

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline

import plotly.graph_objs as go
from plotly.offline import plot

#for offline plotting
from plotly.offline import download_plotlyjs, init_notebook_mode, plot, iplot
init_notebook_mode(connected=True)
```

```
In [2]: tesla = pd.read_csv('tesla.csv')
tesla.head()
```

```
Out[2]:
```

	Date	Open	High	Low	Close	Adj Close	Volume
0	29-06-2010	19.000000	25.00	17.540001	23.889999	23.889999	18766300
1	30-06-2010	25.790001	30.42	23.299999	23.830000	23.830000	17187100
2	01-07-2010	25.000000	25.92	20.270000	21.959999	21.959999	8218800
3	02-07-2010	23.000000	23.10	18.709999	19.200001	19.200001	5139800
4	06-07-2010	20.000000	20.00	15.830000	16.110001	16.110001	6866900

```
In [3]: tesla.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2193 entries, 0 to 2192
Data columns (total 7 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Date        2193 non-null   object
1   Open        2193 non-null   float64
2   High        2193 non-null   float64
3   Low         2193 non-null   float64
4   Close       2193 non-null   float64
5   Adj Close   2193 non-null   float64
6   Volume      2193 non-null   int64
dtypes: float64(5), int64(1), object(1)
memory usage: 120.1+ KB
```

```
In [4]: tesla['Date'] = pd.to_datetime(tesla['Date'])
```

C:\Users\coder\AppData\Local\Temp\ipykernel_16284\3702129700.py:1: UserWarning:

Parsing dates in DD/MM/YYYY format when dayfirst=False (the default) was specified. This may lead to inconsistently parsed dates! Specify a format to ensure consistent parsing.

```
In [5]: print(f'Dataframe contains stock prices between {tesla.Date.min()} {tesla.Date.max()}')
print(f'Total days = {(tesla.Date.max() - tesla.Date.min()).days} days')
```

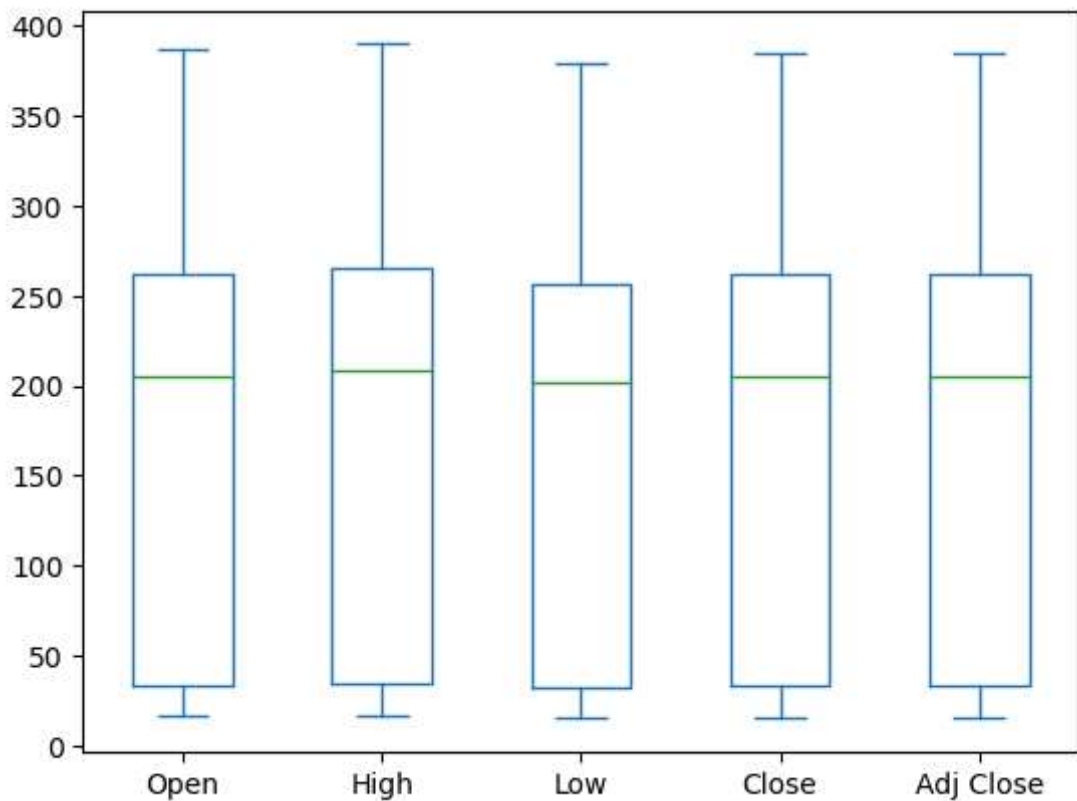
Dataframe contains stock prices between 2010-01-07 00:00:00 2019-12-03 00:00:00
Total days = 3617 days

