

## Logistic Regression + Square Error

Step 1:  $f_{w,b}(x) = \sigma\left(\sum_i w_i x_i + b\right)$

Step 2: Training data:  $(x^n, \hat{y}^n)$ ,  $\hat{y}^n$ : 1 for class 1, 0 for class 2

$$L(f) = \frac{1}{2} \sum_n (f_{w,b}(x^n) - \hat{y}^n)^2$$

Step 3:

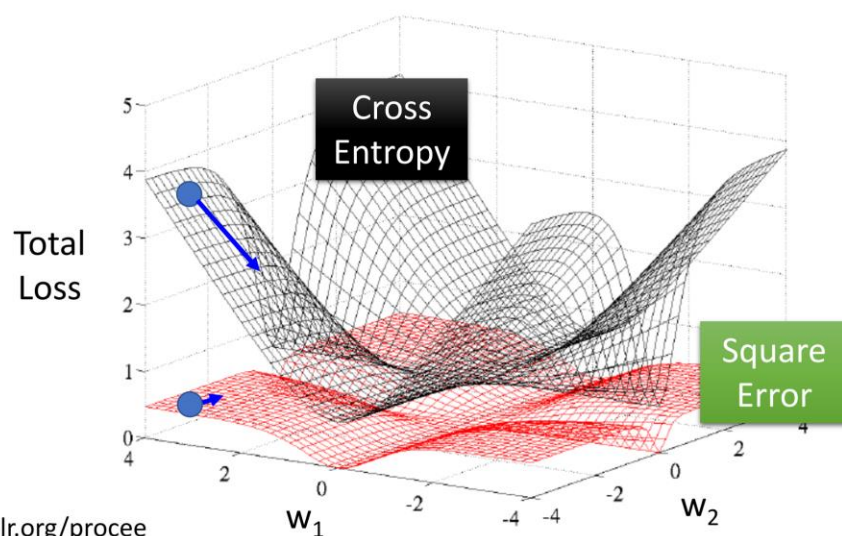
$$\frac{\partial (f_{w,b}(x) - \hat{y})^2}{\partial w_i} = 2(f_{w,b}(x) - \hat{y}) \frac{\partial f_{w,b}(x)}{\partial z} \frac{\partial z}{\partial w_i}$$

$$= 2(f_{w,b}(x) - \hat{y}) f_{w,b}(x) (1 - f_{w,b}(x)) x_i$$

$\hat{y}^n = 1$  If  $f_{w,b}(x^n) = 1$  (close to target)  $\rightarrow \partial L / \partial w_i = 0$

If  $f_{w,b}(x^n) = 0$  (far from target)  $\rightarrow \partial L / \partial w_i = 0$

## Cross Entropy v.s. Square Error



<http://jmlr.org/proceedings/papers/v9/glorot10a/glorot10a.pdf>