

#1 (1) $A\mathbf{x} = \lambda\mathbf{x}$

$$|A - \lambda I| = 0 \Leftrightarrow \begin{vmatrix} 1-\lambda & r \\ r & 1-\lambda \end{vmatrix} = 0$$

$$\Leftrightarrow (1-\lambda)^2 - r^2 = 0$$

$$\Leftrightarrow (1-\lambda+r)(1-\lambda-r) = 0$$

$$\Leftrightarrow \lambda = 1+r \text{ or } 1-r, \text{ where } -1 < r < 1$$

$$A\mathbf{x} = \lambda\mathbf{x} \Leftrightarrow \begin{pmatrix} 1 & r \\ r & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} \lambda x_1 \\ \lambda x_2 \end{pmatrix}$$

$$\Leftrightarrow \begin{cases} (1-\lambda)x_1 + rx_2 = 0 \\ rx_1 + (1-\lambda)x_2 = 0 \end{cases}$$

① $\lambda = 1+r$

$$\begin{cases} -rx_1 + rx_2 = 0 \\ rx_1 - rx_2 = 0 \end{cases} \Leftrightarrow x_1 = x_2 \text{ if } r \neq 0$$

$$\mathbf{p}_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

② $\lambda = 1-r$

$$\begin{cases} rx_1 + rx_2 = 0 \\ rx_1 + rx_2 = 0 \end{cases} \Leftrightarrow x_1 = -x_2 \text{ if } r \neq 0$$

$$\mathbf{p}_2 = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

(2)

$$\mathbf{p}_1^T \mathbf{p}_2 = (1 \ 1) \begin{pmatrix} 1 \\ -1 \end{pmatrix} = 1-1=0$$

(3) $\mathbf{x}^T A \mathbf{x} = 6$ for $r = 1/2$

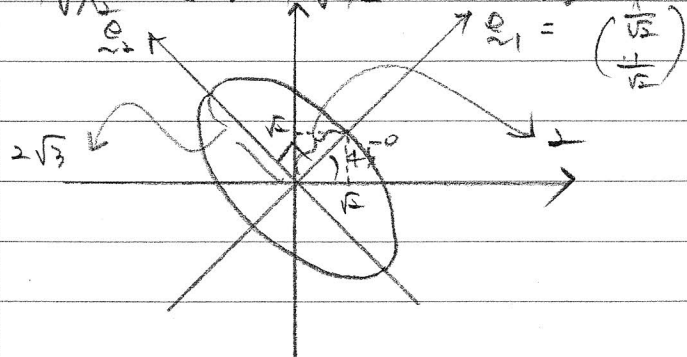
$$\lambda_1 = \frac{3}{2} > \lambda_2 = \frac{1}{2} > 0 \text{ \& } \mathbf{e}_1 = \begin{pmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix}, \mathbf{e}_2 = \begin{pmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{pmatrix}$$

→)

$$\underline{x}^T A \underline{x} = \underline{x}^T \begin{pmatrix} 1 & 1/2 \\ 1/2 & 1 \end{pmatrix} \underline{x} = 6 = c^2$$

$$c/\sqrt{\lambda_1} = \sqrt{6}/\sqrt{3/2} = 2$$

$$c/\sqrt{\lambda_2} = \sqrt{6}/\sqrt{1/2} = 2\sqrt{3}$$



#2 skip.

#4 For exercises 1,

$\lambda_1 = \frac{3}{2} > \lambda_2 = \frac{1}{2}$ & corresponding eigenvectors are

$$\underline{e}_1 = \begin{pmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix}, \quad \underline{e}_2 = \begin{pmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{pmatrix}$$

#7 For (2), use $AB = (I \otimes J)(J \otimes I)$

$$= IJ \otimes JI = J \otimes J$$

$$3. \quad f(x) = 4x_1^2 + 4x_2^2 + 6x_1x_2 = (x_1 \ x_2) \begin{pmatrix} 4 & 3 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

Note that $\begin{pmatrix} 4 & 3 \\ 3 & 4 \end{pmatrix}$ is p.d.

$$(\because) \quad |4| = 4 > 0, \quad \begin{vmatrix} 4 & 3 \\ 3 & 4 \end{vmatrix} = 16 - 9 > 0$$

By the maximization of Quadratic forms for points on the unit sphere,

$$\max_{x \neq 0} \frac{x'Ax}{x'x} = \lambda_1 \quad (\text{attained when } x = e_1)$$

$$\min_{x \neq 0} \frac{x'Ax}{x'x} = \lambda_p \quad (\text{attained when } x = e_p),$$

$$\text{where } \lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_p \geq 0$$

