

Linear Regression:

Objective: to find/ predict the continuous value

TO MAKE THE BEST FITT LINE IN SUCH A WAY TO MINIMIZE ERRORS

Equation of line (simple linear regression):

$$\hat{Y} = W \cdot X + b$$

\hat{Y} : predicted value

X : input

W : Weight/ slope

b : Bias/ intercept

- **Hypothesis Function** : This is our prediction function:

$$h(x) = w \cdot x + b$$

- **Cost Function (Error Measure)** :

Measures how far predictions are from actual values.

The image shows handwritten notes on a blackboard. At the top, it says "Solve { Cost function }". Below this, it says "Minimize $\sum_{i=1}^m \frac{1}{2m} (h_{\theta}(x^{(i)}) - y^{(i)})^2$ ". To the right, there is a derivative calculation: $\frac{\partial}{\partial x} (x^2) = 2x^{2-1} = 2x$. At the bottom, the cost function is boxed: $J(\theta_0, \theta_1) = \frac{1}{2m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)})^2$. Below the box, it says "Squared Error Function".

Goal: **Minimize this cost** to improve model accuracy.

Gradient Descent (Optimization Algorithm) :

- It updates weights to minimize cost function.

5. Learning Rate (α)

- Controls **how fast** weights are updated.
- Small α \rightarrow slow learning
- Large α \rightarrow may overshoot or diverge

Right learning rate = faster and stable training

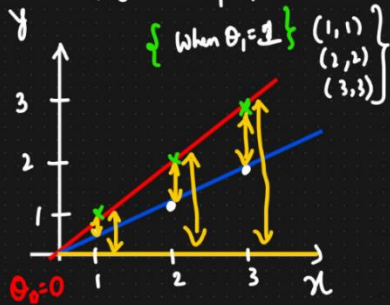
6. Convergence and Global Minima

- Linear regression has **convex cost function**, so only **one global minima** exists.
- Gradient Descent **always reaches the best point** if:
 - Learning rate is proper
 - Sufficient iterations done

Hypothesis

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

$$\text{lets } \theta_0 = 0$$



$$h_{\theta}(x) = 1 \times 1$$

$$h_{\theta}(x) = 1 \times 2$$

$$h_{\theta}(x) = 1 \times 3$$

$$\text{When } \theta_1 = 0.5$$

$$\text{When } \theta_1 = 0.0$$

Cost function

$$J(\theta_0, \theta_1) = \frac{1}{2m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)})^2$$

