

Revision.

CH3. vector, matrix.

— Full rank factorization  $r = \text{rank}(A)$

$$\tilde{A}_{p \times q} = \tilde{K}_{p \times r} \tilde{L}_{r \times q}$$

• left inverse:  $\tilde{U}_{p \times p} \tilde{K}_{p \times r} = I_r$

• right inverse:  $\tilde{L}_{r \times q} \tilde{R}_{q \times r} = I_r$

— Idempotent Matrix ( $A^2 = A$ )

\*  $\text{rank}(A) = \text{tr}(A)$

\* eigenvalues of  $A$ :  $\underbrace{1 \dots 1}_r 0 \dots 0$

— Generalized Inverse.

\* definition:  $\tilde{A} \tilde{A}^+ \tilde{A} = \tilde{A}$

\* how to calculate: e.g.  $\tilde{A} = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}$    
  $\begin{cases} \cdot \text{full rank} \\ \cdot \text{rank}(A_{11}) = \text{rank}(A) \end{cases}$

\* some properties.

\* Moore-Penrose Inverse  $\left\{ \begin{array}{l} \cdot \text{special case of } A^- \\ \cdot \text{unique} \end{array} \right.$

CH3: — Vector and matrix calculus

\*  $u = f(x)$ ,  $x = \begin{pmatrix} x_1 \\ \vdots \\ x_p \end{pmatrix}$ ,  $\frac{\partial u}{\partial x}$  and

$$\frac{\partial u^T x}{\partial x} = u, \quad \frac{\partial x^T A x}{\partial x} = 2 A x$$

\*  $u = f(x)$ ,  $x = \begin{pmatrix} x_{11} & \dots & x_{1p} \\ \vdots & & \vdots \\ x_{p1} & \dots & x_{pp} \end{pmatrix}$  — symmetric positive definite

$$\frac{\partial \text{tr}(xA)}{\partial x} = A + A^T - \text{diag}(A)$$

Condition is not sure.

$$\frac{\partial \ln |x|}{\partial x} = 2 x^{-1} - \text{diag}(x^{-1})$$

\*  $A_{mn} = (a_{ij})_{mn}$ ,  $\frac{\partial A}{\partial x} = \left( \frac{\partial a_{ij}(x)}{\partial x} \right)$

$$\frac{\partial A^{-1}}{\partial x} = -A^{-1} \frac{\partial A}{\partial x} A^{-1}$$

$$\frac{\partial \ln |A|}{\partial x} = \text{tr} \left( A^{-1} \frac{\partial A}{\partial x} \right)$$

#### CH4. Random Vector, Matrix.

$$X = \begin{pmatrix} x_1 \\ \vdots \\ x_p \end{pmatrix} \quad x_1, \dots, x_p \text{ are r.v.s}$$

— mean vector and covariance matrix.

$$* E(X) = \mu_{p \times 1} \quad \Sigma_{p \times p} = \text{Cov}(X) = (\text{Cov}(X_i, X_j))_{p \times p}$$

\* Sample mean and sample covariance matrix.

$$\hat{\mu} = \bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$

observations.

$$\hat{\Sigma} = \hat{S} = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})(x_i - \bar{x})' \quad x_i = \begin{pmatrix} x_{i1} \\ \vdots \\ x_{ip} \end{pmatrix} \quad i=1, \dots, n$$

$$* \text{Cov}(AX, BX) = A \text{Cov}(X, X) B'$$

$$\text{Cov}(AX) = A \text{Cov}(X) A'$$

\* generalized variance  $|\Sigma|$

\* Correlation matrix

\* mgf of  $X$ ,  $M_X(s) = E(e^{s'X})$

\* If  $x_1, \dots, x_p$  are mutually independent.

$\Rightarrow g_1(y_1), g_2(y_2), \dots, g_m(y_m)$  are mutually independent

#### CH5. Multivariate normal

— Density, mgf

— Some important properties.  $X \sim N(\mu, \Sigma)$

$$* BY + C \sim N(B\mu + C, B\Sigma B')$$

$$* X = \begin{pmatrix} x_1 \\ \vdots \\ x_p \end{pmatrix} \sim N\left(\begin{pmatrix} \mu_1 \\ \vdots \\ \mu_p \end{pmatrix}, \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix}\right)$$

\* marginal distribution of  $y_1$  and  $y_2$

\* Conditional distribution  $x_1 | x_2$

\* Independence of  $y_1$  and  $y_2 \Leftrightarrow \text{Cov}(y_1, y_2) = 0$ .

— Partial correlation.

\* definition.

\* meaning.

CH6. Quadratic form.

$$\underline{X} \sim N(\underline{\mu}, \underline{\Sigma}) \quad \underline{X}'A\underline{X}, A - \text{symmetric.}$$

\* mgf of  $\underline{X}'A\underline{X}$

$$* E(\underline{X}'A\underline{X}), \text{Var}(\underline{X}'A\underline{X})$$

— Non-central  $\chi^2$ , F, t distribution.

$$\cdot \underline{X} \sim N(\underline{\mu}, I_n), U = \sum X_i^2 \sim \chi^2(n, \lambda).$$

$$\lambda = \frac{1}{2} \sum \mu_i^2$$

• Non-central F.

$$U_1 \sim \chi^2(p, \lambda), U_2 \sim \chi^2(q). U_1 \text{ and } U_2 \text{ are independent}$$

$$W = \frac{U_1/p}{U_2/q} \sim F(p, q, \lambda)$$

• Non-central t.

$$\underline{Z} \sim N(\underline{\mu}, 1) \quad U \sim \chi^2(n). \quad \underline{Z} \text{ and } U \text{ are independent}$$

$$t = \frac{\underline{Z}}{\sqrt{U/n}} \sim t(n, \lambda)$$



$$\text{Thm 6.1. } \underline{X} \sim N(\underline{\mu}, \underline{\Sigma})$$

$$q = \underline{X}'A\underline{X} \sim \chi^2(r, \lambda), \quad \lambda = \frac{\underline{\mu}'A\underline{\mu}}{2}, r = \text{rank}(A)$$

$$\Leftrightarrow A\underline{\Sigma} \text{ is idempotent}$$

$$\text{Thm 6.2. } \underline{X}'A\underline{X} \text{ and } \underline{X}'B\underline{X} \text{ are independent.}$$