

## CIS 5526: Homework 3

Assigned: September 21<sup>st</sup>, 2022  
Due: September 29<sup>th</sup> at 5pm

### Homework Policy

All assignments are INDIVIDUAL! You may discuss the problems with your colleagues, but you must solve the homework by yourself. Please acknowledge all sources you use in the homework (papers, code or ideas from someone else). Assignments should be submitted in class on the day when they are due. No credit is given for assignments submitted at a later time, unless you have a medical problem.

### Problems: submit as a pdf file through canvas (you can take a photo of your handwriting but save it as pdf)

**Problem 1 (10 points).** For function  $f(x_1, x_2) = x_1^2 + 2x_2^2 + 3x_1x_2 + 2x_1 + 6$  show that you can write it as the quadratic form  $\mathbf{x}'\mathbf{Q}\mathbf{x} + \mathbf{b}\mathbf{x} + c$ , where  $\mathbf{x}$  is a  $2 \times 1$  vector,  $\mathbf{Q}$  is symmetric  $2 \times 2$  matrix  $\mathbf{Q} = \begin{bmatrix} 1 & 3/2 \\ 3/2 & 2 \end{bmatrix}$ ,  $\mathbf{b}$  is  $2 \times 1$  vector  $\mathbf{b} = (2, 0)$ , and  $c$  is a scalar  $c = 6$ .

**Problem 2 (10 points).** Starting from  $f(\mathbf{x}) = \mathbf{x}'\mathbf{Q}\mathbf{x} + \mathbf{b}\mathbf{x} + c$ , where  $\mathbf{x}$  is an  $n \times 1$  vector,  $\mathbf{Q}$  is an  $n \times n$  symmetric matrix,  $\mathbf{b}$  is an  $n \times 1$  vector, and  $c$  is a scalar, show that  $\nabla f(\mathbf{x}) = 2\mathbf{Q}\mathbf{x} + \mathbf{b}$ . **Hint:** Find partial derivatives of the quadratic form representation that uses sums:

$$f(\mathbf{x}) = \sum_{i=1}^n \sum_{j=1}^n Q_{ij} x_i x_j + \sum_{i=1}^n b_i x_i + c$$

**Problem 3 (10 points).** Using the gradient  $\nabla f(\mathbf{x}) = 2\mathbf{Q}\mathbf{x} + \mathbf{b}$  from problem 2, and assuming  $f(\mathbf{x})$  has the minimum, find  $\mathbf{x}$  that minimizes  $f(\mathbf{x})$  as a matrix formula.

**Problem 4 (10 points).** Starting from the definition for Mean Squared Error (MSE) for linear regression ( $N$  is the number of examples and  $M$  is number of features in the training data),

$$MSE(\mathbf{w}) = \frac{1}{N} \sum_{i=1}^N (y_i - \sum_{j=0}^M w_j x_{ij})^2$$

- Show that MSE can be written as a quadratic form  $MSE(\mathbf{w}) = \mathbf{w}'\mathbf{Q}\mathbf{w} + \mathbf{b}\mathbf{w} + c$ , where  $\mathbf{w}$  is an  $(M+1)$ -dimensional vector  $\mathbf{w} = (w_0, w_1, \dots, w_M)$ . Hint: treat  $x$  and  $y$  as constants, and treat  $w_j$ 's as variables.
- What are  $\mathbf{Q}$ ,  $\mathbf{b}$ ,  $c$ ?
- Using the answers from problem 3 and problem 4.b quickly derive  $\mathbf{w}$  that minimizes MSE of linear regression.
- We decided to use gradient descent algorithm to find  $\mathbf{w}$  that minimizes MSE. Derive the update formula in a vector form (**hint:** we showed it in class).

**Problem 5 (5 points).** Starting from the definition of expectation, derive mean and variance formula for a Bernoulli random variable  $X = \{0, 1\}$ , where  $P(X=1)=p$ .

**Problem 6 (5 points).** Starting from the definition of expectation, derive mean and standard deviation of a uniformly distributed random variable  $X$  in range  $[0, 1]$ .

**Problem 7 (5 points).** Starting from the definition of expectation prove that  $E(aX+bY) = aE(X) + bE(Y)$  is correct both in the case when  $X$  and  $Y$  are discrete random variables and when they are continuous random variables.

**Problem 8 (5 points).** Show that if two random variables  $X$  and  $Y$  are independent, their covariance has to be zero. The covariance between random variables  $X$  and  $Y$  is defined as  $E[(X-E(X))(Y-E(Y))]$ . Give an example where covariance between  $X$  and  $Y$  is zero, but they are not independent.

**Problem 9 (5 points).** Assuming that  $X \sim \text{Uni}(0, 1)$ , a uniform random variable in range between 0 and 1, find mean and variance of random variable  $Y = 3X + 2$ .

**Problem 10 (5 points).** The following problem is related to Bayes Theorem that is stated as  $P(A|B) = P(B|A)P(A)/P(B)$ , where  $A$  and  $B$  are random variables. Let us assume  $A$  is a binary random variable stating whether a woman has breast cancer and  $B$  is a binary random variable stating whether a woman tested positive on a mammogram. Let us assume the following background knowledge: 0.1% of women have breast cancer; 90% of women who have breast cancer test positive on mammograms; 8% of healthy women test positive on mammograms. What is the probability that a woman has cancer if she has a positive mammogram result? What is the probability that a woman has cancer if she has a negative mammogram result?