## CS 189 Spring 2022

## Introduction to Machine Learning Marvin Zhang

DISO

1 Back to Basics: Linear Algebra

Let  $X \in \mathbb{R}^{m \times n}$ . We do not assume that X has full rank.

- (a) Give the definition of the rowspace, columnspace, and nullspace of X.
- (b) Check the following facts:
  - (a) The rowspace of X is the columnspace of  $X^{T}$ , and vice versa.
  - (b) The nullspace of *X* and the rowspace of *X* are orthogonal complements.
  - (c) The nullspace of  $X^{T}X$  is the same as the nullspace of X. Hint: if v is in the nullspace of  $X^{T}X$ , then  $v^{T}X^{T}Xv = 0$ .
  - (d) The columnspace and rowspace of  $X^TX$  are the same, and are equal to the rowspace of X. *Hint: Use the relationship between nullspace and rowspace.*

## 2 Probability Review

There are *n* archers all shooting at the same target (bulls-eye) of radius 1. Let the score for a particular archer be defined to be the distance away from the center (the lower the score, the better, and 0 is the optimal score). Each archer's score is independent of the others, and is distributed uniformly between 0 and 1. What is the expected value of the worst (highest) score?

- (a) Define a random variable Z that corresponds with the worst (highest) score.
- (b) Derive the Cumulative Distribution Function (CDF) of *Z*.
- (c) Derive the Probability Density Function (PDF) of Z.
- (d) Calculate the expected value of Z.

## 3 Vector Calculus

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Below,  $\mathbf{x} \in \mathbb{R}^d$  means that  $\mathbf{x}$  is a  $d \times 1$  column vector with real-valued entries. Likewise,  $\mathbf{A} \in \mathbb{R}^{d \times d}$  means that  $\mathbf{A}$  is a  $d \times d$  matrix with real-valued entries. In this course, we will by convention consider vectors to be column vectors.

Consider  $\mathbf{x}, \mathbf{w} \in \mathbb{R}^d$  and  $\mathbf{A} \in \mathbb{R}^{d \times d}$ . In the following questions,  $\frac{\partial}{\partial \mathbf{x}}$  denotes the derivative with respect to  $\mathbf{x}$ , while  $\nabla_{\mathbf{x}}$  denotes the gradient with respect to  $\mathbf{x}$ .

Derive the following derivatives.

(a) 
$$\frac{\partial \mathbf{w}^{\top} \mathbf{x}}{\partial \mathbf{x}}$$
 and  $\nabla_{\mathbf{x}}(\mathbf{w}^{\top} \mathbf{x})$ 

(b) 
$$\frac{\partial (w^\top A x)}{\partial x}$$
 and  $\nabla_x (w^\top A x)$ 

(c) 
$$\frac{\partial (\mathbf{w}^{\top} \mathbf{A} \mathbf{x})}{\partial \mathbf{w}}$$
 and  $\nabla_{\mathbf{w}} (\mathbf{w}^{\top} \mathbf{A} \mathbf{x})$ 

(d) 
$$\frac{\partial (w^\top A x)}{\partial A}$$
 and  $\nabla_A (w^\top A x)$ 

(e) 
$$\frac{\partial (\mathbf{x}^{\top} \mathbf{A} \mathbf{x})}{\partial \mathbf{x}}$$
 and  $\nabla_{\mathbf{x}} (\mathbf{x}^{\top} \mathbf{A} \mathbf{x})$ 

(f) 
$$\nabla_{\mathbf{x}}^2(\mathbf{x}^{\mathsf{T}}\mathbf{A}\mathbf{x})$$

<sup>&</sup>lt;sup>1</sup>Good resources for matrix calculus are:

<sup>•</sup> The Matrix Cookbook: https://www.math.uwaterloo.ca/~hwolkowi/matrixcookbook.pdf

<sup>•</sup> Wikipedia: https://en.wikipedia.org/wiki/Matrix\_calculus

<sup>•</sup> Khan Academy: https://www.khanacademy.org/math/multivariable-calculus/multivariable-derivatives

<sup>•</sup> YouTube: https://www.youtube.com/playlist?list=PLSQl0a2vh4HC5feHa6Rc5c0wbRTx56nF7.