COVID-19 DATA ANALYSIS https://github.com/ashukumar7/Covid19

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Documentation on

COVID-19

DATA ANALYSIS

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(Diginique TechLabs, in association with Cognizance'20 IIT ROORKEE)

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Keyword

Virus, Pandemic, WHO, COVID-19, Machine Learning, Supervised, Unsupervised, Reinforcement.

Abstract

The aim of the project is to provide data analysis of covid-19 (a pandemic started in December 2019). Through plotting of data, various cases have been studied like most affected countries due to this pandemic. Study of data from various countries is combined to show the growth of cases and recovery graph. In this project, the predictions on various cases has been done and finally, the accuracy of the algorithm has been determined. Comparison graphs has also been plotted to analyse how much INDIA is getting affected/recover day by day.

Introduction to Covid-19

On 31st December 2019, in the city of Wuhan (CHINA), a cluster of cases of pneumonia of unknown cause was reported to World Health organisation. In January 2020, a previously unknown new virus was identified, subsequently named 2019 novel corona virus. WHO has declared the COVID-19 as a pandemic. A pandemic is defined as disease spread over a wide range of geographical area and that has affected high proportion of the population.

Problem Statement

The pandemic has already taken grip over peoples' life. Since the start of the pandemic, some countries are facing problem of ever-increasing cases. Through the data analysis of cases one can analyse how countries all over the world are doing in terms of controlling the pandemic. Analysing data leads to adapt the prevention model of the countries that are doing great in terms of lowering the graph. Predictions are made with the dataset available to the individual/country/organisations, thus helping them to decide how far they are able to control the pandemic or up to how much extent they should guide preventive measures.

Through this project, a step towards helping people to understand the spread and predict the cases in their country is done. This project also gives an insight of how a country is doing in terms of limiting the spread.

Technology and Concept

Machine Learning

Machine learning is a field of study or process of teaching a computer to learn the fed data without being explicitly programmed. It makes computer make decisions similar to humans.

Now a days, it is actively being used in various field. E.g. Medical, Industries, Astronomy etc. The major types of Machine learning are Supervised Learning, Unsupervised Learning and Reinforcement Learning.

Supervised Learning

The machine learning task of learning a function that can map an input data to output data and performs analysis based on that input-output pair.

Unsupervised Learning

A type of machine learning that draw an inference from dataset consisting of input data without labelled responses. One of the common unsupervised learning methods called cluster analysis, is used find the hidden pattern or grouping of data.

Reinforcement Learning

A type of machine learning that is bound to learn from experiences. There is no training dataset provided *(such methods work in the absence of datasets). An agent in Reinforcement learning that rewards or penalise for actions done by the algorithm. The task is to find the best possible path to reach the goal.

Some important terms

Data frame

Pandas Data frame is 2D, mutable and heterogeneous tabular data structure with labelled axes. Data frame can be made of more than one series (series can only contain single list with index).

Hypothesis

In Machine learning, Hypothesis is a model that is used to approximate the target function and performs mapping of input with output.

Regression

Regression in Machine Learning is about predicting the continuous value-based learning gained by dataset. The correctness of the output can depend on the size of dataset, features, hypothesis used etc.

Classification

The problem of identifying that in which sub-population a new example/observation belongs to, on the basis of learning obtained through training set containing observations along with the category they belong to.

Dataset Pre-processing

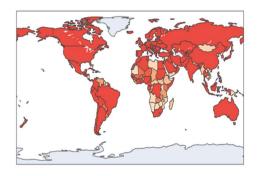
This section include the parsing of date in a proper readable format[1], renaming some columns into short and descriptive names [2], adding new column 'active cases' with the help of other cases available in the dataset [3], creating a data frame that includes the latest cases up to date [4], grouping the data in terms of country and resetting the index [5][6].

