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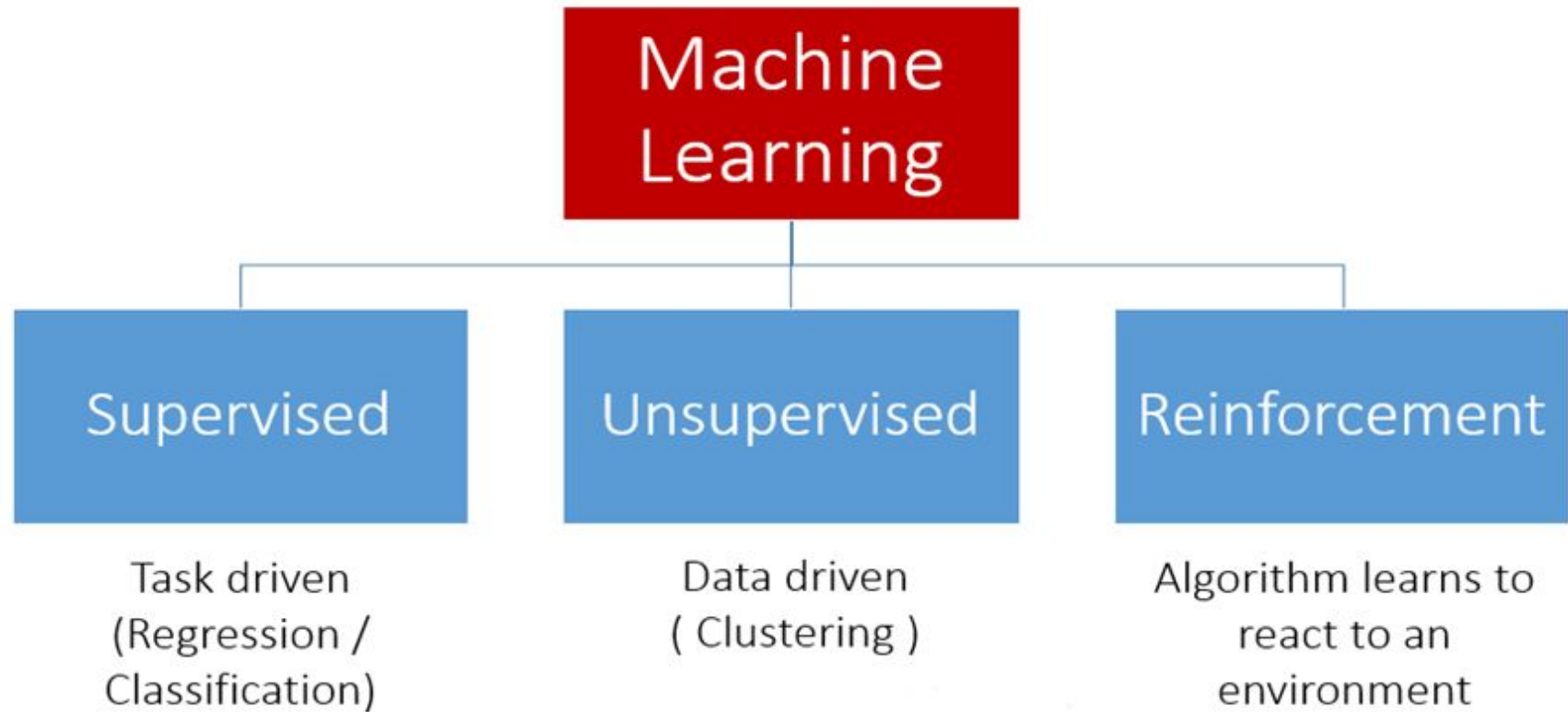
# Linear Regression

