

**HW1**Deadline: OCT 28thEmail: mlfall2023@gmail.com

About submission:

- File's naming format:
 - For Homeworks: "HWX _Student NO."
 - Exe: HW2 _400613198.pdf
 - For MATLAB Homeworks: "PRJX _Student NO."
 - Exe: PRJ1 _400613198.pdf
- About late submissions: %10 of daily reduction of the HW mark.

- 1) We are tasked with estimating the random variable y after observing the random variable x . If the loss function is defined as:

$$\ell(y, \hat{y}) = e^{|y - \hat{y}|}$$

- a. Find the optimal estimator and simplify it as much as possible.
 - b. Assuming $p(y|x) = 2e^{-2y}u(y)$ (where $u(\cdot)$ is Denoted as unit step function), what will be the optimal estimator's prediction for y ?
 - c. Having the same assumption as part b, find \hat{y}_{MMSE} and \hat{y}_{MAP} , the output of MMSE and MAP estimator respectively, for y .
- 2) Check the eigenvalues and eigenvectors of A, B, A^2, A^{-1} and $A + B$. What is the relation between them?

$$A = \begin{bmatrix} 0 & 2 \\ 1 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} 1 & 1 \\ 0 & 3 \end{bmatrix}$$

- 3) For what numbers c and d are S and T positive definite? Test their 3 determinants:

$$S = \begin{bmatrix} c & 1 & 1 \\ 1 & c & 1 \\ 1 & 1 & c \end{bmatrix}, \quad T = \begin{bmatrix} 1 & 2 & 3 \\ 2 & d & 4 \\ 3 & 4 & 5 \end{bmatrix}$$

- 4) When S and T are symmetric positive definite, ST might not even be symmetric. But its eigenvalues are still positive. Start from $STx = \lambda x$ and take dot products with Tx . Then prove $\lambda > 0$.

- 5) Find the eigenvalues and unit eigenvectors of $A^T A$ and AA^T . Keep each $Av = \sigma u$. Then construct the singular value decomposition and verify that A equals $U\Sigma V^T$.

Fibonacci matrix $A = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$

- 6) Under which condition $f(x) = x^T Ax$ is a convex function?
- 7) For $f(x) = h(g(x))$, prove each of these statements:
- a. f is convex, if h is convex and nondecreasing and g is convex
 - b. f is convex, if h is convex and nonincreasing and g is concave
 - c. f is concave, if h is concave and nondecreasing and g is concave
 - d. f is concave, if h is concave and nonincreasing and g is convex
- 8) Newton's method with fixed step size $t = 1$ can diverge if the initial point is not close to x^* . In this problem we consider two examples.
- a. $f(x) = \log(e^x + e^{-x})$ has a unique minimizer $x^* = 0$. Run Newton's method with fixed step size $t = 1$, starting at $x^{(0)} = 1$ and at $x^{(0)} = 1.1$.
 - b. $f(x) = -\log x + x$ has a unique minimizer $x^* = 1$. Run Newton's method with fixed step size $t = 1$, starting at $x^{(0)} = 3$.
 - c. Solve a and b using gradient descent.