

```
In [133.]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sb
from sklearn import preprocessing
from sklearn.model_selection import train_test_split
```

```
In [138.]: data = pd.read_csv('C:/Users/Lenovo/Downloads/new.csv')
data.drop('date',axis=1,inplace=True)
data.drop('symbol',axis=1,inplace=True)
print(data.head(10))
data.head()
```

	open	close	low	high	volume
0	123.430000	125.839996	122.309998	126.250000	2163600
1	125.239998	119.980003	119.940002	125.540001	2386400
2	116.379997	114.949997	114.930000	119.739998	2489500
3	115.480003	116.620003	113.500000	117.440002	2006300
4	117.010002	114.970001	114.089996	117.330002	1408600
5	115.510002	115.550003	114.500000	116.059998	1098000
6	116.459999	112.849998	112.589996	117.070000	949600
7	113.510002	114.379997	110.050003	115.029999	785300
8	113.330002	112.529999	111.919998	114.879997	1093700
9	113.660004	110.379997	109.870003	115.870003	1523500

```
Out[138]:
```

	open	close	low	high	volume
0	123.430000	125.839996	122.309998	126.250000	2163600
1	125.239998	119.980003	119.940002	125.540001	2386400
2	116.379997	114.949997	114.930000	119.739998	2489500
3	115.480003	116.620003	113.500000	117.440002	2006300
4	117.010002	114.970001	114.089996	117.330002	1408600

```
In [123.]: print(df.describe())
```

	open	close	low	high	volume
count	0.0	0.0	0.0	0.0	0.0
mean	NaN	NaN	NaN	NaN	NaN
std	NaN	NaN	NaN	NaN	NaN
min	NaN	NaN	NaN	NaN	NaN
25%	NaN	NaN	NaN	NaN	NaN
50%	NaN	NaN	NaN	NaN	NaN
75%	NaN	NaN	NaN	NaN	NaN
max	NaN	NaN	NaN	NaN	NaN

```
In [124.]: print(data.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3482 entries, 0 to 3481
Data columns (total 5 columns):
 #   Column  Non-Null Count  Dtype
---  -
 0   open    3482 non-null      float64
 1   close   3482 non-null      float64
 2   low     3482 non-null      float64
 3   high    3482 non-null      float64
 4   volume  3482 non-null      int64
dtypes: float64(4), int64(1)
memory usage: 136.1 KB
None
```

```
In [98]: data1 = pd.read_csv('C:/Users/Lenovo/Downloads/new1.csv')
print(data1.head(20))
```

Unnamed: 0	date	symbol	open	close	low	high	volume
0	05-01-2016 00:00	WLTW	123.430000	125.839996	122.309998	126.250000	2163600
1	06-01-2016 00:00	WLTW	125.239998	119.980003	119.940002	125.540001	2386400
2	07-01-2016 00:00	WLTW	116.379997	114.949997	114.930000	119.739998	2489500
3	08-01-2016 00:00	WLTW	115.480003	116.620003	113.500000	117.440002	2006300
4	11-01-2016 00:00	WLTW	117.010002	114.970001	114.089996	117.330002	1408600
5	12-01-2016 00:00	WLTW	115.510002	115.550003	114.500000	116.059998	1098000
6	13-01-2016 00:00	WLTW	116.459999	112.849998	112.589996	117.070000	949600
7	14-01-2016 00:00	WLTW	113.510002	114.379997	110.050003	115.029999	785300
8	15-01-2016 00:00	WLTW	113.330002	112.529999	111.919998	114.879997	1093700
9	19-01-2016 00:00	WLTW	113.660004	110.379997	109.870003	115.870003	1523500
10	20-01-2016 00:00	WLTW	109.059998	109.300003	108.320000	109.730003	110.000000
11	21-01-2016 00:00	WLTW	109.730003	110.000000	108.320000	110.000000	108.320000
12	22-01-2016 00:00	WLTW	111.879997	111.949997	110.190002	112.685001	112.685001
13	25-01-2016 00:00	WLTW	111.320000	110.120003	110.000000	110.000000	107.300000
14	26-01-2016 00:00	WLTW	110.419998	111.000000	107.300000	110.000000	107.300000
15	27-01-2016 00:00	WLTW	110.769997	110.709999	109.019997	110.709999	109.019997
16	28-01-2016 00:00	WLTW	110.900002	NaN	109.900002	110.900002	109.900002
17	29-01-2016 00:00	WLTW	113.349998	114.470001	111.669998	114.470001	111.669998
18	01-02-2016 00:00	WLTW	114.000000	114.500000	112.900002	114.500000	112.900002
19	02-02-2016 00:00	WLTW	113.250000	110.559998	109.750000	110.559998	109.750000