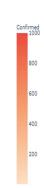
```
[1]
df = pd.read_csv(r'...\covid_19_clean_complete.csv',parse_dates=['Date'])
[2] df.rename(columns={"Country/Region":"country",'Province/State':'state'},inplace=True)
[3] df['active'] = df['Confirmed'] - df['Deaths'] - df['Recovered']
[4] top = df[df['Date'] == df['Date'].max()]
[5]world = top.groupby('country')['Confirmed','active','Deaths'].sum()
[6]
world = top.groupby('country')['Confirmed','active','Deaths'].sum().reset_index()
```

Plotting on world map

Confirmed Cases

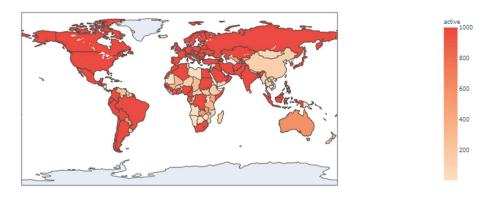
Country with Confirmed Cases





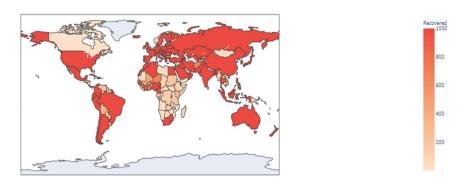
Active Cases

Country with Active Cases



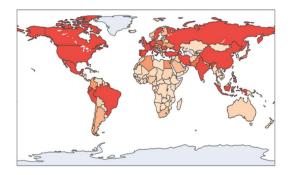
Recovered Cases

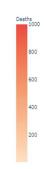
Country with Recovered Cases



Death Cases

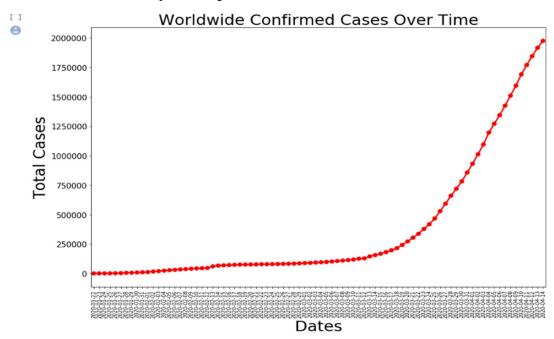
Country with death Cases



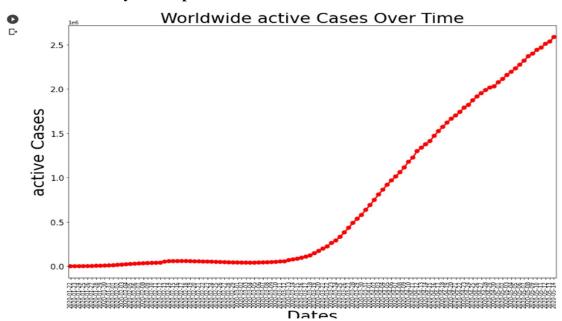


Cases over time

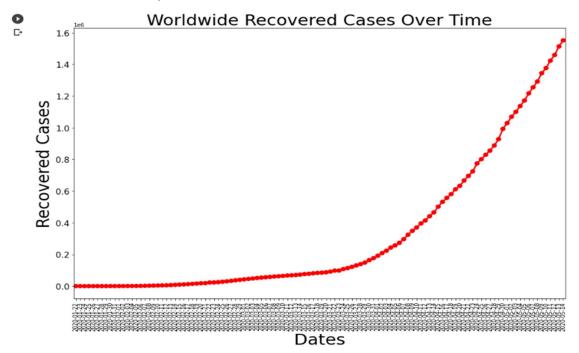
