# Data Analytics I

Create a Linear Regression Model using Python/R to predict home prices using Boston Housing Dataset (https://www.kaggle.com/c/boston-housing). The Boston Housing dataset contains information about various houses in Boston through different parameters. There are 506 samples

and 14 feature variables in this dataset.

The objective is to predict the value of prices of the house using the given features.

#I'm importing the machine learning library sklearn, numpy, and pandas

mport pandas as pd import numpy as np from sklearn import linear\_model from sklearn.model\_selection import train\_test\_split

#Load dataset
df=pd.read\_csv("boston\_housing.csv")

#View dimension of dataset df.shape (506, 14)

#View some statistics of dataset mean,max,min etc df.describe()

		crim	zn	indus	chas	nox	rm	age	dis	rad	tax	ptratio	
	black	Istat	medv										
count	506.00	00000	506.000000		506.000000		506.000000		506.000000		506.000000		
	506.00	506.000000		506.000000		506.000000		506.000000		506.000000		506.000000	
	506.00	00000	506.00	00000									
mean	3.613524		11.363636		11.136779		0.069170		0.554695		6.284634		
	68.574901		3.795043		9.549407		408.237154		18.455534		356.674032		
	12.653	3063	22.532	2806									
std	8.601545		23.322453		6.860353		0.253994		0.115878		0.702617		
	28.148	3861	2.105	710	8.7072	259	168.5	37116	2.1649	946	91.29	4864	
	7.1410	062	9.197	104									
min	0.006320		0.000000		0.460000		0.000000		0.385000		3.561000		
	2.900000		1.129600		1.000000		187.000000		12.600000		0.320000		
	1.7300	000	5.0000	000									
25%	0.0820	045	0.000	000	5.1900	000	0.000	000	0.4490	000	5.885	500	
	45.025000		2.100175		4.000000		279.000000		17.400000		375.3	77500	
	6.9500	000	17.02	5000									
50%	0.256510		0.000000		9.690000		0.000000		0.538000		6.208500		
	77.500	0000	3.207	450	5.0000	000	330.0	00000	19.050	0000	391.4	40000	
	11.360	0000	21.20	0000									

75%	3.677083	12.500000	18.100000	0.000000	0.624000	6.623500				
	94.075000	5.188425	24.000000	666.000000	20.200000	396.225000				
	16.955000	25.000000								
max	88.976200	100.000000	27.740000	1.000000	0.871000	8.780000				
	100.000000	12.126500	24.000000	711.000000	22.000000	396.900000				
	37.970000	50.000000								
#View dataset features and attribute information										
df.head()										
	rim zn	indus chas	nox rm	age dis	rad tax	ptratio black				
	Istat medv									
0	0.00632	18.0 2.31	0 0.538	6.575 65.2	4.0900 1	296.0 15.3				
	396.90 4.98	24.0								
1	0.02731	0.0 7.07	0 0.469	6.421 78.9	4.9671 2	242.0 17.8				
	396.90 9.14	21.6								
2	0.02729	0.0 7.07	0 0.469	7.185 61.1	4.9671 2	242.0 17.8				
	392.83 4.03	34.7								
3	0.03237	0.0 2.18	0 0.458	6.998 45.8	6.06223	222.0 18.7				
	394.63 2.94	33.4								
4	0.06905	0.0 2.18	0 0.458	7.147 54.2	6.06223	222.0 18.7				
	396.90 5.33	36.2								

## # Assign feature variable to df\_x

 $df_x=df$ 

print(df\_x)

crim zn indus chas nox rm age dis rad tax \ 0.00632 18.0 2.31 0 0.538 6.575 65.2 4.0900 1 296.0 1 0.02731 0.0 7.07 0 0.469 6.421 78.9 4.9671 2 242.0 0.02729 0.0 7.07 0 0.469 7.185 61.1 4.9671 2 242.0 3 0.03237 0.0 2.18 0 0.458 6.998 45.8 6.0622 3 222.0 0.06905 0.0 2.18 0 0.458 7.147 54.2 6.0622 3 222.0 501 0.06263 0.0 11.93 0 0.573 6.593 69.1 2.4786 1 273.0 502 0.04527 0.0 11.93 0 0.573 6.120 76.7 2.2875 1 273.0 503 0.06076 0.0 11.93 0 0.573 6.976 91.0 2.1675 1 273.0 504 0.10959 0.0 11.93 0 0.573 6.794 89.3 2.3889 1 273.0 505 0.04741 0.0 11.93 0 0.573 6.030 80.8 2.5050 1 273.0

