

BR

$$\begin{aligned}
 1 \quad \text{Cov}[\underline{X}, \underline{Y}] &= E[(\underline{X} - E\underline{X})(\underline{Y} - E\underline{Y})^T] \\
 &= E[\underline{X}\underline{Y}^T - \underline{X}E\underline{Y}^T - (E\underline{X})\underline{Y}^T + E\underline{X}E\underline{Y}^T] \\
 &= E[\underline{X}\underline{Y}^T] - E\underline{X}E\underline{Y}^T - E\underline{X}E\underline{Y}^T + E\underline{X}E\underline{Y}^T \\
 &= E[\underline{X}\underline{Y}^T] - E\underline{X}E\underline{Y}^T
 \end{aligned}$$

$$\begin{aligned}
 2 \quad \text{Cov}[A\underline{X}, B\underline{Y}] &= E[(A\underline{X} - E[A\underline{X}])(B\underline{Y} - E[B\underline{Y}])^T] \\
 &= E[A(\underline{X} - E\underline{X})(B(\underline{Y} - E\underline{Y}))^T] \\
 &= A E[(\underline{X} - E\underline{X})(\underline{Y} - E\underline{Y})^T] B^T \\
 &= A \text{Cov}[\underline{X}, \underline{Y}] B^T
 \end{aligned}$$

$$\begin{aligned}
 3 \quad \text{Var}[\underline{Y} + \underline{C}] &= \text{Cov}[\underline{Y} + \underline{C}, \underline{Y} + \underline{C}] \\
 &= E[(\underline{Y} + \underline{C} - E[\underline{Y} + \underline{C}])(\underline{Y} + \underline{C} - E[\underline{Y} + \underline{C}])^T] \\
 &= E[(\underline{Y} - E\underline{Y})(\underline{Y} - E\underline{Y})^T] \\
 &= \text{Cov}[\underline{Y}, \underline{Y}] \\
 &= \text{Var} \underline{Y}
 \end{aligned}$$

$$\begin{aligned}
 4 \quad M_{A\underline{Y} + \underline{C}}(\underline{t}) &= E[\exp(\underline{t}^T (A\underline{Y} + \underline{C}))] \\
 &= E[\exp(\underline{t}^T A\underline{Y} + \underline{t}^T \underline{C})] \\
 &= \exp(\underline{t}^T \underline{C}) E[\exp((A^T \underline{t})^T \underline{Y})] \\
 &= \exp(\underline{t}^T \underline{C}) M_{\underline{Y}}(A^T \underline{t})
 \end{aligned}$$

5 $\underline{X}, \underline{Y}$ indep.

$$\begin{aligned}
 M_{\underline{X} + \underline{Y}}(\underline{t}) &= E[\exp(\underline{t}^T (\underline{X} + \underline{Y}))] \\
 &= E[\exp(\underline{t}^T \underline{X} + \underline{t}^T \underline{Y})] \\
 &= E[\exp(\underline{t}^T \underline{X})] E[\exp(\underline{t}^T \underline{Y})] \\
 &= M_{\underline{X}}(\underline{t}) M_{\underline{Y}}(\underline{t})
 \end{aligned}$$

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C orthogonal, dvs. $C^T C = I$

i $C^T C = I \Rightarrow C C^T = I$, if. komp. i lin. alg. s. 15, dvs. systemet af rækkerektorer i C er også ortonormeret

ii $C^T C = C C^T = I \Rightarrow C^{-1} = C^T$, if. def. af C^{-1}

iii $C^T C = I \Rightarrow \det(C^T C) = \det I$

$$\Rightarrow \det C^T \det C = 1$$

$$\Rightarrow (\det C)^2 = 1$$

$$\Rightarrow \det C = \pm 1$$

iv A og B ortogonale

$$(AB)^T (AB) = B^T A^T A B = B^T I B = B^T B = I$$

$\Rightarrow AB$ orthogonal

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A $n \times n$ symmetrisk *

C orthogonal, $C^T A C = \Lambda = \begin{bmatrix} \lambda_1 & & \\ & \dots & \\ & & \lambda_n \end{bmatrix}$ ($C^T = C^{-1}$)

Betragt $\underline{x}^T A \underline{x}$ for alle $\underline{x} \in \mathbb{R}^n$