Confirmed Cases day-wise spread



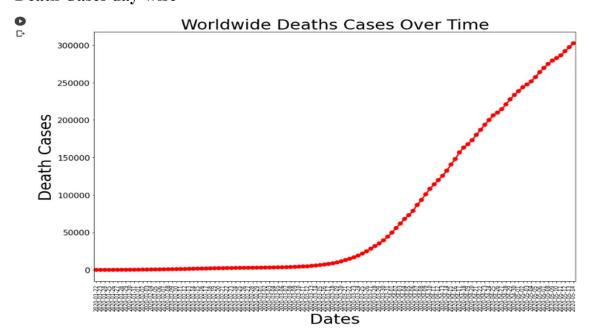
Active Cases day-wise spread



Recovered Cases day-wise



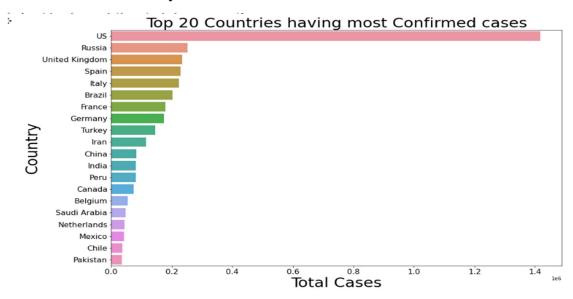
Death Cases day-wise



TOP 20 Countries

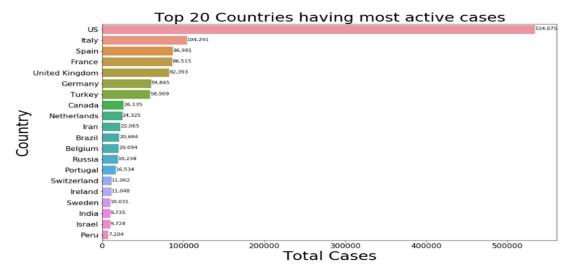
Confirmed

Most affected Country: US



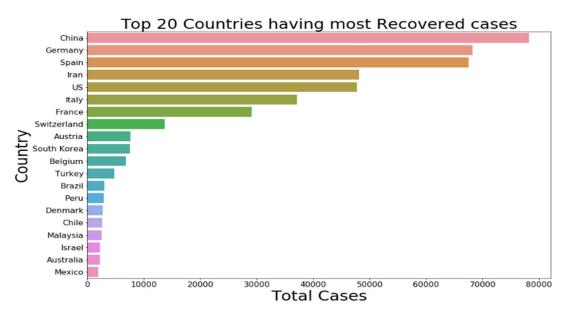
Active

Most affected country: US



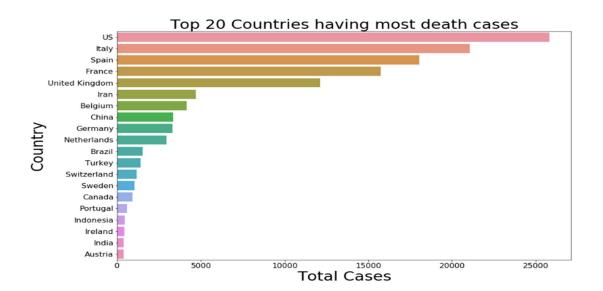
Recovered

Most recoveries: CHINA



Death

Most deaths in: US



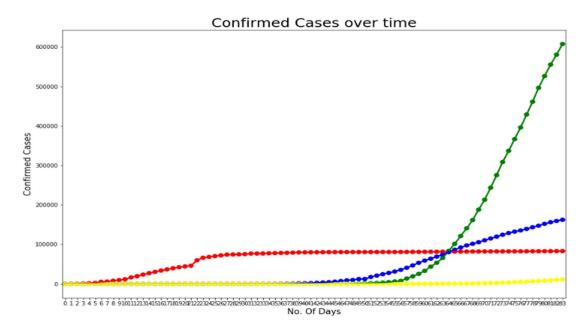
Analysis on some countries

Countries taken for analysis are: CHINA, US, ITALY AND INDIA.

Highest confirmed cases: US.

Lowest Confirmed cases: INDIA.

Confirmed

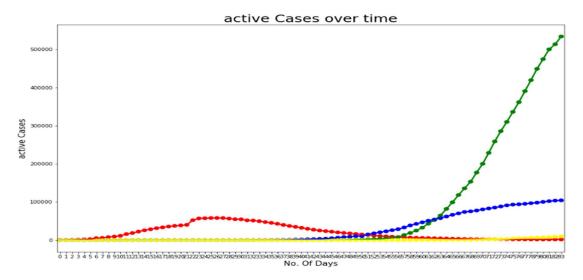


Active

Countries taken for analysis are: CHINA, US, ITALY AND INDIA.

Highest Active cases: US.

Lowest Active cases: CHINA.

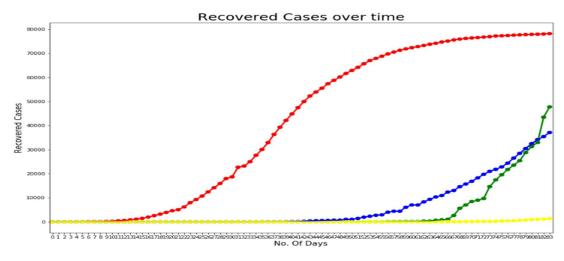


Recovered

Countries taken for analysis are: CHINA, US, ITALY AND INDIA.

