

Homework 6

Problem 3

w_1 has 7 points.

$(0,0), (0,1), (2,0), (3,2), (3,3), (2,2), (2,0)$

w_2 has 7 points.

$(7,7), (8,6), (9,7), (8,10), (7,10), (8,9), (7,11)$

a) Prior probabilities

$$P(w_1) = 7/14 = 1/2$$

$$P(w_2) = 7/14 = 1/2$$

$$b) \quad w_1 \rightarrow \frac{\sum_{i=1}^7 x_i}{7} = 1.7143 = \bar{x}$$

$$\frac{\sum_{i=1}^7 y_i}{7} = 1.428 = \bar{y}$$

$$\frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{7} = 0.8809 = \text{covariance}(x, y)$$

$$\text{Mean} = \begin{bmatrix} 1.7143 & 1.428 \end{bmatrix}$$

$$\text{Covariance matrix} = \begin{bmatrix} 1.5714 & 0.8809 \\ 0.8809 & 1.4762 \end{bmatrix}$$

$$\text{As } \text{var}(x) = \frac{\sum (x_i - \bar{x})^2}{7} = 1.5714$$

$$\text{var}(y) = \frac{\sum (y_i - \bar{y})^2}{7} = 1.4762$$

$$w_2 \rightarrow \frac{\sum_{i=1}^7 x_i^2}{7} = 7.7143$$

$$\frac{\sum_{i=1}^7 y_i^2}{7} = 8.5714$$

$$\text{Covariance} = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{7} = -0.6429$$

Since.

$$\text{Var}(x) = \frac{\sum (x_i - \bar{x})^2}{7} = 0.5714$$

$$\text{Var}(y) = \frac{\sum (y_i - \bar{y})^2}{7} = 3.6190$$

$$\text{mean} = \begin{bmatrix} 7.7143 & 8.5714 \end{bmatrix}$$

$$\text{Covariance matrix} = \begin{bmatrix} 0.5714 & -0.6429 \\ -0.6429 & 3.6190 \end{bmatrix}$$

c) The decision boundary is given as the function of two classes.

$$v_2(x, y)$$

$$P(w_1) \frac{e^{-1/2 (v-\mu_1)^T C_1^{-1} (v-\mu_1)}}{(\sqrt{2\pi})^2 |C_1|^{1/2}} = P(w_2) \frac{e^{-1/2 (v-\mu_2)^T C_2^{-1} (v-\mu_2)}}{(\sqrt{2\pi})^2 |C_2|^{1/2}}$$

$$\ln(P(w_1)) - \frac{1}{2} (v-\mu_1)^T C_1^{-1} (v-\mu_1) - \frac{1}{2} \ln |C_1|$$

$$= \ln(P(w_2)) - \frac{1}{2} (v-\mu_2)^T C_2^{-1} (v-\mu_2) - \frac{1}{2} \ln |C_2|$$

$$\ln(P(\omega_1)) = \ln(P(\omega_2))$$

$$-\frac{1}{2}(\underline{v}-\underline{\mu}_1)^T \underline{C}_1^{-1}(\underline{v}-\underline{\mu}_1) - \frac{1}{2} \ln|\underline{C}_1| = -\frac{1}{2}(\underline{v}-\underline{\mu}_2)^T \underline{C}_2^{-1}(\underline{v}-\underline{\mu}_2) - \frac{1}{2} \ln|\underline{C}_2|$$

$$\frac{1}{2}(\underline{v}-\underline{\mu}_1)^T \underline{C}_1^{-1}(\underline{v}-\underline{\mu}_1) + \frac{1}{2} \ln|\underline{C}_1| = \frac{1}{2}(\underline{v}-\underline{\mu}_2)^T \underline{C}_2^{-1}(\underline{v}-\underline{\mu}_2) + \frac{1}{2} \ln|\underline{C}_2|$$

