

Q2. Microscopic Image $M \times N$

$$\text{Each cell} \begin{cases} \rightarrow BG \Rightarrow \mathcal{N}(\mu_1, \sigma_1^2) \\ \rightarrow FG \Rightarrow \mathcal{N}(\mu_2, \sigma_2^2) \end{cases}$$

$$\mu_1 < \mu_2$$

$$x_{ij} < \theta \rightarrow BG \\ \text{else} \rightarrow FG$$

a) $P(BG) = P(FG)$, $\sigma_1 = \sigma_2$, then show
 $\theta^* = \frac{\mu_1 + \mu_2}{2}$
 optimal.

$$p(x) = \frac{1}{\sigma \sqrt{2\pi}} \exp \left[-\frac{(x-\mu)^2}{2\sigma^2} \right] \quad \boxed{g(x) = P(x|w_1) P(w_1)}$$

$$\boxed{\log: \frac{-1 \log(2\pi)}{2} - \log(\sigma) - \frac{(x-\mu)^2}{2\sigma^2}}$$

$$g(x) = \frac{-(x^2 - 2\mu x + \mu^2)}{2\sigma^2} - \frac{1}{2} \log(2\pi) - \log(\sigma) + \log(P)$$

Decision Boundary: $g_1(x) = g_2(x)$

$$\begin{aligned} & -\frac{(x^2 - 2\mu_1 x + \mu_1^2)}{2\sigma_1^2} - \log(\sigma_1) + \frac{(x^2 - 2\mu_2 x + \mu_2^2)}{2\sigma_2^2} + \log(\sigma_2) \\ & + \log(P(BG)) - \log(P(FG)) \end{aligned}$$

$$\Rightarrow \frac{x^2}{2} \left(\frac{1}{\sigma_2^2} - \frac{1}{\sigma_1^2} \right) + \frac{x}{2} \left(\frac{2\mu_1}{\sigma_1^2} - \frac{2\mu_2}{\sigma_2^2} \right) + \frac{\mu_2^2}{2\sigma_2^2} - \frac{\mu_1^2}{2\sigma_1^2} + \log \left(\frac{\sigma_2}{\sigma_1} \right) + \log \left(\frac{P(B_H)}{P(F_H)} \right)$$

★ $\sigma_1 = \sigma_2 = \sigma$ and $P(B_H) = P(F_H)$
 x^2 terms vanishes.

Decision boundary : $\frac{x}{2\sigma^2} (2\mu_1 - 2\mu_2) + \frac{(\mu_2^2 - \mu_1^2)}{2\sigma^2} = 0$

$$\frac{(\mu_1 - \mu_2)}{\sigma^2} \left[\cancel{x} - \left(\frac{\mu_2 + \mu_1}{2} \right) \right] = 0$$

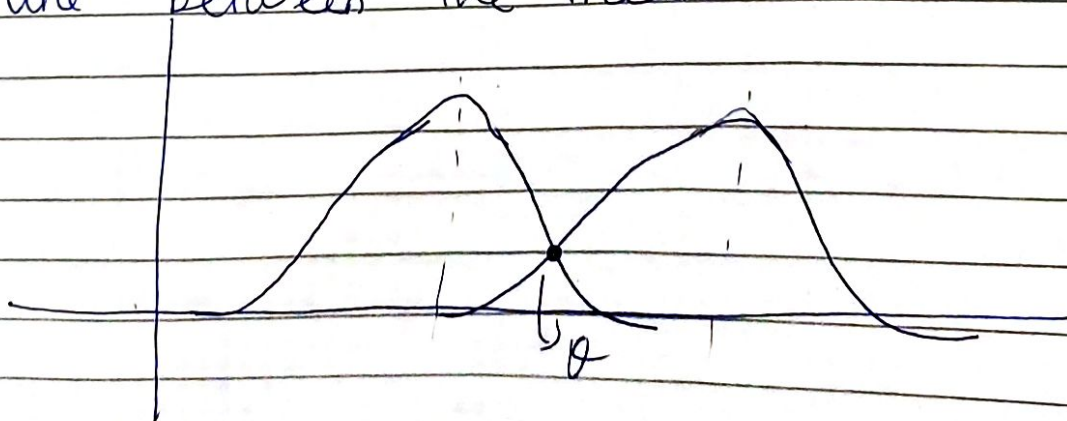
|| 0

now since $\mu_1 < \mu_2$ so $\mu_1 - \mu_2 \neq 0$

~~0~~ ~~0~~ ~~0~~ ~~0~~ ~~0~~

$$\boxed{\theta = \frac{\mu_2 + \mu_1}{2}}$$

This is optimal because ~~the~~ ~~the~~ decision boundary is perpendicular bisector of the line between the means



$$(C) \quad P(BA) = 4 P(FA)$$

$$\mu_1 = 100$$

$$\sigma_1 = \sigma_2$$

$$\mu_2 = 200$$

$$\Rightarrow \frac{2x(\mu_1 - \mu_2)}{2\sigma^2} + \frac{\mu_2^2 - \mu_1^2}{2\sigma^2} + \log\left(\frac{P(BA)}{P(FA)}\right) = 0$$

$$2x(-100) + (200-100)(300) + 2\sigma^2 \log(4) = 0$$

$$x = \frac{(30000 + 2\sigma^2 \log(4))}{200} = 0$$

$$0 = \frac{30000 + 2\sigma^2 \log_e(4)}{200}$$

(b)

PAGE NO.:

$$g_1(x) = -\frac{1}{2} \log(2\pi) - \log(\sigma_1) - \frac{(x - \mu_1)^2}{2\sigma_1^2} + \log(P_1)$$

$$g_1(x) = g_2(x)$$

$$-\log(\sigma_1) - \frac{(x - \mu_1)^2}{2\sigma_1^2} + \log(P(BH))$$

$$= -\log(\sigma_2) - \frac{(x - \mu_2)^2}{2\sigma_2^2} + \log(P(FH))$$

$$\frac{(\mu_1 + \mu_2 - \mu_2)^2}{2\sigma_2^2} - \frac{(\mu_1 + \mu_2 - \mu_1)^2}{2\sigma_1^2} + \log\left(\frac{\sigma_2}{\sigma_1}\right) + \log\left(\frac{P(BH)}{P(FH)}\right) = 0$$

$$\frac{(\mu_1 - \mu_2)^2}{8\sigma_2^2} - \frac{(\mu_2 - \mu_1)^2}{8\sigma_1^2} + \log\left(\frac{\sigma_2}{\sigma_1}\right) + \log\left(\frac{P(BH)}{P(FH)}\right) = 0$$

$$\frac{(\mu_1 - \mu_2)^2}{8} \left[\frac{1}{\sigma_2^2} - \frac{1}{\sigma_1^2} \right] + \log\left(\frac{\sigma_2}{\sigma_1} \times \frac{P(BH)}{P(FH)}\right) = 0$$

$$(\mu_1 - \mu_2)^2 = 8 \log\left(\frac{\sigma_1 \times P(FH)}{\sigma_2 \times P(BH)}\right) \times \frac{\sigma_2^2 \sigma_1^2}{(\sigma_1^2 - \sigma_2^2)}$$

$$\boxed{\mu_1 - \mu_2 = 2\sqrt{2} \sigma_1 \sigma_2 \sqrt{\log\left(\frac{\sigma_1 P(FH)}{\sigma_2 P(BH)}\right)} \times \frac{1}{\sqrt{\sigma_1^2 - \sigma_2^2}}}$$