Highest Recovered cases: CHINA.

Lowest Recovered cases: INDIA

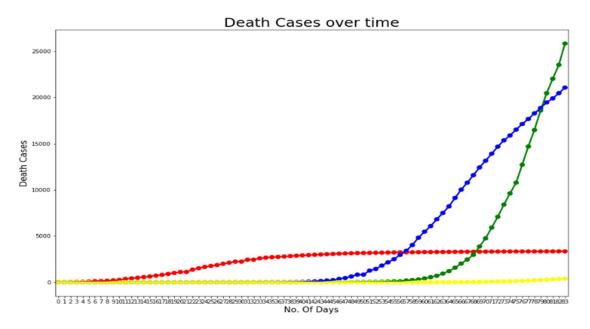


Death

Countries taken for analysis are: CHINA, US, ITALY AND INDIA.

Highest Death cases: US.

Lowest Death cases: INDIA



Prediction and Forecasting

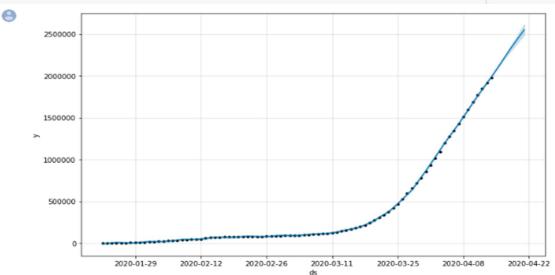
Confirmed Cases (WORLD)

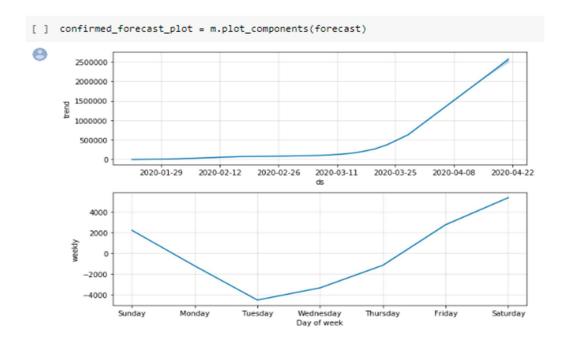
Predicted value : yhat || Actual Value :y

```
Ľ•
       2020-01-22 -5.786925e+03
                                     555.0
       2020-01-23 -6.795861e+02
                                     654.0
       2020-01-24 3.362999e+03
                                    941.0
       2020-01-25 4.558362e+03
                                   1434.0
       2020-01-26 5.294289e+03
                                    2118.0
    109 2020-05-10 4.095126e+06 4101693.0
    110 2020-05-11 4.173778e+06 4177496.0
    111 2020-05-12 4.252689e+06 4261741.0
    112 2020-05-13 4.334106e+06 4347012.0
    113 2020-05-14 4.419698e+06 4442157.0
```

[114 rows x 3 columns]

```
[ ] confirmed_forecast_plot = m.plot(forecast) # ploting predicted value of confirmed cases
  # black dot - actual values
  # blue line = predicted values
```

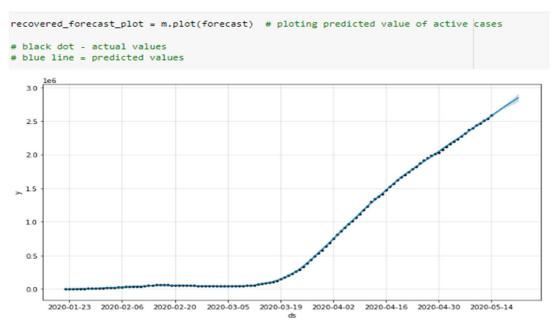


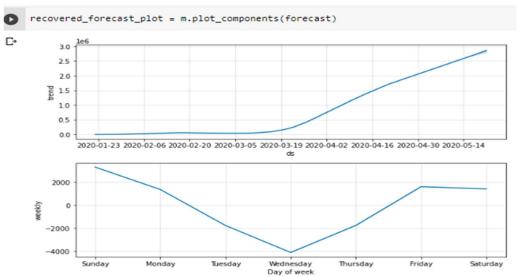


Active Cases (WORLD)