$$(\underline{v}-\underline{\mu}_1)^T \underline{C}_1^{-1}(\underline{v}-\underline{\mu}_1) + \ln|\underline{C}_1| = (\underline{v}-\underline{\mu}_2)^T \underline{C}_2^{-1}(\underline{v}-\underline{\mu}_2) + \ln|\underline{C}_2|$$

$$\ln|\underline{C}_1| - \ln|\underline{C}_2| = 0.0695$$

$$\begin{bmatrix} x-1.714 & y-1.142 \end{bmatrix} \begin{bmatrix} 0.956 & -0.511 \\ -0.511 & 1.018 \end{bmatrix} \begin{bmatrix} x-1.714 \\ y-1.142 \end{bmatrix} = 0.0695$$

$$\begin{bmatrix} x-1.714 & y-1.142 \end{bmatrix} \begin{bmatrix} 0.956x + 1.639 - 0.511y + 0.652 \\ -0.511x + 0.979 + 1.018y - 1.163 \end{bmatrix}$$

$$\begin{bmatrix} x-1.714 & y-1.142 \end{bmatrix} \begin{bmatrix} 0.956x - 0.511y - 0.987 \\ -0.511x + 1.018y - 0.184 \end{bmatrix} = 0.0695$$

$$\Rightarrow 0.956x^2 + 0.987x - 0.511xy - 1.638x + 1.692 + 0.979y - 0.511xy + 1.018y^2 - 0.184y + 0.652x - 1.163y + 0.2101 - 0.0695$$

$$\begin{bmatrix} x-1.7142 & y-8.5114 \end{bmatrix} \begin{bmatrix} 2.187 & 0.388 \\ 0.388 & 0.345 \end{bmatrix} \begin{bmatrix} x-1.7142 \\ y-8.5114 \end{bmatrix}$$

$$\begin{bmatrix} x-1.7142 & y-8.5114 \end{bmatrix} \begin{bmatrix} 2.187x - 16.811 + 0.388y - 3.308 \\ 0.388x - 2.996 + 0.345y - 2.941 \end{bmatrix}$$

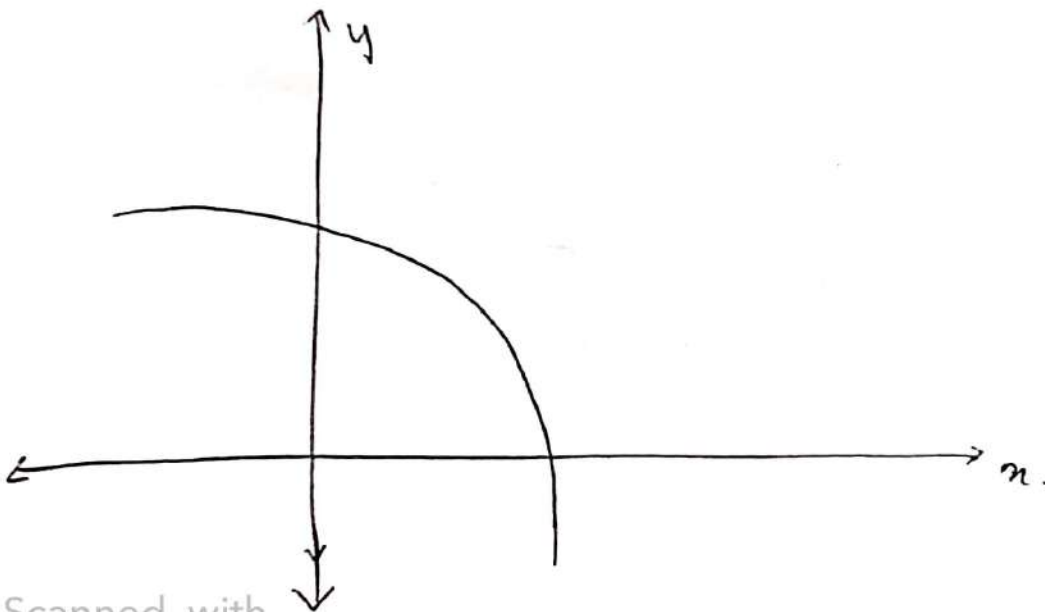
$$[x - 7.7142 \quad y - 8.5711] \begin{bmatrix} 2.181x + 0.388y - 20.179 \\ 0.388x + 0.345y - 5.937 \end{bmatrix}$$