### ptratio black lstat medv

- 0 15.3 396.90 4.98 24.0
- 1 17.8 396.90 9.14 21.6
- 2 17.8 392.83 4.03 34.7
- 3 18.7 394.63 2.94 33.4
- 4 18.7 396.90 5.33 36.2

```
... ... ... ...
    21.0 391.99 9.67 22.4
501
502 21.0 396.90 9.08 20.6
503 21.0 396.90 5.64 23.9
504
    21.0 393.45 6.48 22.0
     21.0 396.90 7.88 11.9
505
# Assign medv variable to df_y
df y=df.medv
print(df_y)
    24.0
0
    21.6
1
2
    34.7
    33.4
3
    36.2
501 22.4
502 20.6
503 23.9
504 22.0
505 11.9
Name: medv, Length: 506, dtype: float64
#Initialize the linear regression model
reg =linear_model.LinearRegression()
#Split the data into 67% training and 33% testing data
#NOTE: We have to split the dependent variables (x) and the target or independent variable (y)
x_train, x_test, y_train, y_test = train_test_split(df_x, df_y, test_size=0.33, random_state=42)
#Train our model with the training data
reg.fit(x_train, y_train)
#Print the coefecients/weights for each feature/column of our model
print(reg.coef_)
-1.37739992e-14 -1.45813673e-15 -3.47378376e-16 -6.97846728e-16
 6.24500451e-16 -3.81639165e-17 3.95462742e-16 2.94902991e-17
 1.08105799e-15 1.00000000e+00]
#print our price predictions on our test data
y_pred = reg.predict(x_test)
print(y_pred)
23.6 32.4 13.6 22.8 16.1 20. 17.8 14. 19.6 16.8 21.5 18.9 7. 21.2
```

```
18.5 29.8 18.8 10.2 50. 14.1 25.2 29.1 12.7 22.4 14.2 13.8 20.3 14.9 21.7 18.3 23.1 23.8 15. 20.8 19.1 19.4 34.7 19.5 24.4 23.4 19.7 28.2 50. 17.4 22.6 15.1 13.1 24.2 19.9 24. 18.9 35.4 15.2 26.5 43.5 21.2 18.4 28.5 23.9 18.5 25. 35.4 31.5 20.2 24.1 20. 13.1 24.8 30.8 12.7 20. 23.7 10.8 20.6 20.8 5. 20.1 48.5 10.9 7. 20.9 17.2 20.9 9.7 19.4 29. 16.4 25. 25. 17.1 23.2 10.4 19.6 17.2 27.5 23. 50. 17.9 9.6 17.2 22.5 21.4 12. 19.9 19.4 13.4 18.2 24.6 21.1 24.7 8.7 27.5 20.7 36.2 31.6 11.7 39.8 13.9 21.8 23.7 17.6 24.4 8.8 19.2 25.3 20.4 23.1 37.9 15.6 45.4 15.7 22.6 14.5 18.7 17.8 16.1 20.6 31.6 29.1 15.6 17.5 22.5 19.4 19.3 8.5 20.6 17. 17.1 14.5 50. 14.3 12.6 28.7 21.2 19.3 23.1 19.1 25. 33.4 5. 29.6 18.7 21.7 23.1 22.8 21. 48.8]
```

#print the predicted price and actual price of houses from the testing data set row 0 y\_pred[0] 23.59999999999994

y\_test[0] 24.0

#### Conclusion:

By looking at the predicted values that the model came up with, and the actual values of the testing data set, it looks like the model is pretty good at making predictions. It's not exact, but it is pretty close and much better than guessing.

To check the model's performance/accuracy I will use a metric called mean squared error (MSE)

print(np.mean((y\_pred-y\_test)\*\*2)) 2.967534119029605e-28

## Conclusion:

The mean squared error function returns a value of about 20.72. This is really good! If our model had predicted the exact value as the actual values then the mean squared error function would have returned a value of 0.