Predicted value: yhat || Actual Value: y

```
# ds - date
# yhat - prediction made
# yhat_lower - lower limit of prediction
# yhoat_upper - upper limit of prediction
metric_df.dropna(inplace=True)
print(metric_df)
             ds
                          yhat
    2020-01-22 -5.659046e+03
                                     510.0
0
    2020-01-23 -2.181894e+03
                                     606.0
    2020-01-24 2.298928e+03
2020-01-25 3.230439e+03
                                     879.0
                                    1353.0
4 2020-01-26 6.247449e+03
                                    2010.0
109 2020-05-10 2.435660e+06 2442197.0
110 2020-05-11 2.471713e+06
                                 2468048.0
111 2020-05-12 2.506587e+06 2510524.0
112 2020-05-13 2.542246e+06 2536535.0
113 2020-05-14 2.582596e+06 2588041.0
```

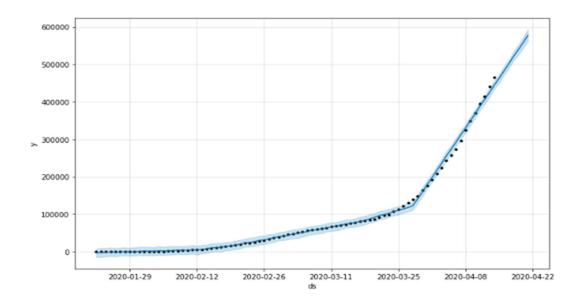


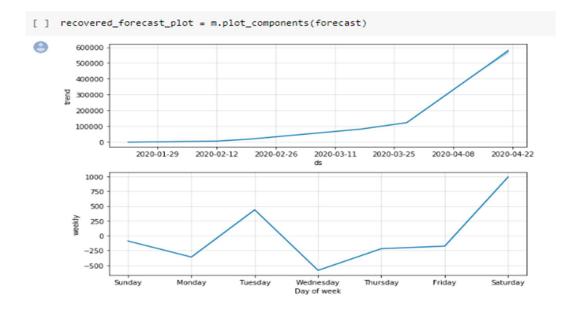


Recovered Cases (WORLD)

Predicted value : yhat || Actual Value :y

```
yhat
C.
               ds
        2020-01-22 -1.340404e+03
                                      28.0
        2020-01-23 3.015232e+02
                                      30.0
        2020-01-24 -2.619864e+02
                                      36.0
        2020-01-25 1.233584e+02
                                      39.0
        2020-01-26 -1.530887e+03
                                      52.0
    109 2020-05-10 1.375868e+06 1376787.0
    110 2020-05-11 1.413454e+06 1423118.0
    111 2020-05-12 1.451859e+06
                                 1459275.0
    112 2020-05-13 1.492197e+06 1513280.0
    113 2020-05-14 1.531985e+06 1551698.0
    [114 rows x 3 columns]
```



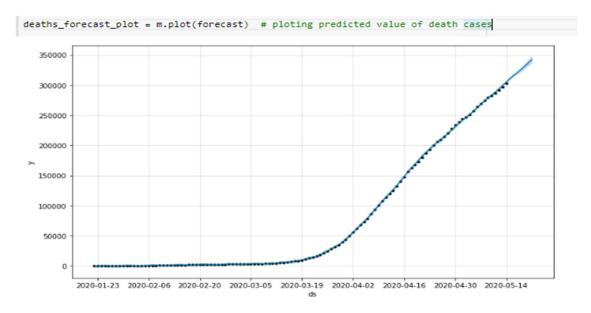


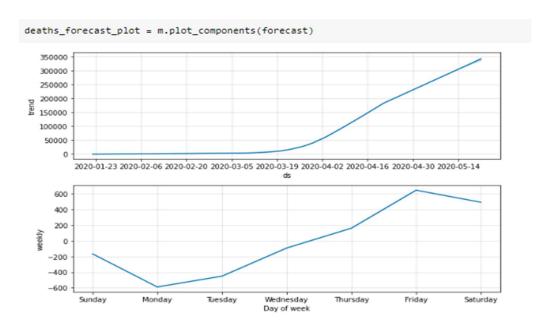
Death Cases (WORLD)

