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# Apple Stock price prediction using Python (Google Colab) and Facebook Prophet

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https://github.com/Nygmet29/AP.git https://youtu.be/\_eJq6bwK2is

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**Apple Stock price prediction using Python (Google Colab)** and Facebook Prophet - is the program written in Google Colab with imported Facebook Prophet library to predict prices. We decided to develop this project, since Stock Prices of world companies are inspected by millions of investors and program that will predict not only Apple's stock prices, but other companies' stock values (if we import history prices) will be quite effective application.

Literature review (links that we used are at the end of the report)

From Nasdaq website we downloaded our dataset of Apple Inc. stock price in (.csv) format since 2018th year till 2025th.

Current work

1) Firstly, we imported pandas library in order to analyze our dataset

```
[ ] #import pandas library
import pandas as pd
```

2) We wrote this code in order to import our dataset from PC

```
#import our dataset from PC
from google.colab import files
files.upload()
```

3) Then we created dataframe that copied all information from our dataset, after that we removed \$ sign from the values of "Close/Last" column, because pandas reads only the numeric values, and then inverted our list of dataset.

```
df = pd.read_csv('AAPL.csv')
 df['Close/Last'] = df['Close/Last'].str.replace('$', '')
 df['Close/Last'] = pd.to numeric(df['Close/Last'])
 df = df.iloc[::-1]
 df
          Date Close/Last
                               Volume
                                                   High
                                          Open
                                                               Low
1258 02/20/2018
                     42.9625 134762640 $43.0125
                                                 $43.565
                                                           $42.855
1257 02/21/2018
                     42.7675 149512280 $43.2075
                                                  $43.53
                                                          $42.7525
1256 02/22/2018
                     43.1250 123815040
                                         $42.95 $43.4875
                                                          $42.9275
1255 02/23/2018
                     43.8750 135088200 $43.4175 $43.9125
                                                           $43.385
1254 02/26/2018
                     44.7425 149414680 $44.0875 $44.8475
                                                          $44.0525
                    153.8500
                                                 $154.26
      02/13/2023
                              62199010 $150.952
                                                           $150.92
  4
      02/14/2023
                    153.2000
                              61707570
                                        $152.12
                                                 $153.77
                                                           $150.86
      02/15/2023
                    155.3300
                              65669250
                                        $153.11
                                                  $155.5
                                                           $152.88
      02/16/2023
                    153.7100
                              68167940
                                        $153.51
                                                 $156.33 $153.3475
      02/17/2023
                    152.5500
                              59144120
                                        $152.35
                                                    $153
                                                           $150.85
1259 rows × 6 columns
```

4) We renamed "Date" and "Close/Last" columns to "ds" and "y", because Facebook prophet reads and outputs in 'ds' and 'y' format.

```
df = df[['Date', 'Close/Last']]
df = df.rename(columns={'Date': 'ds', 'Close/Last': 'y'})
             ds
                      Y
1258 02/20/2018
                 42.9625
1257 02/21/2018
                 42.7675
1256 02/22/2018 43.1250
1255 02/23/2018 43.8750
1254 02/26/2018 44.7425
  4 02/13/2023 153.8500
  3 02/14/2023 153.2000
  2 02/15/2023 155.3300
  1 02/16/2023 153.7100
      02/17/2023 152.5500
1259 rows × 2 columns
```

5) We separated last 30 days and placed them into the array named "last30days\_for\_pred" in order to compare predicted values with the actual values from this array.

```
last30days_for_pred = df[len(df)-30:]
last30days_for_pred
    ds y
29 01/06/2023 129.62
28 01/09/2023 130.15
27 01/10/2023 130.73
26 01/11/2023 133.49
25 01/12/2023 133.41
24 01/13/2023 134.76
23 01/17/2023 135.94
22 01/18/2023 135.21
21 01/19/2023 135.27
20 01/20/2023 137.87
19 01/23/2023 141.11
18 01/24/2023 142.53
17 01/25/2023 141.86
16 01/26/2023 143.96
```

6) Then we removed last 30 days from dataframe.

7) Here we installed fbprophet library in order to forecast prices

```
# installation of pystan
pip install pystan=2.19.1.1

Looking in indexes: https://pyj.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
Collecting pystan=2.19.1.1 pownloading pystan=2.19.1.1-cp38-cp38-manylinuxl_x86_64.whl (62.6 MB)
Downloading pystan=2.19.1.1-cp38-cp38-manylinuxl_x86_64.whl (62.6 MB)
Requirement already satisfied: Cython!=0.25.1,>=0.22 in /usr/local/lib/python3.8/dist-packages (from pystan==2.19.1.1) (0.29.33)
Requirement already satisfied: numpy>=1.7 in /usr/local/lib/python3.8/dist-packages (from pystan==2.19.1.1) (1.21.6)
Installing collected packages: pystan
Found existing installation: pystan 3.3.0
Uninstalling pystan=3.3.0:
Successfully uninstalled pystan=3.3.0
Successfully installed pystan=2.19.1.1

# installation of Facebook prophet
pip install fbprophet

Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
Collecting fbprophet

Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
Collecting fbprophet

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Collecting fbprophet

Requirement already satisfied: cmdstanpy=0.9 5 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (0.29.33)
Requirement already satisfied: cmdstanpy=0.9 5 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (0.9.5)
Requirement already satisfied in mumpy>=1.15.4 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (1.21.6)
Requirement already satisfied in mumpy>=1.0.4 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (1.3.5)
Requirement already satisfied in mumpy>=1.0.2 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (1.3.5)
Requirement already satisfied in mumpy>=1.0.2 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (0.0.9)
Requirement already satisfied in mumpy>=1.0.2 in /usr/local/lib/python3.8/dist-packages (from fbprophet) (2.4.0)
Requirement already satisfied in mumpy>=1.0.2 in /usr/local/lib/python3.8/dist
```

