

Exercise 1

$$\begin{aligned}
 1 \ a) \quad S_B &= \begin{pmatrix} (\mu_{1x} - \mu_{2x}) \\ (\mu_{1y} - \mu_{2y}) \end{pmatrix} \begin{pmatrix} (\mu_{1x} - \mu_{2x}) & (\mu_{1y} - \mu_{2y}) \end{pmatrix} \\
 &= \begin{pmatrix} -2.33 \\ -2.40 \end{pmatrix} \begin{pmatrix} -2.33 & -2.40 \end{pmatrix} \\
 &= \begin{pmatrix} (-2.33)^2 & (-2.33)(-2.40) \\ (-2.40)(-2.33) & (-2.40)^2 \end{pmatrix} \\
 &= \begin{pmatrix} 5.43 & 5.59 \\ 5.59 & 5.76 \end{pmatrix}
 \end{aligned}$$

b) rank of $S_B = 1$

Figure	$(\tilde{\mu}_1 - \tilde{\mu}_2)^2$	$\hat{\sigma}_1^2$	$\hat{\sigma}_2^2$	w	$J(w)$
3 (w_1)	11.23	1.40	2.03	$\begin{pmatrix} 0.70 \\ 0.72 \end{pmatrix}$	3.21
4 (w_2)	0	0.28	0.74	$\begin{pmatrix} 0.72 \\ -0.70 \end{pmatrix}$	0
5 (w_3)	3.04	0.16	0.06	$\begin{pmatrix} 0.25 \\ -0.97 \end{pmatrix}$	13.81

b) Figure 4 has $(\tilde{\mu}_1 - \tilde{\mu}_2)^2 = 0$, as both the projection lies on the same point i.e. (0,0)

Figure 3 has $(\tilde{\mu}_1 - \tilde{\mu}_2)^2 = (\mu_1 - \mu_2)^2$, since the projection is exactly where it lies, thus the original distance is what we need:

$$\therefore \left\| \begin{pmatrix} 2.34 \\ 2.40 \end{pmatrix} \right\|_2^2 = 5.475 + 5.76 = 11.23$$

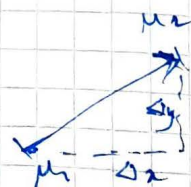
Figs:

$$\begin{aligned}
 \tilde{\mu}_1 &= w^T \cdot \mu_1 = (0.25 - 0.97) \begin{pmatrix} 1.17 \\ 1.20 \end{pmatrix} = -0.872 \\
 \tilde{\mu}_2 &= w^T \cdot \mu_2 = 0.872 \\
 \therefore (\mu_1 - \mu_2)^2 &= (-1.741)^2 = 3.04
 \end{aligned}$$

c) $w_1 = \frac{\Delta x}{\|\mu_2 - \mu_1\|}, \frac{\Delta y}{\|\mu_2 - \mu_1\|} = \frac{2.34}{\sqrt{11.23}}, \frac{2.40}{\sqrt{11.23}}$

$$= 0.698, 0.716$$

w_2 : Since $w_2^T \cdot \mu_1 = w_2^T \cdot \mu_2 = 0$. Thus the $\vec{w}_2 \perp \vec{w}_1$, from perception



chapter.

d) $J(w) = \frac{(\tilde{\mu}_1 - \tilde{\mu}_2)^2}{\tilde{\sigma}_1^2 + \tilde{\sigma}_2^2}$

Values are shown in the table

3) As can be seen in figures 3. Even after achieving a maximum squared mean distance, class separation is not perfect. We do have some misclassification. Therefore as shown in figure 5, we also need to minimize the variances within the class.

4) a) $\lambda_2 = 0$, because there can be at most 1 ~~eigenvalue~~ non-zero eigenvalue as rank of $L_B = 1$.

This is not specific to this subset but rather a general rule.

b) $\tilde{x}_1 = \begin{pmatrix} 0.25 & -0.97 \end{pmatrix} \begin{pmatrix} 0.38 \\ -0.42 \end{pmatrix} = 0.095 + 1.39 = 1.485$

$\tilde{x}_2 = \begin{pmatrix} 0.25 & -0.97 \end{pmatrix} \begin{pmatrix} 1.88 \\ 1.08 \end{pmatrix} = 0.47 - 1.047 = -0.58$

c) y is more important for class discrimination because ~~it produces high resultant value~~ than ~~for same x but different y~~ , the points can belong to different class in ~~proj~~ projected plane, but for same y & different x we cannot say so. Als. because in the resultant \tilde{x}_1 , y 's contribution is more than x 's contribution.

d) $w_0 = (-0.05)$