1. (6 points) Define the extreme mean (EM) of a dataset to be the average of its largest and smallest values. Let

$$f(x) = -3x + 4.$$

Show that for any dataset $x_1 \leq x_2 \leq \cdots \leq x_n$,

$$EM(f(x_1), f(x_2), \dots, f(x_n)) = f(EM(x_1, x_2, \dots, x_n))$$

2. (10 points) Consider a new loss function,

$$L(h,y) = e^{(h-y)^2}.$$

Given a dataset y_1, y_2, \ldots, y_n , let R(h) represent the empirical risk for the dataset using this loss function.

- a) (4 points) For the dataset $\{1, 3, 4\}$, calculate R(2). Simplify your answer as much as possible without a calculator.
- b) (6 points) For the same dataset {1,3,4}, perform one iteration of gradient descent on R(h), starting at an initial prediction of $h_0 = 2$ with a step size of $\alpha = \frac{1}{2}$. Show your work and simplify your answer.
- **3.** (8 points) Suppose you have a dataset

$$\{(x_1, y_1), (x_2, y_2), \dots, (x_8, y_8)\}$$

with n = 8 ordered pairs such that the variance of $\{x_1, x_2, \ldots, x_8\}$ is 50. Let *m* be the slope of the regression line fit to this data.

Suppose now we fit a regression line to the dataset

$$\{(x_1, y_2), (x_2, y_1), \dots, (x_8, y_8)\}$$

where the first two y-values have been swapped. Let m' be the slope of this new regression line.

If $x_1 = 3$, $y_1 = 7$, $x_2 = 8$, and $y_2 = 2$, what is the difference between the new slope and the old slope? That is, what is m' - m? The answer you get should be a number with no variables.

Hint: There are many equivalent formulas for the slope of the regression line. We recommend using the version of the formula without \overline{y} .

4. (9 points) Consider the dataset shown below.

a) (5 points) We want to use multiple regression to fit a prediction rule of the form

$$H(x^{(1)}, x^{(2)}, x^{(3)}) = w_0 + w_1 x^{(1)} x^{(3)} + w_2 (x^{(2)} - x^{(3)})^2.$$

Write down the design matrix X and observation vector \vec{y} for this scenario. No justification needed.

b) (4 points) For the X and \vec{y} that you have written down, let \vec{w} be the optimal parameter vector, which comes from solving the normal equations $X^T X \vec{w} = X^T \vec{y}$. Let $\vec{e} = \vec{y} - X \vec{w}$ be the error vector, and let e_i be the *i*th component of this error vector. Show that

$$4e_1 + e_2 + 4e_3 + e_4 = 0.$$