

CS 58 3A Spring 23

Sanjeet Vinod Jain
CWRD : 20012768

Q1] $x = [5, 0, 1, -2]^T$ $a = [4, -2, 6, -1]^T$

1) $\|x\|_2^2 = \sum_{i=1}^n x_i^2 = 5^2 + 0^2 + 1^2 + (-2)^2$
 $= 25 + 1 + 4$

$= 30$

2) ℓ_1 norm $= \|x\|_1 = \sum_{i=1}^n |x_i| = 5 + 0 + 1 + 2$
 $= 8$

3) inner product of x and $a \rightarrow a^T x$

$$a^T x = ([4, -2, 6, -1]^T)^T \cdot [5, 0, 1, -2]^T$$

$$= [4, -2, 6, -1] \begin{bmatrix} 5 \\ 0 \\ 1 \\ -2 \end{bmatrix} = 20 + 0 + (6)(1) + (-1)(-2)$$
$$= 20 + 6 + 2$$
$$= 28$$

Q2] $A = \begin{bmatrix} 6 & 1 & 2 \\ -5 & 0 & -3 \end{bmatrix}_{2 \times 3}$ $b = \begin{bmatrix} -4 \\ 5 \\ 0 \end{bmatrix}_{3 \times 1}$

$$1) Ab = \begin{bmatrix} 6 & 1 & 2 \\ -5 & 0 & -3 \end{bmatrix} \begin{bmatrix} -4 \\ 5 \\ 0 \end{bmatrix} = \begin{bmatrix} (6)(-4) + (1)(5) + (2)(0) \\ (-5)(-4) + 0 + 0 \end{bmatrix}_{2 \times 1}$$
$$= \begin{bmatrix} -24 + 5 + 0 \\ 20 \end{bmatrix} = \begin{bmatrix} -19 \\ 20 \end{bmatrix}$$

$$2) A A^T = \begin{bmatrix} 6 & 1 & 2 \\ -5 & 0 & -3 \end{bmatrix}_{2 \times 3} \begin{bmatrix} 6 & -5 \\ 1 & 0 \\ 2 & -3 \end{bmatrix}_{3 \times 2}$$

$$= \begin{bmatrix} (6)(6) + (1)(1) + (2)(2) & (6)(-5) + 0 + (-2)(-3) \\ (-5)(6) + 0 + (-3)(2) & (-5)(-5) + 0 + (-3)(-3) \end{bmatrix}$$

$$= \begin{bmatrix} 36 + 1 + 4 & -30 + 0 + 6 \\ -30 + 0 - 6 & 25 + 0 + 9 \end{bmatrix} = \begin{bmatrix} 41 & -36 \\ -36 & 34 \end{bmatrix}$$

Q 3]

$$x = [x_1, x_2, x_3] \quad y = \frac{x_1^2}{2} + \log_e x_2 - \frac{x_1}{x_3}$$

$$\frac{\partial y}{\partial x} \text{ at } x = [9, 1/2, 1/3]$$

$$\begin{aligned} \frac{\partial y}{\partial x} &= \frac{\partial}{\partial x} \left[\frac{x_1^2}{2} + \log_e x_2 - \frac{x_1}{x_3} \right] \\ &= \frac{\partial}{\partial x} \left(\frac{x_1^2}{2} \right) + \frac{\partial}{\partial x} (\log_e x_2) - \frac{\partial}{\partial x} \left(\frac{x_1}{x_3} \right) \end{aligned}$$

$$\frac{\partial y}{\partial x} = \left[\frac{\partial y}{\partial x_1}, \frac{\partial y}{\partial x_2}, \frac{\partial y}{\partial x_3} \right]$$

$$\begin{aligned} \Rightarrow \frac{\partial y}{\partial x_1} &= \frac{\partial}{\partial x_1} \left(\frac{x_1^2}{2} \right) + \frac{\partial}{\partial x_1} (\log_e x_2) - \frac{\partial}{\partial x_1} \left(\frac{x_1}{x_3} \right) \\ &= \frac{2x_1}{2} + 0 - \frac{1}{x_3} = x_1 - \frac{1}{x_3} \end{aligned}$$

