

Matrix Algebra Marathon

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1 Derivatives of Function of Matrices

1.1 Definition

Definition 1.1.1. We denote the derivative of a function $f : \mathbb{R}^{m \times n} \rightarrow \mathbb{R}$ by

$$\nabla_{\mathbf{X}} f(\mathbf{X}) := \begin{bmatrix} \nabla_{X_{1,1}}(f(\mathbf{X})) & \cdots & \nabla_{X_{1,n}}(f(\mathbf{X})) \\ \vdots & \ddots & \vdots \\ \nabla_{X_{m,1}}(f(\mathbf{X})) & \cdots & \nabla_{X_{m,n}}(f(\mathbf{X})) \end{bmatrix}.$$

1.2 Exercise

Exercise 2.14.2.

Let $\mathbf{X}, \mathbf{A} \in \mathbb{R}^{m \times n}$. Show that

$$\nabla_{\mathbf{X}}(\text{tr}(\mathbf{A}^T \mathbf{X})) = \mathbf{A}. \quad (1)$$

Proof. Let $\mathbf{X}, \mathbf{A} \in \mathbb{R}^{m \times n}$. Recall the definition of inner-product of matrices and

$$\langle \mathbf{X}, \mathbf{Y} \rangle = \text{tr}(\mathbf{X}^T \mathbf{Y}).$$

$$\begin{aligned} \nabla_{\mathbf{X}}(\text{tr}(\mathbf{A}^T \mathbf{X})) &= \begin{bmatrix} \nabla_{X_{1,1}} \text{tr}(\mathbf{A}^T \mathbf{X}) & \cdots & \nabla_{X_{1,n}} \text{tr}(\mathbf{A}^T \mathbf{X}) \\ \vdots & \ddots & \vdots \\ \nabla_{X_{m,1}} \text{tr}(\mathbf{A}^T \mathbf{X}) & \cdots & \nabla_{X_{m,n}} \text{tr}(\mathbf{A}^T \mathbf{X}) \end{bmatrix} \\ &= \begin{bmatrix} \nabla_{X_{1,1}} \left(\sum_{i=1}^m \sum_{j=1}^n A_{i,j} X_{i,j} \right) & \cdots & \nabla_{X_{1,n}} \left(\sum_{i=1}^m \sum_{j=1}^n A_{i,j} X_{i,j} \right) \\ \vdots & \ddots & \vdots \\ \nabla_{X_{m,1}} \left(\sum_{i=1}^m \sum_{j=1}^n A_{i,j} X_{i,j} \right) & \cdots & \nabla_{X_{m,n}} \left(\sum_{i=1}^m \sum_{j=1}^n A_{i,j} X_{i,j} \right) \end{bmatrix} \\ &= \begin{bmatrix} A_{1,1} & \cdots & A_{1,n} \\ \vdots & \ddots & \vdots \\ A_{m,1} & \cdots & A_{m,n} \end{bmatrix} \\ &= \mathbf{A} \end{aligned}$$

□

Exercise 2.14.4.

Let $\mathbf{X}, \mathbf{A}, \mathbf{B} \in \mathbb{R}^{m \times n}$. Show that

$$\nabla_{\mathbf{X}}(\text{tr}(\mathbf{A}^T(\mathbf{X} + \mathbf{B}))) = \mathbf{A}. \quad (2)$$

Proof. Let $\mathbf{X}, \mathbf{A}, \mathbf{B} \in \mathbb{R}^{m \times n}$.

