

Netflix

Stock Price Prediction

Netflix, Inc. is an American media company based in Los Gatos, California. Founded in 1997 by Reed Hastings and Marc Randolph in Scotts Valley, California

This dataset contains the stock price values of Netflix from 2018 to 2022.

Feature Description:

Date - Date of open market
 Open - Price when market opens
 High - Highest price during open period
 Low - Lowest price during open period
 Close - Price when market is closed
 Adj Close - Closing price after adjustments
 Volume - number of shares traded

Import data

```
In [3]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
import statistics
import plotly.offline as pyo
import plotly.graph_objs as go
```

Upload Dataset

```
In [5]: df=pd.read_csv("NFLX.csv")
```

```
In [7]: df1=df.copy()
```

```
In [10]: df1.head()
```

```
Out[10]:
```

	Date	Open	High	Low	Close	Adj Close	Volume
0	2018-02-05	262.000000	267.899994	250.029999	254.259995	254.259995	11896100
1	2018-02-06	247.699997	266.700012	245.000000	265.720001	265.720001	12595800
2	2018-02-07	266.579987	272.450012	264.329987	264.559998	264.559998	8981500
3	2018-02-08	267.079987	267.619995	250.000000	250.100006	250.100006	9306700
4	2018-02-09	253.850006	255.800003	236.110001	249.470001	249.470001	16906900

Checking the dataset in rows and columns

```
In [11]: df.shape
```

```
Out[11]: (1009, 7)
```

Checking datatype

```
In [13]: df.dtypes
```

```
Out[13]: Date          object
Open          float64
High          float64
Low           float64
Close         float64
Adj Close     float64
Volume        int64
dtype: object
```

Duplicate data check

```
In [16]: df.duplicated().sum()
```

```
Out[16]: 0
```

Null Values

```
In [18]: df.isnull().sum()
```

```
Out[18]: Date          0
Open          0
High          0
Low           0
Close         0
Adj Close     0
Volume        0
dtype: int64
```

Unique values

```
In [19]: df.nunique()
```

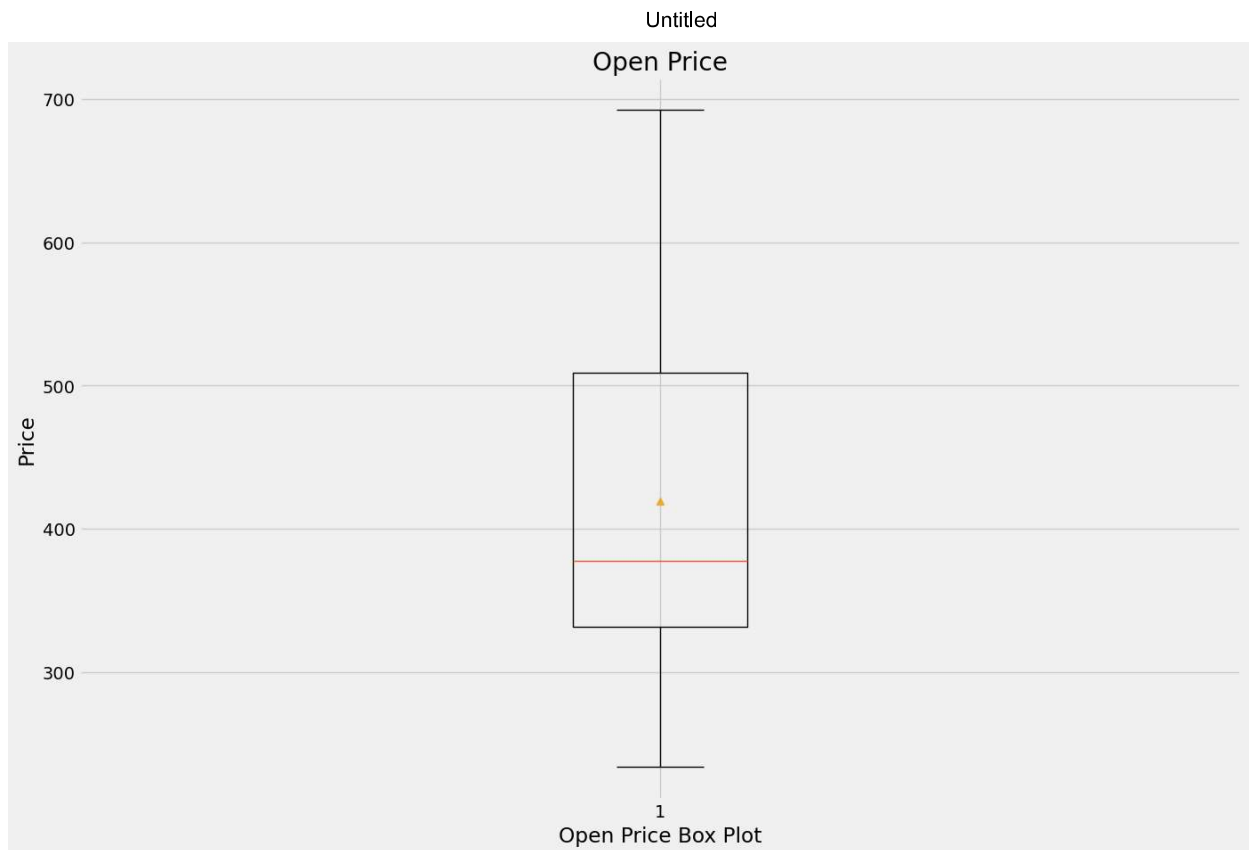
```
Out[19]: Date          1009
Open          976
High          983
Low           989
Close         988
Adj Close     988
Volume        1005
dtype: int64
```

EDA(Exploratory Data Analysis)