Sæt $\underline{y} = C^{-1} \underline{x} \Leftrightarrow \underline{x} = C \underline{y}$, bemærk $\underline{x} \neq \underline{0} \Leftrightarrow \underline{y} \neq \underline{0}$

$$\underline{x}^T A \underline{x} = (C \underline{y})^T A (C \underline{y}) = \underline{y}^T C^T A C \underline{y} = \underline{y}^T \Lambda \underline{y} = \sum_i \lambda_i y_i^2$$

i $\underline{x} = C \underline{e}_j \Rightarrow \underline{y} = \underline{e}_j \Rightarrow \sum_i \lambda_i y_i^2 = \lambda_j$

dvs. $\underline{x}^T A \underline{x} > 0$ for alle $\underline{x} \Rightarrow \lambda_j > 0, j = 1, \dots, n$

ii $\forall \underline{x} \neq \underline{0} : \lambda_i > 0 \Rightarrow \lambda_i y_i^2 > 0 \Rightarrow \sum_i \lambda_i y_i^2 > 0$

$$\Leftrightarrow \underline{x}^T A \underline{x} > 0$$

altså $\forall \underline{x} \neq \underline{0} : \underline{x}^T A \underline{x} > 0 \Leftrightarrow \lambda_i > 0, i = 1, \dots, n$

* Spektralsætningen siger bl. a., at enhver symmetrisk matrix kan diagonaliseres ved en orthogonal matrix.

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A positive definite, C orthogonal, $C^{-1}AC = \Lambda$

$$\Lambda = \Gamma \lambda_1 \dots \lambda_n \Gamma^{-1} = (\Gamma \lambda_1^{\frac{1}{2}} \dots \lambda_n^{\frac{1}{2}} \Gamma^{-1})^2$$

id est $\lambda_i > 0$ for alle i , \forall n .

$$\text{Set } \Lambda^{\frac{1}{2}} = \Gamma \lambda_1^{\frac{1}{2}} \dots \lambda_n^{\frac{1}{2}} \Gamma^{-1}, \text{ des. } (\Lambda^{\frac{1}{2}})^2 = \Lambda$$

$$\text{Set } A^{\frac{1}{2}} = C \Lambda^{\frac{1}{2}} C^{-1}, \text{ hence}$$

$$(A^{\frac{1}{2}})^2 = C \Lambda^{\frac{1}{2}} C^{-1} C \Lambda^{\frac{1}{2}} C^{-1} = C \Lambda C^{-1} = A$$

10

$$\int_{\mathbb{R}^n} f_Y(y_1, \dots, y_n) dV \text{ motives } \int_{\mathbb{R}^n} f_Y(y) d\Omega$$

Incl. for variabeltransformation $y = \Sigma^{\frac{1}{2}} x + \mu$ med
 Jacobideterminant $J(x) = \det \Sigma^{\frac{1}{2}}$:

$$\begin{aligned} \int_{\mathbb{R}^n} f_Y(y) dV &= \int_{\mathbb{R}^n} f_Y(\Sigma^{\frac{1}{2}} x + \mu) |\det \Sigma^{\frac{1}{2}}| d\Omega \\ &= \int_{\mathbb{R}^n} f_X(x) d\Omega, \quad \forall \text{ note s. 3} \\ &= \prod_{i=1}^n \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}} e^{-\frac{x_i^2}{2}} dx_i = \prod_{i=1}^n 1 = 1 \end{aligned}$$

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$$i \quad \underline{Y} \sim N_n(\underline{\mu}, \Sigma) \Rightarrow M_{\underline{Y}}(\underline{t}) = \exp(\underline{t}^T \underline{\mu} + \frac{1}{2} \underline{t}^T \Sigma \underline{t})$$

$$\Rightarrow \forall \underline{l} \neq \underline{0}: M_{\underline{l}^T \underline{Y}}(\underline{t}) = M_{\underline{Y}}(\underline{l} \underline{t}), \quad \forall \text{ note s. 4}$$

$$= \exp(\underline{l}^T \underline{\mu} \underline{t} + \frac{1}{2} \underline{l}^T \Sigma \underline{l} \underline{t}^2)$$

$$\Leftrightarrow \forall \underline{l} \neq \underline{0}: \underline{l}^T \underline{Y} \sim N(\underline{l}^T \underline{\mu}, \frac{1}{2} \underline{l}^T \Sigma \underline{l})$$

$$ii \quad \forall \underline{l} \neq \underline{0}: \underline{l}^T \underline{Y} \sim N(\underline{l}^T \underline{\mu}, \frac{1}{2} \underline{l}^T \Sigma \underline{l})$$