$$\begin{aligned}
& \nabla_{\mathbf{X}}(\text{tr}(\mathbf{A}^T(\mathbf{X} + \mathbf{B}))) \\
&= \begin{bmatrix} \nabla_{X_{1,1}}(\text{tr}(\mathbf{A}^T(\mathbf{X} + \mathbf{B}))) & \cdots & \nabla_{X_{1,n}}(\text{tr}(\mathbf{A}^T(\mathbf{X} + \mathbf{B}))) \\ \vdots & \ddots & \vdots \\ \nabla_{X_{m,1}}(\text{tr}(\mathbf{A}^T(\mathbf{X} + \mathbf{B}))) & \cdots & \nabla_{X_{m,n}}(\text{tr}(\mathbf{A}^T(\mathbf{X} + \mathbf{B}))) \end{bmatrix} \\
&= \begin{bmatrix} \nabla_{X_{1,1}}\left(\sum_{i=1}^m \sum_{j=1}^n A_{i,j}(X_{i,j} + B_{i,j})\right) & \cdots & \nabla_{X_{1,n}}\left(\sum_{i=1}^m \sum_{j=1}^n A_{i,j}(X_{i,j} + B_{i,j})\right) \\ \vdots & \ddots & \vdots \\ \nabla_{X_{m,1}}\left(\sum_{i=1}^m \sum_{j=1}^n A_{i,j}(X_{i,j} + B_{i,j})\right) & \cdots & \nabla_{X_{m,n}}\left(\sum_{i=1}^m \sum_{j=1}^n A_{i,j}(X_{i,j} + B_{i,j})\right) \end{bmatrix} \\
&= \begin{bmatrix} A_{1,1} & \cdots & A_{1,n} \\ \vdots & \ddots & \vdots \\ A_{m,1} & \cdots & A_{m,n} \end{bmatrix} \\
&= \mathbf{A}
\end{aligned}$$

□

2 Derivatives of Matrix-Valued Functions

2.1 Definition

Definition 2.1.1. Consider a matrix-valued function $\mathbf{F} : \mathbb{R} \rightarrow \mathbb{R}^{m \times n}$ expressed as

$$\mathbf{F}(x) := \begin{bmatrix} F_{1,1}(x) & \cdots & F_{1,n}(x) \\ \vdots & \ddots & \vdots \\ F_{m,1}(x) & \cdots & F_{m,n}(x) \end{bmatrix}.$$

The derivative of the function is denoted by

$$\nabla_x \mathbf{F}(x) := \begin{bmatrix} \nabla_x F_{1,1}(x) & \cdots & \nabla_x F_{1,n}(x) \\ \vdots & \ddots & \vdots \\ \nabla_x F_{m,1}(x) & \cdots & \nabla_x F_{m,n}(x) \end{bmatrix}.$$

2.2 Exercise

Exercise 2.15.3.

Let $\mathbf{A} \in \mathbb{R}^{m \times n}$ and $\mathbf{F} : \mathbb{R} \rightarrow \mathbb{R}^{m \times n}$. Show that

$$\nabla_x \langle \mathbf{A}, \mathbf{F}(x) \rangle = \langle \mathbf{A}, \nabla_x \mathbf{F}(x) \rangle. \quad (3)$$

Proof. Let $\mathbf{A} \in \mathbb{R}^{m \times n}$ and $\mathbf{F} : \mathbb{R} \rightarrow \mathbb{R}^{m \times n}$.

$$\begin{aligned} \nabla_x \langle \mathbf{A}, \mathbf{F}(x) \rangle &= \nabla_x \left(\sum_{i=1}^m \sum_{j=1}^n A_{i,j} F_{i,j}(x) \right) \\ &= \left(\sum_{i=1}^m \sum_{j=1}^n A_{i,j} \nabla_x F_{i,j}(x) \right) \\ &= \langle \mathbf{A}, \nabla_x \mathbf{F}(x) \rangle \end{aligned}$$

□

Exercise 2.15.5.

Let $\mathbf{F} : \mathbb{R} \rightarrow \mathbb{R}^{m \times k}$ and $\mathbf{G} : \mathbb{R} \rightarrow \mathbb{R}^{k \times n}$. Show that

$$\nabla_x (\mathbf{F}(x) \mathbf{G}(x)) = (\nabla_x \mathbf{F}(x)) \mathbf{G}(x) + \mathbf{F}(x) \nabla_x \mathbf{G}(x). \quad (4)$$

Proof. Let $\mathbf{F} : \mathbb{R} \rightarrow \mathbb{R}^{m \times k}$ and $\mathbf{G} : \mathbb{R} \rightarrow \mathbb{R}^{k \times n}$. For all $i \in \mathbb{N}_m$ and $j \in \mathbb{N}_n$,

$$\begin{aligned} (\nabla_x (\mathbf{F}(x) \mathbf{G}(x)))_{i,j} &= \nabla_x (\mathbf{F}(x) \mathbf{G}(x))_{i,j} \\ &= \nabla_x \left(\sum_{l=1}^k F_{i,l}(x) G_{l,j}(x) \right) \\ &= \sum_{l=1}^k ((\nabla_x F_{i,l}(x)) G_{l,j}(x) + F_{i,l}(x) (\nabla_x G_{l,j}(x))) \\ &= \sum_{l=1}^k (\nabla_x F_{i,l}(x)) G_{l,j}(x) + \sum_{l=1}^k F_{i,l}(x) \nabla_x G_{l,j}(x) \\ &= \sum_{l=1}^k (\nabla_x \mathbf{F}(x))_{i,l} G_{l,j}(x) + \sum_{l=1}^k F_{i,l}(x) (\nabla_x \mathbf{G}(x))_{l,j} \\ &= ((\nabla_x \mathbf{F}(x)) \mathbf{G}(x))_{i,j} + (\mathbf{F}(x) \nabla_x \mathbf{G}(x))_{i,j} \\ &= ((\nabla_x \mathbf{F}(x)) \mathbf{G}(x) + \mathbf{F}(x) \nabla_x \mathbf{G}(x))_{i,j}. \end{aligned}$$

Therefore, (4) holds.

□

3 Diagonal Matrices

3.1 Definition

Definition 3.1.1. A diagonal matrix is a square matrix whose all the off-diagonal entries are zero. Such a matrix with diagonal entries $\mathbf{d} := [d_1, \dots, d_n]^T \in \mathbb{R}^n$ is denoted by

$$\text{diag}(\mathbf{d}) := \begin{bmatrix} d_1 & 0 & \cdots & 0 \\ 0 & d_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & d_n \end{bmatrix}$$

3.2 Exercise

Exercise 2.16.3.

Let $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$. Show that

$$\text{diag}(\mathbf{x})\text{diag}(\mathbf{y}) = \text{diag}(\mathbf{y})\text{diag}(\mathbf{x}). \quad (5)$$

Proof. Let $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$.

$$\begin{aligned} \text{diag}(\mathbf{x})\text{diag}(\mathbf{y}) &= \begin{bmatrix} x_1 & & \mathbf{O} \\ & \ddots & \\ \mathbf{O} & & x_n \end{bmatrix} \begin{bmatrix} y_1 & & \mathbf{O} \\ & \ddots & \\ \mathbf{O} & & y_n \end{bmatrix} \\ &= \begin{bmatrix} x_1 y_1 & & \mathbf{O} \\ & \ddots & \\ \mathbf{O} & & x_n y_n \end{bmatrix} \\ &= \begin{bmatrix} y_1 x_1 & & \mathbf{O} \\ & \ddots & \\ \mathbf{O} & & y_n x_n \end{bmatrix} \\ &= \begin{bmatrix} y_1 & & \mathbf{O} \\ & \ddots & \\ \mathbf{O} & & y_n \end{bmatrix} \begin{bmatrix} x_1 & & \mathbf{O} \\ & \ddots & \\ \mathbf{O} & & x_n \end{bmatrix} \\ &= \text{diag}(\mathbf{y})\text{diag}(\mathbf{x}) \end{aligned}$$

□

4 Orthonormal Matrices

4.1 Definition

Definition 4.1.1. An $m \times n$ matrix \mathbf{P} is said to be orthonormal if the matrix satisfies $\mathbf{P}^T \mathbf{P} = \mathbf{I}_n$. Symbol $\mathbb{O}^{m \times n}$ is used to denote the set of $m \times n$ orthonormal matrices.

4.2 Exercise

Exercise 2.17.2.

Let $\mathbf{P} = [\mathbf{p}_1, \dots, \mathbf{p}_n] \in \mathbb{O}^{m \times n}$. Show that $\forall i, \forall j \in \mathbb{N}_n$,

$$\langle \mathbf{p}_i, \mathbf{p}_j \rangle = \delta_{i,j} \quad (6)$$

where $\delta_{i,j}$ is the Kronecker delta.

Proof. Let $\mathbf{P} = [\mathbf{p}_1, \dots, \mathbf{p}_n] \in \mathbb{O}^{m \times n}$ and $i, j \in \mathbb{N}_n$.

$$\begin{aligned} (\mathbf{P}^T \mathbf{P})_{i,j} &= \sum_{k=1}^m (\mathbf{P}^T)_{i,k} P_{k,j} \\ &= \sum_{k=1}^m P_{k,i} P_{k,j} \\ &= \langle \mathbf{p}_i, \mathbf{p}_j \rangle \end{aligned}$$

Furthermore,

$$\begin{aligned} (\mathbf{I}_n)_{i,j} &= \begin{cases} 1 & (i = j) \\ 0 & (i \neq j) \end{cases} \\ &= \delta_{i,j} \end{aligned}$$

Thus, since $\mathbf{P}^T \mathbf{P} = \mathbf{I}_n$,

$$\langle \mathbf{p}_i, \mathbf{p}_j \rangle = (\mathbf{P}^T \mathbf{P})_{i,j} = (\mathbf{I}_n)_{i,j} = \delta_{i,j}.$$

Then, we get (6). □

Exercise 2.17.4.

Let $\mathbf{P} \in \mathbb{O}^{n \times n}$. Show that

$$\mathbf{P}^T = \mathbf{P}^{-1}. \quad (7)$$

Proof. Let $\mathbf{P} \in \mathbb{O}^{n \times n}$. Since $\mathbf{P} \in \mathbb{O}^{n \times n}$,

$$\mathbf{P}^T \mathbf{P} = \mathbf{I}_n.$$

Recall the definition of inverse of square matrices. Then, we get

$$\mathbf{P}^{-1} = \mathbf{P}^T.$$

Thus, (7) holds. □

Exercise 2.17.6. —

Let $\mathbf{P} \in \mathbb{O}^{n \times n}$. Show that

$$\det(\mathbf{P}) \in \{\pm 1\}. \quad (8)$$

Proof. Let $\mathbf{P} \in \mathbb{O}^{n \times n}$. Recall that $\det(\mathbf{AB}) = \det(\mathbf{A})\det(\mathbf{B})$ and $\det(\mathbf{A}^T) = \det(\mathbf{A})$ for $\mathbf{A}, \mathbf{B} \in \mathbb{R}^{n \times n}$. Then,

$$\det(\mathbf{P}^T \mathbf{P}) = \det(\mathbf{P}^T)\det(\mathbf{P}) = \det(\mathbf{P})\det(\mathbf{P}) = (\det(\mathbf{P}))^2.$$

Recall that

$$\det(\mathbf{I}_n) = 1.$$

Then, from the definition of orthonormal matrices,

$$(\det(\mathbf{P}))^2 = \det(\mathbf{P}^T \mathbf{P}) = \det(\mathbf{I}_n) = 1.$$

By solving this equation, we have $\det(\mathbf{P}) = \pm 1$. Hence, $\det(\mathbf{P}) \in \{\pm 1\}$. □

Exercise 2.17.8. —

Let $\mathbf{P} \in \mathbb{O}^{m \times k}$ and $\mathbf{Q} \in \mathbb{O}^{k \times n}$ where $m \geq k \geq n$. Show that $\mathbf{PQ} \in \mathbb{O}^{m \times n}$.

Proof. Let $\mathbf{P} \in \mathbb{O}^{m \times k}$ and $\mathbf{Q} \in \mathbb{O}^{k \times n}$ where $m \geq k \geq n$. Since $\mathbf{P} \in \mathbb{O}^{m \times k}$,

$$\mathbf{P}^T \mathbf{P} = \mathbf{I}_k.$$

Similarly,

$$\mathbf{Q}^T \mathbf{Q} = \mathbf{I}_n.$$

Then,

$$\begin{aligned} (\mathbf{PQ})^T \mathbf{PQ} &= \mathbf{Q}^T \mathbf{P}^T \mathbf{PQ} \\ &= \mathbf{Q}^T (\mathbf{P}^T \mathbf{P}) \mathbf{Q} \\ &= \mathbf{Q}^T \mathbf{I}_k \mathbf{Q} \\ &= \mathbf{Q}^T \mathbf{Q} \\ &= \mathbf{I}_n. \end{aligned}$$

Hence, $\mathbf{PQ} \in \mathbb{O}^{m \times n}$. □