

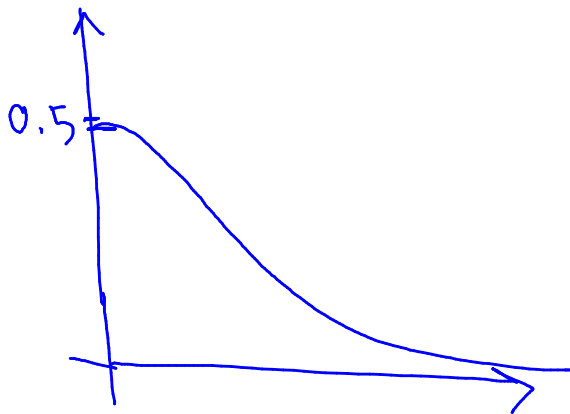
Chapter 2 HW Solutions

2.8) a)

$$P(\text{error}) = \int_{R_2} P(x|w_1) p(w_1) dx + \int_{R_1} P(x|w_2) p(w_2) dx$$

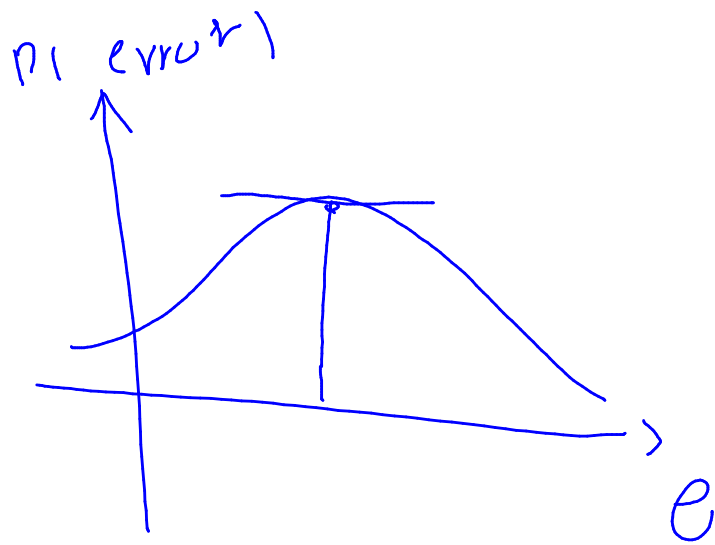
$$= \frac{1}{\pi b} \left[\int_{-\infty}^{\frac{a_2+a_1}{2}} \frac{1}{1 + \left(\frac{x-a_1}{b}\right)^2} dx + \int_{\frac{a_1+a_2}{2}}^{+\infty} \frac{1}{1 + \left(\frac{x-a_2}{b}\right)^2} dx \right]$$

$$= \frac{1}{2} - \frac{1}{\pi} \arctan \left| \frac{a_2-a_1}{b} \right|$$



$$\frac{a_2 - a_1}{b} \rightarrow 0$$

$$a_2 = a_1$$



$$\textcircled{a} R(\alpha^{(n)} | x) = R(\alpha_1 | x) P(\alpha_1 | x) + \dots$$

$$\textcircled{b} = \sum_{i=1}^a R(\alpha_i | x) P(\alpha_i | x) + R(\alpha_a | x) P$$

$$R = \int R(\alpha^{(n)} | x) P(x) dx$$

$$R(\alpha_{i^*} | n) \leq R(\alpha_i | n) \text{ for all } i \neq i^*$$

$$R(\alpha(n) | n) = \sum_{i=1}^a R(\alpha_i | n) P(\alpha_i | n) =$$

$$= \sum_{i=1}^a \left[R(\alpha_i | n) - R(\alpha_{i^*} | n) + R(\alpha_{i^*} | n) \right] P(\alpha_i | n)$$

$$= R(\alpha_{i^*} | n) \sum_{i=1}^a P(\alpha_i | n) + \sum_{i \neq i^*} \left[R(\alpha_i | n) - R(\alpha_{i^*} | n) \right] P(\alpha_i | n)$$

$$= R(\alpha_{i^*} | n) + \sum_{i \neq i^*} \left(R(\alpha_i | n) - R(\alpha_{i^*} | n) \right) P(\alpha_i | n)$$



