

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

boston = pd.read_csv("/content/Boston.csv")

# Printing first 5 records of the dataset
print(dataset.head(5))
```

	Unnamed: 0	crim	zn	indus	chas	nox	rm	age	dis	rad	\
0	1	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	
1	2	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	
2	3	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	
3	4	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	
4	5	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	

	tax	ptratio	black	lstat	medv
0	296	15.3	396.90	4.98	24.0
1	242	17.8	396.90	9.14	21.6
2	242	17.8	392.83	4.03	34.7
3	222	18.7	394.63	2.94	33.4
4	222	18.7	396.90	5.33	36.2

```
boston.shape

(506, 15)

boston.columns

Index(['Unnamed: 0', 'crim', 'zn', 'indus', 'chas', 'nox', 'rm', 'age', 'dis',
      'rad', 'tax', 'ptratio', 'black', 'lstat', 'medv'],
      dtype='object')
```

```
data = pd.DataFrame(boston)
data.head(10)
```

	Unnamed: 0	crim	zn	indus	chas	nox	rm	age	dis	rad	tax	ptratio	bl
0	1	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	296	15.3	396.90
1	2	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	242	17.8	396.90
2	3	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	242	17.8	392.83
3	4	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	222	18.7	394.63
4	5	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	222	18.7	396.90
5	6	0.02985	0.0	2.18	0	0.458	6.430	58.7	6.0622	3	222	18.7	396.90
6	7	0.08829	12.5	7.87	0	0.524	6.012	66.6	5.5605	5	311	15.2	396.90
7	8	0.14455	12.5	7.87	0	0.524	6.172	96.1	5.9505	5	311	15.2	396.90
8	9	0.21124	12.5	7.87	0	0.524	5.631	100.0	6.0821	5	311	15.2	385.10

```
data.describe()
```

```

      zn      indus      chas      nox      rm      age      dis      rad
0 6.000000 506.000000 506.000000 506.000000 506.000000 506.000000 506.000000 506.000000
1 1.363636 11.136779 0.069170 0.554695 6.284634 68.574901 3.795043 9.549407
3 3.322453 6.860353 0.253994 0.115878 0.702617 28.148861 2.105710 8.707259
data.info()

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 506 entries, 0 to 505

```

```
Data columns (total 15 columns):
```

```

#   Column      Non-Null Count  Dtype
---  ---

```

```

0   Unnamed: 0    506 non-null    int64
1   crim         506 non-null    float64
2   zn           506 non-null    float64
3   indus        506 non-null    float64
4   chas         506 non-null    int64
5   nox          506 non-null    float64
6   rm           506 non-null    float64
7   age          506 non-null    float64
8   dis          506 non-null    float64
9   rad          506 non-null    int64
10  tax          506 non-null    int64
11  ptratio      506 non-null    float64
12  black        506 non-null    float64
13  lstat        506 non-null    float64
14  medv         506 non-null    float64

```

```
dtypes: float64(11), int64(4)
```

```
memory usage: 59.4 KB
```

```
# Input Data
```

```
x = boston.drop('medv', axis=1) # Assuming 'medv' is the target variable
```

```
# Output Data
```

```
y = boston['medv']
```

```
# splitting data to training and testing dataset.
```

```
#from sklearn.cross_validation import train_test_split
```

```
#the submodule cross_validation is renamed and deprecated to model_selection
```

```
from sklearn.model_selection import train_test_split
```

```

xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size =0.2,
                                                random_state = 0)

```

```
print("xtrain shape : ", xtrain.shape)
```

```
print("xtest shape : ", xtest.shape)
```

```
print("ytrain shape : ", ytrain.shape)
```

```
print("ytest shape : ", ytest.shape)
```

```
xtrain shape : (404, 14)
```

```
xtest shape : (102, 14)
```

```
ytrain shape : (404,)
```

```
ytest shape : (102,)
```

```
# 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
```

```
# results - 'ytrue' value vs 'y_pred' value
```

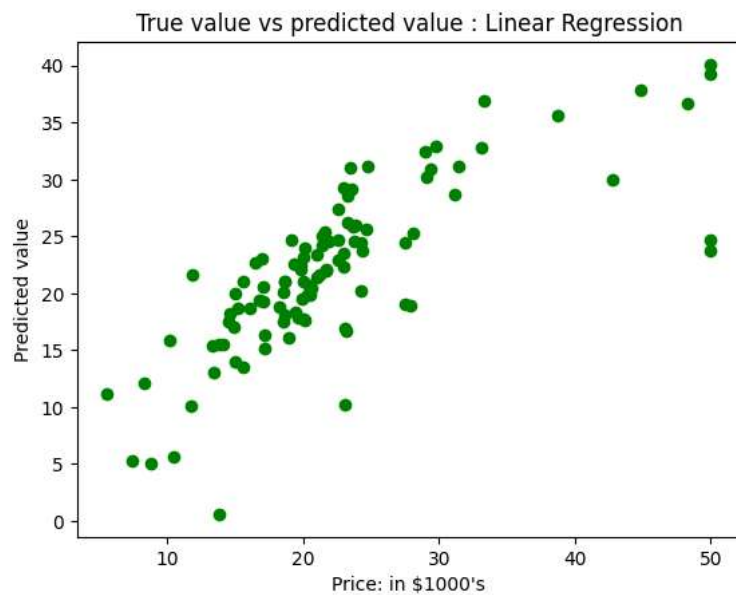
```
plt.scatter(ytest, y_pred, c = 'green')
```

