Homework 1

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I pledge my honor that I have abided by the Stevens Honor System

- **Q1** With $x = \begin{bmatrix} 4 & 0 & -1 & 2 \end{bmatrix}^T$, calculate the following:
 - 1. Squared ℓ_2 -norm of x:

$$||x||_{2}^{2} = \left(\left(\sum_{i=1}^{4} x_{i}^{2}\right)^{1/2}\right)^{2}$$

$$= \sum_{i=1}^{4} x_{i}^{2}$$

$$= 4^{2} + 0^{2} + -1^{2} + 2^{2}$$

$$= 16 + 0 + 1 + 4$$

$$||x||_2^2 = 21$$

2. ℓ_1 -norm of x:

$$||x||_1 = \sum_{i=1}^4 |x_i|$$

= 4 + 0 + | - 1| + 2

$$||x||^1 = 7$$

3. inner product of x and a, $a = \begin{bmatrix} 0 & -2 & 6 & 3 \end{bmatrix}^T$

$$x \cdot a = (4 * 0) + (0 * -2) + (-1 * 6) + (2 * 3)$$
$$= 0 + 0 - 6 + 6$$

$$x \cdot a = 0$$

Q2 Calculate the following with $A \in \mathbb{R}^{2\times 3}$ and $b \in \mathbb{R}^3$:

$$\mathbf{A} = \begin{bmatrix} 6 & 1 & -2 \\ -5 & 0 & 8 \end{bmatrix}$$

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$$\mathbf{b} = \begin{bmatrix} 1\\3\\-2 \end{bmatrix}$$

1.
$$\mathbf{Ab} = \begin{bmatrix} 13 \\ -21 \end{bmatrix}$$

2.
$$\mathbf{A}\mathbf{A}^{\mathbf{T}} = \begin{bmatrix} 6 & 1 & -2 \\ -5 & 0 & 8 \end{bmatrix} \begin{bmatrix} 6 & -5 \\ 1 & 0 \\ -2 & 8 \end{bmatrix} = \begin{bmatrix} 41 & -46 \\ -46 & 89 \end{bmatrix}$$

Q3 With
$$\mathbf{x} = \begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix}$$
 and $y = \frac{x_1^2}{2} + \log_e x_2 - \frac{x_1}{x_3} = \frac{1}{2}x_1 + \log_e x_2 - x_1x_3^{-1}$, calculate $\frac{dy}{d\mathbf{x}}$ at $\mathbf{x} = \begin{bmatrix} 7 & \frac{1}{4} & \frac{1}{3} \end{bmatrix}$:

$$\frac{dy}{d\mathbf{x}} = \begin{bmatrix} \frac{\partial y}{x_1} & \frac{\partial y}{x_2} & \frac{\partial y}{x_3} \end{bmatrix}$$

$$\frac{\partial y}{x_1} = x_1 - x_3^{-1}$$

$$\frac{\partial y}{x_2} = x_2^{-1}$$

$$\frac{\partial y}{x_3} = x_1 x_3^{-2}$$

$$\frac{dy}{d\mathbf{x}} = \begin{bmatrix} x_1 - x_3^{-1} & x_2^{-1} & x_1 x_3^{-2} \end{bmatrix}$$
$$= \begin{bmatrix} 7 - (\frac{1}{3})^{-1} & (\frac{1}{4})^{-1} & 7(\frac{1}{3})^{-2} \end{bmatrix}$$
$$= \begin{bmatrix} 7 - 3 & 4 & 7(9) \end{bmatrix}$$

$$\frac{dy}{d\mathbf{x}} = \begin{bmatrix} 4 & 4 & 63 \end{bmatrix}$$

Q4 X is an
$$n \times d$$
 matrix, **y** is an $n \times 1$ vector, and **w** is a $d \times 1$ vector. Let $f(\mathbf{w}) = \|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2 + \lambda \|\mathbf{w}\|_2^2$. Calculate $\frac{\partial f(\mathbf{w})}{\partial \mathbf{w}}$.

Derivative of some squared ℓ_2 norm of \mathbf{x} :

$$\frac{\partial}{\partial \mathbf{x}}(\|\mathbf{x}\|_2^2) = 2\mathbf{x}$$

Separate the derivatives of each term of $f(\mathbf{w})$:

$$\frac{\partial f(\mathbf{w})}{\partial \mathbf{w}} = \frac{\partial}{\partial \mathbf{w}} (\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_{2}^{2}) + \frac{\partial}{\partial \mathbf{w}} (\lambda \|\mathbf{w}\|_{2}^{2})$$

Starting with $\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$, using the chain rule

$$\frac{\partial}{\partial \mathbf{w}}(\|\mathbf{X}\mathbf{w}-\mathbf{y}\|_2^2) = \mathbf{X}*(2(\mathbf{X}\mathbf{w}-\mathbf{y})) = 2(\mathbf{X}^T\mathbf{X}\mathbf{w}-\mathbf{X}^T\mathbf{y})$$

Now $\lambda \|\mathbf{w}\|_2^2$:

$$\frac{\partial}{\partial \mathbf{w}} (\lambda \|\mathbf{w}\|_2^2) = 2\lambda \mathbf{w}$$

Finally, we have:

$$\frac{\partial f(\mathbf{w})}{\partial \mathbf{w}} = 2(\mathbf{X}^T \mathbf{X} \mathbf{w} - \mathbf{X}^T \mathbf{y}) + 2\lambda \mathbf{w} = 2(\mathbf{X}^T \mathbf{X} \mathbf{w} - \mathbf{X}^T \mathbf{y} + \lambda \mathbf{w})$$