

Matrix Differentiation

前面基本和线的一样

Convention 3

$y = \psi(x)$ y 是 $m \times 1$ 向量, x 是 $n \times 1$ 的向量

$$\frac{\partial y}{\partial x} = \begin{bmatrix} \frac{\partial y_1}{\partial x_1} & \frac{\partial y_1}{\partial x_2} & \dots & \frac{\partial y_1}{\partial x_n} \\ \frac{\partial y_2}{\partial x_1} & \frac{\partial y_2}{\partial x_2} & \dots & \frac{\partial y_2}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial y_m}{\partial x_1} & \frac{\partial y_m}{\partial x_2} & \dots & \frac{\partial y_m}{\partial x_n} \end{bmatrix}$$

Proposition 5

Conclusion 4

$y = Ax$ $y: m \times 1$ 向量 $A: m \times n$ 矩阵 $x: n \times 1$ 向量
 $A \times x$ 无关

则 $\frac{\partial y}{\partial x} = A$

证明: $y_i = A_{ij}x_j$

$$\frac{\partial y_i}{\partial x_j} = A_{ij} \rightarrow \frac{\partial y}{\partial x} = A$$

$$\frac{\partial y}{\partial x}$$

Proposition 6

$y = Ax$ $x = x(z)$ A 是无关

$$\frac{\partial y}{\partial z} = A \frac{\partial x}{\partial z}$$

Proposition 7

① $x = y^T A x$ (x 与 y 无关)

② $y: m \times 1$
 $A: m \times n$
 $x: n \times 1$

$$\begin{cases} \frac{\partial x}{\partial y} = y^T A \\ \frac{\partial y}{\partial x} = x^T A^T \end{cases}$$

证明: 请 Proposition 5 知 $\frac{\partial y}{\partial x} = y^T A$

$$\text{令 } x = w^T x \text{ 且 } w = A^T y \text{ 则 } w^T = y^T A$$

由 $x^T = x = x^T A^T y$, 同由 P 5 得

$$\frac{\partial y}{\partial y} = x^T A^T$$

Proposition 8:

$x = x^T A x$ $x: n \times 1$ $A: n \times n$ A 与 x 无关

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

证明: $x = A_{ij} x_i x_j$ (张量记号)

$$\frac{\partial x}{\partial x_k} = \sum_{j=1}^n A_{kj} x_j + \sum_{i=1}^n A_{ik} x_i = A_{kj} x_j + A_{ik} x_i$$

$$\frac{\partial x}{\partial x} = x^T A^T + x^T A$$

$$= x^T (A + A^T)$$

Proposition 9:

$\frac{\partial x}{\partial x} = 2x^T A$

$$\frac{\partial x}{\partial x} = 2x^T A$$

Proposition 10

$x = y^T x$, $y: n \times 1$ $y = y(z)$ $z: m \times 1$ 向量
 $x: n \times 1$ $x = x(z)$

↓

$$\frac{\partial x}{\partial z} = x^T \frac{\partial y}{\partial z} + y^T \frac{\partial x}{\partial z}$$

证明:

$$\frac{\partial x}{\partial z} = x^T \frac{\partial y}{\partial z} + y^T \frac{\partial x}{\partial z}$$

$$\frac{\partial x}{\partial z} = x^T \frac{\partial y}{\partial z} + y^T \frac{\partial x}{\partial z}$$

Proposition 11:

① $x = x^T x$ ② $x = x(z)$ $z: m \times 1$

$$\frac{\partial x}{\partial z} = 2x^T \frac{\partial x}{\partial z}$$



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Proposition 12:

$$\alpha = y^T A x$$

$$\frac{\partial \alpha}{\partial z} = x^T A^T \frac{\partial y}{\partial z} + y^T A \frac{\partial x}{\partial z}$$

~~Proposition 12~~ 的条件为:

Proposition 7 的条件加上 $x = x(z)$ $y = y(z)$ $z: 1 - \text{num}(\text{vec}(z))$

~~证明~~

证明:

$$\alpha = \sum_i A_{ij} y_i x_j \quad (i: 1-n \quad j: 1-n)$$

$$\frac{\partial \alpha}{\partial z_k} = \sum_j A_{ij} \frac{\partial y_i}{\partial z_k} x_j + \sum_i A_{ij} y_i \frac{\partial x_j}{\partial z_k}$$

$$\alpha = w^T x \quad w^T = y^T A$$

由 Proposition 10:

$$\frac{\partial \alpha}{\partial z} = x^T \frac{\partial w}{\partial z} + w^T \frac{\partial x}{\partial z}$$

$$\frac{\partial \alpha}{\partial z} = \frac{\partial y}{\partial z} \frac{\partial y}{\partial z} + \frac{\partial \alpha}{\partial x} \frac{\partial x}{\partial z}$$

$$= x^T A^T \frac{\partial y}{\partial z} + y^T A \frac{\partial x}{\partial z}$$

