

Date: 17-10-2023

Project Title: Stock Price Prediction

Team ID: 3892

1.Importing required packages:

```
In [51]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
import seaborn as sns
```

2. Loading the dataset:

```
In [2]: data = pd.read_csv("MSFT.csv")
```

3.Display the maximum columns:

```
In [5]: pd.options.display.max_columns = None
```

In [6]: data

Out[6]:

	Date	Open	High	Low	Close	Adj Close	Volume
0	1986-03-13	0.088542	0.101563	0.088542	0.097222	0.062549	1031788800
1	1986-03-14	0.097222	0.102431	0.097222	0.100694	0.064783	308160000
2	1986-03-17	0.100694	0.103299	0.100694	0.102431	0.065899	133171200
3	1986-03-18	0.102431	0.103299	0.098958	0.099826	0.064224	67766400
4	1986-03-19	0.099826	0.100694	0.097222	0.098090	0.063107	47894400
...
8520	2019-12-31	156.770004	157.770004	156.449997	157.699997	157.699997	18369400
8521	2020-01-02	158.779999	160.729996	158.330002	160.619995	160.619995	22622100
8522	2020-01-03	158.320007	159.949997	158.059998	158.619995	158.619995	21116200
8523	2020-01-06	157.080002	159.100006	156.509995	159.029999	159.029999	20813700
8524	2020-01-07	159.320007	159.669998	157.330002	157.580002	157.580002	18017762

8525 rows × 7 columns

4.Displaying top 5 Rows of data:

In [7]: data.head()

Out[7]:

	Date	Open	High	Low	Close	Adj Close	Volume
0	1986-03-13	0.088542	0.101563	0.088542	0.097222	0.062549	1031788800
1	1986-03-14	0.097222	0.102431	0.097222	0.100694	0.064783	308160000
2	1986-03-17	0.100694	0.103299	0.100694	0.102431	0.065899	133171200
3	1986-03-18	0.102431	0.103299	0.098958	0.099826	0.064224	67766400
4	1986-03-19	0.099826	0.100694	0.097222	0.098090	0.063107	47894400

5.Displaying last 5 rows of data:

In [9]: `data.tail()`

Out[9]:

	Date	Open	High	Low	Close	Adj Close	Volume
8520	2019-12-31	156.770004	157.770004	156.449997	157.699997	157.699997	18369400
8521	2020-01-02	158.779999	160.729996	158.330002	160.619995	160.619995	22622100
8522	2020-01-03	158.320007	159.949997	158.059998	158.619995	158.619995	21116200
8523	2020-01-06	157.080002	159.100006	156.509995	159.029999	159.029999	20813700
8524	2020-01-07	159.320007	159.669998	157.330002	157.580002	157.580002	18017762

6.Shows number of rows and columns:

In [10]: `print("Number of columns ",data.shape[1])`
`print("Number of rows ",data.shape[0])`

Number of columns 7
 Number of rows 8525

7.Getting basic information about the dataset:

In [11]: `data.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8525 entries, 0 to 8524
Data columns (total 7 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Date        8525 non-null   object
1   Open        8525 non-null   float64
2   High        8525 non-null   float64
3   Low         8525 non-null   float64
4   Close       8525 non-null   float64
5   Adj Close   8525 non-null   float64
6   Volume      8525 non-null   int64
dtypes: float64(5), int64(1), object(1)
memory usage: 466.3+ KB
```

8.Data Exploration:

In [13]: `data['Date'].unique()`

Out[13]: `array(['1986-03-13', '1986-03-14', '1986-03-17', ..., '2020-01-03',
'2020-01-06', '2020-01-07'], dtype=object)`

```
In [14]: data['Date'].nunique()
```

```
Out[14]: 8525
```

```
In [17]: data['Date'].head(2)
```

```
Out[17]: 0    1986-03-13  
         1    1986-03-14  
         Name: Date, dtype: object
```

9. Checking null values presence:

```
In [20]: data.isnull().sum()
```

```
Out[20]: Date          0  
         Open          0  
         High          0  
         Low           0  
         Close         0  
         Adj Close     0  
         Volume        0  
         dtype: int64
```

10. Handling Missing Values:

