### IFT 6390 Fundamentals of Machine Learning Ioannis Mitliagkas

### Homework 0

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# 1 Theoretical Part [5 points]

- 1. [1 points] Let X be a random variable representing the outcome of a single roll of a 6-sided dice. Show the steps for the calculation of i) the expectation of X and ii) the variance of X.
- 2. [1 points] Let  $u, v \in \mathbb{R}^d$  be two vectors and let  $A \in \mathbb{R}^{n \times d}$  be a matrix. Give the formulas for the euclidean norm of u, for the euclidean inner product (aka dot product) between u and v, and for the matrix-vector product Au.
- 3. [1 points] Consider the two algorithms below. What do they compute and which algorithm is faster?

Observez les deux algorithms ci-dessous. Que calculent-ils et lequel est le plus rapide ?

$$\begin{aligned} \mathbf{ALGO1}(\mathbf{n}) & \mathbf{ALGO2}(\mathbf{n}) \\ \mathbf{result} &= 0 & \mathbf{return} \ (n+1)*n/2 \\ \mathbf{for} \ i &= 1 \dots n \\ \mathbf{result} &= \mathbf{result} + i \\ \mathbf{return} \ \mathbf{result} \end{aligned}$$

4. [1 points] Give the step-by-step derivation of the following derivatives:

i) 
$$\frac{df}{dx} = ?$$
, where  $f(x, \beta) = x^2 \exp(-\beta x)$ 

ii) 
$$\frac{df}{d\beta} = ?$$
, where  $f(x, \beta) = x \exp(-\beta x)$ 

iii) 
$$\frac{df}{dx} = ?$$
, where  $f(x) = \sin(\exp(x^2))$ 

5. [1 points] Let  $X \sim N(\mu, 1)$ , that is the random variable X is distributed according to a Gaussian with mean  $\mu$  and standard deviation 1. Show how you can calculate the second moment of X, given by  $\mathbb{E}[X^2]$ .

## 2 Solutions

1.

$$X = \{1, 2, 3, 4, 5, 6\}$$

Assuming that the dice is fair, each of the 6 outcomes will be equally likely. Hence,

$$p(X = x) = 1/6 \quad \forall x \in X$$

(a) Expectation of X

$$\mu = E[X] = \sum_{x=1}^{6} xp(x)$$
$$= \frac{1}{6} \sum_{x=1}^{6} x$$
$$= \frac{21}{6} = \frac{7}{2} = 3.5$$

(b) Variance of X

$$\begin{split} V[X] &= E[X^2] - E[X]^2 \\ &= \sum_{x=1}^6 x^2 p(x) - \left(\frac{7}{2}\right)^2 \\ &= \frac{1}{6} \sum_{x=1}^6 x^2 - \frac{49}{4} \\ &= \frac{1}{6} \cdot \frac{6 \times 7 \times 13}{6} - \frac{49}{4} \\ &= \frac{35}{12} = 2.917 \end{split}$$

2. (a) Euclidean norm of u

$$||u||_2 = \sqrt{\sum_{i=1}^d u_i^2}$$

(b) Euclidean inner product of u & v

$$u \cdot v = \sum_{i=1}^{d} u_i v_i$$

(c) matrix-vector product Au

$$Au = C$$
 where  $c_i = \sum_{l=1}^{d} a_{il}u_l$   $i \in [1, n]$ 

3. Both the algorithms compute the sum of first n natural numbers. **ALGO1** has loop which runs n times, hence has time complexity O(n).

**ALGO2** has a single operation, hence has time complexity O(1). Therefore, **ALGO2** is faster.

4. (a)

$$f(x,\beta) = x^2 e^{-\beta x}$$
$$\frac{df}{dx} = \frac{d}{dx} (x^2 e^{-\beta x})$$
$$= e^{-\beta x} (2x - \beta)$$

(b)

$$f(x,\beta) = xe^{-\beta x}$$
$$\frac{df}{d\beta} = \frac{d}{d\beta}(xe^{-\beta x})$$
$$= -x^2e^{-\beta x}$$

(c)

$$f(x) = \sin e^{x^2}$$
$$\frac{df}{dx} = \frac{d}{dx}(\sin e^{x^2})$$
$$= 2xe^{x^2}\cos e^{x^2}$$

5.

$$V[X] = \sigma^2 = 1^2 = 1$$

Also,

$$V[X] = E[(X - \mu)^{2}]$$

$$= E[X^{2} + \mu^{2} - 2X\mu]$$

$$= E[X^{2}] - 2\mu E[X] + E[\mu^{2}]$$

$$= E[X^{2}] - 2[E[X]]^{2} + [E[X]]^{2}$$

$$= E[X^{2}] - [E[X]]^{2}$$

$$\Rightarrow E[X^{2}] = V[X] + [E[X]]^{2}$$

$$= 1 + \mu^{2}$$