



Velammal Institute of Technology

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PREDICTING HOUSE PRICES USING MACHINE LEARNING

IBM GROUP 1 ARTIFICIAL INTELLIGENCE – PHASE 3

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EXPLANATION OF THE OUTPUT RESULTS AND THE DATASET

First we import a sample data from scleral library , you can get different types of sample data from Kaggle. The data taken here is the data of various parameters and the house prices in a given city called Boston in the year between 1970 to 2020.

Here the data parameters are explained as follows:

	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	PTRATIO	B	LSTAT	price
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	24.0
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	21.6
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	34.7
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	33.4
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	36.2

- Here the data parameters are explained as follows:

1. CRIM	per capita crime rate by town
2. ZN	proportion of residential land zoned for lots over 25,000 sq.ft.
3. INDUS	proportion of non-retail business acres per town
4. CHAS	Charles River dummy variable (= 1 if tract bounds river; 0 otherwise)
5. NOX	nitric oxides concentration (parts per 10 million)
6. RM	average number of rooms per dwelling
7. AGE	proportion of owner-occupied units built prior to 1940
8. DIS	weighted distances to five Boston employment centres
9. RAD	index of accessibility to radial highways
10. TAX	full-value property-tax rate per \$10,000
11. PTRATIO	pupil-teacher ratio by town
12. B	$1000(B_k - 0.63)^2$ where B_k is the proportion of blacks by town
13. LSTAT	% lower status of the population
14. MEDV	Median value of owner-occupied homes in \$1000's

- Here for understanding purpose we have taken first 5 index/instance of data and printed them.
- In total there are 506 rows of data from the dataset , of which we have printed first 5 rows using head() function.
- There are 14 columns in total, i.e., 13 columns containing data of the place, and the 14th column is the target column which contains the house prices.
- Then we check if our data has some null values i.e. missing values. Since if the data is incomplete , then there will be error during processing state which may lead to loss of accuracy in predicting model. Here in our given data , there is no missing value as we can see.

CRIM	0
ZN	0
INDUS	0
CHAS	0
NOX	0
RM	0
AGE	0
DIS	0
RAD	0
TAX	0
PTRATIO	0
B	0
LSTAT	0
price	0
dtype:	int64

- Since our data contains no missing value, the program will skip the dropping phase in data processing, where data is dropped to increase accuracy and fit missing values in a way so that it is suitable for modeling.

- Next we try to describe the data in such a way so that both people and machine find it easy to understand the given data . In order to do this we use the describe() function.

	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	PTRATIO	B	LSTAT	price
count	506.000000	506.000000	506.000000	506.000000	506.000000	506.000000	506.000000	506.000000	506.000000	506.000000	506.000000	506.000000	506.000000	506.000000
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	12.653063	22.532806
std	8.601545	23.322453	6.860353	0.253994	0.115878	0.702617	28.148861	2.105710	8.707259	168.537116	2.164946	91.294864	7.141062	9.197104
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	1.730000	5.000000
25%	0.082045	0.000000	5.190000	0.000000	0.449000	5.885500	45.025000	2.100175	4.000000	279.000000	17.400000	375.377500	6.950000	17.025000
50%	0.256510	0.000000	9.690000	0.000000	0.538000	6.208500	77.500000	3.207450	5.000000	330.000000	19.050000	391.440000	11.360000	21.200000
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