Predicted value: yhat || Actual Value: y

```
2020-01-22
                 -417.305727
   2020-01-23
                 -94.405048
                                 654.0
    2020-01-24
                 460.126071
                                 941.0
                375.335788
   2020-01-25
                                1434.0
   2020-01-26
               -214.777228
                                2118.0
109 2020-05-10 283459.176501 4101693.0
110 2020-05-11 288318.206060
                             4177496.0
111 2020-05-12 293738.505597 4261741.0
112 2020-05-13 299380.089793 4347012.0
113 2020-05-14 304915.809708 4442157.0
```

[114 rows x 3 columns]



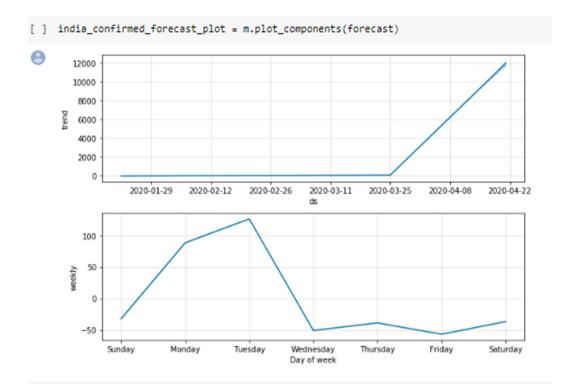


Prediction and Forecasting for India

Confirmed Cases (INDIA)

Predicted value : yhat || Actual Value :y

```
metric_df=forecast.set_index('ds')[['yhat']].join(india_confirmed.set_index('ds').y).reset_index()
     # ds - date
     # yhat - prediction made
     # yhat_lower - lower limit of prediction
     # yhoat_upper - upper limit of prediction
     metric_df.dropna(inplace=True)
     print(metric_df)
                               yhat
C.
                 ds
    0 2020-01-22 -14.291047
1 2020-01-23 76.161076
2 2020-01-24 -376.987538
3 2020-01-25 -371.672414
                                          0.0
     4 2020-01-26 -254.250684
                                         0.0
     109 2020-05-10 64565.901662 67161.0
     110 2020-05-11 67217.116959 70768.0
     111 2020-05-12 69781.644051 74292.0
     112 2020-05-13 72313.937528 78055.0
     113 2020-05-14 74890.626540 81997.0
     [114 rows x 3 columns]
0
                           yhat yhat_lower yhat_upper
     86 2020-04-17 10106.612051 8864.403004 11214.204411
     87 2020-04-18 10565.888551 9482.179100 11686.294913
     88 2020-04-19 11009.571306 9957.137959 12122.438495
     89 2020-04-20 11569.586211 10462.555930 12808.011082
     90 2020-04-21 12046.676495 10947.019012 13223.884052
    india_confirmed_forecast_plot = m.plot(forecast) # ploting predicted value of confirmed cases
0
       12000
       10000
        8000
       6000
        4000
        2000
                  2020-01-29
                             2020-02-12
                                         2020-02-26
                                                     2020-03-11
                                                                 2020-03-25
                                                                           2020-04-08
                                                                                        2020-04-22
```

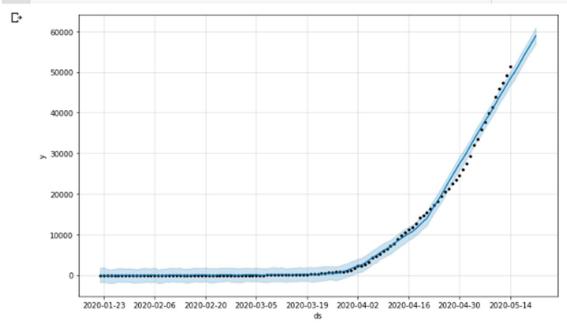


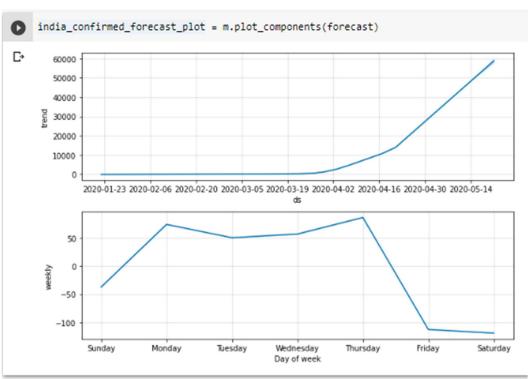
Active Cases (INDIA)

