

Problem Set 0 Solution

Problem 1

(a)

When $f(x) = \frac{1}{2} \vec{x}^T A \vec{x} + \vec{b}^T \vec{x}$, we can derive that

$$\nabla f(x) = \nabla \left(\frac{1}{2} \vec{x}^T A \vec{x} + \vec{b}^T \vec{x} \right) = A \vec{x} + \vec{b}$$

(b)

When $f(x) = g(h(x))$, we can derive that

$$\begin{aligned} \nabla f(x) &= \frac{\partial}{\partial x_i} f(x) = \frac{\partial}{\partial x_i} g(h(x)) = \frac{\partial g(h(x))}{\partial x_i} \\ &= \frac{\partial g(h(x))}{\partial h(x)} \frac{\partial h(x)}{\partial x_i} \\ &= g'(h(x)) \nabla h(x) \end{aligned}$$

(c)

When $f(x) = \frac{1}{2} \vec{x}^T A \vec{x} + \vec{b}^T \vec{x}$, we have obtained that

$$\nabla f(x) = A \vec{x} + \vec{b}$$

For the given matrices

$$A = \begin{pmatrix} A_{11} & \cdots & A_{1n} \\ \vdots & \ddots & \vdots \\ A_{n1} & \cdots & A_{nn} \end{pmatrix}, \quad \vec{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}, \quad \text{and} \quad \vec{b} = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{pmatrix}$$

We can derive that

$$A \vec{x} + \vec{b} = \begin{pmatrix} A_{11}x_1 + A_{12}x_2 + \cdots + A_{1n}x_n + b_1 \\ \vdots \\ A_{n1}x_1 + A_{n2}x_2 + \cdots + A_{nn}x_n + b_n \end{pmatrix} = \begin{pmatrix} \sum_{i=1}^n A_{1i}x_i + b_1 \\ \vdots \\ \sum_{i=1}^n A_{ni}x_i + b_n \end{pmatrix}$$

Thus, from an intuitive view, we can derive that

$$\nabla^2 f(x) = \left(\frac{\partial \nabla f(x)}{\partial x_1} \quad \frac{\partial \nabla f(x)}{\partial x_2} \quad \cdots \quad \frac{\partial \nabla f(x)}{\partial x_n} \right) = \left(\frac{\partial \nabla(A \vec{x} + \vec{b})}{\partial x_1} \quad \frac{\partial \nabla(A \vec{x} + \vec{b})}{\partial x_2} \quad \cdots \quad \frac{\partial \nabla(A \vec{x} + \vec{b})}{\partial x_n} \right)$$

$$= \begin{pmatrix} A_{11} & \cdots & A_{1n} \\ \vdots & \ddots & \vdots \\ A_{n1} & \cdots & A_{nn} \end{pmatrix} = A$$

(d)

When $f(x) = g(\vec{a}^T \vec{x})$, we can derive that

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

To derive the second-order partial derivatives, let $f(x) = g(h(x))$ first

Thus, we can derive that

$$\begin{aligned} \nabla^2 f(x) &= \nabla^2 g(h(x)) = \frac{\partial^2 g(h(x))}{\partial x_i \partial x_j} \\ &= \frac{\partial^2 g(h(x))}{\partial^2 (h(x))^2} \frac{\partial h(x)}{\partial x_i} \frac{\partial h(x)}{\partial x_j} \\ &= g''(h(x)) \frac{\partial h(x)}{\partial x_i} \frac{\partial h(x)}{\partial x_j} \end{aligned}$$

Thus, in this case, we can derive that

$$\nabla^2 f(x) = g''(\vec{a}^T \vec{x}) \frac{\partial(\vec{a}^T \vec{x})}{\partial x_i} \frac{\partial(\vec{a}^T \vec{x})}{\partial x_j} = g''(\vec{a}^T \vec{x}) a_i a_j \quad (1 \leq i \leq n, 1 \leq j \leq n)$$

To span all the available values for i and j , which is the matrix including all the combinations of $a_i a_j$, thus

$$\nabla^2 f(x) = g''(\vec{a}^T \vec{x}) \begin{pmatrix} a_1 a_1 & a_1 a_2 & \cdots & a_n a_1 \\ a_2 a_1 & a_2 a_2 & \cdots & a_n a_2 \\ \vdots & \vdots & \cdots & \vdots \\ a_n a_1 & a_n a_2 & \cdots & a_n a_n \end{pmatrix} = g''(\vec{a}^T \vec{x}) \vec{a} \vec{a}^T$$

