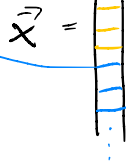
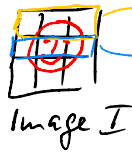


Input  $\vec{x}$

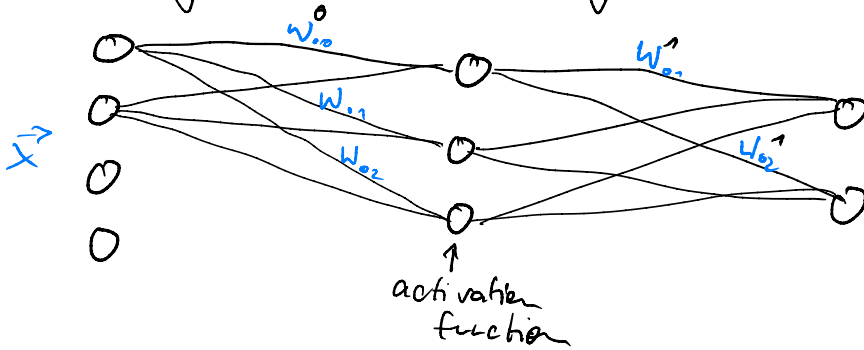


Classifier  $\{\text{face, car}\}$   
 $y = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \dots$

Input Layer

Hidden Layer

Output layer



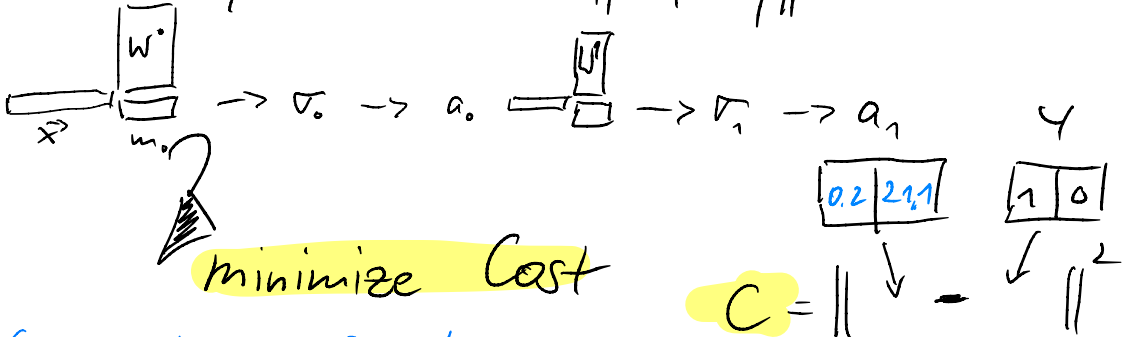
$$\vec{x} \cdot W_0 = m_0$$

$$\vec{a}_0 \cdot W_1 = m_1$$

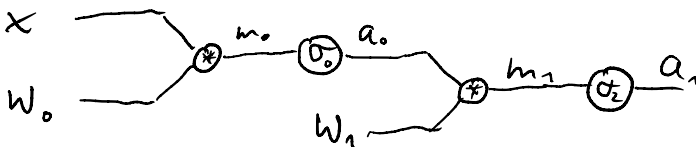
$$\sigma_0(m_0) = a_0$$

$$\sigma_1(m_1) = a_1$$

Loss / Cost  $C = \|a_1 - y\|^2$

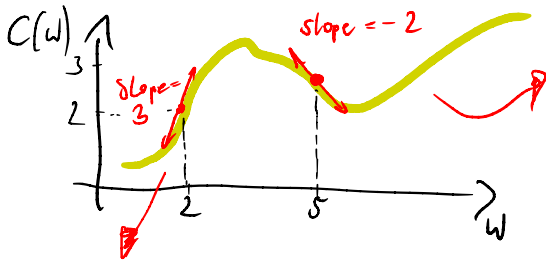


Computational graph



$C(w)$  → Find direction of change

$$\frac{\partial C(w)}{\partial w} \Rightarrow \text{slope}$$



$\frac{\partial C(w)}{\partial w}$  is negative  
when  $C(w)$  is  
decreasing

$\frac{\partial C(w)}{\partial w}$  is positive when  
 $C(w)$  is increasing

**Goal: Minimize  $C(w)$**

Learning  
rate  
↓

if we set  $w = w - \alpha \cdot \text{slope}$   
at points above  $C(w)$  will decrease

$$1.85 = w_{\text{new}} = 2 - \alpha \cdot 3 \quad \alpha = 0.05$$

$$5.1 = w_{\text{new}} = 5 - \alpha \cdot (-2)$$

$$\vec{x} \cdot \vec{w}_0 = m_0$$

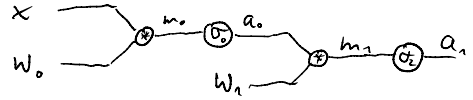
$$\vec{a}_0 \cdot \vec{w}_1 = m_1$$

$$\sigma_0(m_0) = a_0$$

$$\sigma_1(m_1) = a_1$$

$$C = \|a_1 - y\|^2$$

Computational graph



**Chain Rule**  $\frac{\partial C(W)}{\partial W}$

partial derivative  $w_0$ :

$$\boxed{\frac{\partial C}{\partial w_0}} = \frac{\partial C}{\partial a_1} \cdot \frac{\partial a_1}{\partial m_1} \cdot \frac{\partial m_1}{\partial a_0} \cdot \frac{\partial a_0}{\partial m_0} \cdot \frac{\partial m_0}{\partial w_0}$$

partial derivative  $w_1$ :

$$\boxed{\frac{\partial C}{\partial w_1}} = \frac{\partial C}{\partial a_1} \cdot \frac{\partial a_1}{\partial m_1} \cdot \frac{\partial m_1}{\partial w_1}$$

$$\frac{\partial C}{\partial a_1} = 2a_1 - y \quad \frac{\partial a_1}{\partial m_1} = \sigma'_1(m_1) \quad \frac{\partial m_1}{\partial a_0} = w_1$$

$$\frac{\partial a_0}{\partial m_0} = \sigma'_0(m_0) \quad \frac{\partial m_0}{\partial w_0} = x$$

$$\frac{\partial m_1}{\partial w_1} = \overline{a_0} \rightarrow \text{is computed in forward pass} \rightarrow \text{should be cached}$$