

Stock_Market_Prediction

April 30, 2023

0.1 STOCK MARKET PREDICTION

Importing the libraries

```
[ ]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

```
[ ]: from sklearn import preprocessing
from sklearn.ensemble import RandomForestRegressor
from sklearn.model_selection import train_test_split
from matplotlib import style
style.use("fivethirtyeight")
```

Read the csv file into a DataFrame

```
[ ]: df = pd.read_csv(r"tesla.csv")
```

```
[ ]: df.head(10)
```

```
[ ]:
```

	Date	Open	High	Low	Close	Volume	Adj Close
0	6/29/2010	19.000000	25.000000	17.540001	23.889999	18766300	23.889999
1	6/30/2010	25.790001	30.420000	23.299999	23.830000	17187100	23.830000
2	7/1/2010	25.000000	25.920000	20.270000	21.959999	8218800	21.959999
3	7/2/2010	23.000000	23.100000	18.709999	19.200001	5139800	19.200001
4	7/6/2010	20.000000	20.000000	15.830000	16.110001	6866900	16.110001
5	7/7/2010	16.400000	16.629999	14.980000	15.800000	6921700	15.800000
6	7/8/2010	16.139999	17.520000	15.570000	17.459999	7711400	17.459999
7	7/9/2010	17.580000	17.900000	16.549999	17.400000	4050600	17.400000
8	7/12/2010	17.950001	18.070000	17.000000	17.049999	2202500	17.049999
9	7/13/2010	17.389999	18.639999	16.900000	18.139999	2680100	18.139999

```
[ ]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1692 entries, 0 to 1691
Data columns (total 7 columns):
 #   Column      Non-Null Count  Dtype
---  -

```

```

0    Date      1692 non-null  object
1    Open      1692 non-null  float64
2    High      1692 non-null  float64
3    Low       1692 non-null  float64
4    Close     1692 non-null  float64
5    Volume    1692 non-null  int64
6    Adj Close 1692 non-null  float64
dtypes: float64(5), int64(1), object(1)
memory usage: 92.7+ KB

```

```
[ ]: df.dtypes
```

```

[ ]: Date      object
    Open      float64
    High      float64
    Low       float64
    Close     float64
    Volume    int64
    Adj Close  float64
dtype: object

```

Make two new columns which will be used for making predictions.

```
[ ]: df["HL_Perc"] = (df["High"]-df["Low"]) / df["Low"] * 100
    df["CO_Perc"] = (df["Close"] - df["Open"]) / df["Open"] * 100
```

```
[ ]: dates = np.array(df["Date"])
    dates_check = dates[-30:]
    dates = dates[:-30]
```

```
[ ]: df = df[["HL_Perc", "CO_Perc", "Adj Close", "Volume"]]
```

Define the label column

```
[ ]: df["PriceNextMonth"] = df["Adj Close"].shift(-30)
```

```
[ ]: df.tail(10)
```

```

[ ]:
      HL_Perc  CO_Perc  Adj Close  Volume  PriceNextMonth
1682  1.692862  1.331129  251.210007  3351200           NaN
1683  2.243070 -1.321849  248.589996  3449200           NaN
1684  1.936246 -0.052634  246.869995  3725200           NaN
1685  2.329220 -1.102456  244.899994  3861500           NaN
1686  1.440329 -1.023519  243.690002  3057000           NaN
1687  1.676418  0.551422  246.169998  3010700           NaN
1688  4.918296  4.831173  258.000000  7575500           NaN
1689  2.646791 -0.494165  255.729996  4816600           NaN
1690  2.582414 -0.133386  262.049988  7100400           NaN

```

1691 1.581154 -0.946970 261.500000 6475900 NaN

Make feature and label arrays

```
[ ]: X = np.array(df.drop(["PriceNextMonth"], 1))
X = preprocessing.scale(X)
X_Check = X[-30:]
X = X[:-30]
df.dropna(inplace = True)
y = np.array(df["PriceNextMonth"])
```

<ipython-input-14-383e8e913f7f>:1: FutureWarning: In a future version of pandas all arguments of DataFrame.drop except for the argument 'labels' will be keyword-only.

```
X = np.array(df.drop(["PriceNextMonth"], 1))
```

Divide the data set into training data and testing data

```
[ ]: X_train, X_test, y_train, y_test = train_test_split(X,y,test_size = 0.2)
```

Define the prediction model

```
[ ]: model = RandomForestRegressor()
```

Fit the model using training data

```
[ ]: model.fit(X_train, y_train)
```

```
[ ]: RandomForestRegressor()
```

Calculate the confidence value by applying the model to testing data

```
[ ]: conf = model.score(X_test, y_test)
print(conf)
```

0.9599554504935848

Fit the model again using the whole data set

```
[ ]: model.fit(X,y)
```

```
[ ]: RandomForestRegressor()
```

```
[ ]: predictions = model.predict(X_Check)
```

Make the final DataFrame containing Dates, ClosePrices, and Forecast values

```
[ ]: actual = pd.DataFrame(dates, columns = ["Date"])
actual["ClosePrice"] = df["Adj Close"]
actual["Forecast"] = np.nan
```

```

actual.set_index("Date", inplace = True)
forecast = pd.DataFrame(dates_check, columns=["Date"])
forecast["Forecast"] = predictions
forecast["ClosePrice"] = np.nan
forecast.set_index("Date", inplace = True)
var = [actual, forecast]
result = pd.concat(var)

```

Plotting the final results

```

[ ]: result.plot(figsize=(20,10), linewidth=1.5)
plt.legend(loc=2, prop={'size':20})
plt.xlabel('Date')
plt.ylabel('Price')

```

```

[ ]: Text(0, 0.5, 'Price')

```



```

[ ]:

```