

# Lektion 8

TAMS24 – Statistisk teori

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## Lektionsgenomgång

n-dimensionell fördelning,  $\underline{X}_1, \underline{X}_2, \dots, \underline{X}_n$  s.v.

$$C = \left[ \text{cov}(\underline{X}_i, \underline{X}_j) \right]_{i,j=1,\dots,n} \quad \text{Kovariansmatris}$$

$$\mu = \begin{bmatrix} E(\underline{X}_1) \\ \vdots \\ E(\underline{X}_n) \end{bmatrix} \quad \text{värtevärdesvektor}$$

$$\underline{X} = \begin{bmatrix} \underline{X}_1 \\ \vdots \\ \underline{X}_n \end{bmatrix} \quad \text{n-dimensionell normalfördelad}$$

Om

$$f_{\underline{X}}(x_1, \dots, x_n) = f_{\underline{X}}(\underline{x}) = \frac{1}{(2\pi)^{\frac{n}{2}} \sqrt{\det C}} \cdot \left\{ -\frac{1}{2} (\underline{x} - \mu)^T C^{-1} (\underline{x} - \mu) \right\},$$

$$\underline{x} \in \mathbb{R}^n.$$

## Räkneregler

$$(1) \quad \text{Cov}(\underline{X}, \underline{Y}) = E(\underline{X} \cdot \underline{Y}) - E(\underline{X}) \cdot E(\underline{Y}) \quad (\text{förkortningsformel})$$

där

$$E(\underline{X} \cdot \underline{Y}) = \begin{cases} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x \cdot y f(x,y) dx dy, & \underline{X}, \underline{Y} \text{ kontinuerliga} \\ \sum_{\text{alla } j} \sum_{\text{alla } k} j \cdot k p(j,k), & \underline{X}, \underline{Y} \text{ diskreta} \end{cases}$$

$$(2) \quad V(a\underline{X} + b\underline{Y} + c) = C(a\underline{X} + b\underline{Y}, a\underline{X} + b\underline{Y}) =$$

$$= a^2 C(\underline{X}, \underline{X}) + 2ab C(\underline{X}, \underline{Y}) + b^2 C(\underline{Y}, \underline{Y}) =$$

$$= \alpha^2 V(\underline{x}) + 2\alpha b C(\underline{x}, \underline{y}) + b^2 V(\underline{y}).$$

$$(3) \quad \begin{bmatrix} \underline{y}_1 \\ \vdots \\ \underline{y}_k \end{bmatrix} = A \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} + \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} \quad // \quad \underline{y} = A\underline{x} + b$$

där

$$\underline{x} \sim N(C_{\underline{x}}, \mu_{\underline{x}})$$

medför

$$\underline{y} \sim N(C_{\underline{y}}, \mu_{\underline{y}})$$

där

$$C_{\underline{y}} = A C_{\underline{x}} A^T, \quad \mu_{\underline{y}} = A \mu_{\underline{x}} + b$$

(4)  $\underline{x}, \underline{y}$  gemensamt normalfördelade.

Det gäller att

$$\underline{x}, \underline{y} \text{ oberoende} \Leftrightarrow C(\underline{x}, \underline{y}) = 0$$

OBS: " $\Leftarrow$ " är fel om  $\underline{x}, \underline{y}$  inte gemensamt normalfördelad.

G.1.1.

$\Sigma_1, \Sigma_2$  oberoende  $N(0,1)$  medfor

$$\underline{\Sigma} = \begin{bmatrix} \Sigma_1 \\ \Sigma_2 \end{bmatrix} \sim N \left( \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right)$$

$\mu_{\Sigma}$        $C_{\Sigma}$

$$\underline{Y} = A\underline{\Sigma} + b$$

$$\underline{\Sigma} = \begin{bmatrix} \Sigma_1 \\ \Sigma_2 \end{bmatrix}, \quad A = \begin{bmatrix} 1 & 1 \\ 2 & -1 \end{bmatrix}, \quad \underline{\Sigma} = \begin{bmatrix} \Sigma_1 \\ \Sigma_2 \end{bmatrix}, \quad b = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\mu_{\Sigma} = A\mu_{\Sigma} = \begin{bmatrix} 1 & 1 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix} = 0$$

$$C_{\Sigma} = A C_{\Sigma} A^T = \begin{bmatrix} 1 & 1 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 1 & -1 \end{bmatrix} =$$

$$= \begin{bmatrix} 1 & 1 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 1 & -1 \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 1 & 5 \end{bmatrix}$$

$$\det C_{\Sigma} = \frac{1}{9} \begin{bmatrix} 5 & -1 \\ -1 & 2 \end{bmatrix}$$

$$f_{\Sigma} = \frac{1}{\sqrt{(2\pi)^2 \cdot 5}} \cdot \exp \left\{ -\frac{1}{2} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}^T \frac{1}{9} \begin{bmatrix} 5 & -1 \\ -1 & 2 \end{bmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} \right\} =$$

$$= \frac{1}{6\pi} \exp \left\{ \frac{1}{2} (y_1 \ y_2) \cdot \frac{1}{9} \left( \begin{array}{c} 5y_1 - y_2 \\ -y_1 + 2y_2 \end{array} \right) \right\}$$

$$= \frac{1}{6\pi} \exp \left\{ -\frac{1}{18} \left( 5y_1^2 - y_1 y_2 - y_1 y_2 + 2y_2^2 \right) \right\}$$

$$= \frac{1}{6\pi} \exp \left\{ -\frac{1}{18} (5y_1^2 - 2y_1 y_2 + 2y_2^2) \right\}$$

