In [1]: print("Mahima Chauhan")

Mahima Chauhan

In [2]: pip install pandas

Defaulting to user installation because normal site-packages is not writeable Requirement already satisfied: pandas in /usr/local/lib/python3.8/dist-packages (1.4. 2)

Requirement already satisfied: numpy>=1.18.5 in /usr/local/lib/python3.8/dist-package s (from pandas) (1.22.3)

Requirement already satisfied: python-dateutil>=2.8.1 in /usr/local/lib/python3.8/dist-packages (from pandas) (2.8.2)

Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.8/dist-packages (from pandas) (2022.1)

Requirement already satisfied: six>=1.5 in /usr/lib/python3/dist-packages (from pytho n-dateutil>=2.8.1->pandas) (1.14.0)

Note: you may need to restart the kernel to use updated packages.

In [23]: pip install matplotlib

Requirement already satisfied: matplotlib in c:\users\kumar\appdata\local\programs\py thon\python39\lib\site-packages (3.5.1)Note: you may need to restart the kernel to us e updated packages.

Requirement already satisfied: cycler>=0.10 in c:\users\kumar\appdata\local\programs \python\python39\lib\site-packages (from matplotlib) (0.11.0)

Requirement already satisfied: fonttools>=4.22.0 in c:\users\kumar\appdata\local\prog rams\python\python39\lib\site-packages (from matplotlib) (4.32.0)

Requirement already satisfied: python-dateutil>=2.7 in c:\users\kumar\appdata\local\p rograms\python\python39\lib\site-packages (from matplotlib) (2.8.2)

Requirement already satisfied: pyparsing>=2.2.1 in c:\users\kumar\appdata\local\programs\python\python39\lib\site-packages (from matplotlib) (3.0.8)

Requirement already satisfied: numpy>=1.17 in c:\users\kumar\appdata\local\programs\p ython\python39\lib\site-packages (from matplotlib) (1.22.3)

Requirement already satisfied: pillow>=6.2.0 in c:\users\kumar\appdata\local\programs \python\python39\lib\site-packages (from matplotlib) (9.1.0)

Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\kumar\appdata\local\prog rams\python\python39\lib\site-packages (from matplotlib) (1.4.2)

Requirement already satisfied: packaging>=20.0 in c:\users\kumar\appdata\local\progra ms\python\python39\lib\site-packages (from matplotlib) (21.3)

Requirement already satisfied: six>=1.5 in c:\users\kumar\appdata\local\programs\pyth on\python39\lib\site-packages (from python-dateutil>=2.7->matplotlib) (1.15.0)

In [52]: pip install sklearn

Requirement already satisfied: sklearn in c:\users\kumar\appdata\local\programs\python39\lib\site-packages (0.0)

Requirement already satisfied: scikit-learn in c:\users\kumar\appdata\local\programs \python\python39\lib\site-packages (from sklearn) (1.0.2)

Requirement already satisfied: scipy>=1.1.0 in c:\users\kumar\appdata\local\programs \python\python39\lib\site-packages (from scikit-learn->sklearn) (1.8.0)

Requirement already satisfied: numpy>=1.14.6 in c:\users\kumar\appdata\local\programs \python\python39\lib\site-packages (from scikit-learn->sklearn) (1.22.3)

Requirement already satisfied: threadpoolctl>=2.0.0 in c:\users\kumar\appdata\local\p rograms\python\python39\lib\site-packages (from scikit-learn->sklearn) (3.1.0)

Requirement already satisfied: joblib>=0.11 in c:\users\kumar\appdata\local\programs \python\python39\lib\site-packages (from scikit-learn->sklearn) (1.1.0)

Note: you may need to restart the kernel to use updated packages.

