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## Logistic Regression Model

want  $-0 \leq h_{\theta}(x) \leq 1$

$$h_{\theta}(x) = g(\theta^T x)$$

$$g(z) = \frac{1}{1 + e^{-z}}$$

↓

Sigmoid function

or

Logistic function

$$h_{\theta}(x) = \frac{1}{1 + e^{-\theta^T x}} \quad (\text{Hypothesis})$$

Training Set:  $\{(x^{(1)}, y^{(1)}), (x^{(2)}, y^{(2)}), \dots, (x^{(m)}, y^{(m)})\}$

in examples  $x \in \begin{bmatrix} x_0 \\ x_1 \\ \vdots \\ x_n \end{bmatrix}$

$$x_0 = 1, y \in \{0, 1\}$$

$$h_{\theta}(x) = \frac{1}{1 + e^{-\theta^T x}}$$

Cost function is given by

$$J(\theta) = \frac{1}{m} \sum_{i=1}^m \frac{1}{2} (h_{\theta}(x^{(i)}) - y^{(i)})^2$$
$$= \frac{1}{m} \sum_{i=1}^m \text{cost}(h_{\theta}(x^{(i)}), y^{(i)})$$

where,

$$\text{cost}(h_{\theta}(x), y) = \frac{1}{2} (h_{\theta}(x) - y)^2$$

$$\text{cost}(h_{\theta}(x), y) = \begin{cases} -\log h_{\theta}(x) & \text{if } y=1 \\ -\log(1-h_{\theta}(x)) & \text{if } y=0 \end{cases}$$

Note :-  $y=0$  or  $y=1$  always

$$\text{cost}(h_{\theta}(x), y) = -y \log(h_{\theta}(x)) - (1-y) \log(1-h_{\theta}(x))$$

If  $y=1$

$$\text{cost}(h_{\theta}(x), y) = -\log h_{\theta}(x)$$



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$$\text{If } y=0$$

$$\text{Cost}(h_{\theta^0}, y) = -\log(1 - h_{\theta^0})$$

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

$$+ (1 - y^{(i)}) \log(1 - h_{\theta}(x^{(i)})) ]$$

To fit parameters  $\theta$

$$\min_{\theta} J(\theta)$$

Get  $\theta$

To make a prediction given new  $x$ :

$$\text{Output } h_{\theta}(x) = \frac{1}{1 + e^{-\theta^T x}}$$

$$p(y=1 | x; \theta)$$

To minimize Cost function

Apply gradient descent Algorithm

Repeat until convergence

$$\theta_j := \theta_j - \alpha \frac{\partial J}{\partial \theta_j}$$

}

Simultaneously update all  $\theta_j$

$$\frac{\partial J(\theta)}{\partial \theta_j} = \frac{1}{m} \sum_{i=1}^m (h_{\theta}^{(i)} - y^{(i)}) x_j^{(i)}$$

$\therefore$  Repeat 2

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

$$\theta = \begin{bmatrix} \theta_0 \\ \vdots \\ \theta_n \end{bmatrix}$$