

Άσκηση ①

$$P(\vec{X} | w_1) = N(\mu_1, \Sigma_1) = f_{x_1, x_2 | w = w_1}(x_1, x_2)$$

$$\vec{X} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad \left| \quad f_{x_1, x_2 | w = w_1}(x_1, x_2) = f_{\vec{X} | w = w_1}(\vec{y}) \right.$$

$$\vec{y} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$$

$$f_{\vec{X} | w = w_1}(\vec{y}) = \frac{1}{\sqrt{(2\pi)^2 |\Sigma_1|}} e^{-\frac{1}{2} (\vec{y} - \vec{\mu}_1)^T \Sigma_1^{-1} (\vec{y} - \vec{\mu}_1)}$$

$$\vec{\mu}_1 = \begin{bmatrix} 3 \\ 3 \end{bmatrix}, \quad \Sigma_1 = \begin{pmatrix} 1.2 & -0.4 \\ -0.4 & 1.2 \end{pmatrix}$$

①

ομοίως, $f_{\vec{X} | w = w_2}(\vec{y}) = \frac{1}{\sqrt{(2\pi)^2 |\Sigma_2|}} e^{-\frac{1}{2} (\vec{y} - \vec{\mu}_2)^T \Sigma_2^{-1} (\vec{y} - \vec{\mu}_2)}$

Πιθανότητα Σφάλματος :

$$P_e = P(\hat{w} \neq w) = P(w = w_1) p(\hat{w} \neq w | w = w_1) + P(w = w_2) p(\hat{w} \neq w | w = w_2)$$

Αντάδην, η πιθανότητα να εκπληρωθεί λάθος είναι :

Πιθανότητα $w = w_1$ × Πιθανότητα λάθους ($\hat{w} = w_2$), δεδομένου ότι $w = w_1$ ⊕ Πιθανότητα $w = w_2$ × Πιθανότητα $\hat{w} = w_1 | w = w_2$

Πιθανότητα λάθους δεδομένου $w = w_1$: $p(\hat{w} = w_2 | w = w_1)$

_____ || _____ $w = w_2$: $p(\hat{w} = w_1 | w = w_2)$

$$P(\hat{w} \neq w_1 | w = w_1) = 1 - P(\hat{w} = w_1 | w = w_1)$$

$$\begin{aligned}
 \text{Area: } P_e &= P(w=w_1) [1 - P(\hat{w}=w_1 | w=w_1)] + P(w=w_2) P(\hat{w}=w_1 | w=w_2) \\
 &= P(w=w_1) - P(w=w_1) P(\hat{w}=w_1 | w=w_1) + P(w=w_2) P(\hat{w}=w_1 | w=w_2) \\
 &= P(w=w_1) - P(w=w_1) \int_{R_1} f_{\vec{z}|w_1}(\vec{y}) d\vec{y} + P(w=w_2) \int_{R_1} f_{\vec{z}|w_2}(\vec{y}) d\vec{y} = \\
 &P(w=w_1) + \int_{R_1} [f_{\vec{z}|w_2}(\vec{y}) P(w=w_2) - f_{\vec{z}|w_1}(\vec{y}) P(w=w_1)] d\vec{y} = \\
 &P(w=w_1) + \int_{R_1} [P(w=w_2 | \vec{z}=\vec{y}) - P(w=w_1 | \vec{z}=\vec{y})] f_{\vec{z}}(\vec{y}) d\vec{y}
 \end{aligned}$$

$$R_1 = \{ (x_1, x_2) \in \mathbb{R}^2 : P(w=w_2 | \vec{z}=\vec{y}) < P(w=w_1 | \vec{z}=\vec{y}) \}$$

$$R_2 = \{ (x_1, x_2) \in \mathbb{R}^2 : P(w=w_2 | \vec{z}=\vec{y}) > P(w=w_1 | \vec{z}=\vec{y}) \}$$

$$\vec{y} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

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Karónas Aréwagms:

$$P(w=w_2 | \vec{z} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}) \stackrel{w_2}{\underset{w_1}{>}} P(w=w_1 | \vec{z} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix})$$

$$P(w=w_2 | \vec{z} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}) f_{\vec{z}}(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}) = P(w=w_2) f_{\vec{z}|w_2}(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}) \Rightarrow$$

$$P(w=w_2 | \vec{z} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}) = \frac{P(w=w_2) f_{\vec{z}|w_2}(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix})}{f_{\vec{z}}(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix})}$$