Predicted value: yhat || Actual Value: y

```
metric_df=forecast.set_index('ds')[['yhat']].join(india_active.set_index('ds').y).reset_index()
    # ds - date
    # yhat - prediction made
    # yhat_lower - lower limit of prediction
    # yhoat_upper - upper limit of prediction
    metric_df.dropna(inplace=True)
    print(metric_df)
                           yhat
               ds
C.
    0 2020-01-22 -38.880433
                                      0.0
        2020-01-23
                      -5.328541
                                      0.0
       2020-01-24 -200.042815
                                     0.0
       2020-01-25 -202.354706
2020-01-26 -116.416678
                                     0.0
                                     0.0
    109 2020-05-10 42327.947542 43980.0
    110 2020-05-11 43935.575270 45925.0
    111 2020-05-12 45408.209361 47457.0
    112 2020-05-13 46910.992468 49104.0
    113 2020-05-14 48436.613918 51379.0
    [114 rows x 3 columns]
```

india_active_forecast_plot = m.plot(forecast) # ploting predicted value of active cases

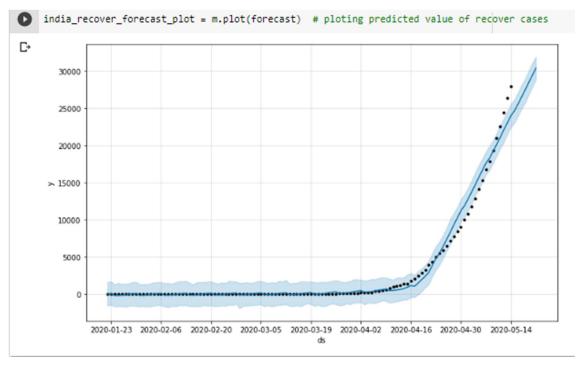


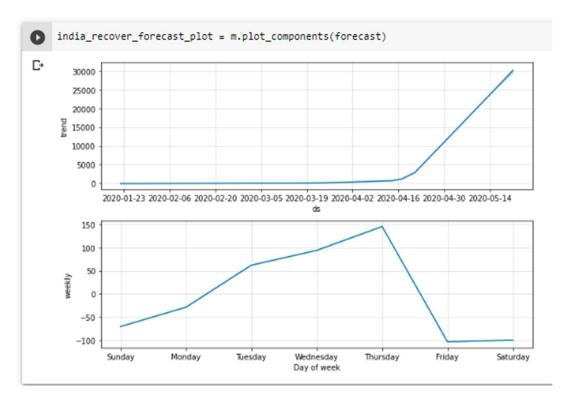


Recovered Cases (INDIA)

Predicted value: yhat || Actual Value: y

```
metric_df=forecast.set_index('ds')[['yhat']].join(india_recovered.set_index('ds').y).reset_index()
    # ds - date
    # yhat - prediction made
    # yhat_lower - lower limit of prediction
    # yhoat_upper - upper limit of prediction
    metric_df.dropna(inplace=True)
    print(metric_df)
C.
    0 2020-01-22 16.394136
1 2020-01-23 70.128292
                                         0.0
                                         0.0
    2 2020-01-24 -176.388463
3 2020-01-25 -170.139918
4 2020-01-26 -138.764944
                                         0.0
                                         0.0
    109 2020-05-10 20108.956349 20969.0
    110 2020-05-11 21062.125678 22549.0
    111 2020-05-12 22064.224745 24420.0
    112 2020-05-13 23007.667178 26400.0
    113 2020-05-14 23970.515173 27969.0
    [114 rows x 3 columns]
```

