

Due date: 22 Dec (2023)

### Part One:

- 1- Given the instance-feature matrix  $X_{N \times D}$ , and a normal direction  $\vec{W}_D$  on which instances are projected to compute the new feature  $\vec{f}_N$ . Compute the variance of  $\vec{f}$ .
- 2- Given grades of a student in  $D$  exams as a vector  $\vec{g}_D$ , it is desired to assign a positive weight ( $w$ ) to each exam in order to maximize the final grade  $\sum g_i \cdot w_i$ . However, it is desired to limit the weights such that  $\sum w_i^2 = 1$  (the weight vector should be normal). What is the best weight vector?
- 3- Having  $N$  instances with  $D$  features in the matrix  $X_{N \times D}$ , it is desired to find a representative  $D$ -vector  $\vec{r}$  in order to minimize degree  $q^{th}$  of norm of distances between instances and the representative. Distance is also computed based on the norm of degree  $p$  (i.e.,  $\|\vec{x}_n - \vec{r}\|_p$ ).
  - a. Model this minimization problem in the form of linear algebra.
  - b. Find the best solution for  $q = p = 1$  and  $q = \infty, p = 2$ .
  - c. Determine  $q$  and  $p$  for which,  $r_i$  is Mode of feature  $\vec{X}_i$ .
- 4- A dice is tossed  $N$  times and the sides 1, 2, ..., & 6 are observed  $N_1, N_2, \dots, N_6$  times, respectively. If the probability of observing  $i^{th}$  side is  $p_i$ , you know that probability of this set of observations is

$$\prod_{i=1}^6 p_i^{N_i}$$

It is desired to model the problem of finding maximum probability in the form of linear algebra with both objective function and constraint(s).

Assume that  $\lg(\vec{a}) = [\lg a_1, \lg a_2, \dots, \lg a_{|\vec{a}|}]^T$ .

- 5- Having a graph with adjacency matrix  $M$  where  $m_{i,j}$  is the probability of moving from city  $i$  to  $j$ . What does  $M^2$  mean? What is Column-Row interpretation of  $M^2$ ?
- 6- Analyze the covariance matrix from viewpoints of elementwise, column-wise and row-wise matrix production. Hint:  $\Sigma = X^T X$  where,  $X$  is the demeaned instance-feature matrix.

## Part Two:

In this part, you will create a Python program to minimize different error norms in the context of linear regression. You must generate a set of data points consisting of  $x$  and their corresponding  $y$  values. The predicted value,  $\hat{y}$ , is defined as  $\hat{y} = x * \alpha + \beta$ , where  $\alpha$  and  $\beta$  are constants.

Your task is to minimize several error norms, which include:

- L0 Norm
- L1 Norm
- L2 Norm
- Infinity Norm

The signed error is calculated using the formula:

$$\text{Signed Error} = \hat{y} - y.$$

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### Submission Guidelines:

- The answers to the exercises of part one must be in PDF file.
  - Allowed programming language: Python
  - Your implementation reports should be in a PDF file including an explanation of your approach, key points of your implementation, and report of your final results.
  - You should upload your submissions at [Quera](#). All of the files should be saved in a ZIP file named in this format: "Lastname-SudentNumber.zip".  
Ex.: "Zamani-40230401.zip"
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