

1.7

 $\underline{x}_1, \dots, \underline{x}_m$ sample

$$\text{Var } \underline{x}_i = \Sigma, \quad i=1, \dots, m$$

$$\underline{\tilde{x}}_r = \underline{x}_r - \underline{\bar{x}}$$

$$\text{Cov}(\underline{\tilde{x}}_r, \underline{\tilde{x}}_s) = \text{Cov}(\underline{x}_r - \underline{\bar{x}}, \underline{x}_s - \underline{\bar{x}})$$

$$= \text{Cov}(\underline{x}_r, \underline{x}_s) - \text{Cov}(\underline{x}_r, \underline{\bar{x}}) - \text{Cov}(\underline{\bar{x}}, \underline{x}_s) + \text{Cov}(\underline{\bar{x}}, \underline{\bar{x}})$$

$$\text{Cov}(\underline{x}_r, \underline{x}_s) = \delta_{rs} \Sigma$$

$$\text{Cov}(\underline{x}_r, \underline{\bar{x}}) = \text{Cov}\left(\underline{x}_r, \frac{1}{n} \sum_i \underline{x}_i\right) = \frac{1}{n} \sum_i \text{Cov}(\underline{x}_r, \underline{x}_i)$$

$$= \frac{1}{n} (\Sigma + (n-1)0) = \frac{1}{n} \Sigma$$

$$\text{Cov}(\underline{\bar{x}}, \underline{x}_s) = \text{Cov}(\underline{\bar{x}}, \underline{x}_s) = \text{Cov}(\underline{x}_r, \underline{\bar{x}}) = \frac{1}{n} \Sigma$$

$$\text{Cov}(\underline{\bar{x}}, \underline{\bar{x}}) = \text{Cov}\left(\frac{1}{n} \sum_i \underline{x}_i, \frac{1}{n} \sum_j \underline{x}_j\right)$$

$$= \frac{1}{n^2} \sum_i \sum_j \text{Cov}(\underline{x}_i, \underline{x}_j)$$

$$= \frac{1}{n^2} (n \Sigma + n(n-1)0) = \frac{1}{n} \Sigma$$

$$\text{Cov}(\underline{\tilde{x}}_r, \underline{\tilde{x}}_s) = \delta_{rs} \Sigma - \frac{1}{n} \Sigma - \frac{1}{n} \Sigma + \frac{1}{n} \Sigma$$

$$= \left(\delta_{rs} - \frac{1}{n}\right) \Sigma$$

1.4

$\underline{y} \sim N_d(\underline{0}, \Sigma)$, A $d \times d$ sym., $\text{rank } A = r$

$\Sigma > 0 \Leftrightarrow \exists R: \Sigma = R^T R$, $\text{rank } R = d$, cf. A5.3

Let $\underline{y} = R^T \underline{z} \Leftrightarrow \underline{z} = (R^T)^{-1} \underline{y}$

$\underline{z} \sim N_d((R^T)^{-1} \underline{0}, (R^T)^{-1} (R^T R) ((R^T)^{-1})^T)$
 $= N_d(\underline{0}, I_d)$

$\underline{y}^T A \underline{y} = (R^T \underline{z})^T A R^T \underline{z} = \underline{z}^T R A R^T \underline{z}$

Note that $R A R^T$ is symmetric

and $\text{rank}(R A R^T) = \text{rank } A = r$

a $\underline{y}^T A \underline{y} \sim \chi^2(r) \Leftrightarrow \underline{z}^T R A R^T \underline{z} \sim \chi^2(r)$

$\Leftrightarrow R A R^T$ idempotent, cf. A6.5

$\Leftrightarrow R A R^T R A R^T = R A R^T$

$\Leftrightarrow R^{-1} R A (R^T R) A R^T (R^T)^{-1} = R^{-1} R A R^T (R^T)^{-1}$

$\Leftrightarrow A \Sigma A = A$

b $\underline{y}^T A \underline{y} \sim \sigma^2 \chi^2(r) \Leftrightarrow \frac{1}{\sigma^2} \underline{y}^T A \underline{y} \sim \chi^2(r)$

$\Leftrightarrow \underline{y}^T \left(\frac{1}{\sigma^2} A\right) \underline{y} \sim \chi^2(r)$

$\Leftrightarrow \left(\frac{1}{\sigma^2} A\right) \Sigma \left(\frac{1}{\sigma^2} A\right) = \frac{1}{\sigma^2} A$, cf. a

$\Leftrightarrow A \Sigma A = \sigma^2 A$

2.1

$$\underline{y} \sim N_d(\underline{\theta}, \Sigma)$$

$$\text{Let } \underline{y} = \Sigma^{\frac{1}{2}} \underline{z} \Leftrightarrow \underline{z} = \Sigma^{-\frac{1}{2}} \underline{y}$$

$$\underline{z} \sim N_d(\Sigma^{-\frac{1}{2}} \underline{\theta}, \Sigma^{-\frac{1}{2}} \Sigma \Sigma^{-\frac{1}{2}}) = N_d(\Sigma^{-\frac{1}{2}} \underline{\theta}, I_d)$$

$$\underline{y}^T \Sigma^{-1} \underline{y} = \underline{z}^T \Sigma^{\frac{1}{2}} \Sigma^{-1} \Sigma^{\frac{1}{2}} \underline{z} = \underline{z}^T \underline{z}$$

$$\begin{aligned} \underline{y}^T \Sigma^{-1} \underline{y} &\sim \chi^2(d; (\Sigma^{-\frac{1}{2}} \underline{\theta})^T (\Sigma^{-\frac{1}{2}} \underline{\theta})) \\ &= \chi^2(d; \underline{\theta}^T \Sigma^{-1} \underline{\theta}) \end{aligned}$$

