

Stanford, CS 229, 2018, autumn, ps 0.

1. $f: \mathbb{R}^n \rightarrow \mathbb{R}$

(a). $f(x) = \frac{1}{2} x^T A x + b^T x$, $A \in \mathbb{S}$, $b \in \mathbb{R}^n$.

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

$$= \begin{bmatrix} \frac{\partial}{\partial x_1} (\frac{1}{2} x^T A x + b^T x) \\ \vdots \\ \frac{\partial}{\partial x_n} (\frac{1}{2} x^T A x + b^T x) \end{bmatrix}$$

$$= \begin{bmatrix} \frac{\partial}{\partial x_1} (\frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n A_{ij} x_i x_j + \sum_{i=1}^n b_i x_i) \\ \vdots \\ \frac{\partial}{\partial x_n} (\frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n A_{ij} x_i x_j + \sum_{i=1}^n b_i x_i) \end{bmatrix}$$

$$= \begin{bmatrix} \frac{\partial}{\partial x_1} (\frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n A_{ij} x_i x_j + \sum_{i=1}^n b_i x_i) \\ \vdots \\ \frac{\partial}{\partial x_n} (\frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n A_{ij} x_i x_j + \sum_{i=1}^n b_i x_i) \end{bmatrix}$$

$$= \begin{bmatrix} \frac{\partial}{\partial x_1} (\frac{1}{2} (\sum_{j=1}^n A_{1j} x_1 x_j + \sum_{i=1}^n A_{i1} x_i x_1 + \sum_{i=2}^n \sum_{j=2}^n A_{ij} x_i x_j) + \sum_{i=1}^n b_i x_i) \\ \vdots \\ \frac{\partial}{\partial x_n} (\frac{1}{2} (\sum_{j=1}^n A_{nj} x_n x_j + \sum_{i=1}^n A_{in} x_i x_n + \sum_{i=2}^n \sum_{j=2}^n A_{ij} x_i x_j) + \sum_{i=1}^n b_i x_i) \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{2} \sum_{j=1}^n A_{1j} x_j + \frac{1}{2} \sum_{i=1}^n A_{i1} x_i + b_1 \\ \vdots \\ \frac{1}{2} \sum_{j=1}^n A_{nj} x_j + \frac{1}{2} \sum_{i=1}^n A_{in} x_i + b_n \end{bmatrix}$$

$$\frac{\partial}{\partial x} A_{11} x_1^2 = 2 A_{11} x_1$$

$$= \begin{bmatrix} \sum_{j=1}^n A_{1j} x_j + b_1 \\ \vdots \\ \sum_{j=1}^n A_{nj} x_j + b_n \end{bmatrix}$$

$$= \begin{bmatrix} \sum_{j=1}^n A_{1j} x_j \\ \vdots \\ \sum_{j=1}^n A_{nj} x_j \end{bmatrix} + \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$$

$$= (x^T A)^T + b$$

$$= A^T x + b$$

$$= A x + b.$$

$$[x_1, x_2, \dots, x_n] \cdot \begin{bmatrix} A_{11} & \dots & A_{1n} \\ \vdots & & \vdots \\ A_{n1} & \dots & A_{nn} \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

$$= \begin{bmatrix} \sum_i^n A_{i1} \cdot x_i & \sum_i^n A_{i2} \cdot x_i & \dots & \sum_i^n A_{in} x_i \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

$$= \sum_i^n \sum_j^n A_{ij} x_i x_j$$

(b). $f(x) = g(h(x))$, $g: \mathbb{R} \rightarrow \mathbb{R}$, $h: \mathbb{R}^n \rightarrow \mathbb{R}$

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

$$= \begin{bmatrix} \frac{\partial}{\partial x_1} g(h(x)) \\ \vdots \\ \frac{\partial}{\partial x_n} g(h(x)) \end{bmatrix}$$

$$\frac{\partial}{\partial x_i} g(h(x)) = \frac{\partial g(h(x))}{\partial h(x)} \cdot \frac{\partial h(x)}{\partial x_i}$$

$$\rightarrow = \begin{bmatrix} \frac{\partial}{\partial h(x)} g(h(x)) \cdot \frac{\partial h(x)}{\partial x_1} \\ \vdots \\ \frac{\partial}{\partial h(x)} g(h(x)) \cdot \frac{\partial h(x)}{\partial x_n} \end{bmatrix}$$