5/19/22, 8:20 AM

```
#pip install seaborn
In [101...
          # Importing Libraries
In [83]:
           import numpy as np
           import pandas as pd
           import matplotlib.pyplot as plt
           # Importing Data
           from sklearn.datasets import load_boston
           boston = load_boston()
In [84]:
          boston.data.shape
          (506, 13)
Out[84]:
In [85]:
          data = pd.DataFrame(boston.data)
           data.columns = boston.feature_names
           data.head(10)
               CRIM
                       ΖN
                           INDUS CHAS
                                          NOX
                                                                DIS RAD
                                                                            TAX PTRATIO
                                                                                                B LSTAT
Out[85]:
                                                  RM
                                                        AGE
          0 0.00632
                      18.0
                              2.31
                                      0.0
                                          0.538
                                                 6.575
                                                        65.2 4.0900
                                                                       1.0
                                                                           296.0
                                                                                      15.3 396.90
                                                                                                    4.98
                              7.07
           1 0.02731
                       0.0
                                          0.469 6.421
                                                        78.9 4.9671
                                                                       2.0 242.0
                                                                                      17.8 396.90
                                                                                                    9.14
                                      0.0
          2 0.02729
                       0.0
                              7.07
                                          0.469 7.185
                                                        61.1 4.9671
                                                                       2.0 242.0
                                                                                      17.8 392.83
                                                                                                    4.03
                                      0.0
                              2.18
                                                                                                     2.94
          3 0.03237
                       0.0
                                      0.0
                                          0.458 6.998
                                                        45.8
                                                             6.0622
                                                                       3.0 222.0
                                                                                      18.7 394.63
           4 0.06905
                       0.0
                              2.18
                                      0.0
                                          0.458 7.147
                                                        54.2 6.0622
                                                                       3.0 222.0
                                                                                      18.7 396.90
                                                                                                     5.33
           5 0.02985
                       0.0
                              2.18
                                      0.0
                                          0.458 6.430
                                                        58.7 6.0622
                                                                       3.0 222.0
                                                                                      18.7 394.12
                                                                                                     5.21
            0.08829
                      12.5
                              7.87
                                          0.524 6.012
                                                        66.6 5.5605
                                                                       5.0 311.0
                                                                                      15.2 395.60
                                                                                                    12.43
                                      0.0
          7 0.14455 12.5
                              7.87
                                                                                      15.2 396.90
                                      0.0
                                          0.524 6.172
                                                        96.1 5.9505
                                                                       5.0 311.0
                                                                                                    19.15
             0.21124 12.5
                              7.87
                                                       100.0
                                                             6.0821
                                                                                      15.2 386.63
                                                                                                   29.93
                                      0.0
                                          0.524 5.631
                                                                       5.0 311.0
          9 0.17004 12.5
                              7.87
                                          0.524 6.004
                                      0.0
                                                        85.9 6.5921
                                                                       5.0 311.0
                                                                                      15.2 386.71
                                                                                                    17.10
          # Adding 'Price' (target) column to the data
In [86]:
           boston.target.shape
          (506,)
Out[86]:
          data = pd.DataFrame(boston.data)
In [102...
           data.columns = boston.feature_names
           data.head(10)
```

5/19/22, 8:20 AM

Out[102]:		CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	PTRATIO	В	LSTAT	
	0	0.00632	18.0	2.31	0.0	0.538	6.575	65.2	4.0900	1.0	296.0	15.3	396.90	4.98	_
	1	0.02731	0.0	7.07	0.0	0.469	6.421	78.9	4.9671	2.0	242.0	17.8	396.90	9.14	
	2	0.02729	0.0	7.07	0.0	0.469	7.185	61.1	4.9671	2.0	242.0	17.8	392.83	4.03	
	3	0.03237	0.0	2.18	0.0	0.458	6.998	45.8	6.0622	3.0	222.0	18.7	394.63	2.94	
	4	0.06905	0.0	2.18	0.0	0.458	7.147	54.2	6.0622	3.0	222.0	18.7	396.90	5.33	,
	5	0.02985	0.0	2.18	0.0	0.458	6.430	58.7	6.0622	3.0	222.0	18.7	394.12	5.21	
	6	0.08829	12.5	7.87	0.0	0.524	6.012	66.6	5.5605	5.0	311.0	15.2	395.60	12.43	
	7	0.14455	12.5	7.87	0.0	0.524	6.172	96.1	5.9505	5.0	311.0	15.2	396.90	19.15	
	8	0.21124	12.5	7.87	0.0	0.524	5.631	100.0	6.0821	5.0	311.0	15.2	386.63	29.93	
	9	0.17004	12.5	7.87	0.0	0.524	6.004	85.9	6.5921	5.0	311.0	15.2	386.71	17.10	
4															•
In [89]:	# Adding 'Price' (target) column to the data boston.target.shape														
Out[89]:	(506,)														
In [90]:	<pre>data['Price'] = boston.target data.head()</pre>														
Out[90]:		CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	PTRATIO	В	LSTAT	Pr
	0	0.00632	18.0	2.31	0.0	0.538	6.575	65.2	4.0900	1.0	296.0	15.3	396.90	4.98	2
	1	0.02731	0.0	7.07	0.0	0.469	6.421	78.9	4.9671	2.0	242.0	17.8	396.90	9.14	2
	2	0.02729	0.0	7.07	0.0	0.469	7.185	61.1	4.9671	2.0	242.0	17.8	392.83	4.03	3
	3	0.03237	0.0	2.18	0.0	0.458	6.998	45.8	6.0622	3.0	222.0	18.7	394.63	2.94	3
	4	0.06905	0.0	2.18	0.0	0.458	7.147	54.2	6.0622	3.0	222.0	18.7	396.90	5.33	3
4															•
In [91]:	da	ıta.descr	ribe()											