$$\text{opoiws } P(w=w_1 | \vec{z} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}) = \frac{P(w=w_1) f_{\vec{z}|w_1}(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix})}{f_{\vec{z}}(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix})}$$

$$P(w=w_1) = P(w_1)$$

$$P(w=w_2) = P(w_2)$$

$$\text{also } P(w_2) f_{\vec{z}|w_2} \left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \right) \stackrel{w_2}{\gtrless} \sum_{w_1} P(w_1) f_{\vec{z}|w_1} \left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \right) \Rightarrow$$

$$P(w_2) \frac{1}{\sqrt{(2\pi)^2 |\Sigma_2|}} e^{-\frac{1}{2} (\vec{y} - \vec{\mu}_2)^T \Sigma_2^{-1} (\vec{y} - \vec{\mu}_2)} \stackrel{w_2}{\gtrless} \sum_{w_1} P(w_1) \frac{1}{\sqrt{(2\pi)^2 |\Sigma_1|}} e^{-\frac{1}{2} (\vec{y} - \vec{\mu}_1)^T \Sigma_1^{-1} (\vec{y} - \vec{\mu}_1)}$$

$$\frac{P(w_2)}{P(w_1)} \sqrt{\frac{|\Sigma_1|}{|\Sigma_2|}} = k \in \mathbb{R}$$

(3)

$$\vec{y} - \vec{\mu}_2 = \begin{bmatrix} x_1 - 6 \\ x_2 - 6 \end{bmatrix}, \quad (\vec{y} - \vec{\mu}_2)^T = \begin{bmatrix} x_1 - 6 & x_2 - 6 \end{bmatrix}$$

$$\Sigma_2^{-1} = \frac{1}{(1.2)^2 - (0.4)^2} \begin{bmatrix} 1.2 & -0.4 \\ -0.4 & 1.2 \end{bmatrix} = \begin{bmatrix} 0.9375 & -0.3125 \\ -0.3125 & 0.9375 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{15}{16} & -5/16 \\ -5/16 & 15/16 \end{bmatrix} = \frac{1}{16} \begin{bmatrix} 15 & -5 \\ -5 & 15 \end{bmatrix}$$

$$(\vec{y} - \vec{\mu}_2)^T \Sigma_2^{-1} = \begin{bmatrix} x_1 - 6 & x_2 - 6 \end{bmatrix}_{1 \times 2} \begin{bmatrix} \frac{15}{16} & -\frac{5}{16} \\ -\frac{5}{16} & \frac{15}{16} \end{bmatrix}_{2 \times 2} =$$

$$\left[(x_1 - 6) \frac{15}{16} - \frac{5}{16} (x_2 - 6), (x_2 - 6) \frac{15}{16} - \frac{5}{16} (x_1 - 6) \right]$$

$$(\vec{y} - \vec{\mu}_2)^T \Sigma_2^{-1} (\vec{y} - \vec{\mu}_2) = \left[(x_1 - 6) \frac{15}{16} - \frac{5}{16} (x_2 - 6), (x_2 - 6) \frac{15}{16} - \frac{5}{16} (x_1 - 6) \right] \begin{bmatrix} x_1 - 6 \\ x_2 - 6 \end{bmatrix} =$$

$$\frac{15}{16} (x_1-6)^2 - \frac{5}{16} (x_1-6)(x_2-6) - \frac{5}{16} (x_1-6)(x_2-6) + \frac{15}{16} (x_2-6)^2 =$$

$$\frac{15}{16} [(x_1-6)^2 + (x_2-6)^2] - \frac{10}{16} (x_1-6)(x_2-6)$$

$$\text{opoints : } \Sigma_1^{-1} = \frac{1}{16} \begin{bmatrix} 15 & 5 \\ 5 & 15 \end{bmatrix} \quad (4)$$

$$\text{και } (\vec{y} - \vec{\mu}_1)^T \Sigma_1^{-1} (\vec{y} - \vec{\mu}_1) = \frac{15}{16} [(x_1-3)^2 + (x_2-3)^2] + \frac{10}{16} (x_1-3)(x_2-3)$$

