## TASK-5 Stock-Price-Prediction

```
In [31]:
           #import packages
           import pandas as pd
           import numpy as np
           #to plot within notebook
           import matplotlib.pyplot as plt
           %matplotlib inline
           #setting figure size
           from matplotlib.pylab import rcParams
           rcParams['figure.figsize'] = 20,10
           #for normalizing data
           from sklearn.preprocessing import MinMaxScaler
           scaler = MinMaxScaler(feature range=(0, 1))
In [32]:
           #read the file
           df = pd.read csv('NSE-TATAGLOBAL11.csv')
In [33]:
           df.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 1235 entries, 0 to 1234
          Data columns (total 8 columns):
           #
               Column
                                       Non-Null Count
                                                        Dtype
           0
               Date
                                                        object
                                       1235 non-null
           1
                                       1235 non-null
                                                        float64
               0pen
           2
                                                        float64
               High
                                       1235 non-null
           3
                                       1235 non-null
                                                        float64
               Low
           4
                                       1235 non-null
                                                        float64
               Last
           5
                                       1235 non-null
                                                        float64
               Close
               Total Trade Quantity 1235 non-null
                                                         int64
               Turnover (Lacs)
                                       1235 non-null
                                                        float64
          dtypes: float64(6), int64(1), object(1)
          memory usage: 77.3+ KB
In [34]:
           df.describe()
Out[34]:
                                                                                Total Trade
                                                                                               Turnover
                       Open
                                   High
                                                Low
                                                            Last
                                                                       Close
                                                                                  Quantity
                                                                                                 (Lacs)
          count 1235.000000
                             1235.000000
                                         1235.000000
                                                     1235.000000 1235.000000 1.235000e+03
                                                                                            1235.000000
                  168.954858
                              171.429069
                                                                   168.731053 2.604151e+06
                                          166.402308
                                                      168.736356
                                                                                            4843.166502
          mean
            std
                   51.499145
                               52.436761
                                           50.542919
                                                       51.587384
                                                                   51.544928 2.277028e+06
                                                                                            5348.919832
                  103.000000
                              104.600000
                                          100.000000
                                                      102.600000
                                                                   102.650000 1.001800e+05
            min
                                                                                             128.040000
           25%
                  137.550000
                              138.925000
                                          135.250000
                                                      137.175000
                                                                   137.225000 1.284482e+06
                                                                                            1801.035000
           50%
                  151.500000
                              153.250000
                                          149.500000
                                                      151.200000
                                                                  151.100000 1.964885e+06
                                                                                            3068.510000
```

```
Total Trade
                                                                                               Turnover
                       Open
                                    High
                                                                        Close
                                                Low
                                                             Last
                                                                                  Quantity
                                                                                                  (Lacs)
            75%
                  169.000000
                              172.325000
                                           166.700000
                                                       169.100000
                                                                   169.500000
                                                                              3.095788e+06
                                                                                             5852.600000
                  327.700000
                              328.750000
                                           321.650000
                                                       325.950000
                                                                   325.750000 2.919102e+07 55755.080000
            max
In [35]:
           #print the head
           df.head()
                         Open
                                 High
                                                      Close
                                                           Total Trade Quantity Turnover (Lacs)
Out[35]:
                   Date
                                         Low
                                                Last
             08-10-2018 208.00 222.25 206.85 216.00 215.15
                                                                       4642146
                                                                                      10062.83
             05-10-2018 217.00 218.60 205.90 210.25 209.20
                                                                       3519515
                                                                                       7407.06
            04-10-2018 223.50 227.80 216.15 217.25 218.20
                                                                       1728786
                                                                                       3815.79
            03-10-2018 230.00 237.50 225.75 226.45 227.60
                                                                       1708590
                                                                                       3960.27
          4 01-10-2018 234.55 234.60 221.05 230.30 230.90
                                                                       1534749
                                                                                       3486.05
In [36]:
           df.shape
Out[36]: (1235, 8)
In [37]:
           df.isnull().values.any()
          False
Out[37]:
In [38]:
           df.isnull().sum()
                                    0
          Date
Out[38]:
          Open
                                    0
          High
                                    0
          Low
                                    0
          Last
                                    0
          Close
                                    0
          Total Trade Quantity
                                    0
          Turnover (Lacs)
                                    0
          dtype: int64
In [39]:
           #setting index as date
           df['Date'] = pd.to_datetime(df.Date,format='%d-%m-%Y')
           df.index = df['Date']
           #plot
           plt.figure(figsize=(16,8))
           plt.plot(df['Close'], label='Close Price history')
Out[39]: [<matplotlib.lines.Line2D at 0x1c1fa873208>]
```

```
300
         250
          200
         150
         100
                   2014
                                  2015
                                                   2016
                                                                  2017
                                                                                  2018
                                                                                                  2019
In [40]:
          #creating dataframe with date and the target variable
           data = df.sort_index(ascending=True, axis=0)
          new_data = pd.DataFrame(index=range(0,len(df)),columns=['Date', 'Close'])
          for i in range(0,len(data)):
                new_data['Date'][i] = data['Date'][i]
                new_data['Close'][i] = data['Close'][i]
In [41]:
          #splitting into train and validation
          train = new_data[:987]
          valid = new data[987:]
In [42]:
          new data.shape, train.shape, valid.shape
          ((1235, 2), (987, 2), (248, 2))
Out[42]:
In [43]:
          train['Date'].min(), train['Date'].max(), valid['Date'].min(), valid['Date'].max()
          (Timestamp('2013-10-08 00:00:00'),
Out[43]:
           Timestamp('2017-10-06 00:00:00'),
           Timestamp('2017-10-09 00:00:00'),
           Timestamp('2018-10-08 00:00:00'))
In [44]:
          #make predictions
          preds = []
          for i in range(0,248):
               a = train['Close'][len(train)-248+i:].sum() + sum(preds)
               b = a/248
               preds.append(b)
In [45]:
           #calculate rmse
          rms=np.sqrt(np.mean(np.power((np.array(valid['Close'])-preds),2)))
```

```
Out[45]: 104.51415465984348
```

```
In [46]: #plot
    valid['Predictions'] = 0
    valid['Predictions'] = preds
    plt.plot(train['Close'])
    plt.plot(valid[['Close', 'Predictions']])
```