4

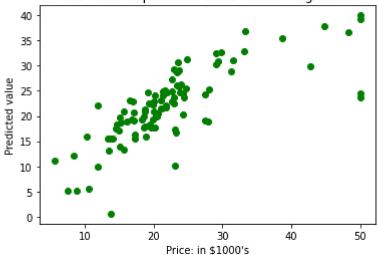
5/19/22, 8:20 AM

CRIM **INDUS** NOX **AGE** DI: Out[91]: ΖN **CHAS** RM **count** 506.000000 506.000000 506.000000 506.000000 506.000000 506.000000 506.000000 506.00000 3.613524 11.363636 11.136779 0.069170 0.554695 6.284634 68.574901 3.79504 mean 8.601545 std 23.322453 6.860353 0.253994 0.115878 0.702617 28.148861 2.10571 min 0.006320 0.000000 0.460000 0.000000 0.385000 3.561000 2.900000 1.12960 25% 0.082045 0.000000 5.190000 0.000000 0.449000 5.885500 45.025000 2.10017 50% 0.256510 0.000000 9.690000 0.000000 0.538000 6.208500 77.500000 3.20745 **75%** 3.677083 12.500000 18.100000 0.000000 0.624000 6.623500 94.075000 5.18842 max 88.976200 100.000000 27.740000 1.000000 0.871000 8.780000 100.000000 12.12650 In [92]: data.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 506 entries, 0 to 505 Data columns (total 14 columns): Column Non-Null Count Dtype ---0 CRIM 506 non-null float64 1 ΖN 506 non-null float64 2 506 non-null float64 **INDUS** 3 506 non-null float64 CHAS 4 NOX 506 non-null float64 5 506 non-null float64 RM6 AGE 506 non-null float64 7 DIS 506 non-null float64 8 RAD 506 non-null float64 9 TAX 506 non-null float64 10 PTRATIO 506 non-null float64 506 non-null float64 11 B float64 12 LSTAT 506 non-null 506 non-null float64 13 Price dtypes: float64(14) memory usage: 55.5 KB # Input Data In [93]: x = boston.data# Output Data y = boston.target # splitting data to training and testing dataset. #from sklearn.cross_validation import train_test_split #the submodule cross validation is renamed and reprecated to model selection from sklearn.model_selection import train_test_split xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size =0.2, print("xtrain shape : ", xtrain.shape) print("xtest shape : ", xtest.shape)

plt.show()

```
print("ytrain shape : ", ytrain.shape)
         print("ytest shape : ", ytest.shape)
         xtrain shape : (404, 13)
         xtest shape : (102, 13)
         ytrain shape : (404,)
         ytest shape : (102,)
In [94]:
         # Fitting Multi Linear regression model to training model
         from sklearn.linear_model import LinearRegression
         regressor = LinearRegression()
         regressor.fit(xtrain, ytrain)
         # predicting the test set results
         y_pred = regressor.predict(xtest)
         # Plotting Scatter graph to show the prediction
In [95]:
         # results - 'ytrue' value vs 'y_pred' value
         plt.scatter(ytest, y pred, c = 'green')
         plt.xlabel("Price: in $1000's")
         plt.ylabel("Predicted value")
         plt.title("True value vs predicted value : Linear Regression")
```

True value vs predicted value : Linear Regression



```
In [96]: # Results of Linear Regression.
from sklearn.metrics import mean_squared_error
mse = mean_squared_error(ytest, y_pred)
print("Mean Square Error : ", mse)
```

Mean Square Error: 33.448979997676474