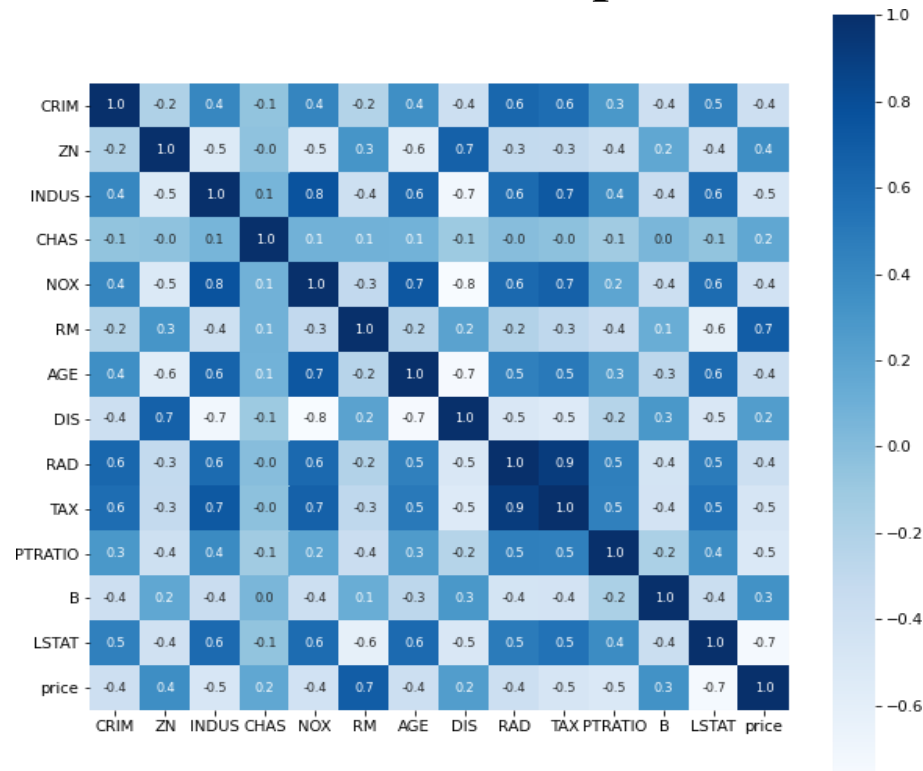
- Counts refers to the number of instances of data in each column i.e 506 since there are 506 rows of data for each column Mean refers to mean value of data in given column.
- Std means the standard value the most common value in given set of data for a particular column.
- Min refers the least data value in each column.
- Max refers to the maximum data value in each column.
- 25% refers that 25 percentile of the data in that column is equal to or below that value.
- Next we try to understand the correlation between the different values, in order to do that, the best way is by using heat map. Heat map is a representation of data in the form of a map or diagram in which data values are represented as colours.

- **Correlation is a statistical measure that expresses the extent to which two variables are linearly related (meaning they change together at a constant rate)**
- **There are two types of correlation, they are:**

- **Positive correlation:** A positive correlation is a relationship between two variables that move in tandem—that is, in the same direction. A positive correlation exists when one variable decreases as the other variable decreases, or one variable increases while the other increases.
- **Negative correlation:** Negative correlation is a relationship between two variables in which one variable increases as the other decreases, and vice versa.

- In statistics, a perfect negative correlation is represented by the value -1.0 , while a 0 indicates no correlation, and $+1.0$ indicates a perfect positive correlation.
- A perfect negative correlation means the relationship that exists between two variables is exactly opposite all of the time.
- These two types of correlation are represented numerically and as well as by shade of color in the heat map.

- HEATMAP – for better understanding of which place is best suited for individual personal preference based on given dataset. This uses correlation concept.



