

Basic knowledge ①

matrix derivation

For variable is a vector



1. constant. $f(x) \cdot x = [x_1, x_2, \dots, x_n]^T$

$$\frac{\partial f}{\partial x} = \left[\frac{\partial f}{\partial x_1}, \dots, \frac{\partial f}{\partial x_n} \right]^T$$

$$\frac{\partial c}{\partial x} = 0_{n \times 1}$$

Prove:

$$\frac{\partial c}{\partial x} = \begin{bmatrix} \frac{\partial c}{\partial x_1} \\ \vdots \\ \frac{\partial c}{\partial x_n} \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix} = 0_{n \times 1}$$

2. Linear Rule.

$$\frac{\partial (c_1 f(x) + c_2 g(x))}{\partial x} = c_1 \frac{\partial f(x)}{\partial x} + c_2 \frac{\partial g(x)}{\partial x}, \text{ here } c_1, c_2 \text{ are constants.}$$

Prove:

$$\frac{\partial (c_1 f(x) + c_2 g(x))}{\partial x} = \begin{bmatrix} \frac{\partial}{\partial x_1} [c_1 f + c_2 g] \\ \vdots \\ \frac{\partial}{\partial x_n} [c_1 f + c_2 g] \end{bmatrix} = \begin{bmatrix} c_1 \frac{\partial f}{\partial x_1} + c_2 \frac{\partial g}{\partial x_1} \\ \vdots \\ c_1 \frac{\partial f}{\partial x_n} + c_2 \frac{\partial g}{\partial x_n} \end{bmatrix}$$

$$= c_1 \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{bmatrix} + c_2 \begin{bmatrix} \frac{\partial g}{\partial x_1} \\ \vdots \\ \frac{\partial g}{\partial x_n} \end{bmatrix} = c_1 \frac{\partial f(x)}{\partial x} + c_2 \frac{\partial g(x)}{\partial x}$$

11

3. Multiply Rule.

$$\frac{d[f(x)g(x)]}{dx} = \frac{df(x)}{dx} \cdot g(x) + f(x) \cdot \frac{dg(x)}{dx}$$

Prove:

$$\frac{d[f(x)g(x)]}{dx} = \begin{bmatrix} \frac{d(fg)}{dx} \\ \frac{d(f+g)}{dx_n} \end{bmatrix} = \begin{bmatrix} \frac{df}{dx} \cdot g + f \cdot \frac{dg}{dx} \\ \frac{df}{dx_n} \cdot g + f \cdot \frac{dg}{dx_n} \end{bmatrix}$$

$$= \begin{bmatrix} \frac{df}{dx} \\ \frac{df}{dx_n} \end{bmatrix} \cdot g + f \cdot \begin{bmatrix} \frac{dg}{dx} \\ \frac{dg}{dx_n} \end{bmatrix} = \frac{df}{dx} \cdot g(x) + f(x) \cdot \frac{dg}{dx}.$$

4. Division rule.

$$\frac{d\left[\frac{f(x)}{g(x)}\right]}{dx} = \frac{\frac{df}{dx} \cdot g(x) - \frac{d(g(x))}{dx} \cdot f(x)}{g(x)^2}, \text{ here } g(x) \neq 0.$$

$$\begin{aligned} \text{Prove: } \frac{d\left[\frac{f(x)}{g(x)}\right]}{dx} &= \begin{bmatrix} \frac{d\left(\frac{f}{g}\right)}{dx} \\ \frac{d\left(\frac{f}{g}\right)}{dx_n} \end{bmatrix} = \begin{bmatrix} \frac{1}{g^2} \left(\frac{df}{dx} \cdot g - \frac{dg}{dx} \cdot f \right) \\ \frac{1}{g^2} \left(\frac{df}{dx_n} \cdot g - \frac{dg}{dx_n} \cdot f \right) \end{bmatrix} \\ &= \frac{1}{g^2} \left(\begin{bmatrix} \frac{df}{dx} \\ \frac{df}{dx_n} \end{bmatrix} \cdot g - \begin{bmatrix} \frac{dg}{dx} \\ \frac{dg}{dx_n} \end{bmatrix} \cdot f \right) \\ &= \frac{1}{g^2} \left(\frac{df}{dx} \cdot g(x) - \frac{dg}{dx} \cdot f(x) \right) \end{aligned}$$

(1)

5. Some equations.

$$\textcircled{1} \quad \frac{\partial(x^T a)}{\partial x} = \frac{\partial(a^T x)}{\partial x} = a, \text{ here } a \text{ is constant vector.}$$

$a = [a_1, \dots, a_n]^T$.