Visualization

```
In [21]: plt.style.use("fivethirtyeight")
```

```
In [24]: 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()
```



```
In [26]: print("Mean price is :",statistics.mean(df["Open"]))
print("Mean price is :",statistics.median(df["Open"]))
```

Mean price is : 419.05967286223984

Mean price is : 377.769989

```
In [27]: 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()
```



We see a price drop in January of 2022 after the 4th quarter

Possible Causes

Price raise 13.99 to 15.49 per month. Streaming competition was rising eating growth in fourth-quarter earnings

Positive Outlook

Being viewed as a value stock compared to a growth stock Price increases should keep revenue growing Solutions

Aim to convert password sharing users into subscribers Add a lower-cost ad-supported tier

Solutions Aim to convert password sharing users into subscribers Add a lower-cost ad-supported tier

Prepare Data

```
In [28]: from sklearn.preprocessing import StandardScaler
```

```
In [29]: # change object to datetime
df['Date'] = pd.to_datetime(df['Date'], format='%Y-%m-%d')

# set date to index
df = df.set_index('Date')
```

```
In [30]: # split data for prediciting january 2022
train = df.loc['2018-02-05':'2021-12-31']
test = df.loc['2022-01-01':'2022-01-31']
```

```
In [31]: # split training data
X_train = train.drop(columns = ['Open'])
y_train = train['Open']
# split testing data
X_test = test.drop(columns = ['Open'])
y_test = test['Open']
```

Random Forest Model

```
In [32]: from sklearn.ensemble import RandomForestRegressor
```

```
In [33]: # build model
rf = RandomForestRegressor(max_depth=20, random_state = 42, n_estimators=150)
rf.fit(X_train, y_train)
```

```
Out[33]: ▼ RandomForestRegressor
RandomForestRegressor(max_depth=20, n_estimators=150, random_state=42)
```

```
In [35]: 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)
```

```
0.9996830741916087
0.9918135946459509
```

```
In [36]: pred = rf.predict(X_test)
train_pred = rf.predict(X_train)
```

```
In [37]: prediction_df = X_test.copy()
prediction_df['Open'] = y_test
prediction_df['Predicted Price'] = pred
prediction_df.head()
```

```
Out[37]:
```

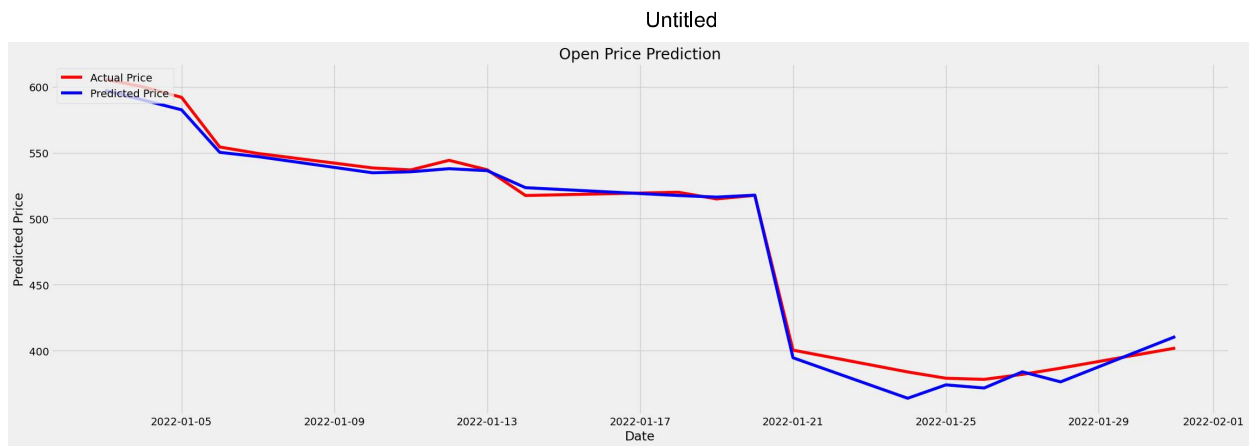
	High	Low	Close	Adj Close	Volume	Open	Predicted Price
Date							
2022-01-03	609.989990	590.559998	597.369995	597.369995	3067500	605.609985	597.302736
2022-01-04	600.409973	581.599976	591.150024	591.150024	4393100	599.909973	589.919729
2022-01-05	592.840027	566.880005	567.520020	567.520020	4148700	592.000000	582.541523
2022-01-06	563.359985	542.010010	553.289978	553.289978	5711800	554.340027	550.311327
2022-01-07	553.429993	538.219971	541.059998	541.059998	3381700	549.460022	547.067863

Result

```
In [39]: 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()
```



```
In [40]: 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()
```



Model Evaluation

```
In [41]: from sklearn import metrics
```

```
In [42]: 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}%')
errors = abs(pred - y_test)
mape = 100 * (errors / y_test)
accuracy = 100 - np.mean(mape)
print('Accuracy:', round(accuracy, 2), '%.')
```

Mean Absolute Error: 5.725

Mean Squared Error: 53.3474

Root Mean Squared Error: 7.3039

(R²) Score: 0.9918

Train Score : 99.97% and Test Score : 99.18% using Random Tree.

Accuracy: 98.75 %.

2023 Company Outlook

Positives

Shares have gained 1.4% this year, outperforming the S&P 500, which is up 0.2%

Removing password sharing can allow a 2-8 billion growth in revenue. Ad tier plan has attracted more than 1 million users after just two months

Cash flow positive in 2022 2 billion in operating cash 1.6 billion in free cash flow

Precautions

The closure of Silicon Valley Bank may affect tech stocks

Fears around higher churn from enforcing password sharing

Gave some money to advertisers when number of viewers expected didn't materialize

```
In [ ]:
```