$$\Rightarrow M_{\underline{l}^T \underline{Y}}(\underline{t}) = \exp(\underline{l}^T \underline{\mu} \underline{t} + \frac{1}{2} \underline{l}^T \Sigma \underline{l} \underline{t}^2), \quad \text{set } \underline{l} = \underline{t} \text{ or } \underline{t} = \underline{l}$$

$$\Rightarrow M_{\underline{t}^T \underline{Y}}(\underline{t}) = \exp(\underline{t}^T \underline{\mu} \underline{t} + \frac{1}{2} \underline{t}^T \Sigma \underline{t} \underline{t}^2)$$

$$\Leftrightarrow M_{\underline{Y}}(\underline{t}) = \exp(\underline{t}^T \underline{\mu} \underline{t} + \frac{1}{2} \underline{t}^T \Sigma \underline{t} \underline{t}^2), \quad \forall \text{ note s. 4}$$

$$\Leftrightarrow \underline{Y} \sim N_n(\underline{\mu}, \Sigma)$$

9. $\det \Sigma = \det(\Sigma^{\frac{1}{2}} \Sigma^{\frac{1}{2}}) = (\det \Sigma^{\frac{1}{2}})^2 \Rightarrow \det \Sigma^{\frac{1}{2}} = (\det \Sigma)^{\frac{1}{2}}$

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12 A $n \times n$ pos. def., C $n \times m$, $\text{rang } C = r$

$$(CACT)^T = (C^T)^T A^T C^T = CACT, \text{ d.h. } CACT \text{ symm.}$$

$$\forall \underline{x} \neq \underline{0}: \underline{x}^T (CACT) \underline{x} = (C^T \underline{x})^T A C^T \underline{x} = \underline{y}^T A \underline{y} > 0,$$

idest $\underline{y} \neq \underline{0} \Leftrightarrow C^T \underline{x} \neq \underline{0} \Leftrightarrow \underline{x} \neq \underline{0}$ ($\text{rang } C^T = \text{rang } C = r$),
d.h. $CACT$ ist pos. def.

(speziell gilt CC^T pos. def., idest $CC^T = C I C^T$)

13 $\langle \underline{u}, \underline{v} \rangle = \underline{u}^T \Sigma \underline{v}$, Σ pos. def.

i $\left. \begin{array}{l} \langle \underline{u}, \underline{u} \rangle \geq 0 \text{ für alle } \underline{u} \\ \langle \underline{u}, \underline{u} \rangle = 0 \Leftrightarrow \underline{u} = \underline{0} \end{array} \right\} \text{Hilf. def. of pos. def. matrix}$

$$\begin{aligned} \text{ii} \quad \langle \underline{u}, \underline{v} \rangle &= \underline{u}^T \Sigma \underline{v} = (\underline{u}^T \Sigma \underline{v})^T = \underline{v}^T \Sigma \underline{u} \\ &= \langle \underline{v}, \underline{u} \rangle \end{aligned}$$

$$\begin{aligned} \text{iii} \quad \langle \underline{u}, \underline{v} + \underline{w} \rangle &= \underline{u}^T \Sigma (\underline{v} + \underline{w}) = \underline{u}^T \Sigma \underline{v} + \underline{u}^T \Sigma \underline{w} \\ &= \langle \underline{u}, \underline{v} \rangle + \langle \underline{u}, \underline{w} \rangle \end{aligned}$$

$$\begin{aligned} \text{iv} \quad r \underline{u}^T \Sigma \underline{v} &= (r \underline{u})^T \Sigma \underline{v} = \underline{u}^T \Sigma (r \underline{v}) \\ \Rightarrow r \langle \underline{u}, \underline{v} \rangle &= \langle r \underline{u}, \underline{v} \rangle = \langle \underline{u}, r \underline{v} \rangle \end{aligned}$$

$\Rightarrow \langle \underline{u}, \underline{v} \rangle$ definiert ein inneres Produkt

