

2.22

$$W \sim W_d(m, \Sigma), \quad \Sigma > 0$$

$$\begin{aligned} \frac{\det W}{\det \Sigma} &= \det \Sigma^{-\frac{1}{2}} \det W \det \Sigma^{-\frac{1}{2}} = \det (\Sigma^{-\frac{1}{2}} W (\Sigma^{-\frac{1}{2}})^T) \\ &= \det Z, \quad Z \sim W_d(m, \Sigma^{-\frac{1}{2}} \Sigma (\Sigma^{-\frac{1}{2}})^T) = W_d(m, I_d) \\ &= z_{11} \tilde{z}_{22} \dots \tilde{z}_{dd}, \quad \text{cf. formula (2.61)} \end{aligned}$$

$$\tilde{z}_{kk} = z_{kk} - \tilde{z}_{k-1}^T Z_{k-1}^{-1} \tilde{z}_{k-1}, \quad Z_1 = [z_{11}]$$

$$Z_k = \begin{bmatrix} Z_{k-1} & z_{k-1} \\ \tilde{z}_{k-1}^T & z_{kk} \end{bmatrix}$$

$$k = 2, \dots, d$$

Covollary 2 to theorem 2.2 shows that

$$z_{11} \sim \chi^2(m)$$

Lemma 2.6 and 2.7 shows for $k = 2, 3, \dots, d$ (cf. lecture note MA5) that

$$\tilde{z}_{kk} \sim \chi^2(m-k+1) \text{ indep. of } Z_{k-1} \quad *$$

$\Rightarrow z_{11}, \tilde{z}_{22}, \dots, \tilde{z}_{dd}$ are independent χ^2 distributed variables with degrees of freedom $m, m-1, \dots, m-d+1$

* Alternatively by use of lemma 2.10:

$$\tilde{z}_{kk} \sim W_1(m-k+1, 1 - \underset{\sim}{0}^T \underset{\sim}{I_{k-1}}^{-1} \underset{\sim}{0}) = \chi^2(m-k+1)$$

indep. of Z_{k-1}

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$$W \sim W_d(m, \Sigma), \quad \Sigma > 0, \quad m \geq d+2$$

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$$\frac{\underline{l}^T \Sigma^{-1} \underline{l}}{\underline{l}^T W^{-1} \underline{l}} \sim \chi^2(m-d+1) \text{ for all } \underline{l} \neq \underline{0} \text{ cf. lemma 2.7}$$

$$\Rightarrow E \left[\left(\frac{\underline{l}^T \Sigma^{-1} \underline{l}}{\underline{l}^T W^{-1} \underline{l}} \right)^{-1} \right] = \frac{1}{(m-d+1)-2} = \frac{1}{m-d-1}$$

$$\Rightarrow E \left[\underline{l}^T W^{-1} \underline{l} \right] = \frac{1}{m-d-1} \underline{l}^T \Sigma^{-1} \underline{l}$$

$$\Rightarrow \underline{l}^T E[W^{-1}] \underline{l} = \underline{l}^T \frac{\Sigma^{-1}}{m-d-1} \underline{l}$$

$$\Rightarrow E[W^{-1}] = \frac{\Sigma^{-1}}{m-d-1}$$

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$$\underline{y} \sim N_d(\underline{0}, \Sigma), \quad \underline{y} \text{ of } W \text{ independent}$$

$$\begin{aligned} E[\underline{y}^T W^{-1} \underline{y}] &= E[\text{tr}(\underline{y}^T W^{-1} \underline{y})] = E[\text{tr} W^{-1} \underline{y} \underline{y}^T] \\ &= \text{tr} [E W^{-1} E(\underline{y} \underline{y}^T)] = \text{tr}(E W^{-1} \text{Var } \underline{y}) \\ &= \text{tr} \left(\frac{\Sigma^{-1}}{m-d-1} \Sigma \right) = \frac{\text{tr } I_d}{m-d-1} \\ &= \frac{d}{m-d-1} \end{aligned}$$

2.24

 $\underline{x}_1, \underline{x}_2 \sim N_d(\underline{0}, \Sigma)$ independent

$$a \underline{x}_1 \underline{x}_1^T + b \underline{x}_1 \underline{x}_2^T + b \underline{x}_2 \underline{x}_1^T + a \underline{x}_2 \underline{x}_2^T = \begin{bmatrix} \underline{x}_1 & \underline{x}_2 \end{bmatrix} \begin{bmatrix} a & b \\ b & a \end{bmatrix} \begin{bmatrix} \underline{x}_1^T \\ \underline{x}_2^T \end{bmatrix} \\ = X^T A X$$

Corollary 1 to theorem 2.4 shows

$$X^T A X \sim W_d(r, \Sigma) \Leftrightarrow A^2 = A$$

$$\Leftrightarrow \begin{bmatrix} a & b \\ b & a \end{bmatrix} \begin{bmatrix} a & b \\ b & a \end{bmatrix} = \begin{bmatrix} a & b \\ b & a \end{bmatrix}$$

$$\Leftrightarrow \begin{bmatrix} a^2 + b^2 & 2ab \\ 2ab & a^2 + b^2 \end{bmatrix} = \begin{bmatrix} a & b \\ b & a \end{bmatrix}$$

$$\Leftrightarrow \begin{cases} a = a^2 + b^2 \\ b = 2ab \end{cases}$$

$$\Leftrightarrow (a, b) = \begin{cases} (\frac{1}{2}, \frac{1}{2}) & (r=1) \\ (\frac{1}{2}, -\frac{1}{2}) & (r=1) \\ (1, 0) & (r=2) \end{cases}$$

2.25

 $\underline{x}_i \sim N_d(\underline{0}, \Sigma)$, $i=1, \dots, m$, independent

$$X = [\underline{x}_1 \dots \underline{x}_m]^T$$

 A_j , $j=1, \dots, r$, sym. og idempotent $d \times d$

$$A_j A_k = 0, \quad j \neq k$$

Corollary 2 to theorem 2.4 shows

$$A_j A_k = 0, \quad j \neq k \Leftrightarrow \left\{ \begin{array}{l} X^T A_j X \sim W_d(m_j, \Sigma) \\ X^T A_k X \sim W_d(m_k, \Sigma) \end{array} \right\} \text{ indep.}$$

$$\Rightarrow X^T A_j X, \quad j=1, \dots, r, \text{ mutually independent}$$