P7

Pro 13: (Pro 12 的特例)

$$\alpha = x^T A x$$

$$\frac{\partial \alpha}{\partial z} = x^T (A + A^T) \frac{\partial x}{\partial z}$$



$$dA = dW = C_{ij} \varepsilon_j d\varepsilon_i = C_{11} \varepsilon_1 d\varepsilon_1 + C_{12} \varepsilon_2 d\varepsilon_1 + C_{21} \varepsilon_1 d\varepsilon_2 + C_{22} \varepsilon_2 d\varepsilon_2 + \dots$$

ε_i

$$\begin{pmatrix} \frac{1}{2} C_{11} \varepsilon_1^2 & C_{12} \varepsilon_1 \varepsilon_2 \\ C_{21} \varepsilon_1 \varepsilon_2 & \frac{1}{2} C_{22} \varepsilon_2^2 \end{pmatrix}$$



$$W = \frac{1}{2} C_{11} \varepsilon_1^2 + C_{12} \varepsilon_1 \varepsilon_2 + C_{21} \varepsilon_1 \varepsilon_2 + \frac{1}{2} C_{22} \varepsilon_2^2 + \dots$$

$$W = \frac{1}{2} [\varepsilon_1 \ \varepsilon_2 \ \varepsilon_3 \ \varepsilon_4 \ \varepsilon_5 \ \varepsilon_6] \begin{bmatrix} C_{11} & C_{12} & \dots & C_{16} \\ C_{21} & C_{22} & \dots & C_{26} \\ \vdots & \vdots & \ddots & \vdots \\ C_{61} & C_{62} & \dots & C_{66} \end{bmatrix} \begin{bmatrix} \varepsilon_1 \\ \varepsilon_2 \\ \vdots \\ \varepsilon_6 \end{bmatrix}$$

$$= \frac{1}{2} \varepsilon^T C \varepsilon$$

$$dW = \frac{1}{2} \varepsilon^T C d\varepsilon$$

$$\frac{\partial W}{\partial \varepsilon_1} = \left[\frac{1}{2} C_{11} \varepsilon_1 + \frac{1}{2} C_{12} \varepsilon_2 + \frac{1}{2} C_{13} \varepsilon_3 + \frac{1}{2} C_{14} \varepsilon_4 + \frac{1}{2} C_{15} \varepsilon_5 + \frac{1}{2} C_{16} \varepsilon_6 + \dots \right]$$

$$\left[\frac{\partial C_{1j} \varepsilon_j \varepsilon_1}{\partial \varepsilon_2}, \frac{\partial C_{2j} \varepsilon_j \varepsilon_2}{\partial \varepsilon_3}, \dots, \frac{\partial C_{6j} \varepsilon_j \varepsilon_6}{\partial \varepsilon_6} \right]$$

$$= \left[C_{11} \varepsilon_1 + \left(\frac{1}{2} C_{12} \varepsilon_2 + \frac{1}{2} C_{21} \varepsilon_1 \right) + \left(\frac{1}{2} C_{13} \varepsilon_3 + \frac{1}{2} C_{31} \varepsilon_1 \right) + \dots \right. \\ \left. + \left(\frac{1}{2} C_{16} \varepsilon_6 + \frac{1}{2} C_{61} \varepsilon_1 \right), C_{22} \varepsilon_2 + \left(\frac{1}{2} C_{21} \varepsilon_1 + \frac{1}{2} C_{12} \varepsilon_2 \right) + \dots + \left(\frac{1}{2} C_{26} \varepsilon_6 + \frac{1}{2} C_{62} \varepsilon_2 \right), \dots \right]$$

$$\text{由 } F(v) = \frac{dF}{dv} = \frac{\partial F(v)}{\partial v^i} g^i \otimes g^i = \frac{\partial F}{\partial v^i} \otimes g^i$$

$$(v = v^i g_i) \quad dv = dv^i g_i; \text{ 由 } dF = F(v) \cdot dv$$

$$dF = \frac{\partial F}{\partial v^i} \otimes g^i (dv^i g_i) = \frac{\partial F}{\partial v^i} dv^i$$

$$dA = dW = \sigma_i d\varepsilon_i = C_{ij} \varepsilon_j d\varepsilon_i \rightarrow W =$$

$$\sigma_i = C_{ij} \varepsilon_j \quad (i, j = 1, 2, \dots, 6)$$

$$W = \frac{1}{2} C_{ij} \varepsilon_i \varepsilon_j$$

$$W = \frac{1}{2} \varepsilon^T C \varepsilon$$

$$+ \frac{1}{2} C_{66} \varepsilon_6^2 + \frac{1}{2} C_{22} \varepsilon_2^2 + \frac{1}{2} C_{12} \varepsilon_1 \varepsilon_2 + \frac{1}{2} C_{13} \varepsilon_1 \varepsilon_3 + \frac{1}{2} C_{14} \varepsilon_1 \varepsilon_4 + \dots$$

$$\left[\frac{\partial W}{\partial \varepsilon} = \frac{1}{2} \varepsilon^T (C + C^T) \right] \text{ (来自知书)}$$



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