

Derivation of Stochastic Gradient Descent

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We want to prove:

$$\begin{aligned}\nabla l(\theta) &= \sum_{i=0}^n [y^{(i)} \frac{1}{h_{\theta}(x^{(i)})} (\frac{\partial}{\partial \theta_j} h_{\theta}(x^{(i)})) + (1 - y^{(i)}) \frac{-1}{(1-h_{\theta}(x^{(i)}))} (\frac{\partial}{\partial \theta_j} h_{\theta}(x^{(i)}))] \\ &= \sum_{i=0}^n [(y^{(i)} - h_{\theta}(x^{(i)})) x_j^{(i)}]\end{aligned}$$

$$\text{we know } l(\theta) = \sum_{i=0}^n [y^{(i)} \log h_{\theta}(x^{(i)}) + (1 - y^{(i)}) \log(1 - h_{\theta}(x^{(i)}))]$$

$$\text{So, } \nabla l(\theta) = \sum_{i=0}^n [y^{(i)} \frac{1}{h_{\theta}(x^{(i)})} (\frac{\partial}{\partial \theta_j} h_{\theta}(x^{(i)})) + (1 - y^{(i)}) \frac{-1}{(1-h_{\theta}(x^{(i)}))} (\frac{\partial}{\partial \theta_j} h_{\theta}(x^{(i)}))]$$

$$\text{since } h_{\theta}(x) = g(\theta^T x) = g(z) = \frac{1}{1+e^{-z}} \dots (1),$$

$$\implies g'(z) = \frac{-e^{-z}}{(1+e^{-z})^2} = \frac{1}{1+e^{-z}} \frac{-e^{-z}}{1+e^{-z}} = g(z)(1 - g(z)) \dots (2)$$

$$\text{Let } \frac{\partial z}{\partial \theta_j} = x_j,$$

$$\text{Then from equation (1) and (2), } \frac{\partial}{\partial \theta_j} h_{\theta}(x) = g'(z) \frac{\partial z}{\partial \theta_j} = g(z)(1 - g(z)) x_j$$

$$\implies \nabla l(\theta) = \sum_{i=0}^n [y^{(i)} \frac{1}{g(z)} (g(z)(1 - g(z)) x_j^{(i)}) + (1 - y^{(i)}) \frac{-1}{(1-g(z))} (g(z)(1 - g(z)) x_j^{(i)})]$$

$$\iff \sum_{i=0}^n [y^{(i)} (1 - g(z)) x_j^{(i)} - (1 - y^{(i)}) (g(z) x_j^{(i)})]$$

$$\Longleftrightarrow \sum_{i=0}^n [y^{(i)} x_j^{(i)} - y^{(i)} g(z) x_j^{(i)} - g(z) x_j^{(i)} + y^{(i)} (g(z) x_j^{(i)})]$$

$$\Longleftrightarrow \sum_{i=0}^n [y^{(i)} x_j^{(i)} - g(z) x_j^{(i)}] \text{ (After cancelling terms)}$$

$$\Longleftrightarrow \sum_{i=0}^n [(y^{(i)} - g(z)) x_j^{(i)}]$$

$$\Longleftrightarrow \sum_{i=0}^n [(y^{(i)} - h_{\theta}(x^{(i)})) x_j^{(i)}]$$

Now, we have proved that

$$\begin{aligned} \nabla l(\theta) &= \sum_{i=0}^n [y^{(i)} \frac{1}{h_{\theta}(x^{(i)})} (\frac{\partial}{\partial \theta_j} h_{\theta}(x^{(i)})) + (1 - y^{(i)}) \frac{-1}{(1 - h_{\theta}(x^{(i)}))} (\frac{\partial}{\partial \theta_j} h_{\theta}(x^{(i)}))] \\ &= \sum_{i=0}^n [(y^{(i)} - h_{\theta}(x^{(i)})) x_j^{(i)}] \end{aligned}$$