

3(a)

To Prove : $\nabla_x (x^T a) = \nabla_x (a^T x) = a$

for any two column vectors $x, a \in \mathbb{R}^n$

Proof : $x = [x_1 \dots x_n]^T \Rightarrow x^T = [x_1 \dots x_n]$
 $a = [a_1 \dots a_n]^T \Rightarrow a^T = [a_1 \dots a_n]$

$$f(x) = (x^T \cdot a) = [x_1 \dots x_n] [a_1 \dots a_n]^T \Rightarrow f(x) = [x_1 a_1 \dots x_n a_n]$$
~~$$\Rightarrow f(x) = x_1 a_1 + \dots + x_n a_n$$~~

~~$$\therefore \text{L.O.S.} \nabla_x f(x) = \left[\frac{\partial f}{\partial x_1} \dots \frac{\partial f}{\partial x_n} \right]^T$$~~

$$= [a_1 \dots a_n]^T$$

$$\Rightarrow \nabla_x (x^T a) = a \quad \text{--- (1)}$$

Say, $\hat{f}(x) = (a^T x) = [a_1 \dots a_n] [x_1 \dots x_n]^T$
 $= [a_1 x_1 \dots a_n x_n] = x^T a = f(x)$

$$\therefore \hat{f}(x) = f(x) \Rightarrow \nabla_x \hat{f}(x) = \nabla_x f(x) \quad \text{--- (2)}$$

from (1) & (2), $\nabla_x (x^T a) = \nabla_x (a^T x) = a \quad \blacksquare$

$$\begin{bmatrix} 2 & 3 \end{bmatrix} \begin{bmatrix} 3 & 4 \\ 5 & 6 \end{bmatrix}$$

3(b)

To Prove: $\nabla_x (x^T A x) = (A + A^T) x$
for col. vector $x \in \mathbb{R}^n$ & $n \times n$ mtr A .

Proof: $x = [x_1, \dots, x_n]^T \Rightarrow x^T = [x_1, \dots, x_n]$

$$A = \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{nn} \end{bmatrix}$$

For $x^T A x$, we multiply left to right.

$$x^T A = [x_1, \dots, x_n]_{1 \times n} \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{nn} \end{bmatrix}_{n \times n}$$

$$= \left[(x_1 a_{11} + \dots + x_n a_{n1}) \quad \dots \quad (x_1 a_{1n} + \dots + x_n a_{nn}) \right]_{1 \times n}$$

$$\text{then, } x^T A x = \underbrace{\left[\downarrow \right]}_{1 \times n} \cdot [x_1, \dots, x_n]^T_{n \times 1}$$

$$\Rightarrow f(x) = x^T A x = \left[x_1 (x_1 a_{11} + \dots + x_n a_{n1}) \right. \\ \left. + \dots + x_n (x_1 a_{1n} + \dots + x_n a_{nn}) \right]_{(x)}$$

$$\therefore f(x) = \begin{bmatrix} (x_1^2 a_{11} + x_2^2 a_{22} + \dots + x_n^2 a_{nn}) \\ + x_1 x_2 a_{12} + x_1 x_2 a_{21} \\ + x_1 x_3 a_{13} + x_1 x_3 a_{31} \\ + \dots \\ + x_1 x_n a_{1n} + x_1 x_n a_{n1} \\ + \dots \end{bmatrix}_{1 \times 1}$$

$$= \begin{bmatrix} (x_1^2 a_{11} + x_2^2 a_{22} + \dots + x_n^2 a_{nn}) + x_1 (x_2 (a_{12} + a_{21}) + \dots + x_n (a_{1n} + a_{n1})) \\ + \dots + x_n (x_1 (a_{n1} + a_{1n}) + \dots + x_{n-1} (a_{nn-1} + a_{n-1n})) \end{bmatrix}_{1 \times 1}$$

$$\nabla_x f(x) = \begin{bmatrix} (2x_1 a_{11} + x_2 (a_{11} + a_{12}) + \dots + x_n (a_{11} + a_{1n})) \\ \vdots \\ (2x_n a_{nn} + x_1 (a_{nn} + a_{1n}) + \dots + x_{n-1} (a_{nn} + a_{n-1,n})) \end{bmatrix}_{n \times 1}$$

→ ②

Solving for $(A + A^T)x = \bar{p}(x)$.

$$A + A^T = \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{nn} \end{bmatrix}_{n \times n} + \begin{bmatrix} a_{11} & \dots & a_{n1} \\ \vdots & \ddots & \vdots \\ a_{1n} & \dots & a_{nn} \end{bmatrix}_{n \times n}$$

$$= \begin{bmatrix} 2a_{11} & \dots & a_{1n} + a_{n1} \\ \vdots & \ddots & \vdots \\ a_{n1} + a_{1n} & \dots & 2a_{nn} \end{bmatrix}_{n \times n}$$

$$(A + A^T)x = \underbrace{\begin{bmatrix} \vdots \\ \vdots \end{bmatrix}}_{n \times 1} \cdot \underbrace{[x_1 \dots x_n]}_{n \times 1}^T$$

$$= \begin{bmatrix} (2a_{11}x_1 + (a_{12} + a_{21})x_2 + \dots + (a_{1n} + a_{n1})x_n) \\ \vdots \\ ((a_{n1} + a_{1n})x_1 + \dots + 2a_{nn}x_n) \end{bmatrix}_{n \times 1}$$

→ ①

from ① & ②, we have,

$$\nabla_x f(x) = \nabla_x (x^T A x) = (A + A^T) x. \quad \blacksquare$$

3(c)

Say, $A \rightarrow$ symmetric.

then by property of symmetric matrices,

$$A + A^T = 2A.$$

$$\text{thus, } \nabla_x f(x) = \nabla_x (x^T A x) = (A + A^T) x = 2Ax.$$

3(d)

$$f(x) = (Ax + b)^T (Ax + b) \\ = (x^T A^T + b^T)(Ax + b)$$

$$= \underline{x^T A^T A x} + \underline{x^T A^T b} + \underline{b^T A x} + b^T b$$

$$= x^T \cdot (A^T A) \cdot x + x^T \cdot (A^T b) + (b^T A) \cdot x + (b^T b)$$

$$\nabla_x f(x) = (A^T A + (A^T A)^T) x + A^T b + b^T A + 0$$

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

$$= 2(A^T A) x + 2(A^T b) \quad (\text{due to symmetry \& transpose properties})$$

$$= 2A^T (Ax + b).$$

Linear Regression

- (i) Training Cost = 39.243 "
- (ii) Testing Cost = 206.796 "