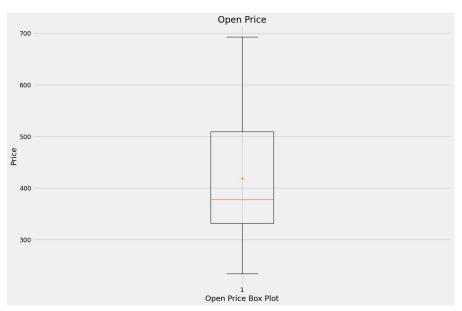
```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
import statistics
df = pd.read_csv("/content/NFLX1.csv")
df1 = df.copy()
df.head()
                                                                                     \blacksquare
             Date
                        0pen
                                   High
                                               Low
                                                        Close Adj Close
                                                                           Volume
            05-02-
                                                                                     ıl.
     0
                  262.000000 267.899994 250.029999 254.259995 254.259995 11896100
             2018
            06-02-
     1
                  247.699997 266.700012 245.000000 265.720001 265.720001 12595800
             2018
            07-02-
     2
                  266.579987 272.450012 264.329987 264.559998 264.559998
                                                                          8981500
             2018
            08-02-
 Next steps:
             Generate code with df

    View recommended plots

df.shape
     (1009, 7)
df.dtypes
     Date
                  object
     0pen
                 float64
     High
                 float64
                 float64
     Low
                 float64
     Close
     Adj Close
                 float64
     Volume
                   int64
     dtype: object
df.duplicated().sum()
     0
df.isna().sum()
     Date
     0pen
                 0
     High
                 0
     Low
                 0
     Close
                 0
     Adj Close
                 0
     Volume
                 0
     dtype: int64
df.nunique()
     Date
                 1009
     0pen
                  976
     High
                  983
                  989
     Low
     Close
     Adj Close
                  988
     Volume
                 1005
     dtype: int64
plt.style.use('fivethirtyeight')
plt.subplots(figsize=(15, 10))
plt.title("Open Price")
plt.boxplot(df['Open'], showmeans=True)
```

plt.xlabel("Open Price Box Plot")

```
plt.ylabel("Price")
plt.show()
```



```
print("Mean price is :", statistics.mean(df['Open']))
print("Median price is :", statistics.median(df['Open']))

    Mean price is : 419.05967286223984
    Median price is : 377.769989

plt.subplots(figsize=(25, 8))
plt.title("Open Price vs Close Price")
plt.plot(df['Open'], color='red', linestyle='solid', label = 'Open Price')
plt.plot(df['Close'], color='green', linestyle='dashed', label = 'Close Price')
plt.xlabel("Date")
plt.ylabel("Open vs Close Price")
plt.legend(loc="upper left")
plt.show()
```



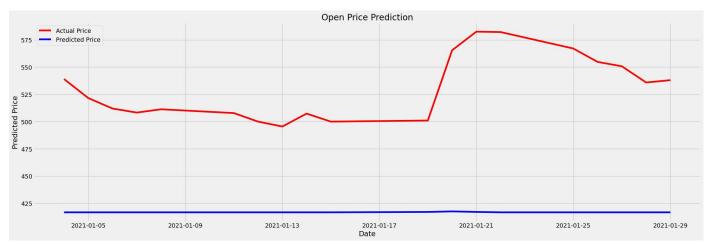
```
from sklearn.preprocessing import StandardScaler
df['Date']=pd.to_datetime(df['Date'],format='%d-%m-%Y')
# set date to index
df = df.set_index('Date')
train = df.loc['2018-02-05':'2018-12-31']
test = df.loc['2021-01-01':'2021-01-31']
X_train = train.drop(columns = ['Open'])
y_train = train['Open']
# split testing data
X_test = test.drop(columns = ['Open'])
y_test = test['Open']
from sklearn.ensemble import RandomForestRegressor
rf = RandomForestRegressor(max_depth=20, random_state = 42, n_estimators=150)
rf.fit(X_train, y_train)
                               RandomForestRegressor
     RandomForestRegressor(max_depth=20, n_estimators=150, random_state=42)
rf_train_score = rf.score(X_train, y_train)
rf_test_score = rf.score(X_test, y_test)
print(rf_train_score)
print(rf_test_score)
     0.9979090389971715
     -15.961320806912266
pred = rf.predict(X_test)
train_pred = rf.predict(X_train)
prediction_df = X_test.copy()
prediction_df['Open'] = y_test
prediction_df['Predicted Price'] = pred
prediction_df.head()
```

```
Predicted
                High
                             Low
                                      Close Adj Close
                                                         Volume
                                                                       0pen
                                                                                 Price
     Date
    2021-
           540.799988 515.090027 522.859985 522.859985 4444400 539.000000 416.850669
    01-04
    2021-
           526.780029 515.890015 520.799988 520.799988 3133900 521.549988 416.850669
    01-05
    2021-
           513.099976 499.500000 500.489990 500.489990 5346200 511.970001 416.850669
    01-06
Next steps:
            Generate code with prediction_df
                                               View recommended plots
```

```
plt.subplots(figsize=(25, 8))
plt.title("Open Price Prediction")
#plt.plot(prediction_df['Open'], color='red', linestyle='solid')
plt.plot(df['Open'], color='red', linestyle='solid', label = 'Actual Price')
plt.plot(prediction_df['Predicted Price'], color='blue', linestyle='solid', label = 'Predicted Price')
plt.xlabel("Date")
plt.ylabel("Predicted Price")
plt.legend(loc="upper left")
plt.show()
```



```
plt.subplots(figsize=(25, 8))
plt.title("Open Price Prediction")
plt.plot(prediction_df['Open'], color='red', linestyle='solid', label = 'Actual Price')
plt.plot(prediction_df['Predicted Price'], color='blue', linestyle='solid', label = 'Predicted Price')
plt.xlabel("Date")
plt.ylabel("Predicted Price")
plt.legend(loc="upper left")
plt.show()
```



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```
print("Mean Absolute Error:",round(metrics.mean_absolute_error(y_test, pred), 4))
print("Mean Squared Error:", round(metrics.mean_squared_error(y_test, pred), 4))
print("Root Mean Squared Error:", round(np.sqrt(metrics.mean_squared_error(y_test, pred)), 4))
print("(R^2) Score:", round(metrics.r2_score(y_test, pred), 4))
print(f'Train Score : {rf.score(X_train, y_train) * 100:.2f}% and Test Score : {rf.score(X_test, y_test) * 100:.2f}% using Random Tree.')
errors = abs(pred - y_test)
mape = 100 * (errors / y_test)
accuracy = 100 - np.mean(mape)
print('Accuracy:', round(accuracy, 2), '%.')

Mean Absolute Error: 113.5985
Mean Squared Error: 13708.832
Root Mean Squared Error: 117.0847
(R^2) Score: -15.9613
Train Score : 99.79% and Test Score : -1596.13% using Random Tree.
Accuracy: 78.81 %.
```