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# Types of Machine Learning



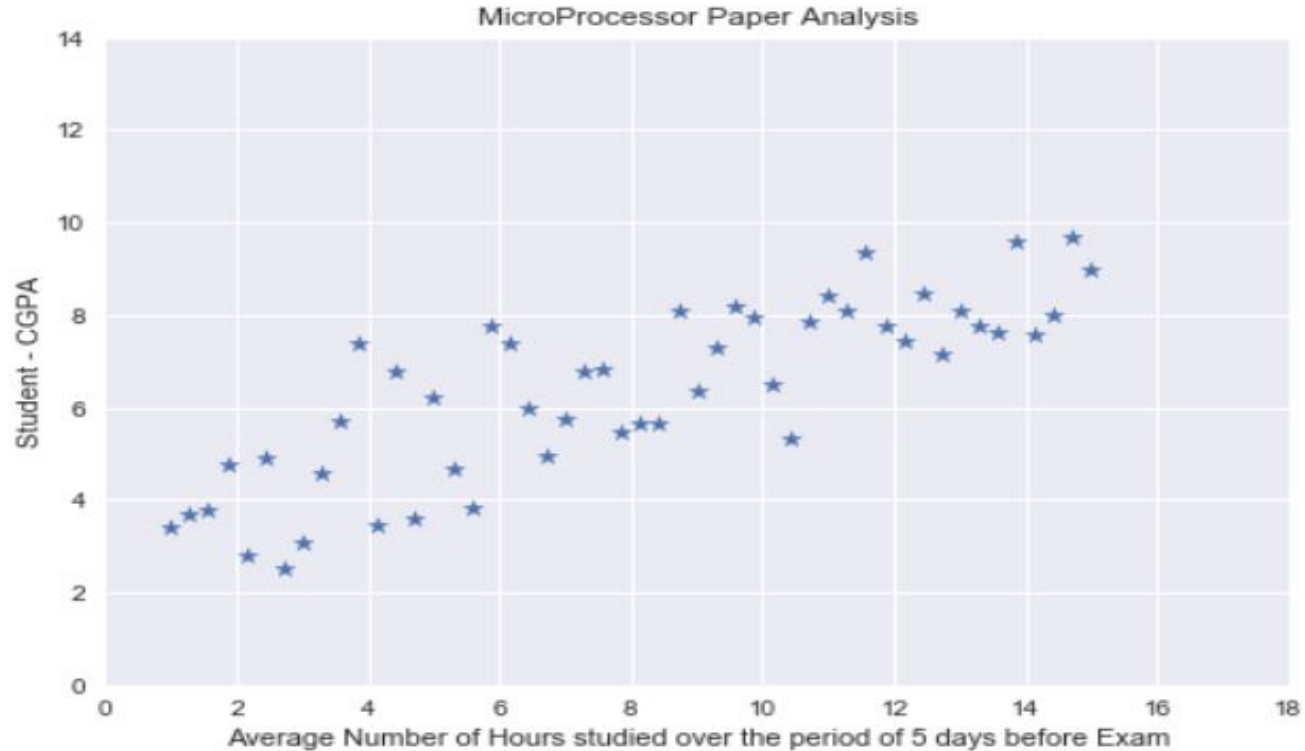
# Univariate Linear Regression

*Univariate linear regression focuses on determining relationship between one independent variable( $X$ ) and one dependent variable( $Y$ ).*

<b>X(CGPA)</b>	<b>Label(Y) (Number of hrs. study)</b>
7.8	6.6
9.4	12
8.8	10
8.3	9.5
7.3	6
.....	.....