```
In [22]: data.dropna(inplace=True)
```

```
In [23]: # Print the first 5 rows of the dataframe  
         print(data.head())
```

	Date	Open	High	Low	Close	Adj Close	Volume
0	1986-03-13	0.088542	0.101563	0.088542	0.097222	0.062549	1031788800
1	1986-03-14	0.097222	0.102431	0.097222	0.100694	0.064783	308160000
2	1986-03-17	0.100694	0.103299	0.100694	0.102431	0.065899	133171200
3	1986-03-18	0.102431	0.103299	0.098958	0.099826	0.064224	67766400
4	1986-03-19	0.099826	0.100694	0.097222	0.098090	0.063107	47894400

In [26]: data

Out[26]:

	Date	Open	High	Low	Close	Adj Close	Volume
0	1986-03-13	0.088542	0.101563	0.088542	0.097222	0.062549	1031788800
1	1986-03-14	0.097222	0.102431	0.097222	0.100694	0.064783	308160000
2	1986-03-17	0.100694	0.103299	0.100694	0.102431	0.065899	133171200
3	1986-03-18	0.102431	0.103299	0.098958	0.099826	0.064224	67766400
4	1986-03-19	0.099826	0.100694	0.097222	0.098090	0.063107	47894400
...
8520	2019-12-31	156.770004	157.770004	156.449997	157.699997	157.699997	18369400
8521	2020-01-02	158.779999	160.729996	158.330002	160.619995	160.619995	22622100
8522	2020-01-03	158.320007	159.949997	158.059998	158.619995	158.619995	21116200
8523	2020-01-06	157.080002	159.100006	156.509995	159.029999	159.029999	20813700
8524	2020-01-07	159.320007	159.669998	157.330002	157.580002	157.580002	18017762

8525 rows × 7 columns

11.Split the dataset into features (X) and target (Y):

```
In [ ]: X = df.drop('Date', axis=1)
        Y = df['Open']
```

12.Print the shape of the training and testing sets:

```
In [43]: print('X_train shape:', X_train.shape)
        print('X_test shape:', X_test.shape)
        print('Y_train shape:', Y_train.shape)
        print('Y_test shape:', Y_test.shape)
```

```
X_train shape: (6820, 6)
X_test shape: (1705, 6)
Y_train shape: (6820,)
Y_test shape: (1705,)
```

13.Feature Scaling.(import StandardScaler):

```
In [44]: import pandas as pd
        from sklearn.preprocessing import StandardScaler
```

```
In [45]: #Select the columns you want to scale  
cols_to_scale = ['Open', 'High', 'Low', 'Close', 'Adj Close', 'Volume']
```

```
In [46]: #Create a Standard Scalar object  
scaler = StandardScaler()
```

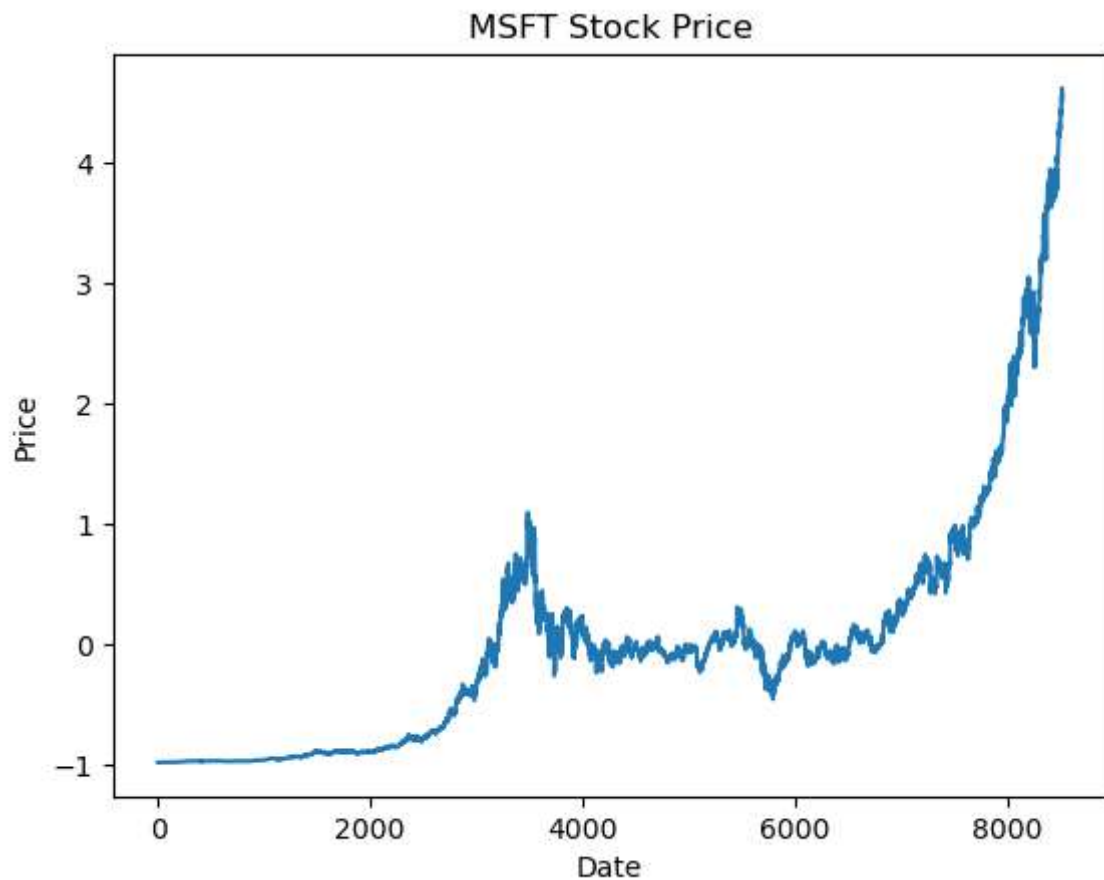
```
In [47]: #Fit the scalar to the selected coloumn  
df[cols_to_scale] = scaler.fit_transform(df[cols_to_scale])
```

