

Machine Learning Processes

Ali Bidaran

Introduction

Preprocessing

Training

Preprocessing

Reducing memory of RAM and dataset.

Improving speed and Performance during Training model

MinMaxScaler

Standard Scaler

Normalization

Training

Hypothesis: $h_{\theta}(x) = \theta^T x = \theta_0 x_0 + \theta_1 x_1 + \theta_2 x_2 + \dots + \theta_n x_n$

Handwritten notes: $x_0 = 1$ (green arrow pointing to x_0), θ (green circle under θ), $n+1$ -dimensional vector (green text)

Parameters: $\theta_0, \theta_1, \dots, \theta_n$

Handwritten notes: θ (green circle under θ), $n+1$ -dimensional vector (green text)

Cost function:

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

Handwritten notes: $J(\theta)$ (green text under $J(\theta_0, \theta_1, \dots, \theta_n)$)

Gradient descent:

Repeat {

$\rightarrow \theta_j := \theta_j - \alpha \left[\frac{\partial}{\partial \theta_j} J(\theta_0, \dots, \theta_n) \right]$

Handwritten notes: $J(\theta)$ (green text under $J(\theta_0, \dots, \theta_n)$), θ (green circle under θ)

}

(simultaneously update for every $j = 0, \dots, n$)

Gradient Descent

Previously ($n=1$):

Repeat {

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

→ $\theta_1 := \theta_1 - \alpha \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)}) \underline{x_1^{(i)}}$
(simultaneously update θ_0, θ_1)

}

New algorithm ($n \geq 1$):

Repeat {

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

(simultaneously update θ_j for $j = 0, \dots, n$)

}

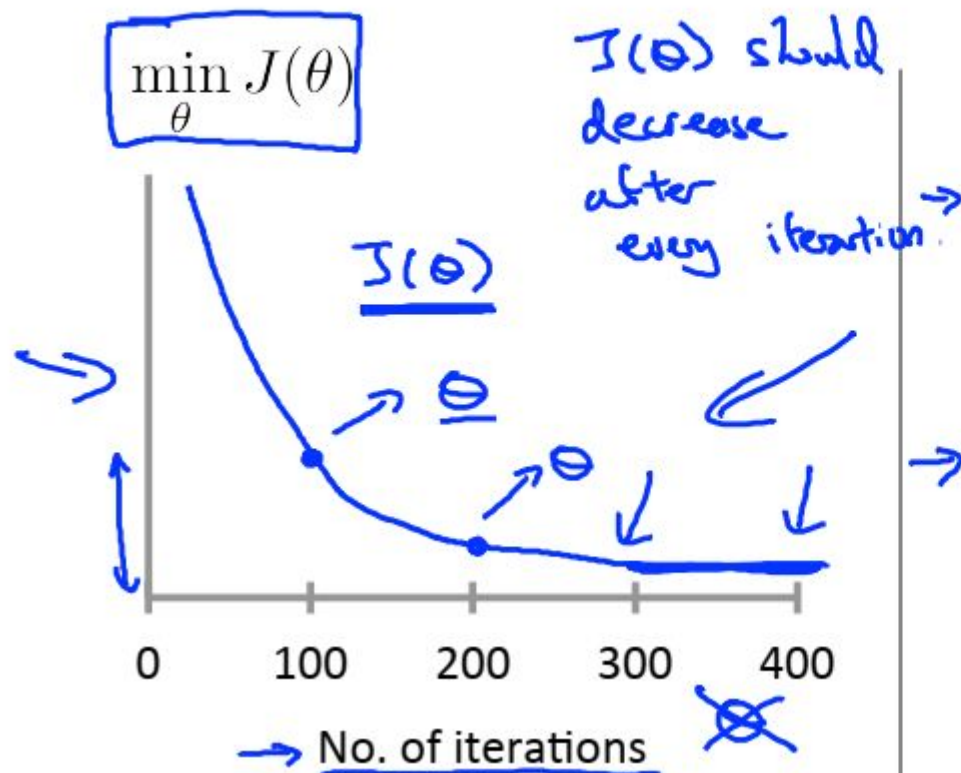
→ $\theta_0 := \theta_0 - \alpha \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)}) \underline{x_0^{(i)}}$

→ $\theta_1 := \theta_1 - \alpha \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)}) \underline{x_1^{(i)}}$

→ $\theta_2 := \theta_2 - \alpha \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)}) \underline{x_2^{(i)}}$

...

Making sure gradient descent is working correctly.



→ Example automatic convergence test:

→ Declare convergence if $J(\theta)$ decreases by less than 10^{-3} in one iteration.

—m