## Dataset - How univariate data Looks like

### Example : - CGPA Vs Hours of Study (Not Attendance)



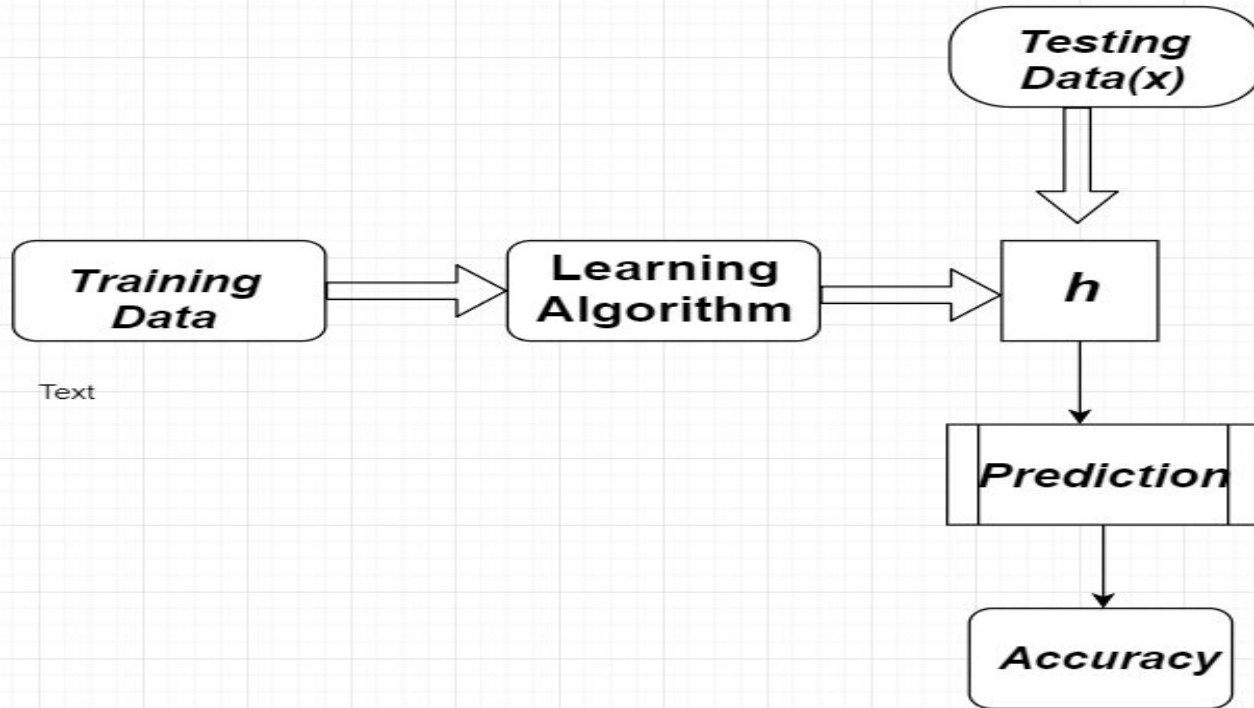
# Dataset Common Terminology

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1. No. of Samples( $m$ )
  2. Input Features( $n$ )
  3. Training set
  4. Test set
  5. Hypothesis
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# Aim and working (Pipeline)

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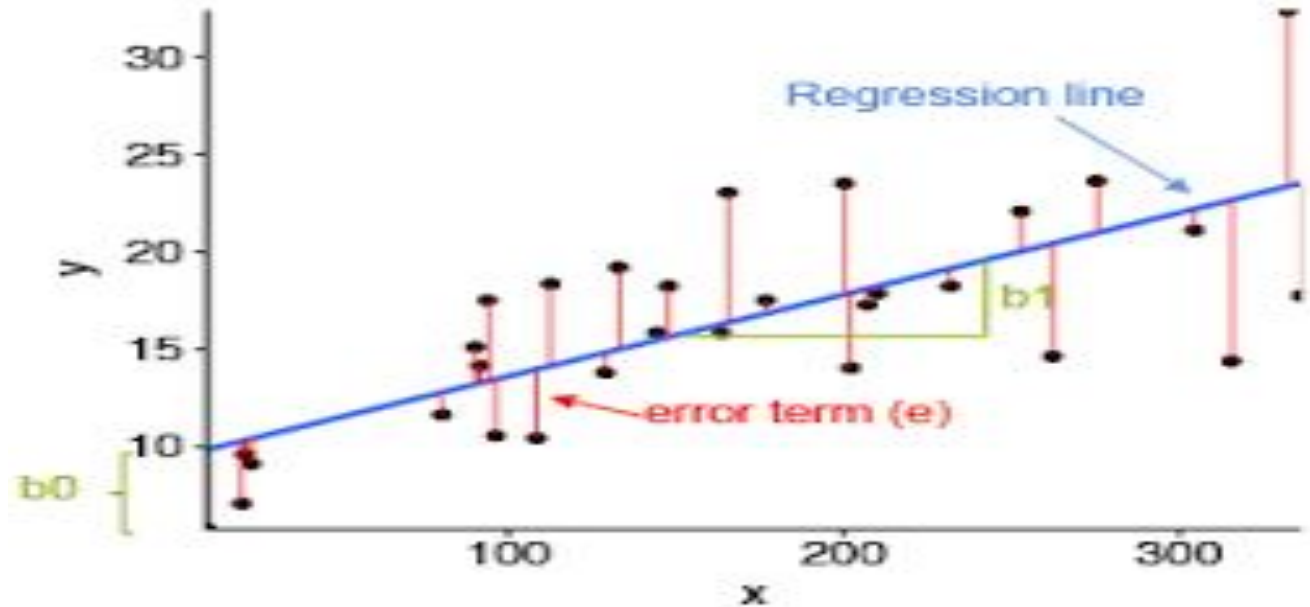
# Which line seems right ?