$$\text{at } \Rightarrow \frac{\partial y}{\partial x_1} = 9 - 1 = 9 - 3 = 6$$

$$\begin{aligned} \text{Now } \frac{\partial y}{\partial x_2} &= \frac{\partial (x_1^2/2)}{\partial x_2} + \frac{\partial \log_e x_2}{\partial x_2} - \frac{\partial (x_1/x_3)}{\partial x_2} \\ &= \frac{\partial \log_e x_2}{\partial x_2} = \frac{1}{x_2} = \frac{1}{1/2} = \underline{\underline{2}} \quad \text{--- (2)} \end{aligned}$$

$$\begin{aligned} \text{Now } \frac{\partial y}{\partial x_3} &= \frac{-\partial (x_1/x_3)}{\partial x_3} = \frac{-x_1}{\cancel{x_3}} \left(\frac{\partial (1/x_3)}{\partial x_3} \right) \\ &= -x_1 \left(\frac{-1}{x_3^2} \right) = \frac{x_1}{x_3^2} = \frac{9}{(1/3)^2} = 9 \times 9 = \underline{\underline{81}} \quad \text{--- (3)} \end{aligned}$$

Combining (1), (2) & (3) we get

$$\frac{\partial y}{\partial x} = [6, 2, 81]$$

$$Q4] f(w) = ||xw - y||_2^2 + \lambda ||w||_2^2$$

Generally ^{squared} l_2 norm for vector y given as

$$||y||_2^2 = \sum_{k=1}^n y_k^2$$

$$\text{Taking derivative } \frac{\partial ||y||_2^2}{\partial y_j} = \frac{\partial}{\partial y_j} \left[\sum_{k=1}^n y_k^2 \right]$$

$$= \sum_{k=1}^n \frac{\partial}{\partial y_j} (y_k^2) = 2y_j$$

$$= \sum_{k=1}^n \frac{\partial}{\partial y_j} (y_k^2) = 2y_j$$

$$= 0 \text{ if } j \neq k$$

$$= 2y_j \text{ else}$$

$$\therefore \frac{\partial ||w||_2^2}{\partial w} = 2w \quad \& \quad \frac{\partial ||w||_2^2}{\partial w_j} = 2w_j$$

$$\Rightarrow \frac{\partial (||xw - y||_2^2)}{\partial w_i} = \frac{\partial}{\partial w_i} \left(\sum_{j=1}^n (x_j w_i - y_j)^2 \right)$$

$$= \frac{\partial}{\partial w_i} \sum_{j=1}^n (x_j (w_i - y_j))^2 =$$

$$= \sum_{j=1}^n \left(\frac{\partial (x_j (w_i - y_j))^2}{\partial w_i} \right) = \sum_{j=1}^n 2(x_j (w_i - y_j)) \frac{\partial (x_j w_i)}{\partial w_i}$$

$$= \sum_{j=1}^n (2x_j (x_j w_i - y_j))$$

$$\Rightarrow \frac{\partial}{\partial \mathbf{w}} \left(\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2 \right) = \begin{bmatrix} \sum_{j=1}^n (2x_j (x_j^0 \mathbf{w}_1 - y_j)) \\ \vdots \\ \sum_{j=1}^n (2x_j (x_j^l \mathbf{w}_l - y_j)) \end{bmatrix}$$

Also $\frac{\partial \lambda \|\mathbf{w}\|_2^2}{\partial \mathbf{w}} = \cancel{2\lambda \mathbf{w}} \quad \lambda \begin{bmatrix} 2\mathbf{w}_0 \\ \vdots \\ 2\mathbf{w}_l \end{bmatrix} = 2\lambda \begin{bmatrix} \mathbf{w}_0 \\ \vdots \\ \mathbf{w}_l \end{bmatrix}$

$$\Rightarrow \frac{\partial f(\mathbf{w})}{\partial \mathbf{w}} = \begin{bmatrix} \sum_{j=1}^n (2(x_j^0)(x_j^0 \mathbf{w}_0 - y_j)) \\ \vdots \\ \sum_{j=1}^n (2x_j (x_j^l \mathbf{w}_l - y_j)) \end{bmatrix} + 2\lambda \begin{bmatrix} \mathbf{w}_0 \\ \vdots \\ \mathbf{w}_l \end{bmatrix}$$