Problem 2

(a) *Proof*

Let $\vec{z} = \begin{pmatrix} z_1 \\ z_2 \\ \vdots \\ z_n \end{pmatrix}$, thus we can obtain that $\vec{z}^T = (z_1 \ z_2 \ \cdots \ z_n)$

Thus, we can derive that

$$\vec{z}\vec{z}^T = \begin{pmatrix} z_1 z_1 & z_1 z_2 & \cdots & z_n z_1 \\ z_2 z_1 & z_2 z_2 & \cdots & z_n z_2 \\ \vdots & \vdots & \cdots & \vdots \\ z_n z_1 & z_n z_2 & \cdots & z_n z_n \end{pmatrix}$$

Since $(\vec{z}\vec{z}^T)^T = (\vec{z}^T)^T \vec{z} = \vec{z}\vec{z}^T$, and for $\vec{x} \in \mathbb{R}^n$,

we have $\vec{x}^T A \vec{x} = \vec{x}^T \vec{z}\vec{z}^T \vec{x} = (\vec{x}^T \vec{z})^2 \geq 0$ (since $\vec{x}^T \vec{z} = \vec{z}^T \vec{x}$)

Thus, $A = \vec{z}\vec{z}^T$ is positive semidefinite.

(b)

Since the vector \vec{z} is a n-by-1 matrix, and \vec{x}^T is a 1-by-n matrix, we can find that

$\vec{x}^T \vec{z}$ is a real number. (1-by-1 matrix)

Thus, the null-space of A is $\text{Nul}(A) = \{x \in \mathbb{R}^n : \vec{x}^T \vec{z} = 0\}$

(c)

BAB^T is positive semidefinite, here is the proof.

Since $(BAB^T)^T = (B^T)^T (BA)^T = BA^T B^T$, and $A^T = A$,

$$(BAB^T)^T = BAB^T$$

For $\vec{x} \in \mathbb{R}^{n \times n}$, we have $\vec{x}^T BAB^T \vec{x} = (\vec{x}^T B) A (\vec{x}^T B)^T \geq 0$

Thus, BAB^T is positive semidefinite.

Problem 3

(a) *Proof*

Since we have $A = T\Lambda T^{-1}$, we can obtain that $AT = T\Lambda T^{-1}T = T\Lambda$.

$$\text{Also, we have } T = (\vec{t}^{(1)} \ \dots \ \vec{t}^{(n)}), \text{ and } \Lambda = \begin{pmatrix} \lambda_1 & 0 & 0 & \dots & 0 \\ 0 & \lambda_2 & 0 & \dots & 0 \\ 0 & 0 & \lambda_3 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \lambda_n \end{pmatrix},$$

we can derive that

$$T\Lambda = (\vec{t}^{(1)} \ \dots \ \vec{t}^{(n)}) \begin{pmatrix} \lambda_1 & 0 & 0 & \dots & 0 \\ 0 & \lambda_2 & 0 & \dots & 0 \\ 0 & 0 & \lambda_3 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \lambda_n \end{pmatrix} = (\lambda_1 \vec{t}^{(1)} \ \lambda_2 \vec{t}^{(2)} \ \dots \ \lambda_n \vec{t}^{(n)})$$

Thus, we can know that $At^{(i)} = \lambda_i t^{(i)}$, and the proof completes.

(b) *Proof*

Omitted, since it is identical to the first one.

(c) *Proof*

Since we have known that for an orthogonal matrix $U = (\vec{u}^{(1)} \ \dots \ \vec{u}^{(n)})$, we can

obtain that $\vec{u}^{(i)}$ is an eigenvector for a matrix $A = U\Lambda U^T$. For the eigenvector, we can

know that $\|\vec{u}^{(i)}\| = 1$. In terms of the conclusion $A\vec{u}^{(i)} = \lambda_i \vec{u}^{(i)}$, we can obtain that

$$(\vec{u}^{(i)})^T A \vec{u}^{(i)} = (\vec{u}^{(i)})^T \lambda_i \vec{u}^{(i)} = \lambda_i \|\vec{u}^{(i)}\|^2 = \lambda_i$$

Since A is positive semi-definite, thus, for $\vec{u}^{(i)} \in \mathbb{R}^n$, we have $(\vec{u}^{(i)})^T A \vec{u}^{(i)} \geq 0$.

Thus, we can conclude that $\lambda_i(A) \geq 0$.