Hypothesis:  $h_{\theta}(x) = \theta_0 + \theta_1 x$



Q - How to find parameters(hypothesis) ?

Q - What is the measure of finding Best line ?

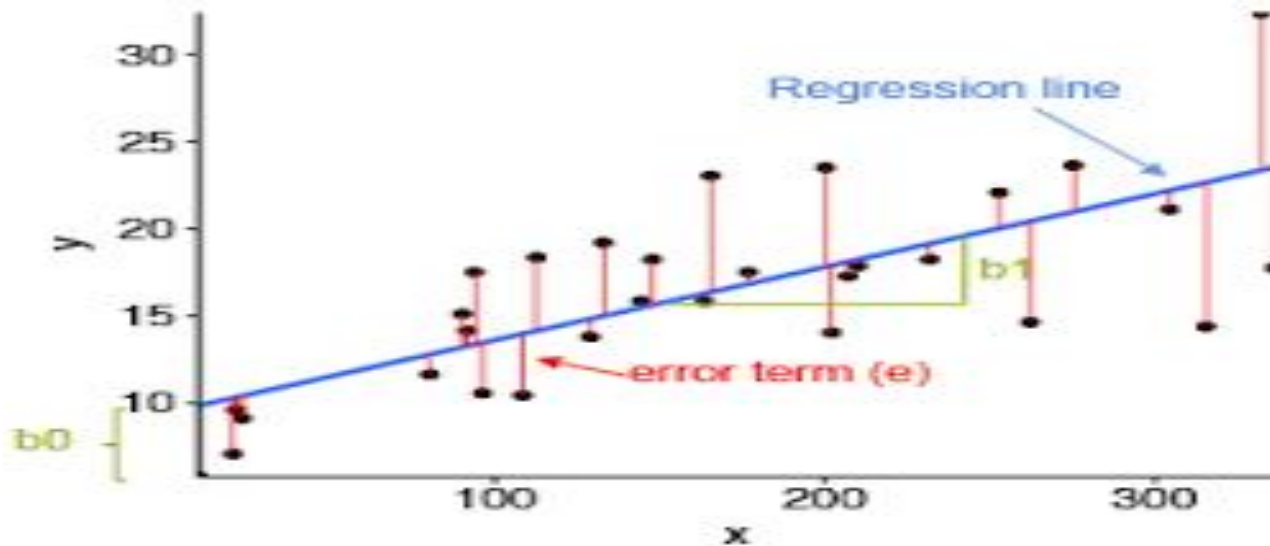


Ans - Error Function (How?)

