```
import numpy as np
import pandas as pd
from sklearn import metrics
%matplotlib inline
import matplotlib.pyplot as plt
import seaborn as sns
```

data=pd.read\_csv('/content/NSE-TATAGLOBAL11[1].csv')

data.head()

	Date	Open	High	Low	Last	Close	Total Trade Quantity	Turnover (Lacs)
0	2018-10-08	208.00	222.25	206.85	216.00	215.15	4642146.0	10062.83
1	2018-10-05	217.00	218.60	205.90	210.25	209.20	3519515.0	7407.06
2	2018-10-04	223.50	227.80	216.15	217.25	218.20	1728786.0	3815.79
3	2018-10-03	230.00	237.50	225.75	226.45	227.60	1708590.0	3960.27
4	2018-10-01	234.55	234.60	221.05	230.30	230.90	1534749.0	3486.05

data['Date']=pd.to\_datetime(data.Date)

data.shape

(1235, 8)

data.isnull().sum()

Date 0
Open 0
High 0

```
Low 0
Last 0
Close 0
Total Trade Quantity 0
Turnover (Lacs) 0
dtype: int64
```

## data.isna().any()

```
False
Date
                       False
0pen
High
                       False
Low
                       False
Last
                       False
Close
                       False
Total Trade Quantity
                       False
Turnover (Lacs)
                       False
dtype: bool
```

## data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1235 entries, 0 to 1234
Data columns (total 8 columns):
```

#	Column	Non-Null Count	Dtype
0	Date	1235 non-null	<pre>datetime64[ns]</pre>
1	Open	1235 non-null	float64
2	High	1235 non-null	float64
3	Low	1235 non-null	float64
4	Last	1235 non-null	float64
5	Close	1235 non-null	float64
6	Total Trade Quantity	1235 non-null	float64
7	Turnover (Lacs)	1235 non-null	float64
<pre>dtypes: datetime64[ns](1),</pre>		float64(7)	
memo	ry usage: 77.3 KB		

data.describe()

	<b>O</b> pen	High	Low	Last	Close	Total Trade Quantity	Turnover (Lacs)
count	1235.000000	1235.000000	1235.000000	1235.000000	1235.000000	1.235000e+03	1235.000000
mean	168.954858	171.429069	166.402308	168.736356	168.731053	2.604151e+06	4843.166502
std	51.499145	52.436761	50.542919	51.587384	51.544928	2.277028e+06	5348.919832
min	103.000000	104.600000	100.000000	102.600000	102.650000	1.001800e+05	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
75%	169.000000	172.325000	166.700000	169.100000	169.500000	3.095788e+06	5852.600000
max	327.700000	328.750000	321.650000	325.950000	325.750000	2.919102e+07	55755.080000

print(len(data))

1235

data['Open'].plot(figsize=(15,5))

```
<Axes: >
      300
      250
x=data[['Open','High','Low','Last']]
y=data['Close']
from sklearn.model selection import train test split
x_train , x_test , y_train , y_test=train_test_split(x,y,random_state=0)
x_train.shape
     (926, 4)
x_test.shape
     (309, 4)
from sklearn.linear model import LinearRegression
from sklearn.metrics import confusion matrix, accuracy score
regressor=LinearRegression()
regressor.fit(x_train,y_train)
      ▼ LinearRegression
     LinearRegression()
print(regressor.coef_)
```

1083

153.45

[-0.06299618 0.09073469 0.08672877 0.88451674]

