

Denote  $\eta(x) = P(Y = 1|X = x)$ . Consider Bayes classifier with a reject option:

$$g_c(x) = \begin{cases} 1, & \eta(x) > 1/2 + c, \\ 0, & \eta(x) < 1/2 - c, \\ r, & \text{otherwise.} \end{cases} \quad (1)$$

**Proposition 1.** For any decision  $g(x)$  such that

$$P(g(X) = r) \leq P(g_c(X) = r), \quad (2)$$

we have

$$P(g(X) \neq Y|g(X) \neq r) \geq P(g_c(X) \neq Y|g_c(X) \neq r). \quad (3)$$

In other words, the Bayes classifier  $g_c(x)$  with a reject option is optimal for fixed rejection rate.

*Proof.* We first clarify some notations. Denote

$$A_c \equiv \{x | \max(\eta(x), 1 - \eta(x)) > 1/2 + c\} = \{x | g_c(x) \neq r\}, \quad (4)$$

which is the decision region. The error rate is

$$P(g_c(X) \neq Y|g_c(X) \neq r) = 1 - P(g_c(X) = Y|g_c(X) \neq r) \quad (5)$$

$$= 1 - \frac{P(g_c(X) = Y, g_c(X) \neq r)}{P(g_c(X) \neq r)} \quad (6)$$

$$= 1 - \frac{1}{P(X \in A_c)} \int_{x \in A_c} \max(\eta(x), 1 - \eta(x)) dP, \quad (7)$$

where  $dP$  is the measure of  $X$ , and  $P(X \in A_c) = \int_{x \in A_c} dP$ .

The strategy is to prove the following two lemmas:

**Lemma 2.** Let  $B$  be any region such that  $P(X \in B) \geq P(X \in A_c)$ , then  $P(g_c(X) \neq Y|X \in B) \geq P(g_c(X) \neq Y|X \in A_c)$ .

**Lemma 3.** Let  $B$  be any region such that  $P(X \in B) \geq P(X \in A_c)$  and  $g$  be any decision, then  $P(g(X) \neq Y|X \in B) \geq P(g_c(X) \neq Y|X \in B)$ .

Lemma 3 is the same as proving the optimality of Bayes decision rule without rejection, hence we omit the proof. We are left to prove lemma 2.

*Proof of Lemma 2.* After arranging terms, it suffices to prove

$$P(X \in A_c)P(X \in B) [P(g_c(X) \neq Y|X \in B) - P(g_c(X) \neq Y|X \in A_c)] \geq 0, \quad (8)$$

where the L.H.S. of the inequality is

$$\text{L.H.S.} = P(X \in B) \int_{x \in A_c} \max(\eta(x), 1 - \eta(x)) dP - P(X \in A_c) \int_{x \in B} \max(\eta(x), 1 - \eta(x)) dP. \quad (9)$$

The key is to decompose  $A_c = U + (A_c - U)$  and  $B = U + (B - U)$ , where  $U = A \cap B$ . By definition, for  $x \in A_c - U$  or  $U$ ,  $\max(\eta(x), 1 - \eta(x)) > 1/2 + c$ ; for  $x \in B - U$ ,  $\max(\eta(x), 1 - \eta(x)) \leq 1/2 + c$ . Thus

$$\int_{x \in A_c} \max(\eta(x), 1 - \eta(x)) dP \geq \left(\frac{1}{2} + c\right) P(X \in A_c - U) + \int_{x \in U} \max(\eta(x), 1 - \eta(x)) dP, \quad (10)$$

$$\int_{x \in B} \max(\eta(x), 1 - \eta(x)) dP \leq \left(\frac{1}{2} + c\right) P(X \in B - U) + \int_{x \in U} \max(\eta(x), 1 - \eta(x)) dP. \quad (11)$$

Insert Eq. (10) and (11) into (9),

$$\begin{aligned} \text{L.H.S.} &\geq \left(\frac{1}{2} + c\right) [P(X \in B)P(X \in A_c - U) - P(X \in A_c)P(X \in B - U)] \\ &\quad + [P(X \in B) - P(X \in A_c)] \int_{x \in U} \max(\eta(x), 1 - \eta(x))dP \end{aligned} \tag{12}$$

$$\geq [P(X \in B) - P(X \in A_c)] \left[ \int_{x \in U} \max(\eta(x), 1 - \eta(x))dP - \left(\frac{1}{2} + c\right) P(X \in U) \right] \tag{13}$$

$$\geq 0, \tag{14}$$

where in the last line we have used the assumption  $P(X \in B) \geq P(X \in A_c)$ . This proves lemma 2.  $\square$

Combine these two lemmas we establish the following inequality chain:

$$P(g(X) \neq Y | X \in B) \geq P(g_c(X) \neq Y | X \in B) \geq P(g_c(X) \neq Y | X \in A_c), \tag{15}$$

which proves the proposition.  $\square$

*Remark.* The proof of lemma 2 is essentially the proof of Neyman–Pearson lemma in statistics.