

## 4.7 Exercises

### Conceptual

2. In LDA method, with  $p=1$ , the Bayesian posterior is given by (4.11) :

$$P_k(x) = \frac{\pi_k \frac{1}{\sqrt{2\pi}\sigma} \exp\left[-\frac{1}{2\sigma^2}(x-\mu_k)^2\right]}{\sum_{k=1}^K \pi_k \frac{1}{\sqrt{2\pi}\sigma} \exp\left[-\frac{1}{2\sigma^2}(x-\mu_k)^2\right]} \quad (4.12)$$

Bayes' classifier assign  $x$  to  $k^{\text{Bayes}}$  such that  $P_{k^{\text{Bayes}}}(x)$  is maximum:

$$k^{\text{Bayes}} = \arg\max_k P_k(x) = \arg\max_k \ln P_k(x)$$

Since  $\ln$  is monotonically increasing.

~~$\ln P_k(x)$~~  because the denominator of (4.12) is constant w.r.t.  $k$ . So

$$\arg\max_k \ln P_k(x) = \arg\max_k \ln \left\{ \pi_k \frac{1}{\sqrt{2\pi}\sigma} \exp\left[-\frac{1}{2\sigma^2}(x-\mu_k)^2\right] \right\}$$

$$= \arg\max_k \left\{ \ln \pi_k - \underbrace{\ln \sqrt{2\pi}\sigma}_{\text{const}} - \frac{1}{2\sigma^2}(x-\mu_k)^2 \right\}$$

$$= \arg\max_k \left\{ \ln \pi_k - \underbrace{\frac{1}{2\sigma^2}}_{\text{const}} (x^2 - 2x\mu_k + \mu_k^2) \right\}$$

$$= \arg\max_k \left\{ \frac{x\mu_k}{\sigma^2} - \frac{\mu_k^2}{2\sigma^2} + \ln \pi_k \right\}$$

Linear discriminant