```
print(regressor.intercept_)
     0.1401341854183329
predicted=regressor.predict(x test)
print(x_test)
                    High
            0pen
                            Low
                                   Last
    1083 152.95 154.85 151.05 153.35
     18
          221.00 224.50 219.10 223.15
         156.50 157.00 151.85 153.00
     1099
          132.95 133.90 132.30 133.20
     818
          314.65 319.20 312.20 317.45
     184
          154.40 154.70 151.55 153.00
     907
     801
          134.90 137.80 134.25 136.80
    1089 145.80 146.70 144.15 144.65
          125.00 126.05 123.80 124.10
     435
     833
          149.40 150.30 143.00 144.60
    [309 rows x 4 columns]
predicted.shape
    (309,)
dframe=pd.DataFrame({'Actual Price':y test,'PredictedPrice':predicted})
print(dframe)
          Actual Price PredictedPrice
```

18	222.95	222.970100
1099	152.95	153.027403
818	132.60	133.206013
184	317.60	317.147460
• • •	• • •	
907	152.40	152.924987
801	136.80	136.790417
1089	144.55	144.713369
435	124.45	124.208268
833	144.90	144.669263

[309 rows x 2 columns]

dframe.head(25)

	Actual Price	PredictedPrice
1083	153.45	153.296158
18	222.95	222.970100
1099	152.95	153.027403
818	132.60	133.206013
184	317.60	317.147460
608	122.40	122.511089
1213	160.35	161.661581
717	128.00	127.727919
1137	141.20	141.178114
1009	162.20	163.373307
1068	174.85	173.933181
624	110 25	110 202077

from sklearn.metrics import confusion\_matrix, accuracy\_score

regressor.score(x\_test,y\_test)

## 0.9999281448580609

1167EE01E11 CO.011

import math

10/1 202.75 202.700127

print('Mean Absolute Error:',metrics.mean\_absolute\_error(y\_test,predicted))

Mean Absolute Error: 0.29103838136150123

print('Mean Squared Error:',metrics.mean\_squared\_error(y\_test,predicted))

Mean Squared Error: 0.1608032139897181

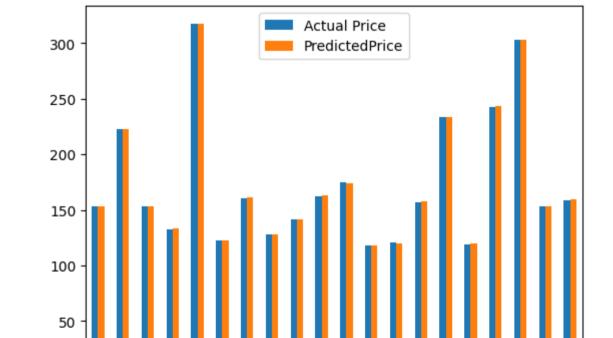
```
print('Root Mean Squared Error:',math.sqrt(metrics.mean_squared_error(y_test,predicted)))
```

Root Mean Squared Error: 0.4010027605761812

graph=dframe.head(20)

graph.plot(kind='bar')

<Axes: >



1009 1068 621 761

717

608

1099

```
#Calculate the daily returns
returns=data['Close'].pct_change()
returns
     0
                  NaN
     1
            -0.027655
     2
             0.043021
     3
             0.043080
     4
             0.014499
     1230
             0.008858
     1231
             0.003763
            0.000625
     1232
            -0.028723
     1233
     1234
             0.001607
     Name: Close, Length: 1235, dtype: float64
#Calculate the moving averages
ma 10=data['Close'].rolling(window=10).mean()
ma 50=data['Close'].rolling(window=50).mean()
ma_10
ma_50
     0
                 NaN
     1
                 NaN
     2
                 NaN
     3
                 NaN
                 NaN
     1230
             153.301
     1231
             153.357
     1232
            153.466
     1233
            153.515
     1234
             153.681
     Name: Close, Length: 1235, dtype: float64
#Calculate the standard deviation
std=data['Close'].rolling(window=10).std()
```

sta						
	0		NaN			
	1		NaN			
	2		NaN			
	3		NaN			
	4		NaN			
			•			
	1230	2.558	8586			
	1231	2.53	5553			
	1232	2.392	2053			
	1233	2.373	3587			
	1234	2.703	1975			
	Name:	Close,	Length:	1235,	dtype:	float64

✓ 0s completed at 1:09 PM

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