Because $R(\alpha_i | x) - R(\alpha_{i^*} | x) \geq 0$

\Rightarrow Maximize can be achieved by

$P(\alpha_i | x) = 0$ for $i \neq i^*$

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$$R(d_i | x) = \sum_{j=1, j \neq i}^c \lambda_s P(w_j | x)$$

$$= \lambda_s \sum_{j=1, j \neq i}^c P(w_j | x) = \lambda_s (1 - P(w_i | x))$$

$$R(d_{c+1} | x) = \lambda_r$$

$$\begin{cases} \lambda_s (1 - P(w_i | x)) \leq \lambda_s (1 - P(w_j | x)) \\ \lambda_s (1 - P(w_i | x)) \leq \lambda_r \end{cases}$$

$$i \left\{ \begin{array}{l} p(\omega_i/x) \geq p(\omega_j/x) \\ p(\omega_i/x) \geq 1 - \frac{\lambda_r}{\lambda_s} \end{array} \right.$$

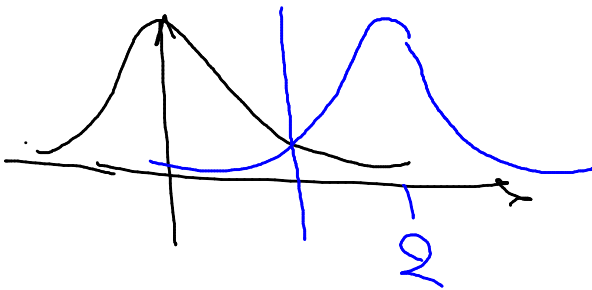
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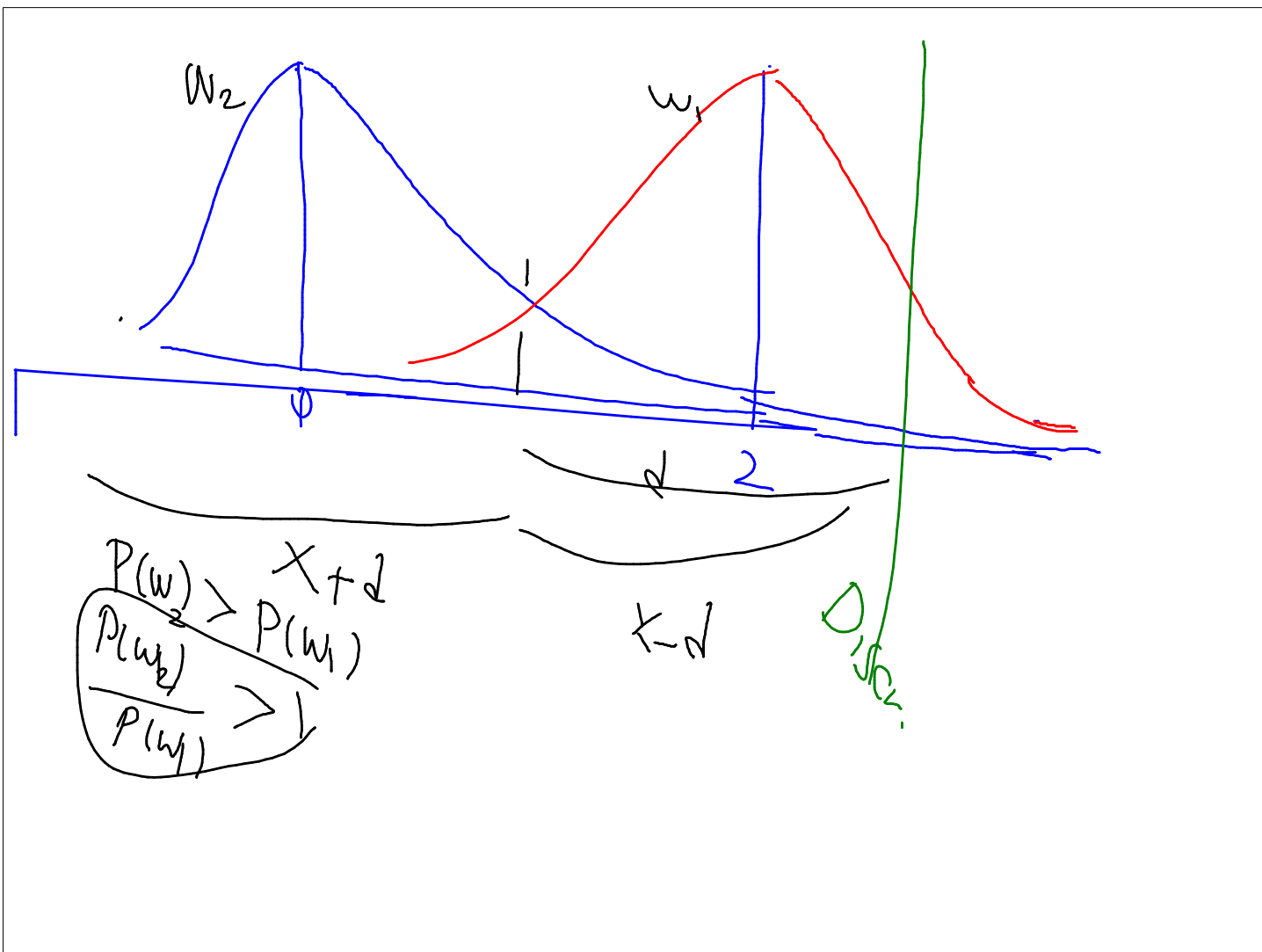
$$\mu_1 > \mu_2$$

$$(63) \Rightarrow \alpha_0 = \frac{1}{2}(\mu_1 + \mu_2) - \frac{\ln \left(\frac{P(w_1)}{P(w_2)} \right)}{(\mu_1 - \mu_2)^T \Sigma^{-1} (\mu_1 - \mu_2)}$$

$$\alpha_0 > \mu_1$$

$$\frac{P(w_1)}{P(w_2)} < \exp \left(-\frac{1}{2} (\mu_1 - \mu_2)^T \Sigma^{-1} (\mu_1 - \mu_2) \right)$$





$P(\omega_k | \underline{x}) > P(\omega_j | \underline{x})$ for all j and k
 $\Rightarrow \underline{x} \in S_k$

$$\begin{aligned} P(\omega_j | \underline{x}) &= P(\underline{x} | \omega_j) P(\omega_j) \\ &= \prod_{i=1}^n P(x_i | \omega_j) P(\omega_j) \end{aligned}$$

$$= \prod_{i=1}^d (P_{ij})^{x_i} (1 - P_{ij})^{1 - x_i} P(\omega_j)$$

$$= (P_{ij})^{\sum x_i} (1 - P_{ij})^{\sum (1 - x_i)} P(\omega_j)$$

$$\Rightarrow \ln(P(\omega_j | x)) = \sum x_i \ln P_{ij} + \dots$$

$$\dots \sum (1 - x_i) \ln(1 - P_{ij}) + \ln P(\omega_j)$$

$$= \sum_{i=1}^d x_i (\ln P_{ij} - \ln(1 - P_{ij})) + \dots$$

$$\sum \ln(1 - P_{ij}) + \ln P(\omega_j)$$

$$\Rightarrow g_j(x) = \sum_{i=1}^d x_i \ln\left(\frac{P_{ij}}{1 - P_{ij}}\right) + \sum_{i=1}^d \ln(1 - P_{ij}) + \ln(P(\omega_j)) \quad Q.E.D.$$