```
In [19]:
         import numpy as np
         import pandas as pd
         from sklearn.preprocessing import StandardScaler
         import statsmodels.api as sm
         from sklearn import linear model
         df = pd.read csv(r'C:\Users\Berk\Desktop\aapl1.csv')
In [20]: | df.tail(2)
Out[20]:
                                                                                     Ask
                    #RIC
                                    Date-Time
                                               Price Volume Numberoftrades Bidprice
                                                                                    Price
                                     2019-12-
                                                      2716.0
                                                                            293.50 293.85
          22346 AAPL.OQ
                                             293.675
                                                                      NaN
                         31T22:00:00.000000000Z
                                     2019-12-
          22347 AAPL.OQ
                                             293.810
                                                      848.0
                                                                      NaN
                                                                            293.77 293.85
                         31T23:00:00.000000000Z
In [21]: | df = df.dropna(subset=['Numberoftrades', 'Depth', 'Volume', 'Spread', 'Price'])
In [22]: | x = df[['Numberoftrades', 'Depth', 'Volume', 'Spread']]
         y = df[['Price']]
In [23]: x normalized = StandardScaler().fit transform(x)
         y normalized = StandardScaler().fit transform(y)
In [24]: #To check if the data is normalized
         x normalized
Out[24]: array([[-0.90541615, -1.04209567, -0.81497772, 0.00374322],
                [-0.90727053, -1.04521684, -0.81509221, 0.00192592],
                [-0.90564795, -1.04258453, -0.81411772, 0.00737782],
                 [-0.48586371, -0.41707072, -0.5324871, -0.0307855],
                 [0.64484171, 0.47528485, 0.37295597, -0.0271509],
                [-0.90842951, -1.03677463, 1.25058628, 0.03100274]])
In [25]: | x e = pd.DataFrame(x normalized, columns = ['Numberoftrades', 'Depth', 'Volum
         e','Spread'])
In [26]: reg = linear model.LinearRegression()
         reg.fit(x_normalized, y_normalized)
Out[26]: LinearRegression(copy X=True, fit intercept=True, n jobs=None, normalize=Fals
         e)
```

```
In [27]: model = sm.OLS(y_normalized, x_normalized).fit()
    predictions = model.predict(x_normalized)

    print_model = model.summary()
    print(print_model)
```

OLS Regression Results

=======										
Dep. Variable	:		У	/	R-squ	uared	(unce	ntered)	:	
0.011				_		_				
Model:			OLS	5	Adj.	R-sq	uared	(uncente	ered)	:
0.011 Method:		l east	: Squares		F-sta	atict	ic·			
39.91		Least	. Jquai es	•	1-300	acisc	ic.			
Date:		Wed, 03	Jun 2020	9	Prob	(F-s	tatist	ic):		
2.66e-33		•				•		•		
Time:			03:31:06	5	Log-l	Likel	ihood:			
-20243.										
No. Observation	ons:		14322	2	AIC:					
4.049e+04 Df Residuals:			14318	2	BIC:					
4.052e+04			14310)	DIC.					
Df Model:			4	1						
Covariance Ty	pe:	r	onrobust	t						
=========	======		======	===	=====	====	=====	======	====	=======
=								F.0		
r1	coet	f std	err		t		P> t	[0]	.025	0.97
5]										
_										
x1	0.047	0.	016	2	.947		0.003	0	.016	0.07
8										
x2	-0.1208	3 0.	014	-8	.381		0.000	-0	.149	-0.09
3	0 070	- 0	010	_	161		0 000	•	051	0.00
x3 0	0.0706	o 0.	010	/	.161		0.000	0	.051	0.09
x4	-0.0125	5 0.	008	-1	.497		0.134	-0	.029	0.00
4	0.012	· · · · · · · · · · · · · · · · · · ·	000		• 427		0.154		.025	0.00
=========	======		======	===	=====		=====	======		=======
=										
Omnibus:			1848.193	3	Durb:	in-Wa	tson:			0.01
4 Dogah (Omodibus)			0.000	,	7	D -	/ 3.0	١.		2644 45
Prob(Omnibus) 1	•		0.000)	Jarqi	re-Re	ra (JB):		2641.15
Skew:			1.030	3	Prob	(JB) ·				0.0
0			2.030	-	05	(55).				0.0
Kurtosis:			3.424	1	Cond	. No.				3.6
3										
==========	======	======	======	===	=====	====	=====	======	====	=======
=										

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```
In [28]: from sklearn.tree import DecisionTreeRegressor
         df = pd.read csv(r'C:\Users\Berk\Desktop\aapl1.csv')
         df.describe()
         df = df.dropna(subset=['Numberoftrades', 'Depth','Volume','Spread','Price'])
         x = df[['Numberoftrades', 'Depth', 'Volume', 'Spread']].values
         y = df[['Price']].values
         from sklearn.model selection import train test split
         x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, rando
         m state=0)
         regr = DecisionTreeRegressor()
         regr.fit(x_train,y_train)
Out[28]: DecisionTreeRegressor(criterion='mse', max_depth=None, max_features=None,
                                max_leaf_nodes=None, min_impurity_decrease=0.0,
                                min_impurity_split=None, min_samples_leaf=1,
                                min_samples_split=2, min_weight_fraction_leaf=0.0,
                                presort=False, random_state=None, splitter='best')
In [29]: | y_pred = regr.predict(x_test)
In [30]: | df2=pd.DataFrame({'y_pred':y_pred})
         df2.assign(y=y_test)
Out[30]:
               y_pred
                          У
             0 180.84 100.67
               125.45 128.60
               105.34
                       94.04
             3
               186.02 166.57
                223.11 155.47
          2860
               101.88 101.82
          2861
                175.42 142.83
          2862
                118.52 113.40
          2863
               165.24 151.68
          2864
               157.51 143.81
         2865 rows × 2 columns
In [31]:
         from sklearn import metrics
         print('Mean Absolute Error:', metrics.mean_absolute_error(y_test, y_pred))
         print('Mean Squared Error:', metrics.mean squared error(y test, y pred))
         print('Root Mean Squared Error:', np.sqrt(metrics.mean_squared_error(y_test, y
          _pred)))
         Mean Absolute Error: 30.67673996509599
```

Mean Absolute Error: 30.67673996509599 Mean Squared Error: 1898.073335445026 Root Mean Squared Error: 43.566883471795705

In []:			