

# ERROR RATE OF A LINEAR

Jain  
CSE 802  
1/27/03

## DISCRIMINANT FUNCTION

- ASSUME A 2-CLASS PROBLEM
- DUE TO THE SYMMETRY OF THE PROBLEM (IDENTICAL  $\Sigma$ ), THE TWO TYPES OF ERRORS ARE IDENTICAL
- DECIDE  $\underline{x} \in \omega_1$  IF  $g_1(\underline{x}) > g_2(\underline{x})$

OR

$$-\frac{1}{2}(\underline{x} - \underline{\mu}_1)^t \underline{\Sigma}^{-1} (\underline{x} - \underline{\mu}_1) + \log [P(\omega_1)] > \\ -\frac{1}{2}(\underline{x} - \underline{\mu}_2)^t \underline{\Sigma}^{-1} (\underline{x} - \underline{\mu}_2) + \log [P(\omega_2)]$$

OR

$$(\underline{\mu}_2 - \underline{\mu}_1)^t \underline{\Sigma}^{-1} \underline{x} + \frac{1}{2} (\underline{\mu}_1^t \underline{\Sigma}^{-1} \underline{\mu}_1 - \underline{\mu}_2^t \underline{\Sigma}^{-1} \underline{\mu}_2) \\ < \log [P(\omega_1)/P(\omega_2)]$$

LET

$$h(\underline{x}) = (\underline{\mu}_2 - \underline{\mu}_1)^t \underline{\Sigma}^{-1} \underline{x} + \frac{1}{2} (\underline{\mu}_1^t \underline{\Sigma}^{-1} \underline{\mu}_1 - \underline{\mu}_2^t \underline{\Sigma}^{-1} \underline{\mu}_2)$$

COMPUTE EXPECTED VALUES & VARIANCES OF  $h(\underline{x})$  WHEN  $\underline{x} \in \omega_1$  &  $\underline{x} \in \omega_2$

$$\eta_1 = E[h(\underline{x}) | \underline{x} \in \omega_1]$$

$$= (\underline{\mu}_2 - \underline{\mu}_1)^t \underline{\Sigma}^{-1} E(\underline{x} | \omega_1) + \frac{1}{2} (\underline{\mu}_1^t \underline{\Sigma}^{-1} \underline{\mu}_1 - \underline{\mu}_2^t \underline{\Sigma}^{-1} \underline{\mu}_2)$$

$$= - \left[ \frac{1}{2} (\underline{\mu}_2 - \underline{\mu}_1)^t \underline{\Sigma}^{-1} (\underline{\mu}_2 - \underline{\mu}_1) \right]$$

$$= - \eta$$

WHERE

$$\eta = (\underline{\mu}_2 - \underline{\mu}_1)^t \underline{\Sigma}^{-1} (\underline{\mu}_2 - \underline{\mu}_1)$$

= SQUARED MAHALANOBIS

DISTANCE BETWEEN

$\underline{\mu}_1$  &  $\underline{\mu}_2$

SIMILARLY,

$$\begin{aligned}\eta_2 &= +\frac{1}{2} (\underline{\mu}_2 - \underline{\mu}_1)^t \underline{\Sigma}^{-1} (\underline{\mu}_2 - \underline{\mu}_1) \\ &= +\eta\end{aligned}$$

$$\sigma_1^2 = E[(h(\underline{x}) - \eta_1)^2 | \omega_1]$$

$$= E\left[\left\{(\underline{\mu}_2 - \underline{\mu}_1)^t \underline{\Sigma}^{-1} (\underline{x} - \underline{\mu}_1)\right\}^2 \mid \underline{x} \in \omega_1\right]$$

