

DIGITAL SIGNAL PROCESSING

THEORY

(Time: 70 minutes. Mark: 3/8)

T1.- We want to estimate a random vector \mathbf{s} from a vector of observations \mathbf{x} which is obtained as:

$$\mathbf{x} = \mathbf{H} \mathbf{s} + \mathbf{r}$$

where \mathbf{H} is a known matrix, \mathbf{r} is a random noise vector with distribution $G(\mathbf{0}, \mathbf{V}_r \mathbf{I})$, and \mathbf{s} is the random vector to be estimated, which follows a distribution $G(E\{\mathbf{s}\}, \mathbf{V}_s)$.

Knowing also that \mathbf{s} and \mathbf{r} are independent random vectors:

- a) Obtain the ML estimator of \mathbf{s} , $\hat{\mathbf{s}}_{\text{ml}}$.
- b) Is $\hat{\mathbf{s}}_{\text{ml}}$ an unbiased estimator? Justify your answer.
- c) As it is known, the MSE estimator of \mathbf{s} is given by:

$$\hat{\mathbf{s}}_{\text{mse}} = (\mathbf{H}^T \mathbf{H} + \mathbf{V}_r \mathbf{V}_s^{-1})^{-1} \mathbf{H}^T \mathbf{x}$$

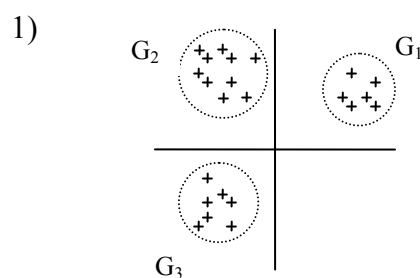
Calculate the bias of $\hat{\mathbf{s}}_{\text{mse}}$, and explain under which conditions the bias tends to 0.

(25 min; 1 p)

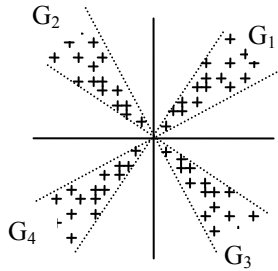
T2.- The following expressions:

- a) $\|\mathbf{x} - \mathbf{y}\|^2$
- b) $\|\mathbf{x}\| \|\mathbf{y}\|$
- c) $(\|\mathbf{x}\| - \|\mathbf{y}\|)^2$
- d) $\|\mathbf{x}\|^2 + \|\mathbf{y}\|^2$
- e) $\frac{\mathbf{x}^T \mathbf{y}}{\|\mathbf{x}\| \|\mathbf{y}\|}$

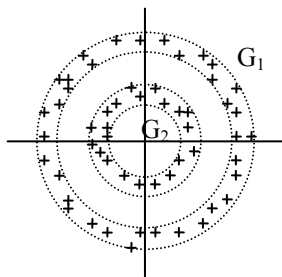
are considered to be used as similarity criteria in a clustering algorithm. Explain which of them you would select to obtain the clusters depicted in the figures below.



2)



3)



(20 min; 1p)

T3.- We have a set of K samples, $\{x^{(k)}\}_{k=1}^K$, which have been independently drawn from a random variable x whose pdf is given by:

$$p_x(x) = \frac{1}{bx^2} \exp\left(-\frac{1}{bx}\right) u(x)$$

with $b > 0$.

- Express \hat{b}_{ml} as a function of the available samples.
- Verify that random variable $y = 1/x$ follows an unilateral exponential pdf, $p_y(y)$, and obtain the mean value of such a distribution.
- In the light of previous results, is \hat{b}_{ml} an unbiased estimator?

(25 min; 1 p)

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PROBLEMS

(Time: 135 minutes. Mark: 5/8)

P1.- In a binary classification problem formulated under the ML criterion, the likelihood of the observations are:

$$p(x|H_0) = \begin{cases} 2(1-x), & 0 < x < 1 \\ 0, & \text{otherwise} \end{cases}$$

$$p(x|H_1) = \begin{cases} 1/a, & 0 < x < a \\ 0, & \text{otherwise} \end{cases}$$

where $a > 1$ is a deterministic parameter.

- a) Design the optimum decider assuming the value of a is known.
- b) Assume now the value of a is unknown. Following a minimax strategy, decisions are taken by comparison with a threshold x_u^* that is fixed to minimize the maximum expected cost (corresponding to the ML criterion), i.e.,

$$x_u^* = \arg \left\{ \min_{x_u} \left\{ \max_a \bar{C}_{ML}(x_u, a) \right\} \right\}$$

where x_u is a generic threshold for the decision

$$\begin{matrix} D_1 \\ x > \\ < \\ D_0 \end{matrix} x_u$$

Obtain the value of threshold x_u^* .

- c) Compute the increment of the expected cost (ML) when applying the minimax strategy, in comparison to the cost that would be obtained if parameter a were known.

P2.- Consider the following family of cost functions:

$$C_N(s, \hat{s}) = \frac{1}{N+1} \hat{s}^{N+1} + \frac{1}{N(N+1)} s^{N+1} - \frac{1}{N} s \hat{s}^N$$

where N is an odd and non-negative integer.

- a) Analyze the validity of the invariance principle for such a family of cost functions, and for an arbitrary joint distribution of s and x (where x is the observation).
- b) Assuming that

$$p(s, x) = \frac{1}{\lambda x} \exp\left(-\frac{s}{x} - \frac{x}{\lambda}\right) u(s) u(x), \quad \lambda > 0$$

Obtain the estimator which incurs in a minimum expected cost.

- c) Obtain the minimum mean cost.
- d) Determine now the coefficient w that minimizes the mean cost of an estimator

$$\hat{s} = wx^m$$

where m is a positive integer.

- e) Noting the dependence of w on the value of N , state if this result is in contradiction with the invariance principle. Justify your answer.

Hint: $\int_0^{\infty} x^N \exp(-x) dx = N!$

(75 min; 2.5 p)