

Stavanger, November 19, 2021

Solutions to theoretical exercise 1

ELE520 Machine learning

Problem 1

a)

$$\begin{aligned}
 \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} p(\mathbf{x}) d\mathbf{x} &= 1 \\
 \int_{a_2}^{b_2} \int_{a_1}^{b_1} c dx_1 dx_2 &= 1 \\
 c \int_{a_2}^{b_2} [x_1]_{a_1}^{b_1} dx_2 &= 1 \\
 c \int_{a_2}^{b_2} (a_1 - b_1) dx_2 &= 1 \\
 c(b_1 - a_1)[x_2]_{a_2}^{b_2} &= 1 \\
 c(b_1 - a_1)(b_2 - a_2) &= 1
 \end{aligned} \tag{1}$$

i.e.

$$c = \frac{1}{(b_1 - a_1)(b_2 - a_2)}.$$

b)

$$\begin{aligned}
 \boldsymbol{\mu} &= \mathbb{E}[\mathbf{x}] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \mathbf{x} p(\mathbf{x}) d\mathbf{x} \\
 &= \int_{a_2}^{b_2} \int_{a_1}^{b_1} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \frac{1}{(b_1 - a_1)(b_2 - a_2)} dx_1 dx_2 \\
 &= \begin{bmatrix} \frac{1}{(b_1 - a_1)(b_2 - a_2)} \left[\frac{x_1^2}{2} \right]_{a_1}^{b_1} [x_2]_{a_2}^{b_2} \\ \frac{1}{(b_1 - a_1)(b_2 - a_2)} \left[\frac{x_2^2}{2} \right]_{a_2}^{b_2} [x_1]_{a_1}^{b_1} \end{bmatrix} \\
 &= \begin{bmatrix} \frac{a_1 + b_1}{2} \\ \frac{a_2 + b_2}{2} \end{bmatrix}.
 \end{aligned}$$

This is the center of the rectangle where $p(\mathbf{x}) \neq 0$ as illustrated in figure 1.

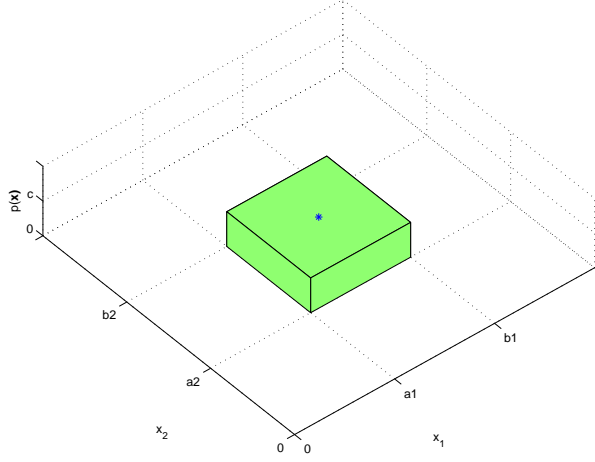


Figure 1: Uniform 2-dimensional probability density function with center of gravity shown raised to level c .

c)

$$\begin{aligned}\mathbf{\Sigma} &= \mathbb{E}[(\mathbf{x} - \boldsymbol{\mu})(\mathbf{x} - \boldsymbol{\mu})^T] = E \begin{bmatrix} (x_1 - \mu_1)^2 & (x_1 - \mu_1)(x_2 - \mu_2) \\ (x_1 - \mu_1)(x_2 - \mu_2) & (x_2 - \mu_2)^2 \end{bmatrix} \\ &= \begin{bmatrix} \mathbb{E}[x_1^2] & \mathbb{E}[x_1 x_2] \\ \mathbb{E}[x_1 x_2] & \mathbb{E}[x_2^2] \end{bmatrix} - \begin{bmatrix} \mu_1^2 & \mu_1 \mu_2 \\ \mu_1 \mu_2 & \mu_2^2 \end{bmatrix},\end{aligned}\quad (2)$$

where

$$\begin{aligned}\mathbb{E}[x_1^2] &= \int_{a_2}^{b_2} \int_{a_1}^{b_1} x_1^2 \frac{1}{(b_1 - a_1)(b_2 - a_2)} dx_1 dx_2 \\ &= \frac{1}{(b_1 - a_1)(b_2 - a_2)} \frac{1}{3} [x_1^3]_{a_1}^{b_1} [x_2]_{a_2}^{b_2} \\ &= \frac{(b_2 - a_2)(b_1^3 - a_1^3)}{3(b_1 - a_1)(b_2 - a_2)} \\ &= \frac{(b_1 - a_1)(b_1^2 + a_1^2 + a_1 b_1)}{3(b_1 - a_1)} \\ &= \frac{b_1^2 + a_1^2 + a_1 b_1}{3}, \\ \mathbb{E}[x_2^2] &= \frac{b_2^2 + a_2^2 + a_2 b_2}{3}\end{aligned}\quad (3)$$

and

$$\begin{aligned}\mathbb{E}[x_1 x_2] &= \frac{1}{(b_1 - a_1)(b_2 - a_2)} \int_{a_2}^{b_2} x_2 dx_2 \int_{a_1}^{b_1} x_1 dx_1 \\ &= \frac{1}{(b_1 - a_1)(b_2 - a_2)(2)(2)} [x_2^2]_{a_2}^{b_2} [x_1^2]_{a_1}^{b_1} \\ &= \frac{(b_1^2 - a_1^2)(b_2^2 - a_2^2)}{4(b_1 - a_1)(b_2 - a_2)} \\ &= \frac{(a_1 + b_1)(a_2 + b_2)}{4}.\end{aligned}\quad (4)$$

Substituting this into (2) we get

$$\Sigma = \begin{bmatrix} \frac{(a_1-b_1)^2}{12} & 0 \\ 0 & \frac{(a_2-b_2)^2}{12} \end{bmatrix} \quad (5)$$

- d) All variables in \mathbf{x} have the same variance.
- e) The variables in \mathbf{x} are uncorrelated.