Αρα :

$$k e^{-\frac{1}{2} (\vec{y} - \vec{\mu}_2)^T \Sigma_2^{-1} (\vec{y} - \vec{\mu}_2)} \omega_2 \gtrless e^{-\frac{1}{2} (\vec{y} - \vec{\mu}_1)^T \Sigma_1^{-1} (\vec{y} - \vec{\mu}_1)} \Rightarrow$$

$$\ln \left(k e^{-\frac{1}{2} (\vec{y} - \vec{\mu}_2)^T \Sigma_2^{-1} (\vec{y} - \vec{\mu}_2)} \right) \omega_2 \gtrless \ln \left(e^{-\frac{1}{2} (\vec{y} - \vec{\mu}_1)^T \Sigma_1^{-1} (\vec{y} - \vec{\mu}_1)} \right) \Rightarrow$$

$$\ln(k) - \frac{1}{2} \left[\frac{15}{16} [(x_1-6)^2 + (x_2-6)^2] - \frac{10}{16} (x_1-6)(x_2-6) \right] \omega_2 \gtrless -\frac{1}{2} \left[\frac{15}{16} [(x_1-3)^2 + (x_2-3)^2] + \frac{10}{16} (x_1-3)(x_2-3) \right] \omega_1$$

$$\begin{aligned}
& \ln(k) + \left[-\frac{15}{32} [(x_1-6)^2 + (x_2-6)^2] + \frac{10}{32} (x_1-6)(x_2-6) \right] \underset{\omega_1}{\overset{\omega_2}{\gtrless}} -\frac{15}{32} [(x_1-3)^2 + (x_2-3)^2] - \frac{10}{32} (x_1-3)(x_2-3) \Rightarrow \\
& \ln(k) - \frac{15}{32} (x_1^2 - 12x_1 + 36) - \frac{15}{32} (x_2^2 - 12x_2 + 36) + \frac{10}{32} (x_1x_2 - 6x_1 - 6x_2 + 36) \underset{\omega_1}{\overset{\omega_2}{\gtrless}} -\frac{15}{32} (x_1^2 - 6x_1 + 9) - \frac{15}{32} (x_2^2 - 6x_2 + 9) - \frac{10}{32} (x_1x_2 - 3x_1 - 3x_2 + 9) \Rightarrow \\
& \ln(k) + \frac{15 \cdot 12 x_1}{32} + \frac{15 \cdot 12 x_2}{32} - \frac{15 \cdot 36}{16} + \frac{10 x_1 x_2}{32} - \frac{6 \cdot 10}{32} x_1 - \frac{6 \cdot 10}{32} x_2 + \frac{10 \cdot 36}{32} \underset{\omega_1}{\overset{\omega_2}{\gtrless}} + \frac{6 \cdot 15 x_1}{32} + \frac{6 \cdot 15 x_2}{32} - \frac{9 \cdot 15}{16} - \frac{10}{32} x_1 x_2 + \frac{3 \cdot 10}{32} x_1 + \frac{3 \cdot 10}{32} x_2 - \frac{9 \cdot 10}{32} \Rightarrow \\
& \ln(k) + \frac{120}{32} x_1 + \frac{120}{32} x_2 + \frac{5 \cdot 36}{16} - \frac{15 \cdot 36}{16} + \frac{10}{32} x_1 x_2 \underset{\omega_1}{\overset{\omega_2}{\gtrless}} + \frac{120}{32} x_1 + \frac{120}{32} x_2 - \frac{9 \cdot 5}{16} - \frac{9 \cdot 15}{16} - \frac{10}{32} x_1 x_2 \Rightarrow \\
& \ln(k) - \frac{10 \cdot 36}{16} + \frac{10}{32} x_1 x_2 \underset{\omega_1}{\overset{\omega_2}{\gtrless}} - \frac{20 \cdot 9}{16} - \frac{10}{32} x_1 x_2 \Rightarrow \quad (5) \\
& \frac{20}{32} x_1 x_2 \underset{\omega_1}{\overset{\omega_2}{\gtrless}} \frac{10 \cdot 36 - 20 \cdot 9}{16} - \ln(k) \Rightarrow \frac{5}{8} x_1 x_2 \underset{\omega_1}{\overset{\omega_2}{\gtrless}} 11.25 - \ln(k) \Rightarrow \\
& x_1 x_2 \underset{\omega_1}{\overset{\omega_2}{\gtrless}} 18 - \frac{8}{5} \ln(k)
\end{aligned}$$

$$\text{για } \Sigma_1 = \Sigma_2 = \begin{pmatrix} 1.2 & 0.4 \\ 0.4 & 1.2 \end{pmatrix} = \Sigma$$

⑥

$$\Sigma^{-1} = \frac{1}{16} \begin{bmatrix} 15 & -5 \\ -5 & 15 \end{bmatrix}$$

$$(\vec{y} - \vec{\mu}_1)^T \Sigma^{-1} (\vec{y} - \vec{\mu}_2) = \frac{15}{16} [(x_1 - 3)^2 + (x_2 - 3)^2] - \frac{10}{16} (x_1 - 3)(x_2 - 3)$$

