

# Linear Discriminant Analysis (LDA)



① Between class separability  $\uparrow - S_B$

② Intra-class separability  $\downarrow - S_W$

$$\text{Cost } J(w) = \frac{(m_0 - m_1)^2}{\alpha_0^2 + \alpha_1^2}$$

$$\alpha_0^2 + \alpha_1^2 = w^T \left[ \sum_{x_i \in C_0} (x_i - M_0)(x_i - M_0)^T + \sum_{x_i \in C_1} (x_i - M_1)(x_i - M_1)^T \right] w$$

$$= w^T S_W w$$

$$(m_0 - m_1)^2 = (w^T M_0 - w^T M_1)^2$$

$$= w^T [(M_0 - M_1)(M_0 - M_1)^T] w$$

$S_B = \text{Scatter matrix}$

$$= w^T S_B w$$

$C_0$   
 $m_0 = \# \text{ of points in } C_0$   
 $\text{mean } M_0 = \frac{1}{m_0} \sum_{x_i \in C_0} x_i$

$C_1$   
 $m_1 = \# \text{ of points in } C_1$   
 $M_1 = \frac{1}{m_1} \sum_{x_j \in C_1} x_j$

Project to mean  $- m_0 = w^T M_0$   
 $\alpha_0^2 = \sum_{x_i \in C_0} (w^T x_i - m_0)^2$

$m = m_0 + m_1$   
 $m_1 = w^T M_1$   
 $\alpha_1^2 = \sum_{x_j \in C_1} (w^T x_j - m_1)^2$

$$\alpha_0^2 = \sum (w^T x_i - m_0)^2 = \sum (w^T x_i - w^T M_0)^2$$

$$= w^T \left[ \sum (x_i - M_0)(x_i - M_0)^T \right] w$$

$$+$$

$$\alpha_1^2 = w^T \left[ \sum (x_j - M_1)(x_j - M_1)^T \right] w$$

$$J(w) = \frac{w^T S_B w}{w^T S_W w}$$

$$\frac{\partial J}{\partial w} = 2 \frac{S_B w}{w^T S_W w} - \frac{2 w^T S_B w}{(w^T S_W w)^2} S_W w$$

$$\Rightarrow S_W w = \lambda w$$

$$\Rightarrow w = \frac{1}{\lambda} S_W w$$

$$J(w) = \frac{w^T S_B w}{w^T S_w w}$$

$$\frac{\partial J}{\partial w} = \frac{2 S_B w}{w^T S_w w} - \frac{(w^T S_B w)(2 S_w w)}{(w^T S_w w)^2} = 0$$

$$\Rightarrow S_w w = \lambda S_B w$$

$$\Rightarrow w = \lambda \underbrace{S_w^{-1} S_B}_{S} w$$

$$\Rightarrow w \times \frac{1}{\lambda} = S w$$

$$\frac{d}{dx} \left( \frac{u}{v} \right) = \frac{v \frac{du}{dx} - u \frac{dv}{dx}}{v^2} = \frac{\frac{du}{dx}}{v} - \frac{u}{v^2} \frac{dv}{dx}$$

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Biological Neuron



- 1957-1985 - Frank Rosenblatt - Perception
- 1943-44 - McCulloch + Pitts (MP model)
- 1965-68 - MLP
- 1969 - Limitations of Perceptron
- 2007 ↓
- 1986 - Backprop
- 1989 →

$$J(w) = \frac{w^T S_B w}{w^T S_w w}$$

$$\frac{\partial J}{\partial w} = 2 \frac{S_B w}{w^T S_w w} - \frac{(w^T S_B w)}{(w^T S_w w)^2} (2 w^T S_w w)$$

$$\Rightarrow S_w w = \lambda \frac{S_B w}{S_w w}$$

$$\Rightarrow w = \lambda \frac{S_w^{-1} S_B w}{S_w w}$$

$$\Rightarrow w \times \frac{1}{\lambda} = S_w w$$