

# 矩阵求导推导

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$$1 \quad \frac{\partial(\mathbf{x}^T \mathbf{a})}{\partial \mathbf{x}} = \frac{\partial(\mathbf{a}^T \mathbf{x})}{\partial \mathbf{x}} = \mathbf{a}$$

其中,  $\mathbf{a}$  为常数向量,  $\mathbf{a} = (a_1, a_2, \dots, a_n)^T$ 。

$$\begin{aligned} \frac{\partial(\mathbf{x}^T \mathbf{a})}{\partial \mathbf{x}} &= \frac{\partial(\mathbf{a}^T \mathbf{x})}{\partial \mathbf{x}} \\ &= \frac{\partial(a_1 x_1 + a_2 x_2 + \dots + a_n x_n)}{\partial \mathbf{x}} \\ &= \begin{bmatrix} \frac{\partial(a_1 x_1 + a_2 x_2 + \dots + a_n x_n)}{\partial x_1} \\ \frac{\partial(a_1 x_1 + a_2 x_2 + \dots + a_n x_n)}{\partial x_2} \\ \vdots \\ \frac{\partial(a_1 x_1 + a_2 x_2 + \dots + a_n x_n)}{\partial x_n} \end{bmatrix} \\ &= \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix} \\ &= \mathbf{a} \end{aligned} \tag{1}$$

$$\mathbf{2} \quad \frac{\partial(\mathbf{x}^T \mathbf{x})}{\partial \mathbf{x}} = 2\mathbf{x}$$

$$\frac{\partial(\mathbf{x}^T \mathbf{x})}{\partial \mathbf{x}} = \frac{\partial(x_1^2 + x_2^2 + \cdots + x_n^2)}{\partial \mathbf{x}}$$

$$= \begin{bmatrix} \frac{\partial(x_1^2 + x_2^2 + \cdots + x_n^2)}{\partial x_1} \\ \frac{\partial(x_1^2 + x_2^2 + \cdots + x_n^2)}{\partial x_2} \\ \vdots \\ \frac{\partial(x_1^2 + x_2^2 + \cdots + x_n^2)}{\partial x_n} \end{bmatrix}$$

$$= \begin{bmatrix} 2x_1 \\ 2x_2 \\ \vdots \\ 2x_n \end{bmatrix} \tag{2}$$

$$= 2 \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

$$= 2\mathbf{x}$$

$$3 \quad \frac{\partial(\mathbf{x}^T \mathbf{A} \mathbf{x})}{\partial \mathbf{x}} = \mathbf{A} \mathbf{x} + \mathbf{A}^T \mathbf{x}$$

其中,  $\mathbf{A}_{n \times n}$  是常数矩阵,  $\mathbf{A}_{n \times n} = (a_{ij})_{i=1, j=1}^{n, n}$ 。

$$\begin{aligned}
 & \partial(a_{11}x_1x_1 + a_{12}x_1x_2 + \cdots + a_{1n}x_1x_n \\
 & \quad + a_{21}x_2x_1 + a_{22}x_2x_2 + \cdots + a_{2n}x_2x_n \\
 \frac{\partial(\mathbf{x}^T \mathbf{A} \mathbf{x})}{\partial \mathbf{x}} &= \frac{+a_{n1}x_nx_1 + a_{n2}x_nx_2 + \cdots + a_{nn}x_nx_n}{\partial \mathbf{x}} \\
 &= \begin{bmatrix} (a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n) + (a_{11}x_1 + a_{21}x_2 + \cdots + a_{n1}x_n) \\ (a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n) + (a_{12}x_1 + a_{22}x_2 + \cdots + a_{n2}x_n) \\ \vdots \\ (a_{n1}x_1 + a_{n2}x_2 + \cdots + a_{nn}x_n) + (a_{1n}x_1 + a_{2n}x_2 + \cdots + a_{nn}x_n) \end{bmatrix} \\
 &= \begin{bmatrix} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n \\ \vdots \\ a_{n1}x_1 + a_{n2}x_2 + \cdots + a_{nn}x_n \end{bmatrix} + \begin{bmatrix} a_{11}x_1 + a_{21}x_2 + \cdots + a_{n1}x_n \\ a_{12}x_1 + a_{22}x_2 + \cdots + a_{n2}x_n \\ \vdots \\ a_{1n}x_1 + a_{2n}x_2 + \cdots + a_{nn}x_n \end{bmatrix} \quad (3) \\
 &= \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} + \begin{bmatrix} a_{11} & a_{21} & \cdots & a_{n1} \\ a_{12} & a_{22} & \cdots & a_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1n} & a_{2n} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \\
 &= \mathbf{A} \mathbf{x} + \mathbf{A}^T \mathbf{x}
 \end{aligned}$$