

Name: Ruifeng Zhang

Student ID #: 861212163

Problem 1. (5 points)

$$p(x|y = A) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2\sigma^2}(x-\mu_A)^T(x-\mu_A)}$$

$$p(x|y = B) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2\sigma^2}(x-\mu_B)^T(x-\mu_B)}$$

Answer:

$$p(y = A|x) = \frac{\frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2\sigma^2}(x-\mu_A)^T(x-\mu_A)} p_A}{p(x)}$$

$$p(y = B|x) = \frac{\frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2\sigma^2}(x-\mu_B)^T(x-\mu_B)} p_B}{p(x)}$$

$$p(y = A|x) > p(y = B|x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2\sigma^2}(x-\mu_A)^T(x-\mu_A)} p_A > \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2\sigma^2}(x-\mu_B)^T(x-\mu_B)} p_B$$

$$= e^{-\frac{1}{2\sigma^2}(x-\mu_A)^T(x-\mu_A)} p_A > e^{-\frac{1}{2\sigma^2}(x-\mu_B)^T(x-\mu_B)} p_B$$

$$= -\frac{1}{2\sigma^2}(x-\mu_A)^T(x-\mu_A) + \ln(p_A) > -\frac{1}{2\sigma^2}(x-\mu_B)^T(x-\mu_B) + \ln(p_B)$$

$$= -\frac{1}{2\sigma^2}(x-\mu_A)^T(x-\mu_A) + \ln(p_A) > -\frac{1}{2\sigma^2}(x-\mu_B)^T(x-\mu_B) + \ln(p_B)$$

$$= -\frac{(x^T - \mu_A^T)(x - \mu_A)}{2\sigma^2} + \ln(p_A) > -\frac{(x^T - \mu_B^T)(x - \mu_B)}{2\sigma^2} + \ln(p_B)$$

$$= (x^T - \mu_A^T)(x - \mu_A) + \ln(p_A) > (x^T - \mu_B^T)(x - \mu_B) + \ln(p_B)$$

$$= x^T x - x^T \mu_A - \mu_A^T x + \mu_A^T \mu_A + \ln(p_A) > x^T x - x^T \mu_B - \mu_B^T x + \mu_B^T \mu_B + \ln(p_B)$$

$$= \ln(p_A) - x^T \mu_A - \mu_A^T x + \mu_A^T \mu_A > \ln(p_B) - x^T \mu_B - \mu_B^T x + \mu_B^T \mu_B$$

When $\ln(p_A) - x^T \mu_A - \mu_A^T x + \mu_A^T \mu_A = \ln(p_B) - x^T \mu_B - \mu_B^T x + \mu_B^T \mu_B$ there is a linear decision boundary