

# Assignment 1 : Linear Regression

Linear regression is based on the relationship of dependent and independent variables. Using linear regression we can predict the dependent variable using the independent variable by using  $Y=MX+C$ .

## Part 1 : Linear Regression using Pseudoinverse

In this we use the formula weights  $m = (X^T X)^{-1} (X^T) Y$  also known as pseudo inverse.

In this I have used the data set which is provided by sklearn which is of **Boston house price prediction**. This can be directly downloaded from sklearn library also for calculating the inverse I have used the PINV function provided by numpy library.

To run the Program, first upload the .ipynb file.

Description of Dataset:

1. Experimental data used for **Prediction** (house price in boston) from different independent features.
2. This dataset contains 13 features and a target variable: 'CRIM' 'ZN' 'INDUS' 'CHAS' 'NOX' 'RM' 'AGE' 'DIS' 'RAD' 'TAX' 'PTRATIO' 'B' 'LSTAT'
3. Target Variable: price or target
4. This Dataset contains 506 rows and 14 columns with 13 features and 1 target.
5. No missing values.

I have used the LSTAT feature as an independent feature for prediction of target; any other feature can also be used. All the required steps are already written in the .ipynb file, Different Splits are also Provided; minimum mean squared error is found as 13%.

This shows pseudo inverse is not a good fit for the predictions.

## Part 1 : Linear Regression using gradient descent

In this we use the formula weights  $Y=MX+C$  and optimize the weight  $w$  or  $m$  using the gradients of  $M$  and  $C$ .

In this I have used the data set which is provided by sklearn which is of **Boston house price prediction**. This can be directly downloaded/loaded from sklearn library also for calculating the inverse I have used the PINV function provided by numpy library.

To run the Program, first upload the .ipynb file.

Description of Dataset:

6. Experimental data used for **Prediction** (house price in boston) from different independent features.
7. This dataset contains 13 features and a target variable: 'CRIM' 'ZN' 'INDUS' 'CHAS' 'NOX' 'RM' 'AGE' 'DIS' 'RAD' 'TAX' 'PTRATIO' 'B' 'LSTAT'
8. Target Variable: price or target
9. This Dataset contains 506 rows and 14 columns with 13 features and 1 target.
10. No missing values.

I have used the LSTAT feature as an independent feature for prediction of target; any other feature can also be used. All the required steps are already written in the .ipynb file, Different Splits are also Provided; minimum mean squared error is found as 1%.

This shows that gradient descent is best fit for predictions.