high	volume
0	126.250000
1	125.540001
2	119.739998
3	117.440002
4	NaN
5	116.059998
6	117.070000
7	115.029999
8	114.879997
9	115.870003
10	NaN
11	110.580002
12	112.949997
13	114.629997
14	111.400002
15	112.570000
16	112.970001
17	114.589996
18	114.849998
19	NaN

```
In [97]: print(data1.describe())
```

Unnamed: 0	open	close	low	high	volume
count	50.000000	50.000000	49.000000	50.000000	47.000000
mean	24.500000	113.906800	113.966122	112.023400	115.461915
std	14.57738	3.892047	3.907816	3.857870	3.718608
min	0.000000	105.629997	107.129997	104.110001	109.260002
25%	12.250000	111.207498	111.000000	109.657499	112.685001
50%	24.500000	113.364998	113.320000	111.180000	114.879997
75%	36.750000	116.439999	116.620003	114.822500	117.405002
max	49.000000	125.239998	125.839996	122.309998	126.250000

high	volume
count	5.000000e+01
mean	9.988980e+05
std	5.144256e+05
min	4.112000e+05
25%	6.806250e+05
50%	8.962000e+05
75%	1.178725e+06
max	2.489500e+06

```
In [98]: print(data1.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 50 entries, 0 to 49
Data columns (total 8 columns):
 #   Column  Non-Null Count  Dtype
---  -
 0   Unnamed: 0    50 non-null      int64
 1   date          50 non-null      object
 2   symbol        50 non-null      object
 3   open          50 non-null      float64
 4   close         49 non-null      float64
 5   low           50 non-null      float64
 6   high          47 non-null      float64
 7   volume        50 non-null      int64
dtypes: float64(4), int64(2), object(2)
memory usage: 3.2+ KB
None
```

```
In [126.]: data1.dropna(inplace=True)
data1.fillna(-9999,inplace=True)
data1['high'] = data1['high'].fillna(data1['high'].mean())
print(data1.head())
```

Unnamed: 0	date	symbol	open	close	low	high	volume
0	05-01-2016 00:00	WLTW	123.430000	125.839996	122.309998	126.250000	2163600
1	06-01-2016 00:00	WLTW	125.239998	119.980003	119.940002	125.540001	2386400
2	07-01-2016 00:00	WLTW	116.379997	114.949997	114.930000	119.739998	2489500
3	08-01-2016 00:00	WLTW	115.480003	116.620003	113.500000	117.440002	2006300
5	12-01-2016 00:00	WLTW	115.510002	115.550003	114.500000	116.059998	1098000

high	volume
0	126.250000
1	125.540001
2	119.739998
3	117.440002
5	116.059998

```
In [91]: #fill the value by fillna()
data1.dropna(inplace=True)
data1.fillna(9999,inplace=True)
print(data1.head(10))
```