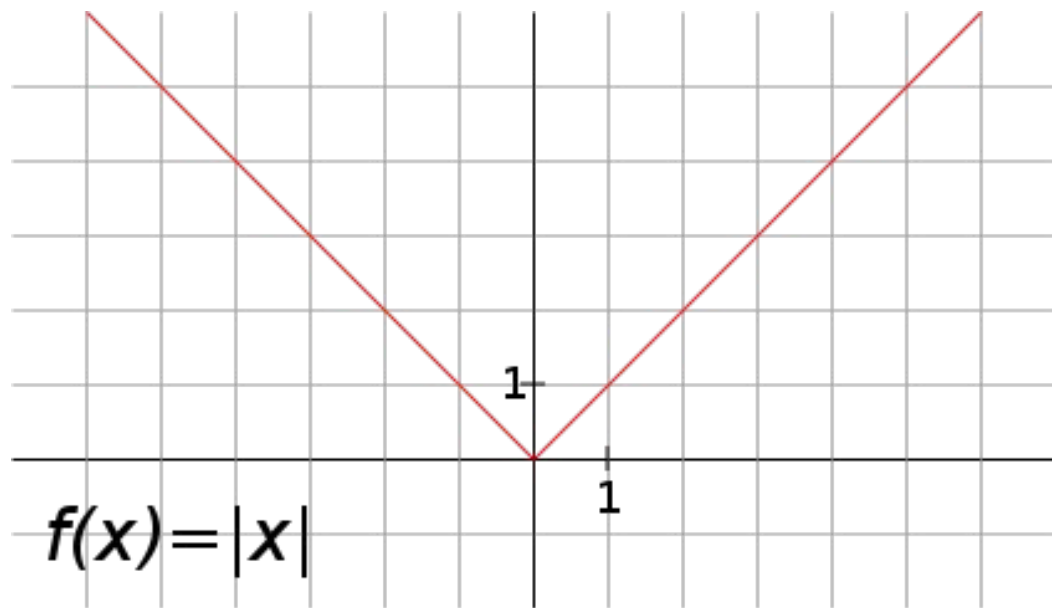


# Choice of Error Function

1.  $\sum (Y_{pred} - Y_{actual})$  -ve and +ve error added up



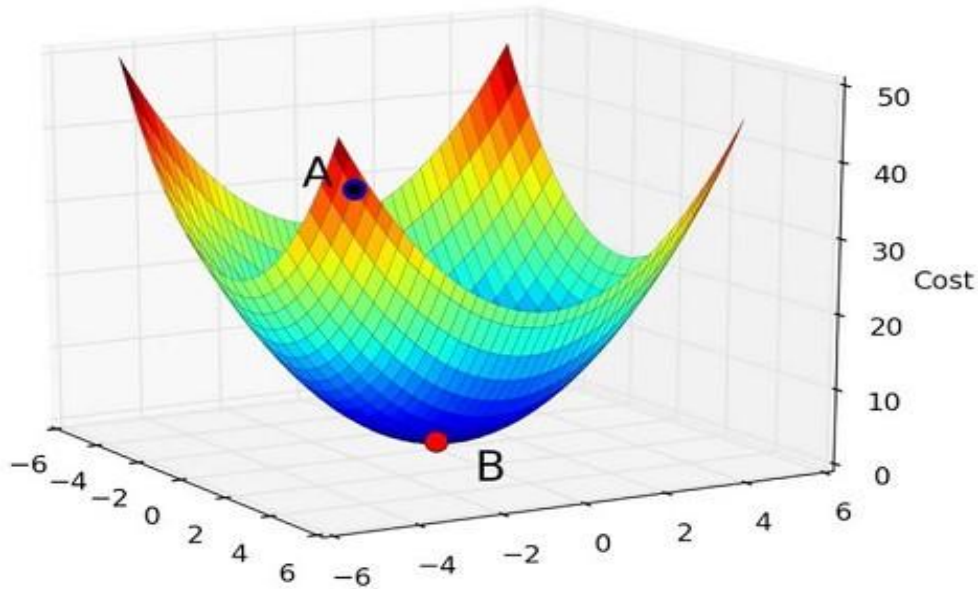
2.  $\sum |Y_{pred} - Y_{actual}|$  Non-differentiable at  $(Y_{pred} == Y_{actual})$



3.

$$\sum (Y_{pred} - Y_{actual})^2$$

Perfect, and this function is Convex in Nature(Helpful)

