

---

## Midterm 1 - DSC 40A, Winter 2024

---

---

### Instructions

- This is a 50-minute exam consisting of 5 questions worth a total of 40 points.
  - The only allowed resource is the provided reference sheet.
  - No calculators.
  - Please write neatly and stay within the provided boxes.
  - You may fill out the **front page only** until you are instructed to start.
- 

### Statement of Academic Integrity

By submitting your exam, you are attesting to the following statement of academic integrity.

*I will act with honesty and integrity during this exam.*

---

Name:

Solutions

PID:

A12345678

Seat you are in:

---

Version - A

---



**Solution:** Since there are 3 possible median values, we will have to discuss each situation separately. In case 1, when  $0 < a \leq 2$ ,  $Median_D = 2$ , therefore we have:

$$3 + \frac{a}{5} < 2$$
$$a < -5$$

But  $a < -5$  is in conflict with the condition  $0 < a \leq 2$ , therefore there is no solution in this situation, and  $Median_D = 2$  is impossible.

In case 2 when  $2 < a < 5$ ,  $Median_D = a$ , therefore we have:

$$3 + \frac{a}{5} < a$$
$$3 < \frac{4}{5}a$$
$$a > \frac{15}{4}$$

So  $a$  has to be larger than  $\frac{15}{4}$ . But remember from the prerequisite condition that  $2 < a < 5$ . To satisfy both conditions, we must have  $\frac{15}{4} < a < 5$ .

In case 3 when  $a \geq 5$ ,  $Median_D = 5$ , therefore we have:

$$3 + \frac{a}{5} < 5$$
$$a < 10$$

combining with the prerequisite condition, we have  $5 \leq a < 10$

Combining the range of case 2 and 3, we have  $\frac{15}{4} < a < 10$  as our final answer.

2. (4 points) Let  $R_{sq}(h)$  represent the mean squared error of a constant prediction  $h$  for a given dataset. For the dataset  $\{3, y_1\}$ , the graph of  $R_{sq}(h)$  has its minimum at the point  $(5, r_1)$ . Find out the value of  $y_1$  and  $r_1$

$$y_1 = \boxed{7} \quad \text{and} \quad r_1 = \boxed{4}$$

**Solution:** The mean square error is written as:

$$R_{sq}(h) = \frac{1}{n} \sum_{i=0}^n (y_i - h)^2$$

Since we only have two data points ( $n = 2$ ), the equation simplifies to:

$$R_{sq}(h) = \frac{1}{2}((y_0 - h)^2 + (y_1 - h)^2)$$

Taking derivative with respect to  $h$ , we have:

$$\frac{dR_{sq}(h)}{dh} = -(y_0 - h) - (y_1 - h)$$

We know that the derivative has to be 0 at the local minima, therefore at  $h = 5$ , we have:

$$\begin{aligned}\frac{dR_{sq}(h)}{dh} &= -(3 - 5) - (y_1 - 5) = 0 \\ y_1 &= 7\end{aligned}$$

So we know that the dataset is 3, 7. Given all these information, we can calculate  $r_1$  with:

$$\begin{aligned}R_{sq}(5) &= \frac{1}{2}((y_0 - 5)^2 + (y_1 - 5)^2) \\ &= \frac{1}{2}((3 - 5)^2 + (7 - 5)^2) \\ &= \frac{1}{2}(4 + 4) = 4\end{aligned}$$

**3.** (10 points) The hyperbolic cosine function is defined as  $\cosh(x) = \frac{1}{2}(e^x + e^{-x})$ . In this problem, we aim to prove the convexity of this function using power series expansion.

**a)** (3 points) Prove that  $f(x) = x^n$  is convex if  $n$  is an even integer.

**Proof:**

**Solution:** Take the second derivative of  $f$ :

$$\begin{aligned}f'(x) &= nx^{n-1} \\ f''(x) &= n(n-1)x^{n-2}\end{aligned}$$

If  $n$  is even, then  $n-2$  must also be even, therefore  $f''(x) = n(n-1)x^{n-2}$  will always be a positive number. This means the second derivative of  $f(x)$  is always larger than 0 and therefore passes the second derivative test.

**b)** (2 points) Power series expansion is a powerful tool to analyze complicated functions. In power series expansion, a function can be written as an infinite sum of polynomial functions with certain

coefficients. For example, the exponential function can be written as:

$$e^x = \sum_{n=0}^{\infty} \frac{x^n}{n!} = 1 + x + \frac{x^2}{2} + \frac{x^3}{6} + \frac{x^4}{24} + \dots \quad (1)$$

where  $n!$  denotes the factorial of  $n$ , defined as the product of all positive integers up to  $n$ , i.e.  $n! = 1 \times 2 \times 3 \times \dots \times (n-1) \times n$ . Given the power series expansion of  $e^x$  above, write the power series expansion of  $e^{-x}$  and explicitly specify the first 5 terms, i.e., similar to the format of Equation 1:

$$e^{-x} = \sum_{n=0}^{\infty} \quad =$$

$$\textbf{Solution: } e^{-x} = \sum_{n=0}^{\infty} \frac{(-x)^n}{n!} = 1 - x + \frac{x^2}{2} - \frac{x^3}{6} + \frac{x^4}{24} + \dots$$

- c) (5 points) Using the conclusions you reached in **a)** and **b)**, prove that  $\cosh(x) = \frac{1}{2}(e^x + e^{-x})$  is convex.

