

$$1. (a) f(x) = \frac{1}{2}x^T A x + b^T x$$

$$\nabla f(x) = \frac{1}{2} \cdot 2Ax + b = Ax + b$$

$$\therefore \nabla f(x) = Ax + b$$

$$(b) f(x) = g(h(x))$$

$$\nabla f(x) = g'(h(x)) \nabla h(x)$$

$$\therefore \nabla f(x) = g'(h(x)) \nabla h(x)$$

$$(c) f(x) = \frac{1}{2}x^T A x + b^T x$$

$$\nabla f(x) = Ax + b$$

$$\nabla^2 f(x) = A$$

$$(d) f(x) = g(a^T x)$$

$$\nabla f(x) = g'(a^T x) a$$

$$\nabla^2 f(x) = g'(a^T x) a \left[ \frac{1}{\partial x_1} \quad \frac{1}{\partial x_2} \quad \dots \quad \frac{1}{\partial x_n} \right]$$

$$= \begin{bmatrix} g''(a^T x) a_1 a_1 & & & g''(a^T x) a_1 a_n \\ \vdots & & & \vdots \\ g''(a^T x) a_n a_1 & & & g''(a^T x) a_n a_n \end{bmatrix}$$

$$= g''(a^T x) a a^T$$

$$\therefore \nabla f(x) = g'(a^T x) a, \quad \nabla^2 f(x) = g''(a^T x) a a^T$$

2. (a)  $A = zz^T$ ,  $z \in \mathbb{R}^n$

i)  $A = A^T$ ?

$$(zz^T)^T = (z^T)^T z^T = zz^T$$

$$\therefore A = A^T$$

ii)  $x^T A x \geq 0$  for  $\forall x \in \mathbb{R}^n$

$$\begin{aligned} x^T A x &= \sum_{j=1}^n \sum_{i=1}^n x_j A_{ij} x_i = \sum_{j=1}^n \sum_{i=1}^n x_j z_i z_j x_i = \sum_{i=1}^n x_i z_i \sum_{j=1}^n x_j z_j \\ &= \left( \sum_{i=1}^n x_i z_i \right)^2 \geq 0 \end{aligned}$$

$$\therefore A \in S_+^n$$

(b)  $N(A) = \{x \mid Ax = 0\}$

$$Ax = zz^T x$$

$$\therefore N(A) = \{x \mid x = \vec{0} \text{ or } z^T x = 0\}$$

$$zz^T = \begin{bmatrix} z_1 \\ \vdots \\ z_n \end{bmatrix} [z_1 \dots z_n] = \begin{bmatrix} z_1 z_1 & z_1 z_2 & \dots & z_1 z_n \\ \vdots & \vdots & \ddots & \vdots \\ z_n z_1 & z_n z_2 & \dots & z_n z_n \end{bmatrix}$$

$$\sim \begin{bmatrix} z_1 z_1 & \dots & z_1 z_n \\ 0 & \dots & 0 \\ 0 & \dots & 0 \end{bmatrix} \quad \therefore \text{Rank}(A) = 1$$

$$\therefore N(A) = \{x \mid x = \vec{0} \text{ or } z^T x = 0\}, \text{Rank}(A) = 1$$

(c)  $BAB^T$

$$(BAB^T)^T = (AB^T)^T B^T = (B^T)^T A^T B^T = BA^T B^T = BAB^T$$

$$(\because A = A^T) \quad \therefore BAB^T = (BAB^T)^T$$

ii)  $x^T BAB^T x \geq 0$  for  $\forall x \in \mathbb{R}^m$

$$(B^T x)^T = x^T B, \quad B^T x \in \mathbb{R}^n, \quad A \in S_+^n \Rightarrow x^T A x \geq 0 \text{ for } \forall x (= B^T x) \in \mathbb{R}^n$$

$$\therefore BAB^T \text{ is PSD.}$$

3.

(a)  $A = T\Lambda T^{-1}$ ,  $TT^{-1} = I = T^{-1}T$

$$AT = T\Lambda T^{-1}T = T\Lambda I = T\Lambda$$

i)  $AT$

$$A[x^1 \ x^2 \ \dots \ x^n] = [Ax^1 \ Ax^2 \ \dots \ Ax^n]$$

ii)  $T\Lambda$

$$[x^1 \ x^2 \ \dots \ x^n] \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{bmatrix}$$

$$= x^1 [\lambda_1 \ 0 \ \dots \ 0] + x^2 [0 \ \lambda_2 \ \dots \ 0] + \dots + x^n [0 \ \dots \ 0 \ \lambda_n]$$

$$= \begin{bmatrix} x^1 \\ \vdots \\ x^n \end{bmatrix} [\lambda_1 \ 0 \ \dots \ 0] + \dots + \begin{bmatrix} x^1 \\ \vdots \\ x^n \end{bmatrix} [0 \ \dots \ 0 \ \lambda_n]$$

$$= \begin{bmatrix} \lambda_1 x^1 & 0 & \dots & 0 \\ \vdots & 0 & \ddots & \vdots \\ \lambda_1 x^n & 0 & \dots & 0 \end{bmatrix} + \dots + \begin{bmatrix} 0 & \dots & 0 & \lambda_n x^n \\ \vdots & \ddots & \vdots & \vdots \\ 0 & \dots & 0 & \lambda_n x^n \end{bmatrix}$$

$$= \begin{bmatrix} \lambda_1 x^1 & \lambda_2 x^2 & \dots \\ \vdots & \vdots & \ddots \\ \lambda_1 x^n & \vdots & \dots \end{bmatrix} = [\lambda_1 x^1 \ \lambda_2 x^2 \ \dots \ \lambda_n x^n]$$

$$\boxed{\therefore Ax^i = \lambda_i x^i}$$

(b)  $U$  is orthogonal  $\Rightarrow U U^T = I = U^T U$   
 is  $A = U \Lambda U^T \Rightarrow A U = U \Lambda U^T U = U \Lambda (U^T U) = U \Lambda I = U \Lambda$   
 $\therefore A U = U \Lambda$