2.2

$$\underline{a} \Rightarrow \underline{b} \quad M_{\underline{y}}(\underline{t}) = \exp(\underline{t}^T \underline{\theta} + \frac{1}{2} \underline{t}^T \Sigma \underline{t}) \quad \text{given}$$

$$\begin{aligned} \forall \underline{l}: M_{\underline{l}^T \underline{y}}(\underline{t}) &= M_{\underline{y}}(\underline{l} \underline{t}) \quad (\text{lemma 1}) \\ &= \exp((\underline{l} \underline{t})^T \underline{\theta} + \frac{1}{2} (\underline{l} \underline{t})^T \Sigma (\underline{l} \underline{t})) \\ &= \exp(\underline{l}^T \underline{\theta} \underline{t} + \frac{1}{2} \underline{l}^T \Sigma \underline{l} \underline{t}^2) \end{aligned}$$

$$\underline{b} \Rightarrow \underline{a} \quad \forall \underline{l}: M_{\underline{l}^T \underline{y}}(\underline{t}) = \exp(\underline{l}^T \underline{\theta} \underline{t} + \frac{1}{2} \underline{l}^T \Sigma \underline{l} \underline{t}^2) \quad \text{given}$$

Choose $\underline{t} = 1$ and $\underline{l} = \underline{t}$:

$$\forall \underline{t}: M_{\underline{t}^T \underline{y}}(\underline{t}) = \exp(\underline{t}^T \underline{\theta} \cdot 1 + \frac{1}{2} \underline{t}^T \Sigma \underline{t} \cdot 1^2)$$

$$\Leftrightarrow M_{\underline{y}}(\underline{t}) = \exp(\underline{t}^T \underline{\theta} + \frac{1}{2} \underline{t}^T \Sigma \underline{t})$$

(lemma 1)

$$2.3 \quad \underline{y} \sim N_d(\underline{\theta}, \Sigma) \Leftrightarrow \forall \underline{c}: \underline{c}^T \underline{y} \sim N(\underline{c}^T \underline{\theta}, \underline{c}^T \Sigma \underline{c}), \quad \Sigma \geq 0$$

ad (i):

$$\begin{aligned} \underline{c}^T (C \underline{y}) &= (\underline{c}^T C) \underline{y} = \underline{c}_1^T \underline{y} \sim N(\underline{c}_1^T \underline{\theta}, \underline{c}_1^T \Sigma \underline{c}_1) \\ &= N(\underline{c}_1^T C \underline{\theta}, \underline{c}_1^T C \Sigma C^T \underline{c}_1) \end{aligned}$$

$$\Rightarrow C \underline{y} \sim N_q(C \underline{\theta}, C \Sigma C^T), \quad q = \text{rank } C \quad *)$$

$$C \Sigma C^T \geq 0 \text{ cf. A4.4}$$

ad (ii):

$$\forall \underline{c}: \underline{c}^T \underline{y} \sim N(\underline{c}^T \underline{\theta}, \underline{c}^T \Sigma \underline{c})$$

$$\Rightarrow \forall \underline{c}: [\underline{c}_1^T \quad \underline{0}^T] \underline{y} \sim N([\underline{c}_1^T \quad \underline{0}^T] \underline{\theta}, [\underline{c}_1^T \quad \underline{0}^T] \Sigma \begin{bmatrix} \underline{c}_1 \\ \underline{0} \end{bmatrix})$$

$$\Leftrightarrow \forall \underline{c}_1: \underline{c}_1^T \underline{y}^{(1)} \sim N(\underline{c}_1^T \underline{\theta}^{(1)}, \underline{c}_1^T \Sigma_{11} \underline{c}_1)$$

$$\Leftrightarrow \underline{y}^{(1)} \sim N_d(\underline{\theta}^{(1)}, \Sigma_{11}), \quad \Sigma_{11} \geq 0 \quad *)$$

ad (iii):

Has been shown in exercise 2.2

ad (iv):

$$\underline{c} = \begin{bmatrix} \underline{c}_1 \\ \underline{c}_2 \end{bmatrix}, \quad \begin{aligned} \underline{c}_1^T \underline{y}^{(1)} &\sim N(\underline{c}_1^T \underline{\theta}^{(1)}, \underline{c}_1^T \Sigma_{11} \underline{c}_1) & \underline{c}_1 & d_1\text{-dim.} \\ \underline{c}_2^T \underline{y}^{(2)} &\sim N(\underline{c}_2^T \underline{\theta}^{(2)}, \underline{c}_2^T \Sigma_{22} \underline{c}_2) & \underline{c}_2 & d_2\text{-dim.} \end{aligned}$$

$\underline{y}^{(1)}$ and $\underline{y}^{(2)}$ independent

$$\Leftrightarrow \forall \underline{c}_1, \underline{c}_2: \underline{c}_1^T \underline{y}^{(1)} \text{ and } \underline{c}_2^T \underline{y}^{(2)} \text{ independent}$$

$$\Leftrightarrow \forall \underline{c}_1, \underline{c}_2: \text{Cov}(\underline{c}_1^T \underline{y}^{(1)}, \underline{c}_2^T \underline{y}^{(2)}) = 0$$

$$\Leftrightarrow \forall \underline{c}_1, \underline{c}_2: \underline{c}_1^T \text{Cov}(\underline{y}^{(1)}, \underline{y}^{(2)}) \underline{c}_2 = 0$$

$$\Leftrightarrow \text{Cov}(\underline{y}^{(1)}, \underline{y}^{(2)}) = 0$$

*) the notation $\underline{y} \sim N(\cdot, \cdot)$ is used although $\text{Var} \underline{y}$ may be singular