```
In [6]: tesla.describe()
```

Out[6]:	Open	High	Low	Close	Adj Close	Volume
count	2193.000000	2193.000000	2193.000000	2193.000000	2193.000000	2.193000e+03
mean	175.652882	178.710262	172.412075	175.648555	175.648555	5.077449e+06
std	115.580903	117.370092	113.654794	115.580771	115.580771	4.545398e+06
min	16.139999	16.629999	14.980000	15.800000	15.800000	1.185000e+05
25%	33.110001	33.910000	32.459999	33.160000	33.160000	1.577800e+06
50%	204.990005	208.160004	201.669998	204.990005	204.990005	4.171700e+06
75%	262.000000	265.329987	256.209991	261.739990	261.739990	6.885600e+06
max	386.690002	389.609985	379.350006	385.000000	385.000000	3.716390e+07

In [7]: `tesla[['Open', 'High', 'Low', 'Close', 'Adj Close']].plot(kind='box')`

Out[7]: <Axes: >



```
In [8]: # Setting the layout for our plot
layout = go.Layout(
    title='Stock Prices of Tesla',
    xaxis=dict(
        title='Date',
        titlefont=dict(
            family='Courier New, monospace',
            size=18,
            color='#7f7f7f'
        )
    ),
    yaxis=dict(
        title='Price',
        titlefont=dict(
            family='Courier New, monospace',
            size=18,

```

```

        color='#7f7f7f'
    )
)
)

tesla_data = [{'x':tesla['Date'], 'y':tesla['Close']}]
plot = go.Figure(data=tesla_data, layout=layout)

```

```

In [9]: #plot(plot) #plotting offline
        iplot(plot)
        # Building the regression model
        from sklearn.model_selection import train_test_split

        #For preprocessing
        from sklearn.preprocessing import MinMaxScaler
        from sklearn.preprocessing import StandardScaler

        #For model evaluation
        from sklearn.metrics import mean_squared_error as mse
        from sklearn.metrics import r2_score
        #Split the data into train and test sets
        X = np.array(tesla.index).reshape(-1,1)
        Y = tesla['Close']
        X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.3, random_state=42)
        # Feature scaling
        scaler = StandardScaler().fit(X_train)
        from sklearn.linear_model import LinearRegression
        #Creating a linear model
        lm = LinearRegression()
        lm.fit(X_train, Y_train)
        LinearRegression()
        #Plot actual and predicted values for train dataset
        trace0 = go.Scatter(
            x = X_train.T[0],
            y = Y_train,
            mode = 'markers',
            name = 'Actual'
        )
        trace1 = go.Scatter(
            x = X_train.T[0],
            y = lm.predict(X_train).T,
            mode = 'lines',
            name = 'Predicted'
        )
        tesla_data = [trace0,trace1]
        layout.xaxis.title.text = 'Day'
        plot2 = go.Figure(data=tesla_data, layout=layout)
        iplot(plot2)
        #Calculate scores for model evaluation
        scores = f'''
        {'Metric'.ljust(10)}{'Train'.center(20)}{'Test'.center(20)}
        {'r2_score'.ljust(10)}{r2_score(Y_train, lm.predict(X_train))}\t{r2_score(Y_test, lm.predict(X_test))}
        {'MSE'.ljust(10)}{mse(Y_train, lm.predict(X_train))}\t{mse(Y_test, lm.predict(X_test))}
        '''
        print(scores)

```


Metric	Train	Test
r2_score	0.8658871776828707	0.8610649253244574
MSE	1821.3833862936174	1780.987539418845

In []: