

# Problem 10

Tuesday, June 11, 2024

6:18 PM

$$4.32 \quad \log \left( \frac{\Pr(Y=k | X=x)}{\Pr(Y=k | X=x)} \right) \quad p > 1$$

$p > 1$   $\mu_k = \text{mean}_k$  is  $p$  dim vector  
 $\Sigma = \text{shared covariance}$   $p \times p$  matrix

$p = 1$   $\mu_1, \dots, \mu_K > \text{scalar}$   
 $\sigma^2$

(4.32) want  $a_k, b_k$  in terms of  $\mu_k, \sigma^2$

$$\Pr(Y=k | X=x) = \frac{\pi_k f_k(x)}{\sum_{i=1}^K \pi_i f_i(x)} \quad (1)$$

$$f_k(x) = \frac{1}{\sqrt{2\pi}\sigma_k} \exp \left( -\frac{1}{2\sigma_k^2} (x - \mu_k)^2 \right) \quad (2)$$

$$\log \left( \frac{\Pr(Y=k | X=x)}{\Pr(Y=k | X=x)} \right) = \left( \frac{\frac{\pi_k f_k(x)}{\sum_{i=1}^K \pi_i f_i(x)}}{\frac{\pi_k f_k(x)}{\sum_{i=1}^K \pi_i f_i(x)}} \right) \log \quad \text{by (1)}$$

$$= \log \left( \frac{\pi_k f_k(x)}{\pi_k f_k(x)} \right) \quad \text{by cancellation of terms}$$

$$= \left( \frac{\frac{1}{\sqrt{2\pi}\sigma_k} \exp \left( -\frac{1}{2\sigma_k^2} (x - \mu_k)^2 \right)}{\frac{1}{\sqrt{2\pi}\sigma_k} \exp \left( -\frac{1}{2\sigma_k^2} (x - \mu_k)^2 \right)} \right) \log \quad \text{by (2)}$$

$$= \log \left( \frac{\pi_k}{\pi_k} \right) - \frac{1}{2\sigma_k^2} (x - \mu_k)^2 + \frac{1}{2\sigma_k^2} (x - \mu_k)^2$$

$$\sigma_k^2 = \sigma^2 \quad \text{based on given}$$

$$= \log \left( \frac{\pi_k}{\pi_k} \right) - \frac{1}{2\sigma^2} ((x - \mu_k)^2 - (x - \mu_k)^2)$$

$$= \log \left( \frac{\pi_k}{\pi_k} \right) - \frac{1}{2\sigma^2} (x^2 - 2\mu_k x + \mu_k^2 - x^2 + 2\mu_k x - \mu_k^2)$$

$$= \log \left( \frac{\pi_k}{\pi_k} \right) - \frac{1}{2\sigma^2} (-2\mu_k x + \mu_k^2 + 2\mu_k x - \mu_k^2)$$

$$= \log \left( \frac{\pi_k}{\pi_k} \right) + \frac{\mu_k - \mu_k}{\sigma^2} (x) + \frac{-\mu_k^2 + \mu_k^2}{2\sigma^2}$$

$$a = \log \left( \frac{\pi_k}{\pi_k} \right) + \frac{-\mu_k^2 + \mu_k^2}{2\sigma^2} \quad \leftarrow \text{based on (x)}$$

$$b = \frac{\mu_k - \mu_k}{\sigma^2} \quad \leftarrow \text{based on (x)}$$