$$\Rightarrow 2.181x^2 + 0.388xy - 20.179x - 16.811x + 155.655 - 2.996y + 0.388xy - 5.937y + 0.345y^2 - 3.329x + 50.888 - 2.9597y$$

The final equation is.

$$1.231x^2 - 0.6727y^2 + 1.9188xy - 40.3801x - 11.5247y + 204.7203 = 0$$

The decision boundary is as follows:



c) Misclassification of w_1 into w_2 is two times costlier than misclassification of w_2 into w_1 .

To fix this issue, we use prior probabilities

$$P(w_1) = 2P(w_2)$$

$$P(w_1) = 2/3, \quad P(w_2) = 1/3$$

From the previous equations.

$$\ln(P(w_1)) - \ln(P(w_2)) = \ln\left(\frac{P(w_1)}{P(w_2)}\right) = \ln(2)$$

The new decision boundary is

$$\left(\frac{1}{2}\right)(1.231x^2 - 0.6727y^2 + 1.9188xy - 40.3801x - 11.5747y + 204.7203) + \ln(2) = 0.$$

Hence the boundary shifts, and the size is contracted, while its shape and orientation is preserved.

Mathematically, define a discriminant function say g :

$$g(x) = \begin{cases} w_2 & \text{when } x \in \Omega_1 \\ w_1 & \text{when } x \in \Omega_2 \end{cases} \quad \left| \begin{array}{l} \text{Are our possible errors defined} \\ \text{in terms of } g. \end{array} \right.$$

w_1, w_2 are 2 classes in pattern space Ω with continuous PDFs $p_1(x)$ and $p_2(x)$ respectively. We can see this as dividing space Ω into Ω_1 & Ω_2 [$\Omega_1 \cup \Omega_2 = \Omega$ & $\Omega_1 \cap \Omega_2 = \emptyset$]

$$\text{Probability of error 1} = \int_{\Omega_2} p_1(x) dx$$

$$\text{Probability of error 2} = \int_{\Omega_1} p_2(x) dx$$

The cost of misclassification of w_1 into w_2 is twice as costly as misclassifying w_2 into w_1 .

$$C_1 = \text{cost of } w_1 \text{ into } w_2 = 2/3$$

$$\text{cost of } w_2 \text{ into } w_1 = 1/3$$

Now, the total expected error.

$$E = \frac{2}{3} P(w_1) \int_{\Omega_2} p_1(x) dx + \frac{1}{3} P(w_2) \int_{\Omega_1} p_2(x) dx$$
$$= \frac{1}{3} \left[2P(w_1) \int_{\Omega_2} p_1(x) dx + (1 - P(w_1)) \int_{\Omega_1} p_2(x) dx \right]$$

which we have to minimize.

The variables here are $P(w_1)$, $P(w_2)$. Ω_1 , Ω_2 depend on P_1 & P_2

We can see from the graphs that a change in decision boundary is caused by the change in error function



(d) Equation: $x_1 = -0.779366368805849 \cdot x_2 - 4.06173842404549e-5 \cdot \sqrt{699416824.0 \cdot x_2^{**2} - 9821504896.0 \cdot x_2 + 62250971881.0} + 16.4013403736799$
(e) Equation: $x_1 = -0.779366368805849 \cdot x_2 - 4.59533245287764e-7 \cdot \sqrt{5464193937500.0 \cdot x_2^{**2} - 76730507000000.0 \cdot x_2 + 481002816699879.0} + 16.4013403736799$