```
plt.xlabel("Price: in $1000's")
```

```
plt.ylabel("Predicted value")
```

```
plt.title("True value vs predicted value : Linear Regression")
```

```
plt.show()
```



```
from sklearn.metrics import mean_squared_error, mean_absolute_error
mse = mean_squared_error(ytest, y_pred)
mae = mean_absolute_error(ytest, y_pred)
print("Mean Square Error : ", mse)
print("Mean Absolute Error : ", mae)
```

```
Mean Square Error : 33.266961459239106
Mean Absolute Error : 3.838476893830883
```

```
from sklearn import svm
from sklearn.svm import SVC
from sklearn.metrics import mean_absolute_percentage_error
```

```
model_SVR = svm.SVR()
model_SVR.fit(X_train, Y_train)
Y_pred = model_SVR.predict(X_valid)
```

```
print(mean_absolute_percentage_error(Y_valid, Y_pred))
```

```
0.1870512931870423
```

```
from sklearn.ensemble import RandomForestRegressor
```

```
model_RFR = RandomForestRegressor(n_estimators=10)
model_RFR.fit(X_train, Y_train)
Y_pred = model_RFR.predict(X_valid)
```

```
mean_absolute_percentage_error(Y_valid, Y_pred)
```

```
0.1871898894607579
```

```
from sklearn.linear_model import LinearRegression
```

```
model_LR = LinearRegression()
model_LR.fit(X_train, Y_train)
Y_pred = model_LR.predict(X_valid)
```

```
print(mean_absolute_percentage_error(Y_valid, Y_pred))
```

```
0.18741683841599854
```

```
!pip3 install catboost
```

```
Collecting catboost
```

```
  Downloading catboost-1.2.2-cp310-cp310-manylinux2014_x86_64.whl (98.7 MB)
```

```
----- 98.7/98.7 MB 9.4 MB/s eta 0:00:00
```

```
Requirement already satisfied: graphviz in /usr/local/lib/python3.10/dist-packages (from catboost) (0.20.1)
```

```
Requirement already satisfied: matplotlib in /usr/local/lib/python3.10/dist-packages (from catboost) (3.7.1)
```

```
Requirement already satisfied: numpy>=1.16.0 in /usr/local/lib/python3.10/dist-packages (from catboost) (1.23.5)
```

```
Requirement already satisfied: pandas>=0.24 in /usr/local/lib/python3.10/dist-packages (from catboost) (1.5.3)
```

```
Requirement already satisfied: scipy in /usr/local/lib/python3.10/dist-packages (from catboost) (1.11.3)
```

```
Requirement already satisfied: plotly in /usr/local/lib/python3.10/dist-packages (from catboost) (5.15.0)
```

```
Requirement already satisfied: six in /usr/local/lib/python3.10/dist-packages (from catboost) (1.16.0)
```

```
Requirement already satisfied: python-dateutil>=2.8.1 in /usr/local/lib/python3.10/dist-packages (from pandas>=0.24->catboost) (2.8.2)
```

```
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-packages (from pandas>=0.24->catboost) (2023.3.post1)
```

```
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib->catboost) (1.1.1)
```

```
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.10/dist-packages (from matplotlib->catboost) (0.12.1)
```

```
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib->catboost) (4.43.1)
```

```
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib->catboost) (1.4.5)
```

```
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib->catboost) (23.2)
```

```
Requirement already satisfied: pillow>=6.2.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib->catboost) (9.4.0)
```

```
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib->catboost) (3.1.1)
```

```
Requirement already satisfied: tenacity>=6.2.0 in /usr/local/lib/python3.10/dist-packages (from plotly->catboost) (8.2.3)
```

```
Installing collected packages: catboost
```

```
Successfully installed catboost-1.2.2
```

```
# This code is contributed by @amartajisce
```

```
from catboost import CatBoostRegressor
```

```
cb_model = CatBoostRegressor()
```

```
cb_model.fit(X_train, Y_train)
```

```
preds = cb_model.predict(X_valid)
```

```
from sklearn.metrics import r2_score
```

```
cb_r2_score=r2_score(Y_valid, preds)
```

```
cb_r2_score
```

```
990:   learn: 24503.8823626   total: 2.01s   remaining: 18.3ms
991:   learn: 24503.6610596   total: 2.01s   remaining: 16.2ms
992:   learn: 24502.7457525   total: 2.01s   remaining: 14.2ms
993:   learn: 24490.1166911   total: 2.02s   remaining: 12.2ms
994:   learn: 24488.5023143   total: 2.02s   remaining: 10.1ms
995:   learn: 24468.9731701   total: 2.02s   remaining: 8.11ms
996:   learn: 24459.0497225   total: 2.02s   remaining: 6.08ms
997:   learn: 24458.7284279   total: 2.02s   remaining: 4.05ms
998:   learn: 24448.9981948   total: 2.02s   remaining: 2.03ms
999:   learn: 24440.7422679   total: 2.03s   remaining: 0us
0.38351169878113034
```