#### Data & Methods

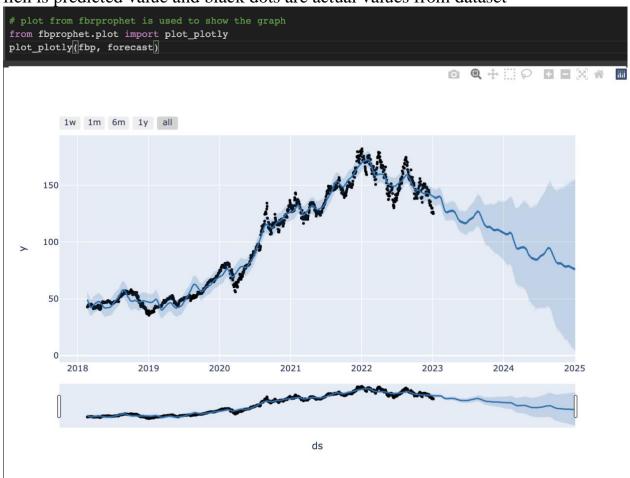
8) Here we fitted our Prophet object, then we created predicted future dataframe for the next 2 years and implemented it to forecast object.

```
[ ] # here prophet library imported, object of prophet, future object is created and predicted and model is trained from fbprophet import Prophet
fbp = Prophet(daily_seasonality = True)
fbp.fit(df)
future = fbp.make_future_dataframe(periods=730)
forecast = fbp.predict(future)
```

In the given code the object of Prophet (fdp) fitted all data from 'df' which is dataframe that consists dataset copy. Then future object trained dataset for the next 730 days according given dataset, and forecast function made a predict of future object.

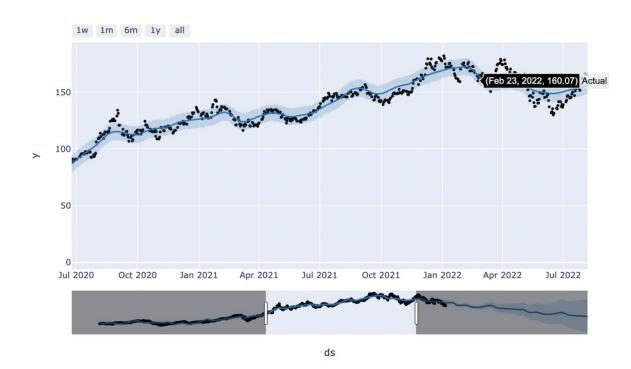
## Results

9) Written code shows the graph of dataframe with predicted values, where blue lien is predicted value and black dots are actual values from dataset

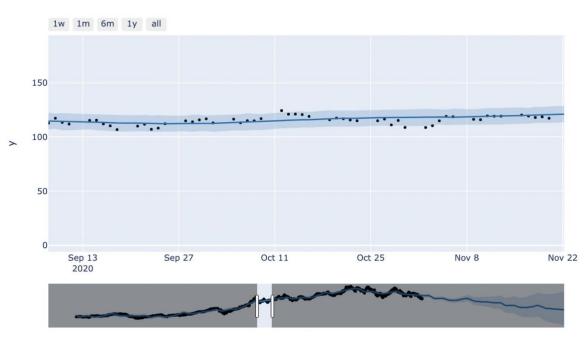


In these graphs you can see how the stock price is changing every day, week, month, year and you also can choose a period of time and see how the stock price changed during a certain period of time.









10) here we can predict any value from 2018 till 2025, it shows our predicted value

```
[ ] # here we can predict any value from 2018 till 2025
   forecast[forecast.ds == '02/14/2023']['yhat']

1268   138.452488
Name: yhat, dtype: float64
```

11) Here you can see the actual value and if you compare with the predicted, you can see the difference

```
# here we can output the actual value in order to compare and check the difference
last30days_for_pred[last30days_for_pred.ds == '02/14/2023']['y']

3    153.2
Name: y, dtype: float64
```

As you can see, our predicted value was "138.452488", but actual value was "153.2". So the difference is 14.747512.

#### Discussion

When we saw the graph, we thought that predicted value will be higher, but it dramatically decreased. Also, as we said in the 5th step, we put away some array with 30 days stock price information, and then deleted it from whole dataframe in order to check the correctness ratio, and difference between predicted price and actual value was quite more than normal. And, surprisingly, our "future" object modeled predicted values for whole dataset from 2018 till 2025 years.

## References

- Website from where we downloaded the dataset of Apple stock price for the last 5 years (2018-2023) - <a href="https://www.nasdaq.com/market-activity/stocks/aapl/historical">https://www.nasdaq.com/market-activity/stocks/aapl/historical</a>
- Website from where we acknowledged how to connect Facebook Prophet library to out Google Colab <a href="https://machinelearningmastery.com/time-series-forecasting-with-prophet-in-python/#:~:text=To%20use%20Prophet%20for%20forecasting,type%20of%20seasonality%2C%20and%20more.">https://machinelearningmastery.com/time-series-forecasting-with-prophet-in-python/#:~:text=To%20use%20Prophet%20for%20forecasting,type%20of%20seasonality%2C%20and%20more.</a>
- From StackOverflow we got aware how to invert the dataset list <a href="https://stackoverflow.com/questions/10933838/how-to-read-a-csv-file-in-reverse-order-in-python">https://stackoverflow.com/questions/10933838/how-to-read-a-csv-file-in-reverse-order-in-python</a>