- Next we split our data into variables x and y , in order to train our model to predict data:

```

      CRIM      ZN      INDUS      CHAS      NOX      RM      AGE      DIS      RAD      TAX \
0      0.00632    18.0      2.31      0.0      0.538    6.575    65.2    4.0900    1.0    296.0
1      0.02731      0.0      7.07      0.0      0.469    6.421    78.9    4.9671    2.0    242.0
2      0.02729      0.0      7.07      0.0      0.469    7.185    61.1    4.9671    2.0    242.0
3      0.03237      0.0      2.18      0.0      0.458    6.998    45.8    6.0622    3.0    222.0
4      0.06905      0.0      2.18      0.0      0.458    7.147    54.2    6.0622    3.0    222.0
...
501     0.06263      0.0     11.93      0.0      0.573    6.593    69.1    2.4786    1.0    273.0
502     0.04527      0.0     11.93      0.0      0.573    6.120    76.7    2.2875    1.0    273.0
503     0.06076      0.0     11.93      0.0      0.573    6.976    91.0    2.1675    1.0    273.0
504     0.10959      0.0     11.93      0.0      0.573    6.794    89.3    2.3889    1.0    273.0
505     0.04741      0.0     11.93      0.0      0.573    6.030    80.8    2.5050    1.0    273.0

      PTRATIO      B      LSTAT
0      15.3    396.90      4.98
1      17.8    396.90      9.14
2      17.8    392.83      4.03
3      18.7    394.63      2.94
4      18.7    396.90      5.33
...
501     21.0    391.99      9.67
502     21.0    396.90      9.08
503     21.0    396.90      5.64
504     21.0    393.45      6.48
505     21.0    396.90      7.88

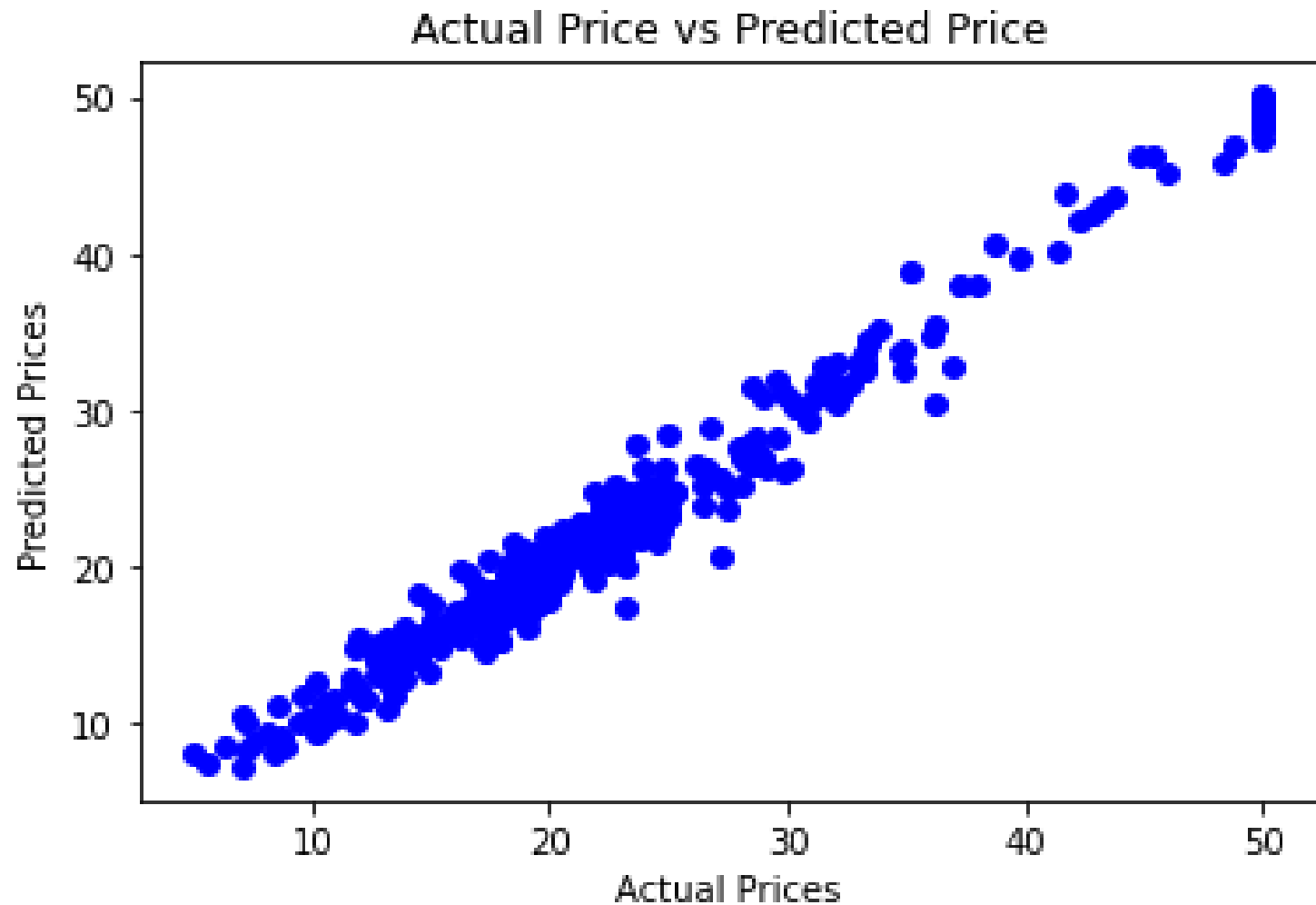
[506 rows x 13 columns]
0      24.0
1      21.6
2      34.7
3      33.4