c:\users\vrinda bajaj\python 3.7.2\lib\site-packages\ipykernel\_launcher.py:2: SettingWit
hCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy

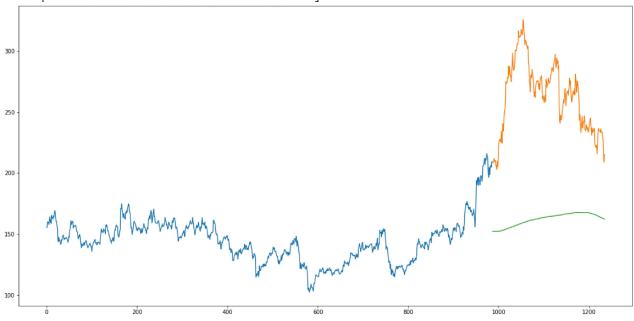
c:\users\vrinda bajaj\python 3.7.2\lib\site-packages\ipykernel\_launcher.py:3: SettingWit
hCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy

This is separate from the ipykernel package so we can avoid doing imports until

Out[46]: [<matplotlib.lines.Line2D at 0x1c1a8f7e5f8>, <matplotlib.lines.Line2D at 0x1c1a8f7e6a0>]



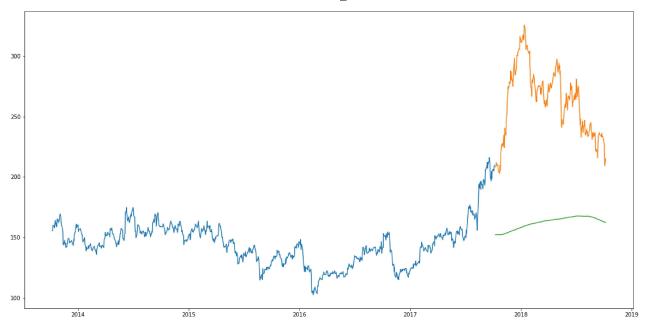
## LINEAR REGRESSION

```
In [47]: #setting index as date values
df['Date'] = pd.to_datetime(df.Date,format='%Y-%m-%d')
df.index = df['Date']

#sorting
data = df.sort_index(ascending=True, axis=0)

#creating a separate dataset
```

```
new data = pd.DataFrame(index=range(0,len(df)),columns=['Date', 'Close'])
          for i in range(0,len(data)):
              new data['Date'][i] = data['Date'][i]
              new_data['Close'][i] = data['Close'][i]
In [48]:
          new data.drop('Date', axis=1, inplace=True) #elapsed will be the time stamp
In [49]:
          #split into train and validation
          train = new_data[:987]
          valid = new data[987:]
          x_train = train.drop('Close', axis=1)
          y train = train['Close']
          x_valid = valid.drop('Close', axis=1)
          y_valid = valid['Close']
          #implement linear regression
          from sklearn.linear model import LinearRegression
          model = LinearRegression()
In [50]:
          #plot
          valid['Predictions'] = 0
          valid['Predictions'] = preds
          valid.index = data[987:].index
          train.index = data[:987].index
          plt.plot(train['Close'])
          plt.plot(valid[['Close', 'Predictions']])
         c:\users\vrinda bajaj\python 3.7.2\lib\site-packages\ipykernel launcher.py:2: SettingWit
         hCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row indexer,col indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user
         guide/indexing.html#returning-a-view-versus-a-copy
         c:\users\vrinda bajaj\python 3.7.2\lib\site-packages\ipykernel launcher.py:3: SettingWit
         hCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row_indexer,col_indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user
         guide/indexing.html#returning-a-view-versus-a-copy
           This is separate from the ipykernel package so we can avoid doing imports until
Out[50]: [<matplotlib.lines.Line2D at 0x1c1a8ff1c18>.
          <matplotlib.lines.Line2D at 0x1c1a8ff1cc0>]
```