ΕΧΟΥΜΕ :

$$l_n(k) = \frac{15}{32} [(x_1 - 6)^2 + (x_2 - 6)^2] + \frac{10}{32} (x_1 - 6)(x_2 - 6) \underset{\omega_1}{\overset{\omega_2}{>}} - \frac{15}{32} [(x_1 - 3)^2 + (x_2 - 3)^2] + \frac{10}{32} (x_1 - 3)(x_2 - 3) \Rightarrow$$

$$32 \ln(k) - 15(x_1^2 - 12x_1 + x_2^2 - 12x_2 + 72) + 10(x_1x_2 - 6x_1 - 6x_2 + 36) \sum_{\omega_1}^{\omega_2} - 15(x_1^2 - 6x_1 - 6x_2 + x_2^2 + 18) + 10(x_1x_2 - 3x_1 - 3x_2) \sum_{\omega_1}^{\omega_2}$$

$$\Rightarrow 32 \ln(k) + 15 \cdot 12x_1 + 15 \cdot 12x_2 - 15 \cdot 72 + 10x_1x_2 - 60x_1 - 60x_2 + 360 \sum_{\omega_1}^{\omega_2} 15 \cdot 6x_1 + 15 \cdot 6x_2 - 18 \cdot 15 + 10x_1x_2 - 30x_1 - 30x_2 \sum_{\omega_1}^{\omega_2}$$

$$\Rightarrow 32 \ln(k) + 120x_1 + 120x_2 - 720 \sum_{\omega_1}^{\omega_2} 60x_1 + 60x_2 - 180 \Rightarrow \textcircled{7}$$

$$32 \ln(k) + 60x_1 + 60x_2 \sum_{\omega_1}^{\omega_2} 540 \Rightarrow$$

$$60(x_1 + x_2) \sum_{\omega_1}^{\omega_2} 540 - 32 \ln(k) \Rightarrow$$

$$x_1 + x_2 \sum_{\omega_1}^{\omega_2} 9 - \frac{32}{60} \ln(k) \Rightarrow$$

$$x_1 + x_2 \sum_{\omega_1}^{\omega_2} 9 - \frac{8}{15} \ln(k)$$

Άσκηση ③

$$P(w_1) = \pi = P(w_2)$$

$$P(x|w_i) = \begin{cases} \frac{x}{\sigma_i^2} e^{-\frac{x^2}{2\sigma_i^2}}, & x \geq 0 \\ 0, & \text{αλλιώς} \end{cases}$$

$$\sigma_1 = 1, \sigma_2 = 2 \quad L = \begin{bmatrix} 0 & 0.5 \\ 1 & 0 \end{bmatrix}$$

$$\begin{cases} t_1 = A_{21} P(w_2|x) \\ t_2 = A_{12} P(w_1|x) \end{cases} \Rightarrow \begin{cases} t_1 = \pi P(x|w_2) \\ t_2 = \frac{\pi}{2} P(x|w_1) \end{cases}$$

$$\begin{aligned} P(w_2|x) &= P(w_2)P(x|w_2) \\ P(w_1|x) &= P(w_1)P(x|w_1) \end{aligned}$$

$$t_2 \underset{w_2}{\underset{w_1}{>}} t_1 \Rightarrow \frac{\pi}{2} P(x|w_1) \underset{w_2}{\underset{w_1}{>}} \pi P(x|w_2) \Rightarrow$$

$$\frac{1}{2} e^{-\frac{x^2}{2}} \underset{w_2}{\underset{w_1}{>}} \frac{1}{4} e^{-\frac{x^2}{8}} \Rightarrow e^{-\frac{x^2}{2}} \underset{w_2}{\underset{w_1}{>}} \frac{1}{2} e^{-\frac{x^2}{8}} \rightarrow$$

$$-\frac{x^2}{2} \underset{w_2}{\underset{w_1}{>}} \ln\left(\frac{1}{2} e^{-\frac{x^2}{8}}\right) \Rightarrow -\frac{x^2}{2} \underset{w_2}{\underset{w_1}{>}} \ln\left(\frac{1}{2}\right) + \ln\left(e^{-\frac{x^2}{8}}\right)$$

$$\Rightarrow -\frac{x^2}{2} \underset{w_2}{\underset{w_1}{>}} -\frac{x^2}{8} - \ln(2) \Rightarrow \ln(2) \underset{w_2}{\underset{w_1}{>}} \frac{x^2}{2} - \frac{x^2}{8} \Rightarrow$$