Prove:

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

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

$$\text{Prove: } \frac{\partial(x^T x)}{\partial x} = \begin{bmatrix} \frac{\partial(x_1^2 + x_2^2 + \dots + x_n^2)}{\partial x_1} \\ \frac{\partial(x_1^2 + \dots + x_n^2)}{\partial x_n} \end{bmatrix} = \begin{bmatrix} 2x_1 \\ \vdots \\ 2x_n \end{bmatrix} = 2x$$

$$\textcircled{3} \quad \frac{\partial(x^T A x)}{\partial x} = Ax + A^T x \text{ . here } A_{n \times n} \text{ is square matrix .}$$

$$A_{n \times n} = (a_{ij})_{i=1, j=1}^{n, n}$$

prove -

$$\begin{bmatrix} x_1 & \dots & x_n \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ \vdots & & & & \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

$$= [a_{11}x_1 + a_{21}x_2 + \dots + a_{n1}x_n, a_{12}x_1 + a_{22}x_2 + \dots + a_{n2}x_n, \dots, a_{1n}x_1 + a_{2n}x_2 + \dots + a_{nn}x_n]$$

$$\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

$$\left(a_{11}x_1x_1 + a_{21}x_1x_2 + a_{31}x_1x_3 + \dots + a_{n1}x_1x_n + \right. \\ \left. a_{12}x_1x_2 + a_{22}x_2x_2 + a_{32}x_2x_3 + \dots + a_{n2}x_2x_n + \right. \\ \left. a_{13}x_1x_3 + a_{23}x_2x_3 + a_{33}x_3x_3 + \dots + a_{n3}x_3x_n + \dots \right) \rightarrow G(x)$$

$$\frac{\partial(x^T A x)}{\partial x} = \begin{bmatrix} \frac{\partial}{\partial x_1} G(x) \\ \vdots \\ \frac{\partial}{\partial x_n} G(x) \end{bmatrix} = \begin{bmatrix} a_{11}x_1 + \boxed{a_{11}x_1} + a_{21}x_2 + a_{31}x_3 + \dots + a_{n1}x_n + \\ a_{12}x_2 + a_{13}x_3 + \dots + a_{n1}x_n \\ a_{12}x_1 + a_{22}x_2 + \boxed{a_{22}x_2} + a_{32}x_3 + \dots + a_{n2}x_n + \\ a_{21}x_1 + a_{23}x_3 + \dots + a_{2n}x_n \\ a_{1n}x_1 + a_{2n}x_2 + a_{3n}x_3 + \dots + \boxed{a_{nn}x_n} + \\ a_{nn}x_1 + a_{nn}x_2 + \dots + a_{n-1}x_{n-1} \end{bmatrix}$$

$$= \begin{bmatrix} a_{11}x_1 + \dots + a_{n1}x_n \\ a_{12}x_1 + \dots + a_{n2}x_n \\ \vdots \\ a_{nn}x_1 + \dots + a_{nn}x_n \end{bmatrix} + \begin{bmatrix} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n \\ \vdots \\ a_{nn}x_1 + a_{nn}x_2 + \dots + a_{nn}x_n \end{bmatrix}$$

$$\stackrel{\text{AX}}{+} \stackrel{\text{AX}}{\downarrow}$$

//

$$\text{④ } \frac{\partial(a^T x x^T b)}{\partial x} = ab^T x + ba^T x$$

Here a, b are constant vector.
 $a = (a_1, \dots, a_n)^T, b = (b_1, \dots, b_n)^T$.

prove:

M_1 :

$$\begin{aligned} \frac{\partial(a^T x x^T b)}{\partial x} &= \left[\begin{array}{c} \frac{\partial}{\partial x_1} [(a_1 x_1 + \dots + a_n x_n)(b_1 x_1 + \dots + b_n x_n)] \\ \frac{\partial}{\partial x_2} [(a_1 x_1 + \dots + a_n x_n)(b_1 x_1 + \dots + b_n x_n)] \\ \vdots \\ \frac{\partial}{\partial x_n} [(a_1 x_1 + \dots + a_n x_n)(b_1 x_1 + \dots + b_n x_n)] \end{array} \right] \\ &= \left[\begin{array}{c} a_1(b_1 x_1 + \dots + b_n x_n) + b_1(a_1 x_1 + \dots + a_n x_n) \\ a_2(b_1 x_1 + \dots + b_n x_n) + b_2(a_1 x_1 + \dots + a_n x_n) \\ \vdots \\ a_n(b_1 x_1 + \dots + b_n x_n) + b_n(a_1 x_1 + \dots + a_n x_n) \end{array} \right] \\ &= \left[\begin{array}{c} a_1(b_1 x_1 + \dots + b_n x_n) \\ \vdots \\ a_n(b_1 x_1 + \dots + b_n x_n) \end{array} \right] + \left[\begin{array}{c} b_1(a_1 x_1 + \dots + a_n x_n) \\ \vdots \\ b_n(a_1 x_1 + \dots + a_n x_n) \end{array} \right] \\ &= \left[\begin{array}{c} a_1 b^T x \\ \vdots \\ a_n b^T x \end{array} \right] + \left[\begin{array}{c} b_1 a^T x \\ \vdots \\ b_n a^T x \end{array} \right] \\ &= ab^T x + a^T b x \end{aligned}$$

M_2 : For $a^T x = x^T a, x^T b = b^T x$.

$$\text{so } \frac{\partial(a^T x x^T b)}{\partial x} = \frac{\partial(x^T a b^T x)}{\partial x}$$

Here ab^T is a $n \times n$ constant matrix, so: (from ③)

$$\frac{\partial(a^T x x^T b)}{\partial x} = \frac{\partial(x^T a b^T x)}{\partial x} = ab^T x + a^T b x$$