**Proof:**

**Solution:** Given that:

$$e^x = \sum_{n=0}^{\infty} \frac{x^n}{n!} = 1 + x + \frac{x^2}{2} + \frac{x^3}{6} + \frac{x^4}{24} + \dots$$
$$e^{-x} = \sum_{n=0}^{\infty} \frac{(-x)^n}{n!} = 1 - x + \frac{x^2}{2} - \frac{x^3}{6} + \frac{x^4}{24} + \dots$$

We can add their power series expansion together, and we will obtain:

$$e^x + e^{-x} = \sum_{n=0}^{\infty} \frac{x^n}{n!} + \sum_{n=0}^{\infty} \frac{x^n}{n!}$$
$$= \sum_{n=0}^{\infty} \frac{(x)^n + (-x)^n}{n!}$$

Within this infinite sum, if  $n$  is even, then the negative sign in  $(-x)^n$  will disappear; if  $n$  is odd, then the negative sign in  $(-x)^n$  will be kept and travel out of the parenthesis. Therefore we have:

$$e^x + e^{-x} = \sum_{n=0}^{\infty} \frac{x^n + x^n}{n!} (\text{for even } n) + \sum_{n=0}^{\infty} \frac{x^n - x^n}{n!} (\text{for odd } n)$$
$$= \sum_{n=0}^{\infty} \frac{2x^n}{n!} (\text{for even } n)$$

Therefore,  $\cosh(x) = \frac{e^x + e^{-x}}{2}$  is a sum of  $x^n$  where  $n$  is even. Since we have already proved in a) that  $x^n$  are always convex for even  $n$ ,  $\cosh(x)$  is an infinite sum of convex function and therefore also convex.

4. (10pt) Note that we have two simplified closed form expressions for the estimated slope  $w$  in simple linear regression that you have already seen in discussions and lectures:

$$w = \frac{\sum_i (x_i - \bar{x})y_i}{\sum_i (x_i - \bar{x})^2} \quad (1)$$

$$w = \frac{\sum_i (y_i - \bar{y})x_i}{\sum_i (x_i - \bar{x})^2} \quad (2)$$

where we have dataset  $D = [(x_1, y_1), \dots, (x_n, y_n)]$ , sample means  $\bar{x} = \frac{1}{n} \sum_i x_i$ ,  $\bar{y} = \frac{1}{n} \sum_i y_i$ . Without further explanation,  $\sum_i$  means  $\sum_{i=1}^n$

a) (6pt) Are (1) and (2) equivalent? That is, is the following equality true? Prove or disprove it.

$$\sum_i (x_i - \bar{x})y_i = \sum_i (y_i - \bar{y})x_i$$

**Proof:**



**Solution:** True.

$$\begin{aligned}\sum_i (x_i - \bar{x})y_i &= \sum_i (y_i - \bar{y})x_i \\ \Leftrightarrow \sum_i x_i y_i - \bar{x} \sum_i y_i &= \sum_i x_i y_i - \bar{y} \sum_i x_i \\ \Leftrightarrow \bar{x} \sum_i y_i &= \bar{y} \sum_i x_i \\ \Leftrightarrow \frac{1}{n} \sum_i x_i \sum_i y_i &= \frac{1}{n} \sum_i y_i \sum_i x_i\end{aligned}$$

In fact, the least square estimator for slope is unique.

- b) (2pt) True or False: If the dataset shifted right by a constant distance  $a$ , that is, we have the new dataset  $D_a = (x_1 + a, y_1), \dots, (x_n + a, y_n)$ , then will the estimated slope  $w$  change or not?

True       False

**Solution:** False. By (1), the only term affecting  $w$  is  $x_i - \bar{x}$ , which is unchanged after shifting. Therefore,  $w$  is unchanged.

- c) (2pt) True or False: If the dataset shifted up by a constant distance  $b$ , that is, we have the new dataset  $D_b = [(x_1, y_1 + b), \dots, (x_n, y_n + b)]$ , then will the estimated slope  $w$  change or not?

True       False

**Solution:** False. By (2).

5. (6 points)

Suppose the following information is given for a linear regression:

$$X = \begin{bmatrix} 1 & 2 \\ 1 & -1 \end{bmatrix} \quad \vec{y} = \begin{bmatrix} a \\ b \end{bmatrix} \quad \vec{w}^* = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

Where  $X$  is the design matrix,  $\vec{y}$  is the observation vector, and  $\vec{w}^*$  is the optimal parameter vector. Solve for parameter  $a$  and  $b$  using the normal equation, show your work.

**Answer:**

**Supporting Work:**

**Solution:** Since  $\vec{w}^*$  is the optimal parameter vector, it must satisfy the Normal Equation:

$$X^T X \vec{w} = X^T \vec{y}$$

The left hand side of the equation will read:

$$X^T X \vec{w} = \begin{bmatrix} 1 & 1 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 1 & 5 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 4 \\ 11 \end{bmatrix}$$

The right hand side of the equation is given by:

$$X^T \vec{y} = \begin{bmatrix} 1 & 1 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} a + b \\ 2a - b \end{bmatrix}$$

By setting the left hand side and right hand side equal to each other, we will obtain the following system of equations:

$$\begin{bmatrix} 4 \\ 11 \end{bmatrix} = \begin{bmatrix} a + b \\ 2a - b \end{bmatrix}$$

So we obtained this set of equations:

$$\begin{aligned} 4 &= a + b \\ 11 &= 2a - b \end{aligned}$$

To solve this equation set, we can add them together:

$$\begin{aligned} 4 + 11 &= a + b + 2a - b \\ 3a &= 15 \\ a &= 5 \\ b &= -1 \end{aligned}$$