Eigenvalues of $A = \begin{pmatrix} 4 & 3 \\ 3 & 4 \end{pmatrix}$ are $\lambda_1 = 7 > 1 = \lambda_2$

$$\text{Hence } \max_{x \neq 0} f(x) = \lambda_1 = 7 \quad \leftarrow$$

$$\min_{x \neq 0} f(x) = \lambda_2 = 1$$

$$5. n=3, S = I - \frac{1}{n} U = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} - \begin{pmatrix} \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \end{pmatrix} = \begin{pmatrix} \frac{2}{3} & -\frac{1}{3} & -\frac{1}{3} \\ -\frac{1}{3} & \frac{2}{3} & -\frac{1}{3} \\ -\frac{1}{3} & -\frac{1}{3} & \frac{2}{3} \end{pmatrix} \quad 4)$$

$$|A - \lambda I| = \begin{vmatrix} \frac{2}{3} - \lambda & -\frac{1}{3} & -\frac{1}{3} \\ -\frac{1}{3} & \frac{2}{3} - \lambda & -\frac{1}{3} \\ -\frac{1}{3} & -\frac{1}{3} & \frac{2}{3} - \lambda \end{vmatrix} = 0$$

$$\Leftrightarrow \lambda(\lambda-1)^2 = 0$$

$$i) \lambda = 0 \quad A \underline{x} = \lambda \underline{x}$$

$$x_1 = x_2 = x_3 \quad \underline{x}^T = (1 \ 1 \ 1)$$

$$ii) \lambda = 1 \quad A \underline{x} = \lambda \underline{x}$$

$$x_1 + x_2 + x_3 = 0 \quad \begin{cases} \underline{x}^T = (1 \ -1 \ 0) \\ \underline{x}^T = (1 \ 0 \ -1) \end{cases} \text{ or } \underline{x}^T = (1 \ 1 \ -2)$$

Note orthonormalized eigenvectors

$$\underline{x}_1 = \begin{pmatrix} \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix} \quad \underline{x}_2 = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \\ 0 \end{pmatrix} \quad \underline{x}_3 = \begin{pmatrix} \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{6}} \\ -\frac{2}{\sqrt{6}} \end{pmatrix}$$

HW #2 Solution

1)

$$1. \quad l = x_1 - x_2 = (1 \ -1) \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad \& \quad m = -x_1 + 3x_2 = (-1 \ 3) \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$El = (1 \ -1) \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix} = \mu_1 - \mu_2$$

$$Vl = (1 \ -1) \begin{pmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{pmatrix} \begin{pmatrix} 1 \\ -1 \end{pmatrix} = \sigma_{11} + \sigma_{22} - (\sigma_{12} + \sigma_{21}) \\ = \sigma_{11} + \sigma_{22} - 2\sigma_{12}$$

$$\text{Cov}(l, m) = (1 \ -1) \begin{pmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{pmatrix} \begin{pmatrix} -1 \\ 3 \end{pmatrix} = -\sigma_{11} - 3\sigma_{22} + 4\sigma_{12}$$

$$\text{Corr}(l, m) = \frac{\text{Cov}(l, m)}{\sqrt{Vl} \sqrt{Vm}} = \frac{-\sigma_{11} - 3\sigma_{22} + 4\sigma_{12}}{\sqrt{\sigma_{11} + \sigma_{22} - 2\sigma_{12}} \sqrt{\sigma_{11} + 9\sigma_{22} - 6\sigma_{12}}}$$

2. Do similar stuff as #1.

$$3. \quad (i) \quad \bar{x} = \begin{pmatrix} 2 \\ 3 \\ 1 \end{pmatrix}$$

$$(ii) \quad X_c = \begin{pmatrix} -3 & 0 & -3 \\ 0 & 1 & 1 \\ 3 & -1 & 2 \end{pmatrix}$$

$$(iii) \quad S = \frac{1}{2} X_c^T X_c, \text{ where } X_c \text{ is set (ii)}$$

$$R = \hat{P} = \text{Diag} \left(\frac{1}{\sqrt{s_{ii}}} \right) S \text{Diag} \left(\frac{1}{\sqrt{s_{ii}}} \right),$$

where s_{ii} is the i th main diagonal element of S .

$$4. (i) \underline{\hat{l}} = X \underline{b} = \begin{pmatrix} 1 & 4 & 3 \\ 6 & 2 & 6 \\ 8 & 3 & 3 \\ 1 & 3 & 4 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 8 \\ 14 \\ 14 \\ 8 \end{pmatrix}$$

$$\bar{l} = (8 + 14 + 14 + 8) / 4 = 11$$

$$S_e^2 = \frac{1}{n-1} \sum (l_i - \bar{l})^2 = \dots$$

$$\underline{\hat{m}} = X \underline{c} = \begin{pmatrix} 1 & 4 & 3 \\ 6 & 2 & 6 \\ 8 & 3 & 3 \\ 1 & 3 & 4 \end{pmatrix} \begin{pmatrix} 1 \\ -2 \\ -3 \end{pmatrix} = \begin{pmatrix} 0 \\ -8 \\ 5 \\ -5 \end{pmatrix}$$

$$\bar{m} = -2$$

$$S_m^2 = \frac{1}{n-1} \sum (m_i - \bar{m})^2 = \dots$$

$$S_{ij} = \frac{1}{n-1} \sum (l_i - \bar{l})(m_i - \bar{m})$$

(iii)

$$\underline{\hat{l}} = X \underline{b} \quad \& \quad \underline{\hat{m}} = X \underline{c}$$

$$\bar{l} = \frac{1}{n} \underline{1}^T \underline{\hat{l}} = \frac{1}{n} \underline{1}^T X \underline{b} = \underline{\bar{x}}^T \underline{b}$$

$$S_{ij} = \widehat{\text{cov}(\underline{\hat{l}}, \underline{\hat{m}})} = \underline{b}^T S \underline{c}$$