4      36.2
...
501     22.4
502     20.6
503     23.9
504     22.0
505     11.9
Name: price, Length: 506, dtype: float64

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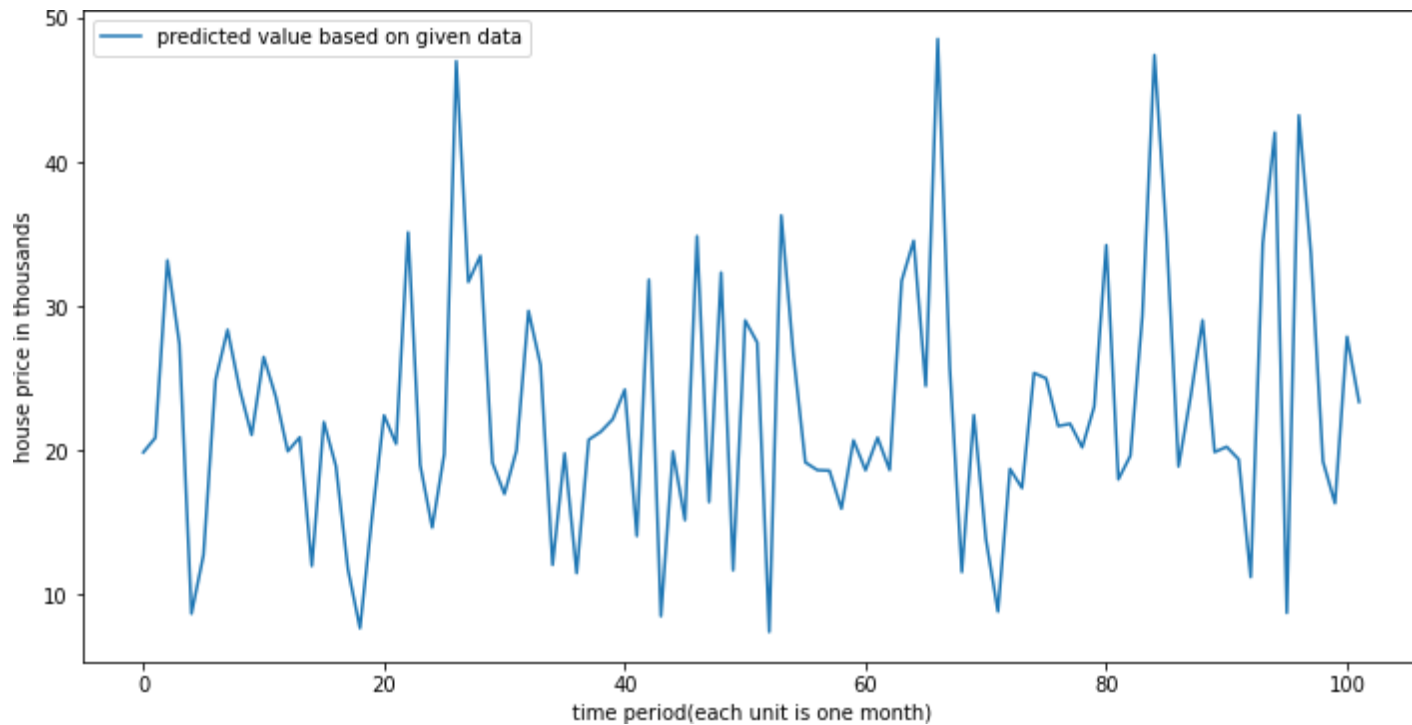

- Here the variable x contains the value of the first 13 columns i.e the parameters that are required for calculating and predicting the house prices. The variable y contains the 14th column values which are the house prices.
- First we predict the values in y using the values in x . Then we compare the actual prices and predicted prices by using scatter plot. Then we find the r square error and mean square error between them . If the errors is less enough then we proceed for testing of the model since the training phase is over. If the error is large , then we use optimizers like adam, and repeat drop and fitting process for a set number of epochs to reduce the error.
- The r square error or mean square error for good accuracy of the model in predicting the data is indicated numerically also.

- A model is good if these error values are less than 5.
- Then during testing process we predict the future house prices using present and past data parameters of houses in an location. Then we plot this graphically as a house price over time graph.
- For training the model , the error needs to be minimum for greater accuracy of model. The error between the actual and predicted price is plotted graphically using scatter plot. Here we can see that error is minimum since the data points of actual and predicted value are close to each other

Graphical Analysis



PREDICTED VALUE OF HOUSE PRICE BASED ON TEST SAMPLE DATA



THANK YOU