Unnamed: 0	date	symbol	open	close	low	high	volume
0	05-01-2016 00:00	WLTW	123.430000	125.839996	122.309998	126.250000	2163600
1	06-01-2016 00:00	WLTW	125.239998	119.980003	119.940002	125.540001	2386400
2	07-01-2016 00:00	WLTW	116.379997	114.949997	114.930000	119.739998	2489500
3	08-01-2016 00:00	WLTW	115.480003	116.620003	113.500000	117.440002	2006300
5	12-01-2016 00:00	WLTW	115.510002	115.550003	114.500000	116.059998	1098000
6	13-01-2016 00:00	WLTW	116.459999	112.849998	112.589996	117.070000	949600
7	14-01-2016 00:00	WLTW	113.510002	114.379997	110.050003	115.029999	785300
8	15-01-2016 00:00	WLTW	113.330002	112.529999	111.919998	114.879997	1093700
9	19-01-2016 00:00	WLTW	113.660004	110.379997	109.870003	115.870003	1523500
11	21-01-2016 00:00	WLTW	109.730003	110.000000	108.320000	110.000000	108.320000

high	volume
0	126.250000
1	125.540001
2	119.739998
3	117.440002
5	116.059998
6	117.070000
7	115.029999
8	114.879997
9	115.870003
11	110.580002

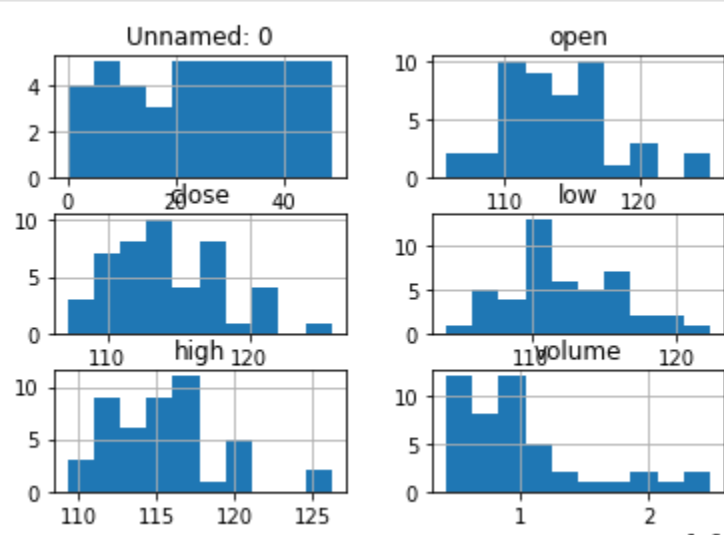
```
In [100.]: #fill the value with mean
data1['high'] = data1['high'].fillna(data1['high'].mean())
print(data1.head())
```

Unnamed: 0	date	symbol	open	close	low	high	volume
0	05-01-2016 00:00	WLTW	123.430000	125.839996	122.309998	126.250000	2163600
1	06-01-2016 00:00	WLTW	125.239998	119.980003	119.940002	125.540001	2386400
2	07-01-2016 00:00	WLTW	116.379997	114.949997	114.930000	119.739998	2489500
3	08-01-2016 00:00	WLTW	115.480003	116.620003	113.500000	117.440002	2006300
5	12-01-2016 00:00	WLTW	115.510002	115.550003	114.500000	116.059998	1098000

high	volume
0	126.250000
1	125.540001
2	119.739998
3	117.440002
5	116.059998

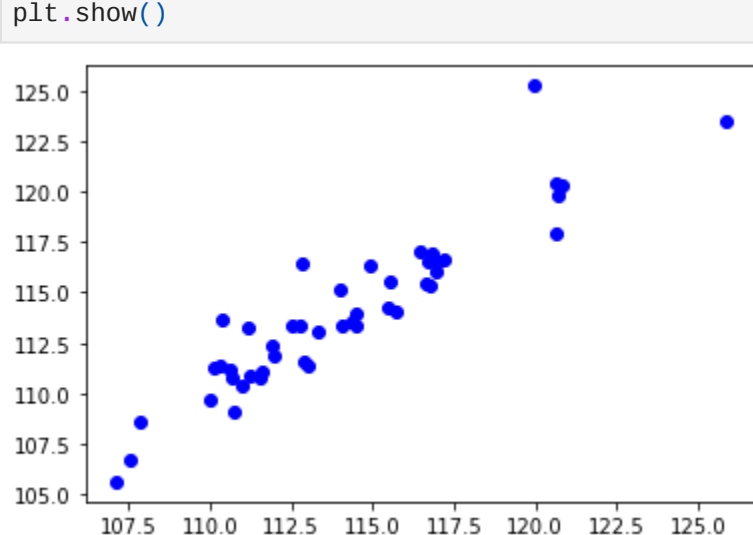
```
In [ ]: 
```

```
In [110.]: data.hist()
plt.show()
```