Use $\underline{\bar{x}}$ & S which are calculated at (ii).

$$5. \underline{x} \sim N_p(\underline{\mu}, \Sigma)$$

$$(i) \underset{\substack{\uparrow \\ t\text{-dim}}}{\underline{x}_2} | \underline{x}_1 \sim N_{p-t}(\underline{\mu}_{2.1}, \Sigma_{2.1}), \text{ where}$$

$$\underline{\mu}_{2.1} = \underline{\mu}_2 + \Sigma_{21} \Sigma_{11}^{-1} (\underline{x}_1 - \underline{\mu}_1) \quad \&$$

$$\Sigma_{2.1} = \Sigma_{22} - \Sigma_{21} \Sigma_{11}^{-1} \Sigma_{12}$$

$$\underline{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_{p-1} \\ x_p \end{pmatrix} = \begin{pmatrix} \underline{x}_1 \\ \underline{x}_2 \end{pmatrix} \sim N \left(\begin{pmatrix} \underline{\mu}_1 \\ \underline{\mu}_2 \end{pmatrix}, \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix} \right)$$

$$\therefore x_p | x_1, \dots, x_{p-1} \sim N(\underline{\mu}_{2.1}, \Sigma_{2.1}),$$

Note that $\underline{\mu}_2 = \mu_p$, $\Sigma_{22} = \sigma_{pp}$ &

$$\Sigma_{12} = \underline{\Sigma}_{1p} = \begin{pmatrix} \sigma_{1p} \\ \sigma_{2p} \\ \vdots \\ \sigma_{p-1,p} \end{pmatrix}$$

where $\underline{\mu}_{2.1} = \mu_p + \underline{\Sigma}_{1p}^T \Sigma_{11}^{-1} (\underline{x}_1 - \underline{\mu}_1)$ &

$$\Sigma_{2.1} = \sigma_{pp} - \underline{\Sigma}_{1p}^T \Sigma_{11}^{-1} \underline{\Sigma}_{1p}$$

$$(ii) \quad \underline{y} = \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = B \underline{x}$$

$\sim N_2(B\underline{\mu}, B\Sigma B^T)$, where

$$B\underline{\mu} = \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix} = \begin{pmatrix} \mu_1 + \mu_2 \\ -\mu_1 + \mu_2 \end{pmatrix} \&$$

$$B\Sigma B^T = \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} = \dots$$

(iii) $\underline{y} \sim N_2(B\underline{\mu}, B\Sigma B^T)$, where

$$B\underline{\mu} = \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix} = \begin{pmatrix} \mu_1 + \mu_2 \\ -\mu_1 + \mu_2 \end{pmatrix} \&$$

$$B\Sigma B^T = \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}$$

$$= \begin{pmatrix} \sigma_{11} + \sigma_{21} & \sigma_{12} + \sigma_{22} \\ -\sigma_{11} + \sigma_{21} & -\sigma_{12} + \sigma_{22} \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}$$

$$= \begin{pmatrix} \sigma_{11} + \sigma_{22} + 2\sigma_{12} & \sigma_{22} - \sigma_{11} \\ \sigma_{22} - \sigma_{11} & \sigma_{11} + \sigma_{22} - 2\sigma_{12} \end{pmatrix}$$

$$\begin{pmatrix} y_1 \\ y_2 \end{pmatrix} \sim N_2 \left(\begin{pmatrix} \mu_1 + \mu_2 \\ \mu_2 - \mu_1 \end{pmatrix}, \begin{pmatrix} \sigma_{11} + \sigma_{22} + 2\sigma_{12} & \sigma_{22} - \sigma_{11} \\ \sigma_{22} - \sigma_{11} & \sigma_{11} + \sigma_{22} - 2\sigma_{12} \end{pmatrix} \right) \quad 4)$$

$$\Rightarrow y_2 | y_1 \sim N(\mu_{2.1}, \Sigma_{2.1}), \text{ where}$$

$$\begin{aligned} \mu_{2.1} &= (\mu_2 - \mu_1) + (\sigma_{22} - \sigma_{11}) (\sigma_{11} + \sigma_{22} + 2\sigma_{12})^{-1} (y_1 - (\mu_1 + \mu_2)) \\ &= \mu_2 - \mu_1 + \frac{\sigma_{22} - \sigma_{11}}{\sigma_{11} + \sigma_{22} + 2\sigma_{12}} (y_1 - (\mu_1 + \mu_2)) \end{aligned}$$

$$\Sigma_{2.1} = \sigma_{11} + \sigma_{22} - 2\sigma_{12} - \frac{(\sigma_{22} - \sigma_{11})^2}{\sigma_{11} + \sigma_{22} + 2\sigma_{12}} \quad \neq$$