## LONG SHORT TERM MEMORY(LSTM)

```
In [51]:
          #importing required libraries
          from sklearn.preprocessing import MinMaxScaler
          from keras.models import Sequential
          from keras.layers import Dense, Dropout, LSTM
          #creating dataframe
          data = df.sort_index(ascending=True, axis=0)
          new data = pd.DataFrame(index=range(0,len(df)),columns=['Date', 'Close'])
          for i in range(0,len(data)):
              new data['Date'][i] = data['Date'][i]
              new_data['Close'][i] = data['Close'][i]
          #setting index
          new data.index = new data.Date
          new_data.drop('Date', axis=1, inplace=True)
          #creating train and test sets
          dataset = new_data.values
          train = dataset[0:987,:]
          valid = dataset[987:,:]
          #converting dataset into x_train and y_train
          scaler = MinMaxScaler(feature_range=(0, 1))
          scaled_data = scaler.fit_transform(dataset)
          x_train, y_train = [], []
          for i in range(60,len(train)):
              x_train.append(scaled_data[i-60:i,0])
              y train.append(scaled data[i,0])
          x_train, y_train = np.array(x_train), np.array(y_train)
          x_train = np.reshape(x_train, (x_train.shape[0],x_train.shape[1],1))
          # create and fit the LSTM network
          model = Sequential()
```

```
model.add(LSTM(units=50, return sequences=True, input shape=(x train.shape[1],1)))
model.add(LSTM(units=50))
model.add(Dense(1))
model.compile(loss='mean_squared_error', optimizer='adam')
model.fit(x train, y train, epochs=1, batch size=1, verbose=2)
#predicting 246 values, using past 60 from the train data
inputs = new_data[len(new_data) - len(valid) - 60:].values
inputs = inputs.reshape(-1,1)
inputs = scaler.transform(inputs)
X test = []
for i in range(60,inputs.shape[0]):
    X_test.append(inputs[i-60:i,0])
X_test = np.array(X_test)
X_test = np.reshape(X_test, (X_test.shape[0], X_test.shape[1],1))
closing price = model.predict(X test)
closing price = scaler.inverse transform(closing price)
```

927/927 - 13s - loss: 0.0012

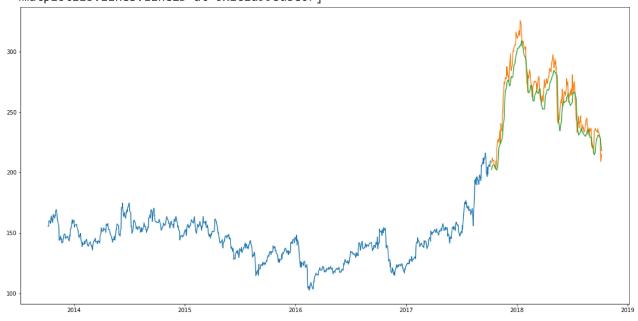
```
In [52]: #for plotting
    train = new_data[:987]
    valid = new_data[987:]
    valid['Predictions'] = closing_price
    plt.plot(train['Close'])
    plt.plot(valid[['Close', 'Predictions']])
```

c:\users\vrinda bajaj\python 3.7.2\lib\site-packages\ipykernel\_launcher.py:4: SettingWit
hCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row indexer,col indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy after removing the cwd from sys.path.

Out[52]: [<matplotlib.lines.Line2D at 0x1c1a99baa20>, <matplotlib.lines.Line2D at 0x1c1a998dbe0>]



In [ ]: