5004 Homework 2

RONG Shuo

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Question 1:

1. Let $T: \mathbb{R}^3 \to \mathbb{R}^2$ be a linear transformation. If T satisfies

$$T\begin{bmatrix}1\\0\\-1\end{bmatrix}=\begin{bmatrix}2\\3\end{bmatrix}$$
, and $T\begin{bmatrix}2\\1\\3\end{bmatrix}=\begin{bmatrix}-1\\0\end{bmatrix}$,

then find

 $T \begin{bmatrix} 8 \\ 3 \\ 7 \end{bmatrix}$

.

Answer

We know that

$$2\begin{bmatrix} 1\\0\\-1 \end{bmatrix} + 3\begin{bmatrix} 2\\1\\3 \end{bmatrix} = \begin{bmatrix} 2\\0\\-2 \end{bmatrix} + \begin{bmatrix} 6\\3\\9 \end{bmatrix} = \begin{bmatrix} 8\\1\\7 \end{bmatrix}$$

So in conclusion,

$$2T \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} + 3T \begin{bmatrix} 2 \\ 1 \\ 3 \end{bmatrix} = 2 \begin{bmatrix} 2 \\ 3 \end{bmatrix} + 3 \begin{bmatrix} -1 \\ 0 \end{bmatrix} = \begin{bmatrix} 4-3 \\ 6+0 \end{bmatrix} = \begin{bmatrix} 1 \\ 6 \end{bmatrix}$$



Question 2:

- 2. Find the Jacobian matrix of the following vector-valued multi-variable functions.
- (a) $f: \mathbb{R}^n \to \mathbb{R}^m$ is defined by $f(\boldsymbol{x}) = A\boldsymbol{x} \boldsymbol{b}$, where $\boldsymbol{x} \in \mathbb{R}^n$, $\boldsymbol{A} \in \mathbb{R}^{m \times n}$, $\boldsymbol{b} \in \mathbb{R}^n$.
- (b) $f: \mathbb{R}^n \to \mathbb{R}^n$ is defined by $f(\boldsymbol{x}) = \boldsymbol{x} \boldsymbol{x}^T \boldsymbol{a}$, where $\boldsymbol{x} \in \mathbb{R}^n$, $\boldsymbol{a} \in \mathbb{R}^n$.

Answer

Assume that:

$$\mathbf{A}_i = [A_{ij}]_{j=1}^n, \mathbf{A}_i \in \mathbb{R}^n$$
$$b_i = \mathbf{b}_i, b_i \in \mathbb{R}$$
$$f_i(\mathbf{x}) = \langle \mathbf{A}_i, \mathbf{x} \rangle - b_i$$

We know that:

$$f(oldsymbol{x}) = oldsymbol{A}oldsymbol{x} - oldsymbol{b}$$
 $f(oldsymbol{x}) = egin{bmatrix} f_1(oldsymbol{x}) & & & \\ f_1(oldsymbol{x}) & & & & \\ \vdots & & & & \\ \nabla f_1(oldsymbol{x})^T & & & \\ \nabla f_n(oldsymbol{x})^T \end{bmatrix}$ $Df(oldsymbol{x}) = egin{bmatrix} oldsymbol{A}_1^T & & & & \\ \nabla f_n(oldsymbol{x})^T \end{bmatrix}$ $Df(oldsymbol{x}) = egin{bmatrix} oldsymbol{A}_1^T & & & & \\ oldsymbol{A}_1^T & & & & \\ oldsymbol{A}_n^T & & & & \\ \end{array}$

We know that Jacobian matrix is the differentiation of $f: \mathbb{R}^n \to \mathbb{R}^m$. So In conclusion, the Jacobian matrix is A.

(b) Assume that:

$$m{X} = m{x}m{x}^T = [x_ix_j]_{i=1,j=1}^{n,n}, m{X} \in \mathbb{R}^{n \times n}$$
 $m{X}_i = [x_ix_j]_{j=1}^n, m{X}_i \in \mathbb{R}^n$
 $f_i(m{x}) = \langle m{X}_i, m{a} \rangle$

if i = k

$$\frac{\partial f_i}{x_k} = \sum_{j=1}^n x_j a_j + x_k a_k$$

if $i \neq k$

$$\frac{\partial f_i}{\partial x_k} = x_i a_k$$

$$f(\boldsymbol{x}) = \begin{bmatrix} f_1(\boldsymbol{x}) \\ \vdots \\ f_n(\boldsymbol{x}) \end{bmatrix}$$

$$Df(\boldsymbol{x}) = \begin{bmatrix} \nabla f_1(\boldsymbol{x})^T \\ \vdots \\ \nabla f_n(\boldsymbol{x})^T \end{bmatrix} = \begin{bmatrix} x_1 a_1 & x_1 a_2 & \cdots & x_1 a_n \\ \vdots & \vdots & \cdots & \vdots \\ x_n a_1 & x_n a_2 & \cdots & x_n a_n \end{bmatrix} + \begin{bmatrix} \sum_{j=1}^n x_j a_j & 0 & \cdots & 0 \\ \vdots & \vdots & \cdots & \vdots \\ 0 & 0 & \cdots & \sum_{j=1}^n x_j a_j \end{bmatrix} = \boldsymbol{x} \boldsymbol{a}^T + (\boldsymbol{x}^T \boldsymbol{a}) \boldsymbol{I}$$

Question 3:

3. Let $f: \mathbb{R}^2 \to \mathbb{R}$, $g: \mathbb{R}^2 \to \mathbb{R}^2$, $g(x,y) = (x^2y, x-y)$ and $h = f \circ g = f(g(x,y))$. Find $\frac{\partial h}{\partial x}|_{x=1,y=2}$ if $\frac{\partial f}{\partial x}|_{x=2,y=-1} = 3$ and $\frac{\partial f}{\partial y}|_{x=2,y=-1} = -2$. (Hint: use the chain rule)

Answer:

We know that:

$$g_{1}(x,y) = x^{2}y$$

$$g_{2}(x,y) = x - y$$

$$h(x,y) = f(g_{1}(x,y), g_{2}(x,y))$$

$$\frac{\partial h}{\partial x}|_{x=1,y=2} = \frac{\partial f}{\partial x}|_{x=2,y=-1} \frac{\partial g_{1}}{\partial x}|_{x=1,y=2} + \frac{\partial f}{\partial y}|_{x=2,y=-1} \frac{\partial g_{2}}{\partial x}|_{x=1,y=2}$$

$$\frac{\partial h}{\partial x}|_{x=1,y=2} = 3\frac{\partial g_{1}}{\partial x} - 2\frac{\partial g_{2}}{\partial x} = 3*(2xy)|_{x=1,y=2} - 2*1 = 10$$

Question 4:

Let $f(t) = f_1(t) * f_2(t)$ be the convolution of two functions $f_1(t)$ and $f_2(t)$ on \mathbb{R} , i.e.,

$$f(t) = \int_{-\infty}^{+\infty} f_1(t-s) f_2(s) ds$$

Let a, a_1, a_2 be real number.

(i) Prove the following identity:

$$f_1(t-a) * f_2(t) = f_1(t) * f_2(t-a) = f(t-a).$$

(ii) Prove the following identity:

$$f_1(t - a_1) * f_2(t - a_2) = f(t - a_1 - a_2).$$

Answer:

(i) Prove $f_1(t-a) * f_2(t) = f(t-a)$:

$$f(t-a) = \int_{-\infty}^{+\infty} f_1(t-s-a) f_2(s) ds$$
Assume $f_1'(t) = f_1(t-a)$

$$f(t-a) = \int_{-\infty}^{+\infty} f_1'(t-s) f_2(s) ds = f_1'(t) * f_2(t) = f_1(t-a) * f_2(t)$$

Prove $f_1(t) * f_2(t - a) = f(t - a)$:

$$f(t-a) = \int_{-\infty}^{+\infty} f_1(t-s-a)f_2(s)ds$$

Assume $f_2'(t) = f_2(t-a)$

$$f(t-a) = \int_{-\infty}^{+\infty} f_1(t-(s+a))f_2(s)d(s+a) = \int_{-\infty}^{+\infty} f_1(t-(s+a))f_2'(s+a)d(s+a)$$

$$= \int_{-\infty}^{+\infty} f_1(t-u)f_2'(u)du$$

$$= f_1(t) * f_2'(t)$$

$$= f_1(t) * f_2(t-a)$$

(ii)

$$f_1(t - a_1) * f_2(t - a_2) = \int_{-\infty}^{+\infty} f_1(t - a_1 - s) f_2(s - a_2) ds$$

$$= \int_{-\infty}^{+\infty} f_1(t - a_1 - s) f_2(s - a_2) d(s - a_2)$$

$$= \int_{-\infty}^{+\infty} f_1(t - a_1 - a_2 - (s - a_2)) f_2(s - a_2) d(s - a_2)$$

$$= \int_{-\infty}^{+\infty} f_1(t - (a_1 + a_2) - u) f_2(u) du$$

$$= f(t - a_1 - a_2).$$

Question 5:

5. Let V_1 and V_2 be two Hilbert spaces with the inner products $\langle \cdot, \cdot \rangle_{V_1}$, and $\langle \cdot, \cdot \rangle_{V_2}$, respectively. Let $T \in \mathcal{L}(V_1, V_2)$, i.e., $T: V_1 \to V_2$ be a bounded linear operator.

(a) Let $S: V_2 \to V_1$ be an operator satisfying $\langle T\boldsymbol{x}, \boldsymbol{y} \rangle_{V_2} = \langle \boldsymbol{x}, S\boldsymbol{y} \rangle_{V_1}$ for any $\boldsymbol{x} \in V_1$ and $\boldsymbol{y} \in V_2$. Prove that S is a bounded linear operator. (Consequently, S is the adjoint of T, i.e., $S = T^*$)

- (b) Prove that $(T^*)^* = T$.
- (c) Prove that $||T|| = ||T^*||$.

Answer

(a)We know

$$\|\mathbf{A}\mathbf{x}\|_{V_2} \le \|\mathbf{A}\| \|\mathbf{x}\|_{V_1} \tag{1}$$

We know, by using Cauchy–Schwarz inequality and (1):

$$\langle Tx, y \rangle_{V_2} \leq \|Tx\|_{V_2} \|y\|_{V_2} \leq \|T\| \|x\|_{V_1} \|y\|_{V_2}$$

We choose \boldsymbol{x} as $\boldsymbol{S}\boldsymbol{y}$

$$egin{aligned} \langle m{T}m{S}m{y},m{y}
angle_{V_2} &= \langle m{S}m{y},m{S}m{y}
angle_{V_1} = \|m{S}m{y}\|_{V_1}^2 \leq \|m{T}m{S}m{y}\|_{V_2} \|m{y}\|_{V_2} \leq \|m{T}\|\|m{S}m{y}\|_{V_1}\|m{y}\|_{V_2} \ \|m{S}m{y}\|_{V_1} &\leq \|m{T}\|\|m{y}\|_{V_2} \leq \infty \end{aligned}$$

Thus, S is a bounded linear operator.

(b) We know:

$$\langle oldsymbol{T}oldsymbol{x},oldsymbol{y}
angle_{V_2}=\langle oldsymbol{x},oldsymbol{T}^*oldsymbol{y}
angle_{V_1}$$

$$egin{aligned} \langle oldsymbol{x}, oldsymbol{T}^* oldsymbol{y}
angle_{V_1} &= \langle (oldsymbol{T}^*)^* oldsymbol{x}, oldsymbol{y}
angle_{V_2} \ \langle oldsymbol{T} oldsymbol{x}, oldsymbol{y}
angle_{V_2} &= \langle (oldsymbol{T}^*)^* oldsymbol{x}, oldsymbol{y}
angle_{V_2} \end{aligned}$$

Therefore

$$\boldsymbol{T} = (\boldsymbol{T}^*)^*$$

(c) Proof of

$$\|oldsymbol{T}\| = \sup_{\|oldsymbol{y}\|_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{V}}}} \|oldsymbol{T}oldsymbol{y}\|_{oldsymbol{oldsymbol{y}}} \sup_{\|oldsymbol{y}\|=1} \langle oldsymbol{x}, oldsymbol{T}oldsymbol{y}
angle_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{Y}}_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{y}}}} \|oldsymbol{T}oldsymbol{y}\|_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{V}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{Y}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{oldsymbol{W}}_{oldsymbol{W}}_{oldsymbol{W}}_{oldsymbol{W}}_{oldsymbol{W}}_{oldsymbol{W}}_{oldsymbol{W}}_{oldsymbol{W}}_{oldsymbol{W}}_{oldsymbol{W}_{oldsymbol{W}_{oldsymbol{W}}_{oldsymbol{W}}_{oldsymbol{W}_{oldsymbol{W}}_{oldsymbol{W}_{oldsymbol{W}_{oldsymbol{W}_{oldsymbol{W}}}_{oldsymbol{W}_{oldsymbol{$$

By using Cauchy Schwarz inequality

$$\langle x, Ty \rangle \leq \|x\| \|Ty\|_{V_{\lambda}}$$

Assume $\|\boldsymbol{x}\| = 1$, s.t.

Therefore

$$\|\boldsymbol{T}\| = \sup_{\|\boldsymbol{y}\|=1} \|\boldsymbol{T}\boldsymbol{y}\|_{\text{t}} = \sup_{\|\boldsymbol{y}\|=1} \sup_{\|\boldsymbol{x}\|=1} \langle \boldsymbol{x}, \boldsymbol{T}\boldsymbol{y} \rangle_{\text{t}}$$