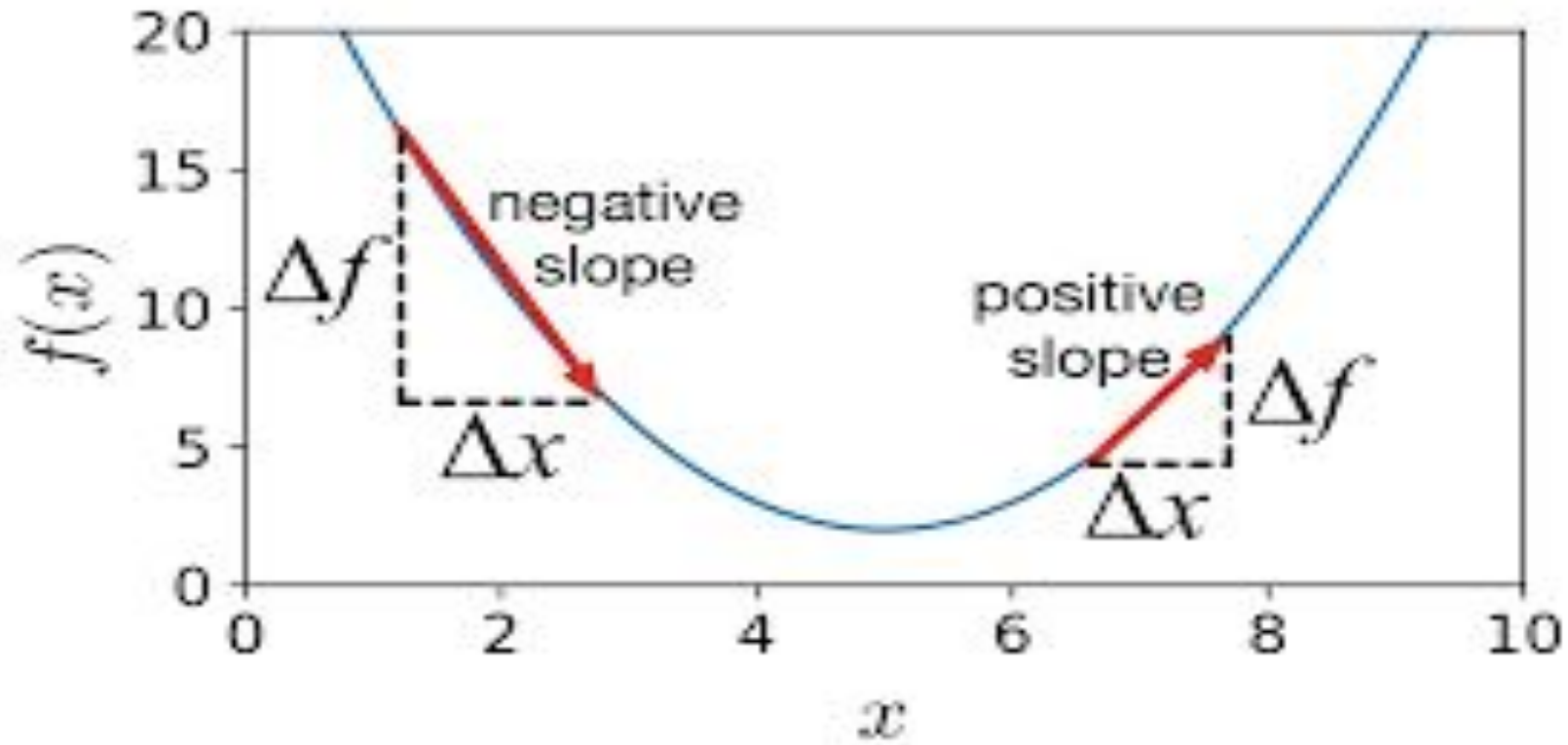


# Goal - Minimizing the Error Function

## Gradient Descent

1. *An Iterative method to move toward the minimum point(using Gradient).*
2. **Gradient descent** is an optimization algorithm used to minimize some function by iteratively moving in the direction of **steepest descent** as defined by the negative of the **gradient**. In machine learning, we use **gradient descent** to update the parameters of our model.

# How we got the formulas of Gradient Descent ?



# Update Rule

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Hypothesis:  $h_{\theta}(x) = \theta_0 + \theta_1 x$

repeat until convergence {  
     $\theta_j := \theta_j - \alpha \frac{\partial}{\partial \theta_j} J(\theta_0, \theta_1)$   
    (for  $j = 1$  and  $j = 0$ )  
}

# Derivative Derivation

repeat until convergence {

$$\theta_0 := \theta_0 - \alpha \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)})$$

$$\theta_1 := \theta_1 - \alpha \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)}) \cdot x^{(i)}$$

}

# Convergence Criteria

- Number of Iteration
- Change in Error



## Let's See the code (Working!)

- Visualizing Line(Hypothesis)
- Behaviour of Error Function using no. of iteration, change in error.
- Variation of Learning rate
- Visualizing Convex Function