The figure displays five histograms arranged in a 2x3 grid (with the last cell empty). The histograms represent the distribution of different stock market variables: 'Unnamed: 0' (index), 'open' (opening price), 'close' (closing price), 'high' (daily high), and 'volume' (trading volume). Each histogram shows the frequency of values on the x-axis and the count on the y-axis. The 'open' and 'close' plots have a y-axis scale up to 10, while the others have a scale up to 4. The 'volume' plot has a y-axis scale up to 10 and an x-axis scale up to 1e6.

```
In [111.]: plt.scatter(data.close,data.open,color="blue")
plt.show()
```



The scatter plot shows the relationship between the opening price ('open') on the x-axis and the closing price ('close') on the y-axis. The data points are colored blue. The x-axis ranges from approximately 107.5 to 125.0, and the y-axis ranges from approximately 105.0 to 125.0. The plot shows a positive correlation between the two variables, with most points clustered between 110 and 120 on both axes.

```
In [115.]: #to find the correlation value for each column
cor =data.corr()
print(cor['open'].sort_values(ascending=True))
```

```
Unnamed: 0    0.070603
volume        0.161164
close         0.927620
low           0.949430
high          0.966265
open          1.000000
Name: open, dtype: float64
```

TO PREDICT THE STOCK MARKET AFTER 10 DAYS

```
In [120.]: forecast =10
data['label']=data['open'].shift(-forecast)
print(data.head())
```

Unnamed: 0	open	close	low	high	volume	label
0	123.430000	125.839996	122.309998	126.250000	2163600	109.059998
1	125.239998	119.980003	119.940002	125.540001	2386400	109.730003
2	116.379997	114.949997	114.930000	119.739998	2489500	111.879997
3	115.480003	116.620003	113.500000	117.440002	2006300	111.320000
5	115.510002	115.550003	114.500000	116.059998	1098000	110.419998

label	
0	111.879997
1	111.320000
2	110.419998
3	110.769997
5	113.349998

```
In [127.]: data.dropna(inplace=True)
print(data.head())
```

	open	close	low	high	volume
0	123.430000	125.839996	122.309998	126.250000	2163600
1	125.239998	119.980003	119.940002	125.540001	2386400
2	116.379997	114.949997	114.930000	119.739998	2489500
3	115.480003	116.620003	113.500000	117.440002	2006300
4	117.010002	114.970001	114.089996	117.330002	1408600

```
In [137.]: forecast =10
data['label']=data['open'].shift(-forecast)
print(data.head())
x = np.array(data.drop(['label'],1))
y = np.array(data['label'])
x =preprocessing.scale(x)
#print(x)
#print(y)
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.15,random_state=4)
print(len(x_train))
print(len(x_test))
```

	open	close	low	high	volume	label
0	123.430000	125.839996	122.309998	126.250000	2163600	109.059998
1	125.239998	119.980003	119.940002	125.540001	2386400	109.730003
2	116.379997	114.949997	114.930000	119.739998	2489500	111.879997
3	115.480003	116.620003	113.500000	117.440002	2006300	111.320000
4	117.010002	114.970001	114.089996	117.330002	1408600	110.419998
2959						
523						

C:\Users\Lenovo\AppData\Local\Temp\ipykernel_64300\1823541486.py:4: FutureWarning: In a future version of pandas all arguments of DataFrame.drop except for the argument 'labels' will be keyword-only.

```
x = np.array(data.drop(['label'],1))
```

```
In [ ]: 
```