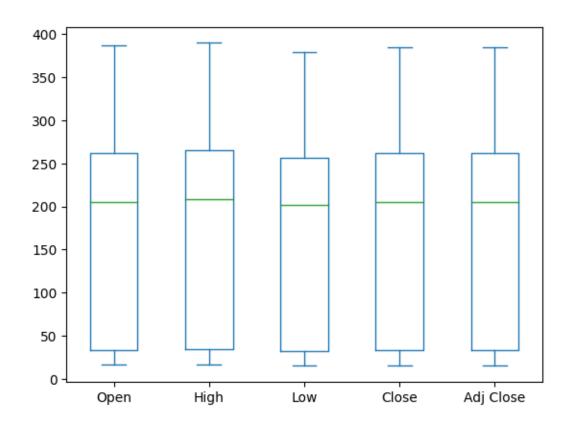
## Untitled

## October 30, 2022

```
[121]: import numpy as np
      import pandas as pd
      import matplotlib.pyplot as plt
      %matplotlib inline
      import chart studio.plotly as py
      import plotly.graph_objs as go
      from plotly.offline import plot
       #for offine plotting
      from plotly.offline import download_plotlyjs, init_notebook_mode , plot, iplot
      init_notebook_mode(connected = True)
[122]: | tesla = pd.read_csv("C:/Users/Swaraj/Downloads/tesla.csv")
      tesla.head()
[122]:
                                                             Adj Close
               Date
                          Open
                                 High
                                             Low
                                                      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
[123]: 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
                      2193 non-null float64
           High
       3
           Low
                      2193 non-null
                                     float64
           Close
                      2193 non-null float64
       5
           Adj Close
                     2193 non-null
                                     float64
           Volume
                      2193 non-null
                                      int64
      dtypes: float64(5), int64(1), object(1)
```

```
memory usage: 120.1+ KB
[124]: tesla['Date'] = pd.to_datetime(tesla['Date'], dayfirst = True)
[125]: print(f'Dataframes contains stock price between {tesla.Date.min()} {tesla.Date.
        →max()}')
       print(f'Total days = {(tesla.Date.max() - tesla.Date.min()).days} days')
      Dataframes contains stock price between 2010-06-29 00:00:00 2019-03-15 00:00:00
      Total days = 3181 days
[126]:
      tesla.describe()
[126]:
                                                                      Adj Close
                     Open
                                  High
                                                 Low
                                                            Close
              2193.000000
                           2193.000000
                                         2193.000000
                                                      2193.000000
                                                                   2193.000000
       count
       mean
               175.652882
                            178.710262
                                          172.412075
                                                       175.648555
                                                                     175.648555
       std
               115.580903
                            117.370092
                                          113.654794
                                                       115.580771
                                                                    115.580771
      min
                16.139999
                             16.629999
                                           14.980000
                                                        15.800000
                                                                      15.800000
       25%
                                                        33.160000
                33.110001
                             33.910000
                                           32.459999
                                                                      33.160000
       50%
               204.990005
                            208.160004
                                          201.669998
                                                       204.990005
                                                                    204.990005
       75%
               262.000000
                            265.329987
                                                       261.739990
                                                                     261.739990
                                          256.209991
       max
               386.690002
                            389.609985
                                          379.350006
                                                       385.000000
                                                                    385.000000
                    Volume
              2.193000e+03
       count
       mean
              5.077449e+06
       std
              4.545398e+06
              1.185000e+05
      min
       25%
              1.577800e+06
       50%
              4.171700e+06
       75%
              6.885600e+06
              3.716390e+07
      max
[127]: tesla[['Open', 'High', 'Low', 'Close', 'Adj Close']].plot(kind = 'box')
[127]: <AxesSubplot: >
```



```
[128]: #Setting the Layout for our graph
       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)
```

```
[129]: iplot(plot)
[130]: # 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
[131]: #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=101)
[132]: scaler = StandardScaler().fit(X_train)
[133]: from sklearn.linear_model import LinearRegression
[134]: | lm = LinearRegression()
       lm.fit(X_train,Y_train)
[134]: LinearRegression()
[135]: #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)
[136]: iplot(plot2)
```

```
[137]: #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
  []:
  []:
  []:
```