Task-1: BEGINNER LEVEL TASK

Task_1-2: Stock Market Prediction And Forecasting Using Stacked LSTM

Datasetlinks: : https://raw.githubusercontent.com/mwitiderrick/stockprice/master/NSE-TATAGLOBAL.csv https://raw.githubusercontent.com/mwitiderrick/stockprice/master/NSE-TATAGLOBAL.csv)

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Importing Libraries

In [1]:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import tensorflow as tf
from sklearn.preprocessing import MinMaxScaler
import warnings
warnings.filterwarnings("ignore")
```

Reading the DataSet

In [2]:

```
#Import the data and remove rows containing NAN values
stk = pd.read_csv("stock.csv")
```

In [3]:

```
stk.head() # Prints first Five Rows
```

Out[3]:

	Date	Open	High	Low	Last	Close	Total Trade Quantity	Turnover (Lacs)
0	2018-09-28	234.05	235.95	230.20	233.50	233.75	3069914	7162.35
1	2018-09-27	234.55	236.80	231.10	233.80	233.25	5082859	11859.95
2	2018-09-26	240.00	240.00	232.50	235.00	234.25	2240909	5248.60
3	2018-09-25	233.30	236.75	232.00	236.25	236.10	2349368	5503.90
4	2018-09-24	233.55	239.20	230.75	234.00	233.30	3423509	7999.55

In [4]:

stk.tail() # Prints Last five rows

Out[4]:

	Date	Open	High	Low	Last	Close	Total Trade Quantity	Turnover (Lacs)
2030	2010-07-27	117.6	119.50	112.00	118.80	118.65	586100	694.98
2031	2010-07-26	120.1	121.00	117.10	117.10	117.60	658440	780.01
2032	2010-07-23	121.8	121.95	120.25	120.35	120.65	281312	340.31
2033	2010-07-22	120.3	122.00	120.25	120.75	120.90	293312	355.17
2034	2010-07-21	122.1	123.00	121.05	121.10	121.55	658666	803.56

In [5]:

stk.describe()

Out[5]:

	Open	High	Low	Last	Close	Total Trade Quantity	Turr (l
count	2035.000000	2035.000000	2035.000000	2035.000000	2035.00000	2.035000e+03	2035.00
mean	149.713735	151.992826	147.293931	149.474251	149.45027	2.335681e+06	3899.98
std	48.664509	49.413109	47.931958	48.732570	48.71204	2.091778e+06	4570.76
min	81.100000	82.800000	80.000000	81.000000	80.95000	3.961000e+04	37.04
25%	120.025000	122.100000	118.300000	120.075000	120.05000	1.146444e+06	1427.46
50%	141.500000	143.400000	139.600000	141.100000	141.25000	1.783456e+06	2512.03
75%	157.175000	159.400000	155.150000	156.925000	156.90000	2.813594e+06	4539.01
max	327.700000	328.750000	321.650000	325.950000	325.75000	2.919102e+07	55755.08
4							

In [6]:

stk.isnull()

Out[6]:

	Date	Open	High	Low	Last	Close	Total Trade Quantity	Turnover (Lacs)
0	False	False						
1	False	False						
2	False	False						
3	False	False						
4	False	False						
2030	False	False						
2031	False	False						
2032	False	False						
2033	False	False						
2034	False	False						

2035 rows × 8 columns

Sorting the Data

In [7]:

```
stk['Date']=pd.to_datetime(stk['Date'])
print(type(stk.Date[0]))
```

<class 'pandas._libs.tslibs.timestamps.Timestamp'>

In [8]:

```
df=stk.sort_values(by='Date')
df.head()
```

Out[8]:

	Date	Open	High	Low	Last	Close	Total Trade Quantity	Turnover (Lacs)
2034	2010-07-21	122.1	123.00	121.05	121.10	121.55	658666	803.56
2033	2010-07-22	120.3	122.00	120.25	120.75	120.90	293312	355.17
2032	2010-07-23	121.8	121.95	120.25	120.35	120.65	281312	340.31
2031	2010-07-26	120.1	121.00	117.10	117.10	117.60	658440	780.01
2030	2010-07-27	117.6	119.50	112.00	118.80	118.65	586100	694.98

In [9]:

```
df.reset_index(inplace=True)
```

In [10]:

df.head()

Out[10]:

	index	Date	Open	High	Low	Last	Close	Total Trade Quantity	Turnover (Lacs)
0	2034	2010-07- 21	122.1	123.00	121.05	121.10	121.55	658666	803.56
1	2033	2010-07- 22	120.3	122.00	120.25	120.75	120.90	293312	355.17
2	2032	2010-07- 23	121.8	121.95	120.25	120.35	120.65	281312	340.31
3	2031	2010-07- 26	120.1	121.00	117.10	117.10	117.60	658440	780.01
4	2030	2010-07- 27	117.6	119.50	112.00	118.80	118.65	586100	694.98

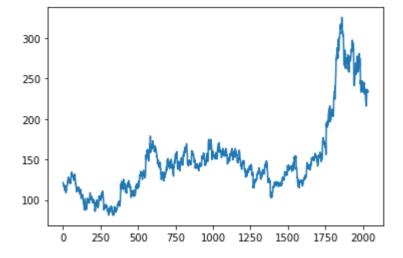
Data Visualization

In [11]:

plt.plot(df['Close'])

Out[11]:

[<matplotlib.lines.Line2D at 0x7f79f8ed05d0>]



```
In [12]:
dff=df['Close']
dff
Out[12]:
        121.55
1
        120.90
2
        120.65
3
        117.60
4
        118.65
         . . .
        233.30
2030
2031
        236.10
        234.25
2032
        233.25
2033
2034
        233.75
Name: Close, Length: 2035, dtype: float64
```

Min Max Scaler

```
In [13]:
```

Spliting the Dataset

[0.62418301]])

```
In [14]:
```

```
training_size=int(len(dff)*0.70)
test_size=len(dff)-training_size
train_data,test_data=dff[0:training_size,:],dff[training_size:len(dff),:1]
```

Converting an Array of values into a Dataset matrix

```
In [15]:
```

```
def create_dataset(dataset, time_step=1):
    dataX, dataY = [], []
    for i in range(len(dataset)-time_step-1):
        a = dataset[i:(i+time_step), 0]
        dataX.append(a)
        dataY.append(dataset[i + time_step, 0])
    return np.array(dataX), np.array(dataY)
```

Spliting the Data into Train and Test

```
In [16]:
time_step = 100
X_train, y_train = create_dataset(train_data, time_step)
X_test, ytest = create_dataset(test_data, time_step)
In [17]:
print(X_train.shape), print(y_train.shape)
(1323, 100)
(1323,)
Out[17]:
(None, None)
In [18]:
print(X_test.shape), print(ytest.shape)
(510, 100)
(510,)
Out[18]:
(None, None)
In [19]:
X_train =X_train.reshape(X_train.shape[0],X_train.shape[1] , 1)
X_test = X_test.reshape(X_test.shape[0], X_test.shape[1] , 1)
```

Creating the stacked LSTM Model

```
In [20]:
```

```
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import LSTM
```

In [21]:

```
model=Sequential()
model.add(LSTM(50, return_sequences=True, input_shape=(100,1)))
model.add(LSTM(50, return_sequences=True))
model.add(LSTM(50))
model.add(Dense(1))
model.compile(loss='mean_squared_error',optimizer='adam')
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
lstm (LSTM)	(None, 100, 50)	10400
lstm_1 (LSTM)	(None, 100, 50)	20200
lstm_2 (LSTM)	(None, 50)	20200
dense (Dense)	(None, 1)	51
=======================================		

Total params: 50,851 Trainable params: 50,851 Non-trainable params: 0

In [22]:

```
model.fit(X_train,y_train,validation_split=0.1,epochs=60,batch_size=64,verbose=1)
Epoch 1/60
al_loss: 0.0022
Epoch 2/60
l_loss: 0.0014
Epoch 3/60
19/19 [================= ] - 7s 387ms/step - loss: 9.0782e-04
val loss: 0.0012
Epoch 4/60
19/19 [================= ] - 7s 357ms/step - loss: 8.4069e-04
- val loss: 0.0011
Epoch 5/60
19/19 [================= ] - 5s 276ms/step - loss: 8.1129e-04
- val_loss: 0.0011
Epoch 6/60
19/19 [================ ] - 3s 180ms/step - loss: 8.2129e-04
- val loss: 0.0011
```

Prediction and Checking Performance

In [23]:

Epoch 7/60

```
test predict=model.predict(X test)
```

In [24]:

```
test predicted=scaler.inverse transform(test predict)
test_predicted
       [226.03673],
      [228.11343],
      [228.45093],
       [229.2298],
       [222.62845],
      [216.06458],
      [212.7183],
       [213.67584],
      [215.61908],
      [214.01822],
      [215.49237],
      [223.54834],
      [231.23088],
      [233.15855],
       [231.00076],
      [227.71596],
      [224.64618],
      [224.62502],
       [224.81697 ]], dtype=float32)
```

Calculating the Performance

```
In [25]:
```

```
import math
from sklearn.metrics import mean_squared_error
```

In [26]:

```
performance = math.sqrt(mean_squared_error(ytest,test_predict))
performance
```

Out[26]:

0.04607845860418348