```
In [ ]:
          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
          from plotly.offline import download_plotlyjs, init_notebook_mode, plot, iplot
          init_notebook_mode(connected= True)
In [60]:
          sm = pd.read csv("stock market.csv")
In [61]:
Out[61]:
                    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
         1004 2022-01-31 401,970001 427,700012 398,200012 427,140015 427,140015 20047500
```

```
1007 2022-02-03 421.440002 429.260010 404.279999 405.600006 405.600006 9905200
           1008 2022-02-04 407.309998 412.769989 396.640015 410.170013 410.170013 7782400
          1009 rows × 7 columns
In [62]:
           sm.columns
Out[62]: Index(['Date', 'Open', 'High', 'Low', 'Close', 'Adj Close', 'Volume'], dtype='object')
In [63]: sm.info()
         <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 1009 entries, 0 to 1008
        Data columns (total 7 columns):
         # Column
                        Non-Null Count Dtype
         0 Date
                         1009 non-null object
         1 Open 1009 non-null float64
2 High 1009 non-null float64
3 Low 1009 non-null float64
4 Close 1009 non-null float64
         5 Adj Close 1009 non-null float64
6 Volume 1009 non-null int64
        dtypes: float64(5), int64(1), object(1)
        memory usage: 55.3+ KB
In [64]: sm.describe()
```

1005 2022-02-01 432.959991 458.480011 425.540009 457.130005 457.130005 22542300 **1006** 2022-02-02 448.250000 451.980011 426.480011 429.480011 429.48011 14346000

Open 0 High 0 Low 0 Close 0 Adj Close 0 Volume 0 dtype: int64 [66]: sm.duplicated().sum()	Out[64]:		Open	High	Low	Close	Adj Close	Volume	
std 108.537532 109.262960 107.555867 108.289999 108.289999 5.465535e+06 min 233.919998 250.649994 231.229996 233.880005 233.880005 1.144000e+06 25% 331.489990 336.299988 326.000000 331.619995 331.619995 4.091900e+06 50% 377.769989 383.010010 370.880005 378.670013 378.670013 5.934500e+06 75% 509.130005 515.630005 502.529999 509.079987 509.079987 9.322400e+06 max 692.349976 700.989990 686.090027 691.690002 691.690002 5.890430e+07 [65]: sm.isna().sum() sm.isna().sum() 661.69002 691.690002 5.890430e+07		count	1009.000000	1009,000000	1009.000000	1009.000000	1009.000000	1.009000e+03	
min 233.919998 250.649994 231.229996 233.880005 233.880005 1.144000e+06 25% 331.489990 336.299988 326.000000 331.619995 331.619995 4.091900e+06 50% 377.769989 383.010010 370.880005 378.670013 378.670013 5.934500e+06 75% 509.130005 515.630005 502.529999 509.079987 509.079987 9.322400e+06 max 692.349976 700.989990 686.090027 691.690002 691.690002 5.890430e+07 [65]: Date 0 <		mean	419.059673	425.320703	412,374044	419.000733	419.000733	7.570685e+06	
25% 331.489990 336.299988 326.00000 331.619995 331.619995 4.091900e+06 50% 377.769989 383.010010 370.880005 378.670013 378.670013 5.934500e+06 75% 509.130005 515.630005 502.529999 509.079987 509.079987 9.322400e+06 max 692.349976 700.989990 686.090027 691.690002 691.690002 5.890430e+07 [65]: sm.isna().sum() [65]: Date 0 Open 0 High 0 Low 0 Close 0 Adj Close 0 Volume 0 dtype: int64 [66]: sm.duplicated().sum()		std	108.537532	109.262960	107.555867	108.289999	108.289999	5.465535e+06	
50% 377.769989 383.010010 370.880005 378.670013 378.670013 5.934500e+06 75% 509.130005 515.630005 502.529999 509.079987 509.079987 9.322400e+06 max 692.349976 700.989990 686.090027 691.690002 5.890430e+07 [65]: sm.isna().sum() [65]: Date		min	233.919998	250.649994	231.229996	233.880005	233.880005	1.144000e+06	
75% 509.130005 515.630005 502.529999 509.079987 509.079987 9.322400e+06 max 692.349976 700.989990 686.090027 691.690002 691.690002 5.890430e+07 [65]: sm.isna().sum() [65]: Date 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		25%	331.489990	336,299988	326.000000	331.619995	331.619995	4.091900e+06	
max 692.349976 700.989990 686.090027 691.690002 5.890430e+07 [65]: sm.isna().sum() [65]: Date		50%	377.769989	383.010010	370.880005	378.670013	378.670013	5.934500e+06	
[65]: sm.isna().sum() [65]: Date		75%	509.130005	515.630005	502,529999	509.079987	509.079987	9.322 <mark>4</mark> 00e+06	
Sm.1sha().sum() Date 0 Open 0 High 0 Low 0 Close 0 Adj Close 0 Volume 0 dtype: int64 [66]: sm.duplicated().sum()		max	692.349976	700.989990	686.090027	691.690002	691.690002	5.890430e+07	
[66]: 0	ot[65];	Date Open High Low Close Adj Cl Volume	0 0 0 0 0 0 0 0						
	[66]:]: sm.duplicated().sum()							
F1.	Out[66]:	0							
	In []:								

```
In [ ]:
In [67]:
           sm['Date'] = pd.to_datetime(sm['Date'])
In [68]:
          print(f'Dataframe\ contains\ stock\ prices\ between\ \{sm.Date.min()\}\ \{sm.Date.max()\}')
          print(f'Total days = {(sm.Date.max() - sm.Date.min()).days} Days')
        Dataframe contains stock prices between 2018-02-05 00:00:00 2022-02-04 00:00:00
        Total days = 1460 Days
In [69];
          sm[['Date','Open','High','Low','Close','Adj Close','Volume']].plot(kind = 'box')
Out[69]: <AxesSubplot:>
        6
                                                      0
        5
        4
                                                      0
                                                      ₿
        3
        2
        1
        0
                    High
                                    Close Adj Close Volume
            Open
                             Low
```

```
In [70]: layout = go.Layout(
               title = 'Stock Prices',
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'
                )
            )
            sm_data = [{'x':sm['Date'] , 'y':sm['Close']}]
plot = go.Figure(data = sm_data, layout =layout)
In [71]: iplot(plot)
In [72]:
            from sklearn.model_selection import train_test_split
            from sklearn.preprocessing import MinMaxScaler
            from sklearn.preprocessing import StandardScaler
            from sklearn.metrics import mean_squared_error as mse
```

```
In [73]:
           X = np.array(sm.index).reshape(-1,1)
           Y = sm['Close']
           X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.3, random_state = 101)
In [74]: scaler = StandardScaler().fit(X_train)
In [75]: from sklearn.linear_model import LinearRegression
In [76]:
           # lm stands for linear regression
           lm = LinearRegression()
           lm.fit(X_train, Y_train)
Out[76]: LinearRegression()
In [82]:
           #actual and predicted values on graph
           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'
           sm_data = [trace0,trace1]
layout.xaxis.title.text = 'Day'
```

plot2 = go.Figure(data = sm data, layout = layout)