$$\ln(2) \underset{w_2}{\underset{w_1}{>}} 3\frac{x^2}{8} \Rightarrow x \underset{w_2}{\underset{w_1}{>}} \sqrt{\frac{8\ln(2)}{3}} \quad (x < 0 \text{ απερρίκνεται})$$

$$\text{όρα το όριο είναι } x_0 = \sqrt{\frac{8\ln(2)}{3}}$$

Exercise 4

(1)

$$X = \begin{bmatrix} 1 & 0 & 1 \\ -1 & 1 & 0 \end{bmatrix}$$

$$X^T X = \begin{bmatrix} 1 & -1 \\ 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ -1 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 2 & -1 & 1 \\ -1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}$$

Finding the eigenvalues :

$$\det(s\mathbf{1} - X^T X) = 0 \implies \begin{vmatrix} s-2 & -1 & 1 \\ -1 & s-1 & 0 \\ 1 & 0 & s-1 \end{vmatrix} = 0 \implies s^3 - 4s^2 + 3s = 0 \implies$$

$$s(s^2 - 4s + 3) = 0 \implies s = 0 \text{ or } s = 1 \text{ or } s = 3$$

Non-zero (and normalized) eigenvectors :

$$\text{for } s = 1 : V_1 = \begin{bmatrix} 0 \\ 0.7071 \\ 0.7071 \end{bmatrix}$$

$$\text{for } s = 3 : V_2 = \begin{bmatrix} 0.8165 \\ -0.4082 \\ 0.4082 \end{bmatrix}$$

(2)

$$\sigma_1 = 1, \sigma_2 = \sqrt{3}$$

(3)

$$X X^T = \begin{bmatrix} 1 & 0 & 1 \\ -1 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 0 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix}$$

$$\det(s\mathbf{1} - X X^T) = 0 \implies \begin{vmatrix} s-2 & -1 \\ -1 & s-2 \end{vmatrix} = 0 \implies s^2 - 4s + 3 = 0 \implies$$

$$s = 1 \text{ or } s = 3$$

$$\text{for } s_1 = 1 : v_1 = \begin{bmatrix} -0.7071 \\ -0.7071 \end{bmatrix}$$

$$\text{for } s_2 = 3 : v_2 = \begin{bmatrix} -0.7071 \\ 0.7071 \end{bmatrix}$$

(4)

$$\hat{X} = v_2 \sigma_2 V_2^T = \begin{bmatrix} 1 & -0.5 & 0.5 \\ -1 & 0.5 & -0.5 \end{bmatrix}$$

Exercise 6

$$P(\vec{X}|\omega_i) = \frac{1}{\sqrt{(2\pi^2|\Sigma_i|)}} e^{-\frac{1}{2}(\vec{x}-\mu_i)^T \Sigma_i^{-1}(\vec{x}-\mu_i)}$$

$$m_1 = E\{\vec{y}\} = E\{\vec{w}^T \vec{X}\} = \vec{w}^T E\{\vec{X}\} = \vec{w}^T \vec{\mu}_1$$

$$m_2 = \vec{w}^T \vec{\mu}_2$$

$$S_1^2 = E\{(\vec{y} - \vec{m}_1)(\vec{y} - \vec{m}_1)^T\} = E\{\vec{w}^T(\vec{x} - \vec{\mu}_1)(\vec{x} - \vec{\mu}_1)\vec{w}\} = \vec{w}^T \Sigma_1 \vec{w}$$

$$S_2^2 = \vec{w}^T \Sigma_2 \vec{w}$$

$$J(w) = \frac{(m_1 - m_2)^2}{S_1^2 + S_2^2} = \frac{\vec{w}^T S_b \vec{w}}{\vec{w}^T S_w \vec{w}} \implies \vec{w} = \beta S_w^{-1} S_b$$

$$S_w = \Sigma_1 + \Sigma_2, \quad S_b = \vec{\mu}_1 - \vec{\mu}_2$$

We want to minimize the function $J(w)$:

$$\vec{w} = \beta(\Sigma_1 + \Sigma_2)^{-1}(\vec{\mu}_1 - \vec{\mu}_2), \text{ for } b > 0$$

$$\vec{w} = \beta \begin{bmatrix} 13 & 9 \\ 9 & 13 \end{bmatrix}^{-1} \begin{bmatrix} -15 \\ -10 \end{bmatrix}$$

$$\text{choose } \beta = ad - bc = 13 * 13 - 9 * 9 = 88$$

$$\vec{w} = \begin{bmatrix} -114 \\ 18 \end{bmatrix}$$