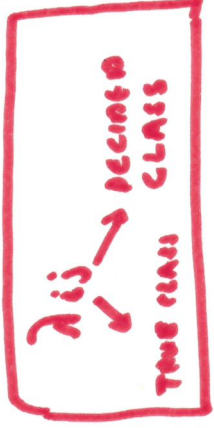


Average Risk Minimization

- 1: Each decision is assigned with a penalty term.
- 2: λ_{ij} : penalty term when deciding in favor of w_j when the true class is w_i .



Penalty Matrix: $\frac{1}{2} = \begin{bmatrix} \lambda_{11} & \lambda_{12} \\ \lambda_{21} & \lambda_{22} \end{bmatrix}$

4: Average Risk w.r.t. ω_1 :

← AVERAGE RISK ON A_1 →

$$r_2 = \int_{A_1} \lambda_{11} p(x|\omega_1) p(\omega_1) dx +$$

$$\int_{A_2} \lambda_{12} p(x|\omega_1) p(\omega_1) dx \quad (I)$$

← AVERAGE RISK ON A_2 →

Probability of wrong decisions in favor of ω_2

5: Average Risk w.r.t. ω_2 :

$$r_2 = \int_{A_1} \lambda_{21} p(x|\omega_2) p(\omega_2) dx +$$

$$\int_{A_2} \lambda_{22} p(x|\omega_2) p(\omega_2) dx \quad (II)$$

Probability of wrong decisions in favor of ω_1

6: Average Risk: $r = r_1 + r_2 \quad (III)$

7: Average Risk on R_1 :

$$\int_{R_1} \underbrace{[\lambda_{11} p(x|w_1) p(w_1) + \lambda_{21} p(x|w_2) p(w_2)]}_{e_1(x)} dx \quad (EV)$$

8: Average Risk on R_2 :

$$\int_{R_2} \underbrace{[\lambda_{12} p(x|w_1) p(w_1) + \lambda_{22} p(x|w_2) p(w_2)]}_{e_2(x)} dx \quad (V)$$

q: Modified Decision Rule:

$w_1, c_1(x) < c_2(x);$

$g(x) =$

$$\begin{cases} w_1, & c_1(x) < c_2(x); \\ w_2, & c_2(x) < c_1(x). \end{cases} \quad (VI)$$

w: Modified Decision Boundary: x_0^* ; $c_1(x) = c_2(x) \Leftrightarrow$

$$\lambda_1 p(x|w_1) p(w_1) + \lambda_{21} p(x|w_2) p(w_2) = \lambda_{12} p(x|w_1) p(w_1) + \lambda_{22} p(x|w_2) p(w_2) \Leftrightarrow$$

$$(\lambda_{11} - \lambda_{12}) p(x|w_1) p(w_1) = (\lambda_{22} - \lambda_{21}) p(x|w_2) p(w_2) \Leftrightarrow$$

$$\frac{p(x|w_1)}{p(x|w_2)} = \frac{p(w_1)}{p(w_2)} \cdot \frac{\lambda_{22} - \lambda_{21}}{\lambda_{11} - \lambda_{12}} \quad \text{or}$$

$$\frac{p(x|w_1)}{p(x|w_2)} = \frac{p(w_2)}{p(w_1)} \cdot \frac{\lambda_{21} - \lambda_{22}}{\lambda_{12} - \lambda_{11}} \quad (VII)$$

#5

N: It is reasonable to assign zero risk when making the correct decision (that is, the penalty matrix is diagonal):

$$L = \begin{bmatrix} 0 & \lambda_{12} \\ \lambda_{21} & 0 \end{bmatrix} \quad (\lambda_{11} = \lambda_{22} = 0)$$

Thus, Eq. VII will be written as:

$$\frac{P(x_1|w_1)}{P(x_1|w_2)} = \frac{f(w_2)}{f(w_1)} \cdot \frac{\lambda_{21}}{\lambda_{12}} \quad (VIII)$$

12: Learning $f(w_2) = \kappa \cdot f(w_1)$ where $\kappa > 1$,

how can we modify the penalty terms $\{\lambda_{12}, \lambda_{21}\}$ so that the decision boundary does not move?

$$x_0^{(1)} = x_0^{(2)}, \kappa > 1$$

(i): when $\kappa = 1$: $\frac{P(x_0^{(1)}|w_1)}{P(x_0^{(1)}|w_2)} = \frac{f(w_1)}{f(w_2)}$ { No Penalty Terms } \rightarrow

$$\frac{P(x_0^{(1)}|w_1)}{P(x_0^{(1)}|w_2)} = 1$$

(ii): when $\kappa > 1$: $\frac{P(x_0^{(1)}|w_1)}{P(x_0^{(1)}|w_2)} = \frac{f(w_1)}{\kappa \cdot f(w_2)} \cdot \frac{\lambda_{21}}{\lambda_{12}}$ { Penalty Terms } \rightarrow

$$\frac{P(x_0^{(1)}|w_1)}{P(x_0^{(1)}|w_2)} = \frac{1}{\kappa} \cdot \frac{\lambda_{21}}{\lambda_{12}} = 1 \Rightarrow \frac{\lambda_{21}}{\lambda_{12}} = \kappa \Rightarrow \boxed{\lambda_{21} = \kappa \cdot \lambda_{12}}$$