$$= \frac{\partial}{\partial h(x)} g(h(x)) \cdot \begin{bmatrix} \frac{\partial}{\partial x_1} h(x) \\ \vdots \\ \frac{\partial}{\partial x_n} h(x) \end{bmatrix}$$

$$= \frac{\partial}{\partial h(x)} g(h(x)) \nabla h(x)$$

(c)

$$f(x) = \frac{1}{2} x^T A x + b^T x, \quad A \in \mathbb{S}^{n \times n}, \quad b \in \mathbb{R}^n$$

$$\begin{aligned} \nabla^2 f(x) &= \begin{bmatrix} \frac{\partial^2}{\partial x_1 \partial x_1} f(x) & \dots & \frac{\partial^2}{\partial x_1 \partial x_n} f(x) \\ \vdots & & \vdots \\ \frac{\partial^2}{\partial x_n \partial x_1} f(x) & \dots & \frac{\partial^2}{\partial x_n \partial x_n} f(x) \end{bmatrix} \\ &= \begin{bmatrix} \frac{\partial}{\partial x_1} \frac{\partial}{\partial x_1} f(x) & \dots & \frac{\partial}{\partial x_1} \frac{\partial}{\partial x_n} f(x) \\ \vdots & & \vdots \\ \frac{\partial}{\partial x_n} \frac{\partial}{\partial x_1} f(x) & \dots & \frac{\partial}{\partial x_n} \frac{\partial}{\partial x_n} f(x) \end{bmatrix} \\ &= \begin{bmatrix} A_{11} & \dots & A_{1n} \\ \vdots & & \vdots \\ A_{n1} & \dots & A_{nn} \end{bmatrix} \\ &= A. \end{aligned}$$

(d)

$$f(x) = g(a^T x), \quad g: \mathbb{R} \rightarrow \mathbb{R}, \quad a \in \mathbb{R}^n$$

$$\begin{aligned} \nabla f(x) &= \begin{bmatrix} \frac{\partial}{\partial x_1} g(a^T x) \\ \vdots \\ \frac{\partial}{\partial x_n} g(a^T x) \end{bmatrix} \\ \left(\frac{\partial}{\partial x_i} g(a^T x) \right) &= \frac{\partial g(a^T x)}{\partial a^T x} \cdot \frac{\partial a^T x}{\partial x_i} \\ &= \frac{\partial g(a^T x)}{\partial a^T x} \cdot a_i \\ &\rightarrow \begin{bmatrix} g'(a^T x) \cdot a_1 \\ \vdots \\ g'(a^T x) \cdot a_n \end{bmatrix} \\ &= g'(a^T x) \cdot \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \\ &= g'(a^T x) \cdot a \end{aligned}$$

$$\nabla^2 f(x) = \begin{bmatrix} \frac{\partial^2}{\partial x_1 \partial x_1} g(a^T x) & \dots & \frac{\partial^2}{\partial x_1 \partial x_n} g(a^T x) \\ \vdots & & \vdots \\ \frac{\partial^2}{\partial x_n \partial x_1} g(a^T x) & \dots & \frac{\partial^2}{\partial x_n \partial x_n} g(a^T x) \end{bmatrix}$$

$$\frac{\partial^2}{\partial x_i \partial x_j} g(a^T x) = \frac{\partial}{\partial x_i} \cdot \frac{\partial}{\partial x_j} g(a^T x)$$

$$= \frac{\partial}{\partial x_i} \cdot g'(a^T x) \cdot a_j$$

$$= a_j \frac{\partial}{\partial x_i} g'(a^T x)$$

$$= a_j \frac{\partial}{\partial a^T x} \cdot g'(a^T x) \cdot \frac{\partial}{\partial x_i} a^T x$$

$$= a_j \cdot g''(a^T x) \cdot a_i$$

$$= a_i \cdot a_j \cdot g''(a^T x)$$

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

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

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

2.