G1.2

$$\begin{pmatrix} \bar{X}_1 \\ \bar{X}_2 \\ \bar{X}_3 \end{pmatrix} \sim N \left( \begin{pmatrix} 60 \\ 60 \\ 60 \end{pmatrix}, \begin{pmatrix} 100 & 80 & 20 \\ 80 & 100 & 10 \\ 20 & 10 & 80 \end{pmatrix} \right)$$

$$\bar{Y} = (\bar{X}_1 + \bar{X}_2 + 2\bar{X}_3)/4$$

a) Find the distribution for  $\bar{Y}$

$$\bar{X}_1 \sim N(60, 10)$$

$$\bar{X}_2 \sim N(60, 10)$$

$$\bar{X}_3 \sim N(60, 20)$$

$$\bar{Y} = (\bar{X}_1 + \bar{X}_2 + 2\bar{X}_3)/4 = \frac{1}{4} (1 \ 1 \ 2) \begin{pmatrix} \bar{X}_1 \\ \bar{X}_2 \\ \bar{X}_3 \end{pmatrix} =$$

$$= \left( \frac{1}{4} \ \frac{1}{4} \ \frac{1}{2} \right) \bar{X} \sim N.$$

$$\mu_{\bar{Y}} = \left( \frac{1}{4} \ \frac{1}{4} \ \frac{1}{2} \right) \mu_{\bar{X}} = \left( \frac{1}{4} \ \frac{1}{4} \ \frac{1}{2} \right) \begin{pmatrix} 60 \\ 60 \\ 60 \end{pmatrix} = 60$$

$$\sigma^2_{\bar{Y}} = V(\bar{Y}) = C \left( \frac{1}{4} \bar{X}_1 + \frac{1}{4} \bar{X}_2 + \frac{1}{2} \bar{X}_3, \ \frac{1}{4} \bar{X}_1 + \frac{1}{4} \bar{X}_2 + \frac{1}{2} \bar{X}_3 \right) =$$

$$= \frac{1}{16} C(\bar{x}_1, \bar{x}_1) + \frac{1}{16} C(\bar{x}_1, \bar{x}_2) + \frac{1}{8} C(\bar{x}_1, \bar{x}_3) + \dots + \dots$$

= 50

$$\mathbb{I} \sim N(60, \sqrt{50})$$

$$0.9 = P(\mathbb{I} > a) = P(N(0,1) > \frac{a-60}{\sqrt{50}}) = 1 - \Phi\left(\frac{a-60}{\sqrt{50}}\right) \Rightarrow$$

$$- \Phi\left(\frac{a-60}{\sqrt{50}}\right) = 0.9 \Rightarrow - \frac{a-60}{\sqrt{50}} = 1.28$$

$$\Leftrightarrow a = -\sqrt{50} \cdot 1.28 + 60$$

b)  $\bar{x}_1, \bar{x}_2, \bar{x}_3$  oberoende

$$\bar{x} \sim N \left( \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \right) = N(\mu_x, C_x)$$

$$U = \bar{x}_1 - 2\bar{x}_2 + \bar{x}_3$$

$$V = c_1 \bar{x}_1 + c_2 \bar{x}_2 + c_3 \bar{x}_3$$

$$\begin{bmatrix} U \\ V \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & -2 & 1 \\ c_1 & c_2 & c_3 \end{bmatrix}}_A \begin{bmatrix} \bar{x}_1 \\ \bar{x}_2 \\ \bar{x}_3 \end{bmatrix}$$

Sökt är villkoret sådant att

$$\text{Cov}(U, V) = 0$$

$$\mu_0 = A\mu_{\infty} + b = A \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$C_0 = AC_{\infty}A^T = \underbrace{\begin{bmatrix} 1 & -2 & 1 \\ C_1 & C_2 & C_3 \end{bmatrix}}_{= \begin{bmatrix} 6 & C_1 - 2C_2 + C_3 \\ C_1 - 2C_2 + C_3 & C_1^2 + C_2^2 + C_3^2 \end{bmatrix}} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & C_1 \\ -2 & C_2 \\ 1 & C_3 \end{bmatrix} =$$

Det sätta villkoret

$$C_1 - 2C_2 + C_3 = 0.$$

G.1.6.