```
In [48]: print(df.head())
```

	Date	Open	High	Low	Close	Adj Close	Volume
0	1986-03-13	-0.982764	-0.984942	-0.981026	-0.982615	-0.828391	24.963577
1	1986-03-14	-0.982461	-0.984912	-0.980720	-0.982494	-0.828312	6.366058
2	1986-03-17	-0.982340	-0.984882	-0.980598	-0.982433	-0.828272	1.868783
3	1986-03-18	-0.982279	-0.984882	-0.980659	-0.982524	-0.828331	0.187856
4	1986-03-19	-0.982370	-0.984972	-0.980720	-0.982585	-0.828371	-0.322861

14. Finding relationship each column using heatmap:

```
In [49]: # Plot the 'Close' column  
plt.plot(df['Close'])  
plt.title('MSFT Stock Price')  
plt.xlabel('Date')  
plt.ylabel('Price')  
plt.show()
```



```
In [52]: # Load the dataset into a pandas dataframe:

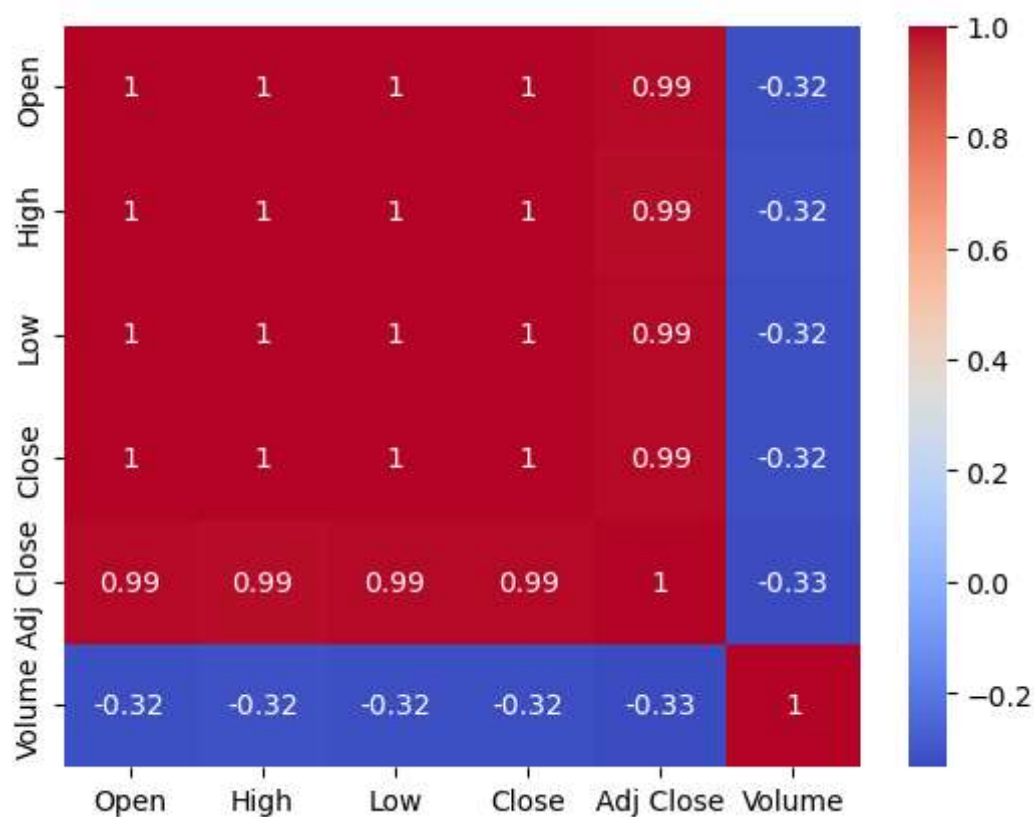
df = pd.read_csv('MSFT.csv', index_col=0, parse_dates=True)

# Create a correlation matrix:

corr_matrix = df.corr()

# Create a heatmap using seaborn:

sns.heatmap(corr_matrix, annot=True, cmap='coolwarm')
plt.show()
```



```
In [ ]:
```