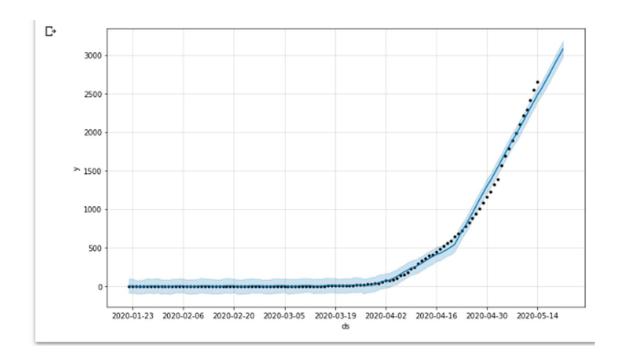


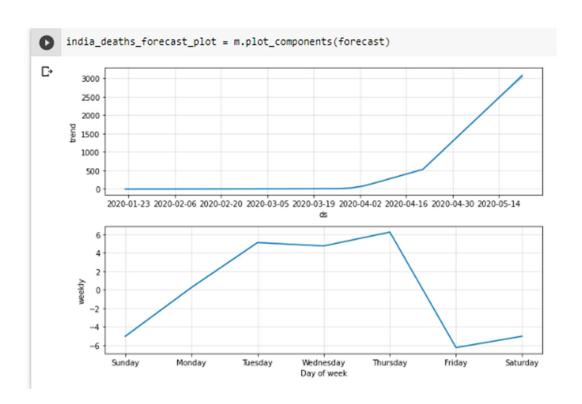


Death Cases (INDIA)

Predicted value: yhat || Actual Value: y

```
metric_df=forecast.set_index('ds')[['yhat']].join(india_deaths.set_index('ds').y).reset_index()
   # ds - date
   # yhat - prediction made
   # yhat_lower - lower limit of prediction
   # yhoat_upper - upper limit of prediction
   metric_df.dropna(inplace=True)
   print(metric_df)
C.
               ds
                         yhat
                   1.513042
   0 2020-01-22
                                  0.0
                   3.132882
-9.289070
       2020-01-23
                                  0.0
       2020-01-24
                                  0.0
   3 2020-01-25 -7.938807
   4 2020-01-26
                   -7.838539
                                  0.0
   109 2020-05-10 2135.318807 2212.0
   110 2020-05-11 2225.249303 2294.0
   111 2020-05-12 2314.767694 2415.0
   112 2020-05-13 2399.031581 2551.0
   113 2020-05-14 2485.169217 2649.0
   [114 rows x 3 columns]
```





Accuracy Check

R2 Score: R-squared values range from 0 to 1 and are commonly stated as percentages from 0% to 100%. An R-squared of 100% means that all movements of a security (or another dependent variable) are completely explained by movements in the index (or the independent variable(s) you are interested in).

Mean Square Error: In statistics, the mean squared error or mean squared deviation of an estimator measures the average of the squares of the errors—that is, the average squared difference between the estimated values and the actual value. MSE is a risk function, corresponding to the expected value of the squared error loss

Mean Absolute Error: In statistics, mean absolute error is a measure of errors between paired observations expressing the same phenomenon. Examples of Y versus X include comparisons of predicted versus observed, subsequent time versus initial time, and one technique of measurement versus an alternative technique of measurement.

On prediction of cases around world

	R2 score	Mean square	Mean absolute
Confirmed	0.99997727504771	43988753.2606206 9	4776.93977517155 5
Active	0.99995581764507	32950433.4746968	4229.98083222683
	2	37	2
Recovered	0.99977223478631	43611232.2349777	603.037756126866
	56	7	5
Deaths	0.99992557004894	724820.399591596	3814.22698518035
	9	2	26

On prediction of cases in INDIA

	R2 score	Mean square	Mean absolute
Confirmed	0.99331190040362	2704197.57931742	861.705369521990
	2	8	6
Active	0.99608041989093	695454.590166281	442.715523953252
	99	2	8
Recovered	0.98265378183551	673471.166019352	425.514801558385
	47	3	3
Deaths	0.99476982387658	2291.27096809311	26.0715598353158
	3	16	67

View publication str

Conclusion

Through this project, the analysis on COVID-19 data has been performed successfully. The analysis on this pandemic spread has been done and compared between different countries. The analysis of confirmed cases, active cases, recovered cases and deaths are done separately to give a clear look on how the virus is spreading, which countries are getting affected mostly and how different countries are recovering. A separate analysis on cases of INDIA has been done and predictions of different cases both around the world and INDIA has been done. At last, the accuracy check using different metrics is performed over all the analysis done in this project.