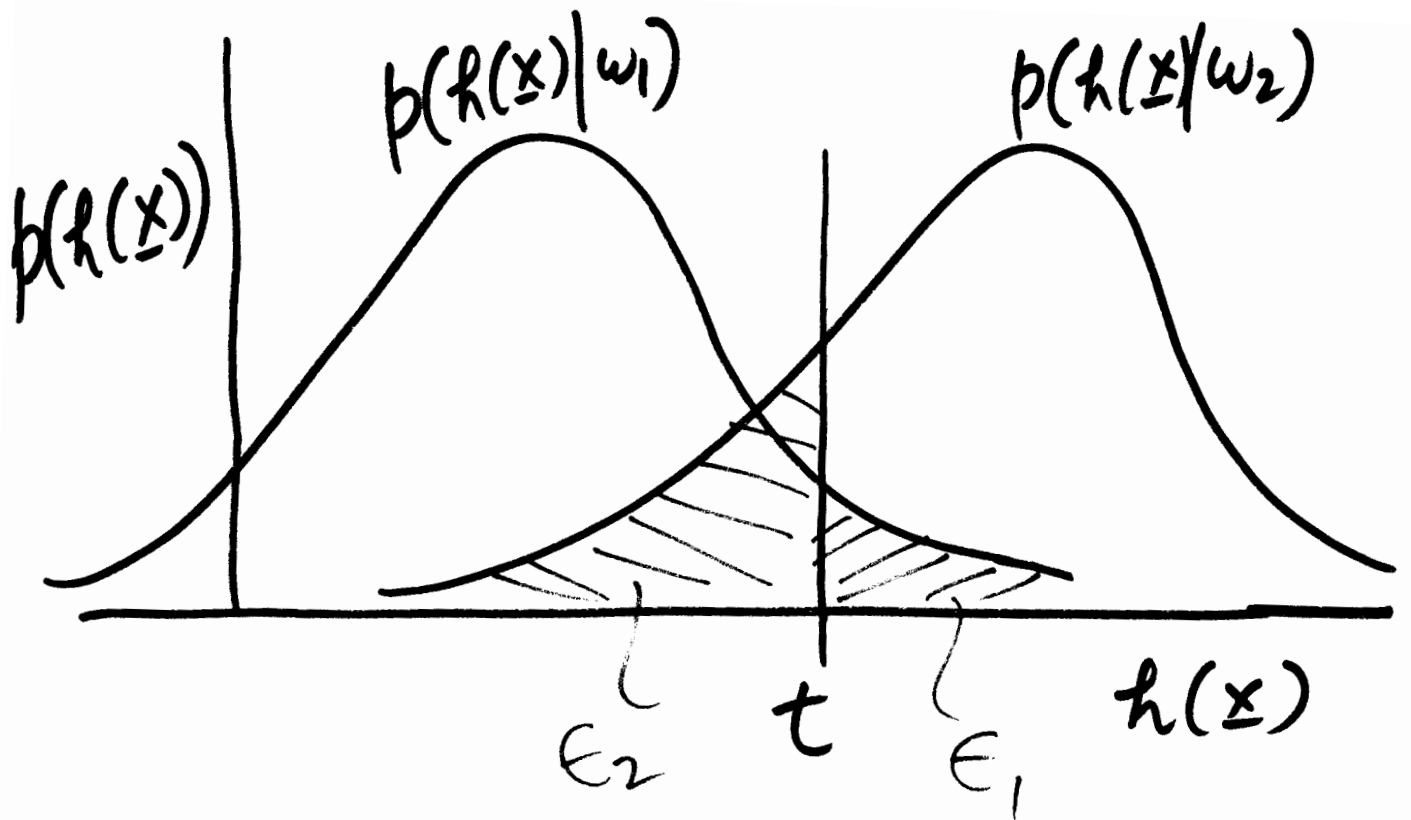
$$= (\underline{\mu}_2 - \underline{\mu}_1)^t \underline{\Sigma}^{-1} (\underline{\mu}_2 - \underline{\mu}_1)$$

$$= 2\eta$$

$$\sigma_2^2 = 2\eta$$

$$p(h(\underline{x}) | \underline{x} \in \omega_1) \sim N(-\eta, 2\eta)$$

$$p(h(\underline{x}) | \underline{x} \in \omega_2) \sim N(+\eta, 2\eta)$$



$$\epsilon_1 = P_r (g_1(x) < g_2(x) | x \in w_1)$$

$$= \int_t^{\infty} p(h(x)|w_1) dh \quad h(x) \sim \frac{1}{\sqrt{2\pi} \cdot \sigma} e^{-\frac{1}{2} \left(\frac{h-\mu}{\sigma}\right)^2}$$

$$= \int_{\frac{\eta+t}{\sqrt{2}n}}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2} \zeta^2} d\zeta$$

$$= \frac{1}{2} - \frac{1}{2} \operatorname{erf} \left( \frac{\eta+t}{\sqrt{4n}} \right)$$

$$\cdot P(\omega_1) = P(\omega_2) = \frac{1}{2} \Rightarrow t = 0$$

$$\epsilon_1 = \epsilon_2 = \frac{1}{2} - \frac{1}{2} \operatorname{erf}\left(\frac{\eta}{\sqrt{4n}}\right)$$

$$= \frac{1}{2} - \frac{1}{2} \operatorname{erf}\left[\frac{\left(\underline{\mu}_1 - \underline{\mu}_2\right)^t \Sigma^{-1} \left(\underline{\mu}_1 - \underline{\mu}_2\right)}{2\sqrt{2}}\right]$$

(i) NO CLASS SEPARATION

$$\underline{\mu}_1 = \underline{\mu}_2$$

$$\bullet \left(\underline{\mu}_1 - \underline{\mu}_2\right)^t \Sigma^{-1} \left(\underline{\mu}_1 - \underline{\mu}_2\right) = 0$$

$$\Rightarrow \epsilon_1 = \epsilon_2 = \frac{1}{2}$$

(ii) PERFECT CLASS SEPARATION

$$\left(\underline{\mu}_1 - \underline{\mu}_2\right)^t \Sigma^{-1} \left(\underline{\mu}_1 - \underline{\mu}_2\right) \rightarrow \infty$$

$$\epsilon_1 = \epsilon_2 \rightarrow 0$$

$$t = \log \left[ \frac{P(w_1)}{P(w_2)} \right]$$

$$\operatorname{erf}(r) = \frac{2}{\sqrt{\pi}} \int_0^r e^{-x^2} dx$$

$$\epsilon_2 = \frac{1}{2} - \frac{1}{2} \operatorname{erf} \left[ \frac{(\eta - t)}{\sqrt{4\eta}} \right]$$

TOTAL PROB. OF ERROR

$$P_e = P(w_1) \epsilon_1 + P(w_2) \epsilon_2$$