Random variables

$\bar{X}_1, \bar{X}_2, \dots, \bar{X}_5$ , independent

$$\bar{X} \sim N(10, 4)$$

$$\bar{Y}_1 = \frac{1}{5} (\bar{X}_1 + \bar{X}_2 + \bar{X}_3 + \bar{X}_4 + \bar{X}_5)$$

$$\bar{Y}_2 = 4\bar{X}_1 + \bar{X}_2 - \bar{X}_3 - \bar{X}_4 - \bar{X}_5$$

a)

$$\begin{bmatrix} \bar{Y}_1 \\ \bar{Y}_2 \end{bmatrix} = \underbrace{\begin{bmatrix} 1/5 & 1/5 & 1/5 & 1/5 & 1/5 \\ 4 & 1 & -1 & -1 & -1 \end{bmatrix}}_{A} \begin{bmatrix} \bar{X}_1 \\ \vdots \\ \bar{X}_5 \end{bmatrix} \quad b = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\mu_{\bar{X}} = \begin{bmatrix} 10 \\ \vdots \\ 10 \end{bmatrix}, \quad C_{\bar{X}} = \begin{bmatrix} 4 & & & & \\ & \ddots & & & \\ & & \ddots & & 4 \end{bmatrix}$$

$$\mu_{\bar{Y}} = A\mu_{\bar{x}} + b = \begin{bmatrix} 1/5 & 1/5 & 1/5 & 1/5 & 1/5 \\ 4 & 1 & -1 & -1 & -1 \end{bmatrix} \begin{bmatrix} 10 \\ \vdots \\ 10 \end{bmatrix} = \begin{bmatrix} 10 \\ \underline{\underline{20}} \end{bmatrix}$$

$$C_{\bar{Y}} = A C_{\bar{x}} A^T = \dots = \begin{bmatrix} 0.8 & 1.6 \\ 1.6 & 80 \end{bmatrix}$$

$$\bar{Y} \sim N \left( \begin{bmatrix} 10 \\ 20 \end{bmatrix}, \begin{bmatrix} 0.8 & 1.6 \\ 1.6 & 80 \end{bmatrix} \right)$$

b)  $P(\bar{Y}_1 > \bar{Y}_2) = P(\bar{Y}_1 - \bar{Y}_2 > 0)$

Vi behöver fördelningen  $\bar{Y}_1 - \bar{Y}_2$

Som linjär kombination av de gemensamt normalfördelade

$\bar{Y}_1, \bar{Y}_2$  är  $\bar{Y}_1 - \bar{Y}_2$  också normalfördelad

$$N(\mu_{\bar{Y}_1 - \bar{Y}_2}, \sigma_{\bar{Y}_1 - \bar{Y}_2})$$

$$\mu_{\bar{Y}_1 - \bar{Y}_2} = E(\bar{Y}_1 - \bar{Y}_2) = E(\bar{Y}_1) - E(\bar{Y}_2) = -10$$

$$\sigma_{\bar{Y}_1 - \bar{Y}_2}^2 = V(\bar{Y}_1 - \bar{Y}_2) = V(\bar{Y}_1) - 2C(\bar{Y}_1, \bar{Y}_2) + V(\bar{Y}_2) =$$

$$= 0.8 - 2 \cdot 1.6 + 80 = 77.6$$

$$P(\bar{Y}_1 > \bar{Y}_2) = P(\bar{Y}_1 - \bar{Y}_2 > 0) = P\left(\frac{\bar{Y}_1 - \bar{Y}_2 - (-10)}{\sqrt{77.6}} > \frac{-(-10)}{\sqrt{77.6}}\right) =$$

$$= P(N(0,1) > \frac{10}{\sqrt{77.6}}) = 1 - \Phi\left(\frac{10}{\sqrt{77.6}}\right) = 1 - 0.8686 =$$

$$= \underline{\underline{0.1314}}$$

$$c) \quad \rho = \frac{C(\underline{\Sigma}_1, \underline{\Sigma}_2)}{\sqrt{V(\underline{\Sigma}_1) \cdot V(\underline{\Sigma}_2)}} = \frac{1.6}{\sqrt{6.8 \cdot 8.0}} = \frac{1.6}{8} = 0.2$$

G.1.7II

Three dimension normal distribution with mean vector

$$\mu = \begin{pmatrix} 71 \\ 53 \\ 18 \end{pmatrix} \quad C = \begin{pmatrix} 100 & 64 & 38 \\ 64 & 64 & 28.8 \\ 38 & 28.8 & 16 \end{pmatrix}$$

$$\begin{pmatrix} \underline{\Sigma}_1 \\ \underline{\Sigma}_2 \\ \underline{\Sigma}_3 \end{pmatrix} \sim N \left( \begin{bmatrix} 71 \\ 53 \\ 18 \end{bmatrix}, \begin{bmatrix} 100 & 64 & 38 \\ 64 & 64 & 28.8 \\ 38 & 28.8 & 16 \end{bmatrix} \right)$$

$$\hat{\underline{\Sigma}}_3 = a + b\underline{\Sigma}_1 + c\underline{\Sigma}_2$$

$$(i) \quad E(\hat{\underline{\Sigma}}_3) = E(\underline{\Sigma}_1)$$

(ii)  $V(\underline{\Sigma}_3 - \hat{\underline{\Sigma}}_3)$  is minimized.