14 Betrachte $(\underline{Y} - \underline{\mu})^T \Sigma^{-1} (\underline{Y} - \underline{\mu})$.

$$\text{Set } \underline{Y} = \Sigma^{\frac{1}{2}} \underline{X} + \underline{\mu}:$$

$$\begin{aligned} (\underline{Y} - \underline{\mu})^T \Sigma^{-1} (\underline{Y} - \underline{\mu}) &= (\Sigma^{\frac{1}{2}} \underline{X})^T \Sigma^{-1} \Sigma^{\frac{1}{2}} \underline{X} \\ &= \underline{X}^T \Sigma^{\frac{1}{2}} \Sigma^{-1} \Sigma^{\frac{1}{2}} \underline{X}, \text{ da } \Sigma^{\frac{1}{2}} \text{ symm.} \\ &= \underline{X}^T \underline{X} = \sum_{i=1}^n X_i^2 \sim \chi^2(n), \end{aligned}$$

da $X_i \sim N(0,1)$, $i=1, \dots, n$, unabh.,

Hilf. note s. 3

15 $Y_i \sim N(\mu, \sigma^2)$, $i = 1, \dots, n$, uafh.

$$X_i = \frac{Y_i - \mu}{\sigma} \sim N(0, 1), \quad i = 1, \dots, n, \quad \text{uafh.}$$

$$\Rightarrow \underline{X} \sim N_n(\underline{0}, \mathbf{I})$$

$$\underline{U} = \mathbf{C} \underline{X}, \quad \mathbf{C} \text{ er orthogonal med første} \\ \text{rækkevektor } \left[\frac{1}{\sqrt{n}} \dots \frac{1}{\sqrt{n}} \right]$$

$$\underline{U} \sim N_n(\underline{0}, \mathbf{C} \mathbf{I} \mathbf{C}^T) = N_n(\underline{0}, \mathbf{I})$$

$$\Rightarrow U_i \sim N(0, 1), \quad i = 1, \dots, n, \quad \text{uafh.}$$

$$U_1 = \left[\frac{1}{\sqrt{n}} \dots \frac{1}{\sqrt{n}} \right] \underline{X} = \frac{1}{\sqrt{n}} \sum_i X_i = \sqrt{n} \bar{X} \Rightarrow \bar{X} = \frac{U_1}{\sqrt{n}}$$

$$\begin{aligned} \frac{(n-1)S^2}{\sigma^2} &= \sum_i \left(\frac{Y_i - \bar{Y}}{\sigma} \right)^2 = \sum_i \left(\frac{Y_i - \mu}{\sigma} - \frac{\bar{Y} - \mu}{\sigma} \right)^2 \\ &= \sum_i (X_i - \bar{X})^2 = \sum_i X_i^2 - n\bar{X}^2 = \sum_i U_i^2 - n \left(\frac{U_1}{\sqrt{n}} \right)^2 \\ &= U_2^2 + \dots + U_n^2 \sim \chi^2(n-1) \quad * \end{aligned}$$

16 $\text{Var}(\underline{Y}, \underline{Z}) = \begin{bmatrix} \mathbf{I}_{n_1} & \mathbf{0} \\ -\Sigma_{21} \Sigma_{11}^{-1} & \mathbf{I}_{n_2} \end{bmatrix} \begin{bmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{bmatrix} \begin{bmatrix} \mathbf{I}_{n_1} & -\Sigma_{11}^{-1} \Sigma_{12} \\ \mathbf{0} & \mathbf{I}_{n_2} \end{bmatrix}$

$$= \begin{bmatrix} \Sigma_{11} & \Sigma_{12} \\ \mathbf{0} & \Sigma_{22} - \Sigma_{21} \Sigma_{11}^{-1} \Sigma_{12} \end{bmatrix} \begin{bmatrix} \mathbf{I}_{n_1} & -\Sigma_{11}^{-1} \Sigma_{12} \\ \mathbf{0} & \mathbf{I}_{n_2} \end{bmatrix}$$

$$= \begin{bmatrix} \Sigma_{11} & \mathbf{0} \\ \mathbf{0} & \Sigma_{22} - \Sigma_{21} \Sigma_{11}^{-1} \Sigma_{12} \end{bmatrix}$$

$$= \begin{bmatrix} \Sigma_{11} & \mathbf{0} \\ \mathbf{0} & \Sigma_{22-1} \end{bmatrix}$$

$$\Rightarrow \underline{Y}_1 \text{ og } \underline{Z} \text{ er uafh.}$$

* Bemærk, at $\bar{Y} = \frac{\sigma U_1}{\sqrt{n}} + \mu$ og $S^2 = \frac{\sigma^2}{n-1} (U_2^2 + \dots + U_n^2)$,
dvs. det ses også her, at \bar{Y} og S^2 er uafhængige.