## Exercise 1 - Rewriting the Fisher criterion for LDA

Let

$$J(w) = \frac{\langle w, \mu_{+} - \mu_{-} \rangle^{2}}{\sigma_{w,+}^{2} + \sigma_{w,-}^{2}} = \frac{N}{D}$$

First let us evaluate the numerator.

$$N = \langle w, \mu_{+} - \mu_{-} \rangle^{2}$$

$$= \langle w, \mu_{+} - \mu_{-} \rangle \langle w, \mu_{+} - \mu_{-} \rangle$$

$$= \langle w, \mu_{+} - \mu_{-} \rangle \langle \mu_{+} - \mu_{-}, w \rangle \quad \text{Since } \langle a, b \rangle = \langle b, a \rangle$$

$$= w^{T} (\mu_{+} - \mu_{-}) (\mu_{+} - \mu_{-})^{T} w$$

$$= w^{T} C_{B} w$$

$$= \langle w, C_{B} w \rangle$$

Moving on the the denominator.

$$\sigma_{w,+}^{2} = \frac{1}{m_{+}} \sum_{i:Y_{i}=+1} \left( \langle w, X_{i} - \mu_{+} \rangle^{2} \right)$$

$$= \frac{1}{m_{+}} \sum_{i:Y_{i}=+1} \left( w^{T} \left( X_{i} - \mu_{+} \right) \left( X_{i} - \mu_{+} \right)^{T} w \right) \quad \text{See evaluation for } N$$

$$= w^{T} \left( \frac{1}{m_{+}} \sum_{i:Y_{i}=+1} \left( X_{i} - \mu_{+} \right) \left( X_{i} - \mu_{+} \right)^{T} \right) w$$

$$\sim \sigma_{w,-}^{2} = w^{T} \left( \frac{1}{m_{-}} \sum_{i:Y_{i}=-1} \left( X_{i} - \mu_{-} \right) \left( X_{i} - \mu_{-} \right)^{T} \right) w$$

$$\implies D = \sigma_{w,+}^{2} + \sigma_{w,-}^{2} = w^{T} \left[ \left\{ \frac{1}{m_{+}} \sum_{i:Y_{i}=+1} \left( X_{i} - \mu_{+} \right) \left( X_{i} - \mu_{+} \right)^{T} \right\} + \left\{ \frac{1}{m_{-}} \sum_{i:Y_{i}=-1} \left( X_{i} - \mu_{-} \right) \left( X_{i} - \mu_{-} \right)^{T} \right\} \right] w$$

$$= w^{T} C_{W} w$$

$$= \langle w, C_{W} w \rangle$$

Therefore,

$$J\left(w\right) = \frac{\left\langle w, \mu_{+} - \mu_{-} \right\rangle^{2}}{\sigma_{w,+}^{2} + \sigma_{w,-}^{2}} = \frac{\left\langle w, C_{B}w \right\rangle}{\left\langle w, C_{W}w \right\rangle}$$

## Exercise 4 - Complexity of multiclass classification

Part (a) - Learning complexity =  $c(m_1^2 + m_2^2)$ 

• One-vs-All: 
$$\binom{k}{1} \times c \left[ \left( \frac{1}{k} n \right)^2 + \left\{ \left( 1 - \frac{1}{k} \right) n \right\}^2 \right] = \frac{cn^2}{k} \left[ (k-1)^2 + 1 \right]$$

• One-vs-One: 
$$\binom{k}{2} \times c \left[ \left( \frac{1}{k} n \right)^2 + \left( \frac{1}{k} n \right)^2 \right] = \frac{cn^2}{k} (k-1)$$

Part (b) - Learning complexity =  $c(m_1 + m_2)$ 

• One-vs-All: 
$$\binom{k}{1} \times c \left[ \left( \frac{1}{k} n \right) + \left( 1 - \frac{1}{k} \right) n \right] = cnk$$

• One-vs-One: 
$$\binom{k}{2} \times c \left[ \left( \frac{1}{k} n \right) + \left( \frac{1}{k} n \right) \right] = cn \left( k - 1 \right)$$

## Comments

- One-vs-One is faster in both cases.
- ullet For scaling n, the algorithm for part (b) is preferable.

## Exercise 5 - Parameter selection by the training error

Overfitting