

$$z = -\log\left(\frac{1}{y} - 1\right) \quad e^{-z} = \frac{1}{y} - 1 \quad \cancel{e^{-z} = \frac{1-y}{1}} \quad h_\theta(x) = \frac{1}{1 + e^{-\theta^T x}} = \frac{1}{1 + e^{-z}}$$

$$e^{-z+1} = \frac{1}{y} \quad y = \frac{1}{e^{-z} + 1}$$

$z = \theta^T x$   
 $z = w_0 x_0 + \dots + w_n x_n$

$$J(\theta) = -\frac{1}{n} \sum_{i=1}^n \left[ y^{(i)} \log(h_\theta(x^{(i)})) + (1 - y^{(i)}) \log(1 - h_\theta(x^{(i)})) \right]$$

$$y^{(i)} \log\left(\frac{1}{1 + e^{-\theta^T x}}\right) = -y^{(i)} \log(1 + e^{-\theta^T x})$$

$$(1 - y^{(i)}) \cdot \left[ \log(e^{-\theta^T x}) - \log(1 + e^{-\theta^T x}) \right] = (1 - y^{(i)}) \left[ -\theta^T x - \log(1 + e^{-\theta^T x}) \right]$$

$$-\frac{1}{n} \sum_{i=1}^n \left[ -y^{(i)} \log(1 + e^{-\theta^T x}) - \theta^T x - \log(1 + e^{-\theta^T x}) + y^{(i)} \theta^T x + y^{(i)} \log(1 + e^{-\theta^T x}) \right]$$

$$= -\frac{1}{n} \sum_{i=1}^n \left[ y^{(i)} \theta^T x - (\theta^T x + \log(1 + e^{-\theta^T x})) \right] \quad \theta^T x = \log_e e^{\theta^T x}$$

$$= -\frac{1}{n} \sum_{i=1}^n \left[ y^{(i)} \theta^T x - \log(e^{\theta^T x} + 1) \right] \quad \log_e e^{\theta^T x} + \log_e (1 + e^{-\theta^T x}) = \log_e (e^{\theta^T x} + 1)$$

$$\therefore J(\theta) = -\frac{1}{n} \sum_{i=1}^n \left[ y^{(i)} \theta^T x^{(i)} - \ln(e^{\theta^T x^{(i)}} + 1) \right]$$

Cost function with (optimization)

$$w_0 x_0 + w_1 x_1 + \dots + w_n x_n$$

$$\frac{\partial}{\partial \theta_j} \cdot (y^{(i)} \cdot x_j^{(i)}) = y^{(i)} \cdot x_j^{(i)} \quad , \quad \frac{\partial}{\partial \theta_j} \cdot (\ln(e^{\theta^T x^{(i)}} + 1)) = \frac{x_j^{(i)} e^{\theta^T x^{(i)}}}{e^{\theta^T x^{(i)}} + 1}$$

$$= x_j^{(i)} \cdot \left( \frac{1}{1 + e^{-\theta^T x^{(i)}}} \right) = x_j^{(i)} \cdot h_\theta(x^{(i)})$$

$$\therefore \frac{\partial J(\theta)}{\partial \theta_j} = -\frac{1}{n} \sum_{i=1}^n \left[ y^{(i)} \cdot x_j^{(i)} - x_j^{(i)} \cdot h_\theta(x^{(i)}) \right]$$

$$= \frac{1}{n} \sum_{i=1}^n \left[ x_j^{(i)} (h_\theta(x^{(i)}) - y^{(i)}) \right]$$

weight update.

$$\theta_j := \theta_j - \alpha \frac{\partial J(\theta)}{\partial \theta_j} \quad \text{is } \theta \text{ total update.}$$

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