(i)  $A U$

$$= A [u^1 \dots u^n] = [A u^1 \quad A u^2 \quad \dots \quad A u^n]$$

(ii)  $U \Lambda$

$$[u^1 \dots u^n] \begin{bmatrix} \lambda_1 & \dots & & 0 \\ 0 & \lambda_2 & & 0 \\ \vdots & & & \vdots \\ 0 & & & \lambda_n \end{bmatrix}$$

$$= u^1 [\lambda_1 \dots 0] + u^2 [0 \lambda_2 \dots 0] + \dots + u^n [0 \dots \lambda_n]$$

$$= \begin{bmatrix} u^1 \lambda_1 & \dots & 0 \\ \vdots & & \\ u^n \lambda_2 & \dots & 0 \end{bmatrix} + \dots + \begin{bmatrix} 0 & \dots & \lambda_n u^n \\ \vdots & & \vdots \\ 0 & \dots & \lambda_n u^n \end{bmatrix}$$

$$= \begin{bmatrix} \lambda_1 u^1 & \dots & \lambda_n u^n \\ \vdots & & \vdots \\ \lambda_1 u^n & \dots & \lambda_n u^n \end{bmatrix} = [\lambda_1 u^1 \quad \lambda_2 u^2 \quad \dots \quad \lambda_n u^n]$$

Since  $A U = U \Lambda$ , each element of matrices is identical.

$$\boxed{\therefore A u^i = \lambda_i u^i}$$

(c)  $A$  is PSD  $\Rightarrow x^T A x \geq 0$  for  $\forall x \in \mathbb{R}^n$ .

$$Ax = \lambda x \Rightarrow x^T A x = x^T \lambda x$$

~~is/when~~

$$x^T \lambda x = \sum_{i=1}^n \sum_{j=1}^n x_j \lambda_{ij} x_i$$

$$\lambda = \text{diag}(\lambda_1, \lambda_2, \dots, \lambda_n)$$

So, when  $i \neq j$ ,  $x_j \lambda_{ij} x_i = 0$ .

When  $i = j$ ,  $x_j \lambda_{ij} x_i = x_i \lambda_{ii} x_i = \{x_i\}^2 \lambda_i(A)$

$$\text{Thus, } x^T \lambda x = \sum_{i=1}^n \sum_{j=1}^n x_j \lambda_{ij} x_i = \sum_{i=1}^n \{x_i\}^2 \lambda_i(A) \geq 0.$$

Since  $\{x_i\}^2 \lambda_i(A) \geq 0$ ,  $\lambda_i(A) \geq 0$ .

$$\therefore \lambda_i(A) \geq 0 \text{ for each } i$$

$$4. (a) Y = X_1 + \dots + X_n$$

$$E[Y] = E[X_1 + \dots + X_n] = E[X_1] + E[X_2] + \dots + E[X_n]$$

$$= \mu_1 + \mu_2 + \dots + \mu_n = \mathbf{1}^T \boldsymbol{\mu}$$

$$\text{Var}[Y] = \text{Var}[X_1 + X_2 + \dots + X_n]$$

$$= \sum_{i=1}^n \text{Var}(X_i) + \sum_{i,j} \text{Cov}(X_i, X_j) \quad (i \neq j)$$

$$= \sum_{i=1}^n \sum_{j=1}^n (\Sigma)_{ij} = \mathbf{1}^T \Sigma \mathbf{1}$$

$$\therefore E[Y] = \mathbf{1}^T \boldsymbol{\mu}, \text{Var}[Y] = \mathbf{1}^T \Sigma \mathbf{1}$$

$Y$  is normal distribution.

$$(b) E[X^T \Sigma^{-1} X] = E[\text{tr}(X^T \Sigma^{-1} X)]$$

$$= E[\text{tr}(\Sigma^{-1} X X^T)] = \text{tr}(E[\Sigma^{-1} X X^T])$$

$$= \text{tr}(\Sigma^{-1} E[X X^T])$$

$$= \text{tr}(\Sigma^{-1} (\Sigma + \boldsymbol{\mu} \boldsymbol{\mu}^T)) = \text{tr}(\mathbf{I} + \Sigma^{-1} \boldsymbol{\mu} \boldsymbol{\mu}^T)$$

$$= \text{tr}(\mathbf{I}) + \text{tr}(\Sigma^{-1} \boldsymbol{\mu} \boldsymbol{\mu}^T)$$

$$= n + \boldsymbol{\mu}^T \Sigma^{-1} \boldsymbol{\mu}$$

$$\therefore E[X^T \Sigma^{-1} X] = n + \boldsymbol{\mu}^T \Sigma^{-1} \boldsymbol{\mu}$$