

## Lecture 18: 06 November 2023

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## 18.1 DME With Known Distribution on Inputs

Consider the DME problem where users  $U_i$  have samples  $X_i$  which are iid from a distribution with unknown mean. Suppose that each user sends one bit of information  $Y_i$ . We require to minimize the maximum MSE, defined as

$$\text{MaxMSE} \triangleq \max_{\mu \in \mathbb{M}} |\hat{\mu} - \mu|^2. \quad (18.1)$$

### 18.1.1 Location Family

Let  $f_X$  denote a pdf with zero mean. Then, the location family corresponding to  $f_X$  is given by.

$$\mathcal{L}(f_X) \triangleq \{f_{X,\mu}(x) = f_X(x - \mu)\}, \mu \in \mathbb{M}. \quad (18.2)$$

Thus,  $cL(f_X)$  is a collection of pdfs where  $\mu$  is the *only* unknown variable.

### 18.1.2 DME From MSE

In this situation, we have

$$\text{MSE}(\mu) = \frac{1}{m^2} \mathbb{E} \left[ \sum_{i=1}^m (\hat{X}_i - \mu) \right]^2 \quad (18.3)$$

$$= \frac{1}{m^2} \mathbb{E} \left[ \sum_{i=1}^m (\hat{X}_i - \mu)^2 + \sum_{i=1}^m \sum_{\substack{j=1 \\ j \neq i}}^m (\hat{X}_i - \mu)(\hat{X}_j - \mu) \right] \quad (18.4)$$

$$= \frac{1}{m} \mathbb{E} \left[ (\hat{X}_1 - \mu)^2 \right] + \frac{m-1}{m} \left( \mathbb{E} [\hat{X}_1 - \mu] \right)^2. \quad (18.5)$$

Consider an encoder

$$Y_i = \begin{cases} 1 & X_i \leq \theta \\ 0 & X_i > \theta \end{cases}. \quad (18.6)$$

Thus, if  $F_X$  is invertible,

$$\frac{1}{m} \sum_{i=1}^m Y_i \xrightarrow{\mathbb{P}} F_X(\theta - \mu) \quad (18.7)$$

$$\implies \hat{\mu} = \theta - F_X^{-1} \left( \frac{1}{m} \sum_{i=1}^m Y_i \right) \quad (18.8)$$