Problem 2

- a) We first find the eigenvalues by solving the characteristic equation given as

$$\begin{aligned} |(\mathbf{M} - \lambda \mathbf{I})| &= 0 \\ \begin{vmatrix} 5 - \lambda & 3 \\ 3 & 5 - \lambda \end{vmatrix} &= 0 \\ (5 - \lambda)^2 - 9 &= 0 \\ &\Downarrow \\ \lambda_1 = 8 &\quad \lambda_2 = 2. \end{aligned} \quad (1)$$

Then we find the corresponding eigenvectors by substituting the eigenvalues we found into $\mathbf{M}\mathbf{x} = \lambda\mathbf{x}$ and solve for \mathbf{x} .

Substitute $\lambda_1 = 8$:

$$\begin{aligned} \mathbf{M}\mathbf{x} &= \lambda_1\mathbf{x} \\ (\mathbf{M} - \lambda_1\mathbf{I})\mathbf{x} &= \mathbf{0} \\ \begin{pmatrix} 5 - \lambda_1 & 3 \\ 3 & 5 - \lambda_1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} &= \mathbf{0} \\ \begin{pmatrix} -3 & 3 \\ 3 & -3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} &= \mathbf{0} \\ \begin{aligned} -3x_1 &\quad 3x_2 &= 0 \\ 3x_1 &\quad -3x_2 &= 0 \end{aligned} & \\ &\Downarrow \\ x_2 = x_1 &\quad x_1 \text{ arbitrary} \\ &\Downarrow \\ x_1 = t &\quad x_2 = t \end{aligned} \quad (2)$$

Correspondingly we solve for $\lambda_2 = 2$:

$$\begin{aligned}
\mathbf{M}\mathbf{x} &= \lambda_2\mathbf{x} \\
(\mathbf{M} - \lambda_2\mathbf{I})\mathbf{x} &= \mathbf{0} \\
\begin{pmatrix} 5 - \lambda_2 & 3 \\ 3 & 5 - \lambda_2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} &= \mathbf{0} \\
\begin{pmatrix} 3 & 3 \\ 3 & 3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} &= \mathbf{0} \\
3x_1 &= 3x_2 \\
3x_1 - 3x_2 &= 0 \\
&\Downarrow \\
x_2 &= x_1 \quad x_1 \text{ arbitrary} \\
&\Downarrow \\
x_1 &= t \quad x_2 = t
\end{aligned} \tag{3}$$

Thus we get the eigenvectors $\mathbf{e}_1 = (t \ t)^t$ and $\mathbf{e}_2 = (t \ -t)^t$ for λ_1 and λ_2 respectively.

We can substitute arbitrary variables, e.g. $\mathbf{e}_1 = 1/\sqrt{2} (1 \ 1)^t$ og $\mathbf{e}_2 = 1/\sqrt{2} (-1 \ 1)^t$ (the scaling ensures unity length).

Thus we get the eigenvector- and eigenvalue-matrices $\mathbf{\Phi}$ and $\mathbf{\Lambda}$:

$$\begin{aligned}
\mathbf{\Phi} &= \begin{pmatrix} 1/\sqrt{2} & -1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{2} \end{pmatrix} \\
\mathbf{\Lambda} &= \begin{pmatrix} 8 & 0 \\ 0 & 2 \end{pmatrix}
\end{aligned} \tag{4}$$

The principal axes are drawn as a vector between $\boldsymbol{\mu}$ and $\mathbf{p}_i = \boldsymbol{\mu} + \sqrt{\lambda_i}\mathbf{e}_i, i = 1, 2$ which are found to be $(3 \ 3)^t$ and $(0 \ 2)^t$ respectively.

- b) The contour line of the probability density function are shown in figure 2 along with the principal axes.

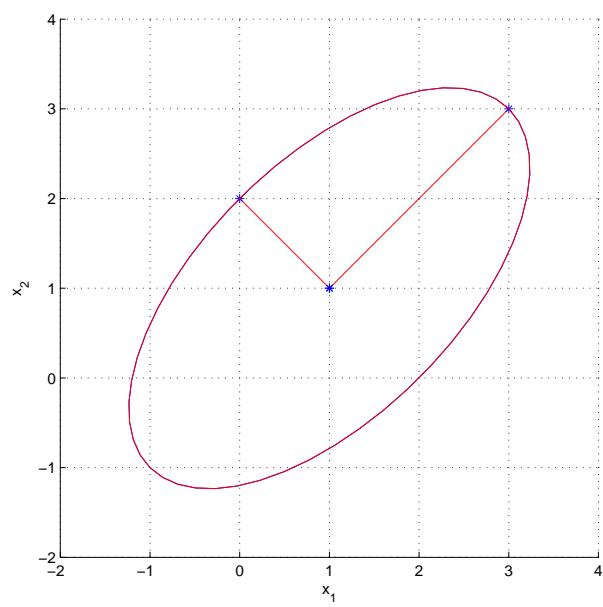


Figure 2: Contour line for 2D distribution.