

repeat {

$$\theta_j := \theta_j - \alpha \frac{\partial}{\partial \theta_j} J(\theta_0, \theta_1)$$

$j = \{0, 1\}$

α is very small

learning will be done slowly



$$\alpha = 0.1$$

α is very high

may overshoot the minima
may fail to converge, even diverge



The purpose of GD is to optimize the cost func associated with linear reg.

$$\text{Cost } f: J_{\theta} = \frac{1}{2n} \sum_{i=1}^n (\hat{y}_i - y_i)^2$$

where $\hat{y} = h_{\theta}(x_i) = \theta_0 + \theta_1 x_i$ (for single var. lin. reg)

$$\begin{aligned} \frac{\text{Aim}}{\frac{\partial}{\partial \theta_0}} J(\theta_0, \theta_1) &= \frac{\partial}{\partial \theta_0} \left[\frac{1}{2n} \sum_{i=1}^n (\theta_0 + \theta_1 x_i - y_i)^2 \right] \\ &= \frac{1}{n} \sum_{i=1}^n (\theta_0 + \theta_1 x_i - y_i) \leftarrow \end{aligned}$$

$$\frac{\partial}{\partial \theta_1} J(\theta_0, \theta_1) = \frac{1}{n} \sum_{i=1}^n (\theta_0 + \theta_1 x_i - y_i) x_i$$

GD for single variable

$$\text{Repeat} \left\{ \begin{array}{l} \theta_0 := \theta_0 - \alpha \frac{1}{n} \sum_{i=1}^n (\theta_0 + \theta_1 x_i - y_i) \end{array} \right.$$

$$\left. \begin{array}{l} \theta_1 = \theta_1 - \alpha \frac{1}{n} \sum_{i=1}^n (\theta_0 + \theta_1 x_i - y_i) x_i \end{array} \right\}$$

for multiple variable

Repeat ()

$$\left\{ \begin{array}{l} \theta_j := \theta_j - \alpha \frac{1}{n} \sum_{i=1}^n (\theta_0 + \theta_1 x_j^i - y^i) x_j^i \\ x_0^i = 1 \quad ; \quad \forall j \in \{0, 1, 2, \dots, k\} \end{array} \right.$$

Training Set

$\{x^i, y^i\}$

Cost function
 J

find θ which minimizes
cost on training data



Termination Criteria in GD

- 1) fixed # of iteration
- 2) min gradient is reached
- ⋮

Polynomial Regression

Data Mining reference book: han, kamber, Pei

