

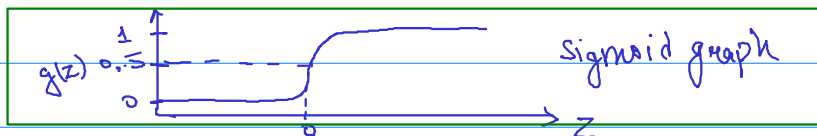
## Logistic Regression Steps

Hypothesis Function -  $h_{\theta}(x) = \theta^T x$  [Linear Regression Hypothesis]

In logistic regression, we modify the above equation a bit by adding sigmoid  $f^n$  to it

$$h_{\theta}(x) = g(\theta^T x), \quad \boxed{g(x) = \frac{1}{1 + e^{-x}}} \rightarrow \text{sigmoid } f^n$$

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



## Logistic Regression Cost function

$$J(\theta) = -[y \log(h_{\theta}(x)) + (1-y) \log(1-h_{\theta}(x))]$$

$$\text{if } h_{\theta}(x) = \hat{y}, \quad J(\theta) = -[y \log \hat{y} + (1-y) \log(1-\hat{y})]$$

$$\begin{aligned} \log 0 &= \text{undefined} / -\infty \\ \log 1 &= 0 \end{aligned}$$

$$J(\theta) = \begin{cases} \text{if } y=0 \text{ \& } \hat{y}=0, & J(\theta) = -(1-y) \log(1-\hat{y}) = 0 \\ \text{if } y=0 \text{ \& } \hat{y}=1, & J(\theta) = -(1-y) \log(1-\hat{y}) = -[1] * -\infty = \infty \\ & \rightarrow 0, \log 0 = -\infty \\ \text{if } y=1, \hat{y}=1, & J(\theta) = -[y \log(\hat{y})] = 0 \\ & \rightarrow \log 1 = 0 \\ \text{if } y=1, \hat{y}=0, & J(\theta) = -[y \log(\hat{y})] = -1 \times [-\infty] \\ & = -1 \times -\infty = \infty \end{cases}$$

cost  $f^n$  heads to  $\infty$ , when the predictor values is misclassifying  
 Logistic regression cost  $f^n$  is complex, to compute the right value of  $\theta$ , we need to minimize  $J(\theta)$

$$\frac{\partial J(\theta)}{\partial \theta} = \frac{1}{n} (h_{\theta}(x) - y) x ; \quad \left\{ \theta_j = \theta_j - \alpha (h_{\theta}(x) - y) x \right\}$$

Iterate until convergence  
 $\alpha$   $\rightarrow$  learning rate