Then

$$\begin{aligned} \|\boldsymbol{T}^*\| &= \sup_{\|\boldsymbol{y}\|_{\stackrel{\sim}{\boldsymbol{V}_{\boldsymbol{\lambda}}}}} \|\boldsymbol{T}^*\boldsymbol{y}\|_{\stackrel{\sim}{\boldsymbol{V}_{\boldsymbol{\lambda}}}} \sup_{\|\boldsymbol{y}\|=1} \sup_{\|\boldsymbol{x}\|_{\stackrel{\sim}{\boldsymbol{V}_{\boldsymbol{\lambda}}}}} \langle \boldsymbol{x}, \boldsymbol{T}^*\boldsymbol{y} \rangle_{\stackrel{\sim}{\boldsymbol{V}_{\boldsymbol{\lambda}}}} \\ &= \sup_{\|\boldsymbol{y}\|=1} \sup_{\|\boldsymbol{x}\|_{\stackrel{\sim}{\boldsymbol{V}_{\boldsymbol{\lambda}}}}} \langle \boldsymbol{T}\boldsymbol{x}, \boldsymbol{y} \rangle_{\stackrel{\sim}{\boldsymbol{V}_{\boldsymbol{\lambda}}}} \sup_{\|\boldsymbol{x}\|=1} \|\boldsymbol{T}\boldsymbol{x}\|_{\stackrel{\sim}{\boldsymbol{V}_{\boldsymbol{\lambda}}}} \|\boldsymbol{T}\| \\ &= \sup_{\|\boldsymbol{y}\|=1} \|\boldsymbol{x}\|_{\stackrel{\sim}{\boldsymbol{V}_{\boldsymbol{\lambda}}}} \|\boldsymbol{x}\|_{\stackrel{\sim}{\boldsymbol{V}_{\boldsymbol{\lambda}}}} \|\boldsymbol{T}\|_{\stackrel{\sim}{\boldsymbol{V}_{\boldsymbol{\lambda}}}} \|\boldsymbol{T}\|_{\stackrel{\sim}{\boldsymbol{V}_{\boldsymbol{\lambda}$$

In conclusion, $\|T^*\| = \|T\|$

Question 6:

Consider the vector space ℓ_{∞} equipped with the norm $\|\cdot\|_{\infty}$. Define the operator $T:\ell_{\infty}\to\ell_{\infty}$ by $T(\{x_n\}_{n\in\mathbb{N}})=\{y_n\}_{n\in\mathbb{N}}$ where $y_n=x_{n+1}$.

- (a) Prove that T is a linear operator.
- (b) Prove that T is a bounded operator.
- (c) Prove that ||T|| = 1.

Answer

(a) Additivity: Proof:

For any $\{x_n\}_{n\in\mathbb{N}}, \{z_n\}_{n\in\mathbb{N}}\in\ell_{\infty}$

$$T(\{x_n\}_{n\in\mathbb{N}} + \{z_n\}_{n\in\mathbb{N}}) = T(\{x_n + z_n\}_{n\in\mathbb{N}}) = \{(x_n + z_n)_{n+1}\}_{n\in\mathbb{N}} = \{x_{n+1} + z_{n+1}\}_{n\in\mathbb{N}} = \{x_{n+1}\}_{n\in\mathbb{N}} + \{z_{n+1}\}_{n\in\mathbb{N}} = T(\{x_n\}_{n\in\mathbb{N}}) + T(\{z_n\}_{n\in\mathbb{N}})$$

Homogeneity Proof:

$$T(\alpha \{x_n\}_{n \in \mathbb{N}}) = T(\{\alpha x_n\}_{n \in \mathbb{N}}) = \{\alpha x_{n+1}\}_{n \in \mathbb{N}} = \alpha \{x_{n+1}\}_{n \in \mathbb{N}} = \alpha T(\{x_n\}_{n \in \mathbb{N}})$$

(b) To prove T is bounded operator, we can prove, for a constance C, that :

$$||T(\{x_n\}_{n\in\mathbb{N}})||_{\infty} \le C||\{x_n\}_{n\in\mathbb{N}}||_{\infty}, \forall \{x_n\}_{n\in\mathbb{N}} \in \ell_{\infty}$$

$$||T(\{x_n\}_{n\in\mathbb{N}})||_{\infty} = \sup_{n\in\mathbb{N}} |y_n| = \sup_{n\in\mathbb{N}} |x_{n+1}| \le \sup_{n\in\mathbb{N}} |x_n| = ||\{x_n\}_{n\in\mathbb{N}}||_{\infty}$$

$$||T(\{x_n\})||_{\infty} \le ||\{x_n\}||_{\infty}$$

In conclusion, T is a bounded operator.

(c) From (b), we know that $||T(\{x_n\})||_{\infty} \leq ||\{x_n\}||_{\infty}$, s.t. $||T|| \leq 1$. Consider the sequence $\{x_n\}_{n\in\mathbb{N}} = \{1,1,\cdots\} = \{x_{n+1}\}_{n\in\mathbb{N}}$ By definition, we know:

$$||T|| \ge \frac{||T(\{x_n\})||_{\infty}}{||\{x_n\}||_{\infty}} = 1$$
$$1 \le ||T|| \le 1$$

Thus, we conclude that ||T|| = 1