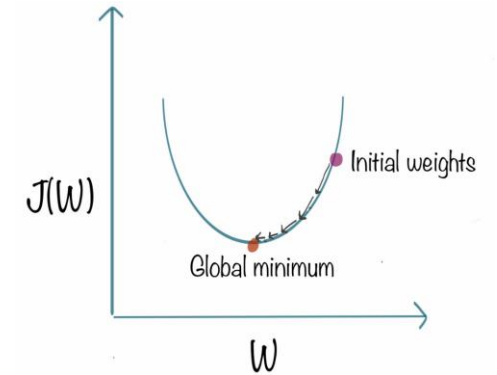


Siddhardhan

# Gradient Descent for Logistic Regression



# Logistic Regression

## About Logistic Regression:



1. Supervised Learning Model
2. Classification model
3. Best for Binary Classification Problem
4. Uses Sigmoid function
5. Binary Cross Entropy Loss Function (or) Log Loss

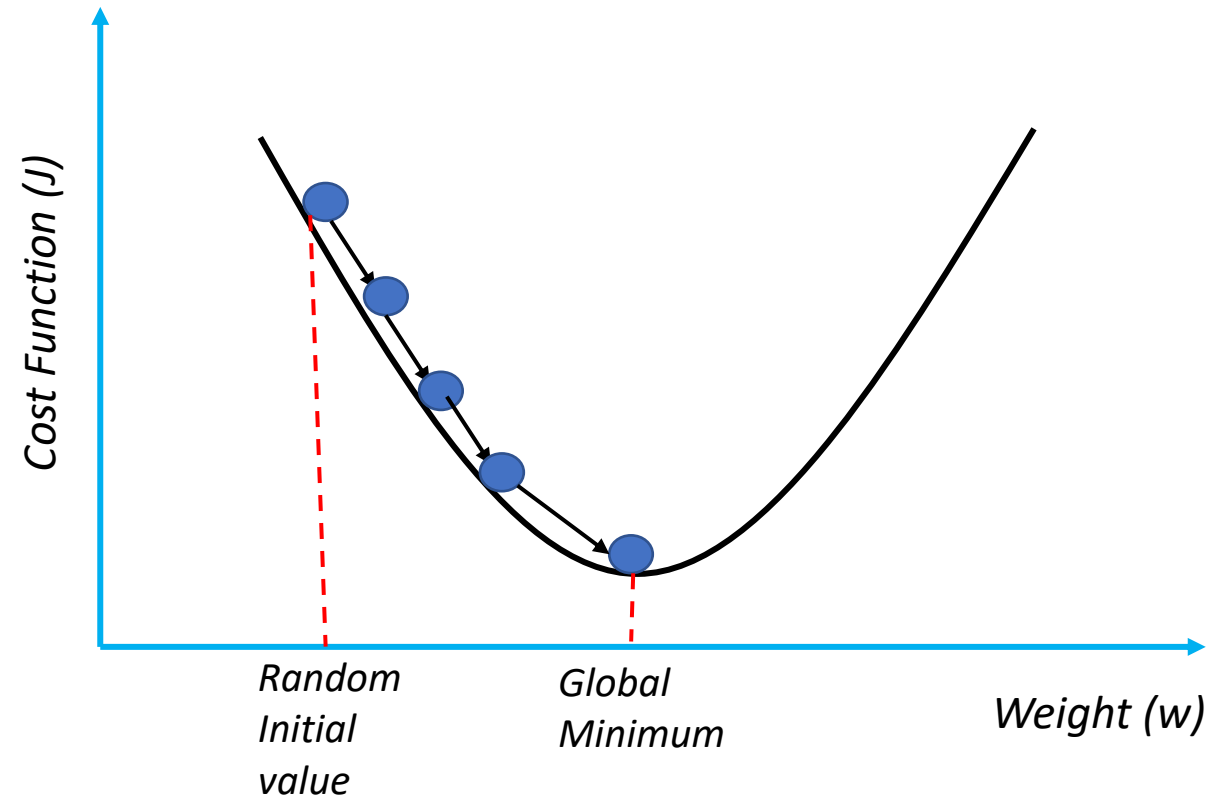
$$\hat{Y} = \frac{1}{1 + e^{-Z}}$$

$$Z = w.X + b$$

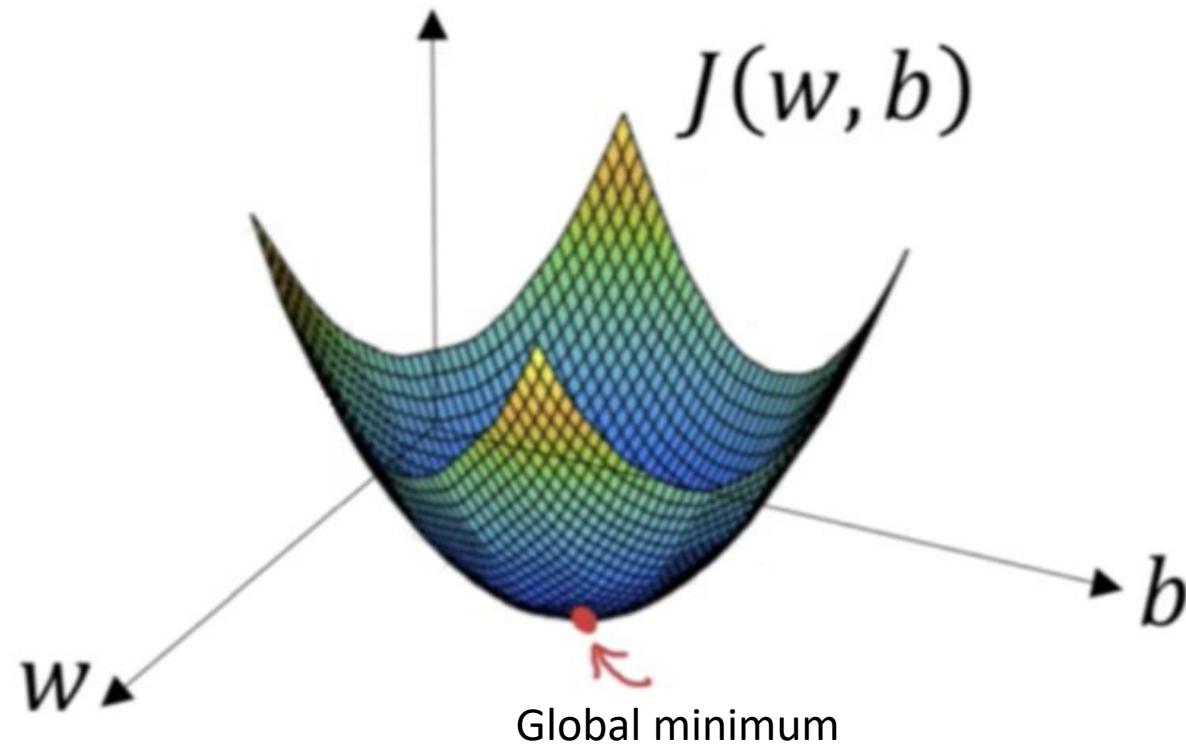
*Sigmoid Function*

$$J(w, b) = \frac{1}{m} \sum (L(y^{(i)}, \hat{y}^{(i)})) = - \frac{1}{m} \sum (y^{(i)} \log \hat{y}^{(i)} + (1 - y^{(i)}) \log (1 - \hat{y}^{(i)}))$$

# Gradient Descent



## Gradient Descent in 3 Dimension



# Gradient Descent

Gradient Descent is an optimization algorithm used for minimizing the cost function in various machine learning algorithms. It is used for updating the parameters of the learning model.

$$w_2 = w_1 - L * dw$$

$$b_2 = b_1 - L * db$$

w --> weight

b --> bias

L --> Learning Rate

dw --> Partial Derivative of cost function with respect to w

db --> Partial Derivative of cost function with respect to b

$$dw = \frac{1}{m} * ( \hat{Y} - Y ). X$$

$$db = \frac{1}{m} * ( \hat{Y} - Y )$$

# Logistic Regression

## ***Logistic Regression model:***

❖ Sigmoid Function

$$\hat{Y} = \frac{1}{1 + e^{-Z}} \quad Z = w.X + b$$

❖ Updating weights  
through Gradient Descent

$$w_2 = w_1 - L * dw$$

❖ Derivatives

$$b_2 = b_1 - L * db$$

$$dw = \frac{1}{m} * ( \hat{Y} - Y ).X$$

$$db = \frac{1}{m} * ( \hat{Y} - Y )$$