

Vectorization of Gradient Descent for Logistic Regression. ($\theta_0 = \theta - \alpha \delta$)

$$X = \begin{bmatrix} x_0^{(1)} & x_1^{(1)} & x_2^{(1)} & \dots & x_n^{(1)} \\ x_0^{(2)} & x_1^{(2)} & x_2^{(2)} & \dots & x_n^{(2)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ x_0^{(m)} & x_1^{(m)} & x_2^{(m)} & \dots & x_n^{(m)} \end{bmatrix} \quad \theta = \begin{bmatrix} \theta_0 \\ \theta_1 \\ \vdots \\ \theta_n \end{bmatrix} \Rightarrow \mathbb{R}^{n+1 \times 1}$$

$\searrow \mathbb{R}^{m \times n+1}$

$$h_{\theta}(x) = \begin{bmatrix} h_{\theta}(x^{(1)}) \\ h_{\theta}(x^{(2)}) \\ \vdots \\ h_{\theta}(x^{(m)}) \end{bmatrix} \Rightarrow \mathbb{R}^m \quad y = \begin{bmatrix} y^{(1)} \\ y^{(2)} \\ \vdots \\ y^{(m)} \end{bmatrix} \Rightarrow \mathbb{R}^m$$

$$J(\theta) = \frac{1}{m} (-y^T \log(h_{\theta}(x)) - (1-y)^T \log(1-h_{\theta}(x)))$$

$$h_{\theta}(x) = X\theta = \begin{bmatrix} \theta_0 x_0^{(1)} & \theta_1 x_1^{(1)} & \dots & \theta_n x_n^{(1)} \\ \vdots & \vdots & \ddots & \vdots \\ \theta_0 x_0^{(m)} & \theta_1 x_1^{(m)} & \dots & \theta_n x_n^{(m)} \end{bmatrix} \quad m \times 1$$

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

code \Rightarrow

$$\theta = \theta - (\alpha/m) * X' * (X * \theta - y)$$

Mathematical \Rightarrow

$$\theta_0 = \theta - \frac{\alpha}{m} X^T (g(X\theta) - y)$$