572390e-03	6_16079131e+02]				



import tensorflow as tf from numpy import array from sklearn.model_selection import train_test_split from sklearn.preprocessing import MinMaxScaler from sklearn.preprocessing import LabelEncoder from sklearn.metrics import mean_squared_error from tensorflow.keras.models import Sequential from tensorflow.keras.layers import Dense from tensorflow.keras.callbacks import ModelCheckpoint, ReduceLROnPlateau, EarlyStopping from tensorflow.keras.layers import LSTM from tensorflow.keras.utils import

plot_model from tensorflow.keras import

Input

from tensorflow.keras.layers import BatchNormalizationfrom fbprophet import

Prophet fb_data = df1.copy()

fb_confirm = fb_data[['Date', 'Confirmed']]

fb_confirm = fb_confirm.rename(columns={'Date':

'ds', 'Confirmed': 'y'})

fb_confirm.head()

ds	у	
0	2020-01-30	1
1	2020-01-31	1
2	2020-02-01	2
3	2020-02-02	3
4	2020-02-03	3

future_dates = my_model.make_future_dataframe(periods=15,

freq='MS') future_dates.tail()



ds	
571	2022-08-01
572	2022-09-01
573	2022-10-01
574	2022-11-01

forecast = my_model.predict(future_dates)

forecast[['ds', 'yhat', 'yhat_lower',

'yhat_upper']].tail()

ds	yhat	yhat_lower	yhat_upper	
570	2022-07-01	8.982125e+07	6.995001e+07	1.103366e+08
571	2022-08-01	9.492832e+07	7.359266e+07	1.186771e+08
572	2022-09-01	1.000809e+08	7.473027e+07	1.260847e+08
573	2022-10-01	1.050324e+08	7.775718e+07	1.342115e+08
574	2022-11-01	1.101297e+08	7.905184e+07	1.422771e+08

from fbprophet.plot import plot_plotly,

add_changepoints_to_plot fig = plot_plotly(my_model,

forecast)

pyo.iplot(fig)



fig = my_model.plot(forecast,xlabel='Date',ylabel='Confirmed Count')

forecast_components=my_model.plot_components(forecast)



Chapter -6 Diagram for proposed System



System Arcitecture Diagram



Algorithms Used



Chapter 7

CONCLUSION

Coronavirus disease (COVID-19) is an infectious disease caused by the SARS-CoV- 2 virus. It is spreading continuously among the population of a particular locality. This spread of virus has almost stopped the whole world and is also responsible for decrease in economy of our country and many other countries of the world. Lockdown were imposed to contain the spread of this virus. There were a lot of cases increasing on daily basis and fatality rate also increased. The government had taken various measures to protect our population from this pandemic. This has not only affected the whole country economically, but has also caused financial as well as health problems on the individual level. Many lost their jobs and a lot of people have faced losses in business. The overall point is that due to this Corona virus pandemic, people have lost many things. Its affect reduces for sometimes but it again comes back in the form of wave and lefts people losing lives, jobs, business etc. Everyone wants to know the future that when will this pandemic end, when will be the peak phase to get prepared for the next waves to come.

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