Before getting into linear Regression, firstly let us understand the perspective in which data can be understood.

1. Statistical Perspective of data

output=funtion(input)

Or

DependentVariable=f(IndependentVariable)

2. Algorithmic view

Model=Algorithm(Data)

(Model: Specific Patterns Learned from Data

Algorithm: The process of learning/understanding Data)

So, here our model is Linear Regression and we write a certain algorithm from scratch which predicts outcomes for given specific inputs.

## LINEAR REGRESSION:

- 1. This examines the relationship between two variables by determining the line of best fit
- 2. The essential requirements for this fit are
  - Slope: m
  - Intercept: c

The equation of linear regression can be written as such,

$$y = m.x + c$$

(where n= number of values in x or y)

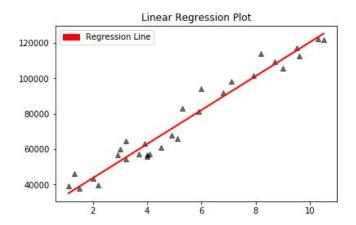
The slope of the given regression line can be determined by the formula,

$$m = (n.\sum_{i}^{n} x.y - \sum_{i}^{n} x.\sum_{i}^{n} y)/(n.\sum_{i}^{n} x^{2} - (\sum_{i}^{n} x)^{2})$$

(where n = number of values in x or y)

The intercept of the given regression line can be determined by the formula,

$$c = \overline{y} - m.\overline{x}$$



So, after the prediction, we need to measure the goodness of fit and for that, we use regression metric named coefficient of determination or R-squared measure.

## R-Squared(Coefficient of determination):

• If our dataset has n values marked  $y_1, y_2, ..., y_i$  and for that values resulted in certain predicted values  $f_1, f_2, ..., f_i$ 

So, 
$$\overline{y} = \frac{1}{n} \sum_{i=1}^{n} y_i$$

Let us calculate the sum of squares,

$$SS_{tot} = \sum_{i=1}^{n} (y_i - \overline{y})^2$$

Next, calculate the regression sum of squares,

$$SS_{reg} = \sum_{i=1}^{n} (f_i - \overline{y})^2$$

Next, calculate the residual sum of squares,

$$SS_{res} = \sum_{i=1}^{n} (y_i - f_i)^2$$

Finally, calculate the coefficient of determination,

$$R^2 = 1 - \left(\frac{SS_{reg}}{SS_{tot}}\right)$$

It provides a measure of how well-observed outcomes are replicated by the model, based on the proportion of total variation of outcomes explained by the model.