(a).

$$z \in \mathbb{R}^n$$

$$A = zz^T \succeq 0?$$

$$\text{let } x \in \mathbb{R}^n.$$

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

$$\text{so, } A \succeq 0.$$

(b)

~~$$z \in \mathbb{R}^n, z_i \neq 0 \text{ for } i \in [n], A = zz^T$$~~

~~$$\text{let } x \in \mathbb{R}^n, x_i \neq 0 \text{ for } i \in [n].$$~~

~~$$x^T A x = x^T z z^T x = \left(\sum_{i=1}^n x_i z_i \right)^2 > 0$$~~

~~$$\text{so, } A \succ 0, A \text{ is positive.}$$~~

nullspace :

$$\begin{aligned} \text{let } x \in \mathbb{R}^n, Ax &= zz^T x \\ &= z \cdot \sum_{i=1}^n x_i z_i \end{aligned}$$

$$\downarrow$$

$$\text{if } Ax = 0 \rightarrow \sum_{i=1}^n x_i z_i = 0$$

$$z^T x = 0$$

rank :

~~because A is positive~~
~~so A is full rank~~

$$\text{rank}(A) = \text{rank}(zz^T)$$

$$\leq \min(\text{rank}(z), \text{rank}(z^T))$$

$$= 1$$

$$N(A) = \{x \in \mathbb{R}^n : z^T x = 0\}$$

(c).

$$A \in \mathbb{R}^{n \times n}, \quad A \succeq 0, \quad B \in \mathbb{R}^{m \times n}$$

$$BAB^T : \text{~~PSD~~} \quad \text{PSD?}$$

$$\text{let } x \in \mathbb{R}^n,$$

$$x^T A x \geq 0.$$

if BAB^T is PSD:

$$x \in \mathbb{R}^m, \quad x^T BAB^T x \geq 0$$

$$(x^T B) A (B^T x) \geq 0$$

$$(B^T x)^T A (B^T x) \geq 0$$

$$B^T x = y \in \mathbb{R}^n, \quad y^T A y \geq 0.$$

So BAB^T is PSD.

3.

$$(a). A \in \mathbb{R}^{n \times n}, A = T \Lambda T^{-1}$$

$$T \in \mathbb{R}^{n \times n}, \Lambda = \text{diag}(\lambda_1, \dots, \lambda_n)$$

$$T = [t^{(1)} \dots t^{(n)}], t^{(i)} \in \mathbb{R}^n$$

$$\text{show: } A t^{(i)} = \lambda_i t^{(i)}$$

$$A = T \Lambda T^{-1}$$

$$A T = T \Lambda$$

$$\begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nn} \end{bmatrix} \cdot [t^{(1)} \dots t^{(n)}] = [t^{(1)} \dots t^{(n)}] \begin{bmatrix} \lambda_1 & \dots & 0 \\ \vdots & & \vdots \\ 0 & \dots & \lambda_n \end{bmatrix}$$

$$A t^{(i)} = t^{(i)} \lambda_i \quad (\text{for each } i \text{ in } n.)$$

$$A t^{(i)} = \lambda_i t^{(i)}$$

(b).

$$A^T = A, \quad U = [u^{(1)} \dots u^{(n)}], \quad U^T U = I$$

$$A = U \Lambda U^T, \quad u^{(i)} \in \mathbb{R}^n$$

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

$$\text{show: } A u^{(i)} = \lambda_i u^{(i)}$$

$$A = U \Lambda U^T$$

$$A U = U \Lambda$$

$$A [u^{(1)} \dots u^{(n)}] = [u^{(1)} \dots u^{(n)}] \cdot \text{diag}(\lambda_1, \dots, \lambda_n)$$

$$[A u^{(1)} \dots A u^{(n)}] = [\lambda_1 u^{(1)} \dots \lambda_n u^{(n)}]$$

$$A u^{(i)} = \lambda_i u^{(i)}$$

so $u^{(i)}$ is an eigenvector of A .

(c) A is PSD, $A \succeq 0$.

$$x \in \mathbb{R}^n, \quad x^T A x \geq 0.$$

$\lambda_i(A)$ means i th eigenvalue of A .

show: $\lambda_i(A) \geq 0$ for each i , $\lambda_i(A) = \lambda_i$

$$A b^{(i)} = \lambda_i b^{(i)}$$

$$(b^{(i)})^T A b^{(i)} \geq 0.$$

$$(b^{(i)})^T \lambda_i b^{(i)} \geq 0$$

$$\lambda_i (b^{(i)})^T b^{(i)} \geq 0.$$

$$\lambda_i \|b^{(i)}\|_2^2 \geq 0.$$

$$\lambda_i \geq 0$$