

## Lecture 21: 13 November 2023

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## 21.1 General Single-Bit Quantization Schemes

Any one-bit quantization scheme takes the form

$$Y_i = \mathbb{1}_{\{X_i \in \mathcal{A}_i\}}. \quad (21.1)$$

For a fully distributed scheme, the  $\mathcal{A}_i$  are fixed and independent of the samples  $X_i$ . However, in an interactive scheme the  $\mathcal{A}_i$  can depend on the  $X_i$ .

### 21.1.1 A Potential One-Bit Scheme

The mean of a random variable  $X$  can be written in another form as

$$\mathbb{E}[X] = \int_{-\infty}^{\infty} x f_X(x) dx \quad (21.2)$$

$$= \int_0^{\infty} \Pr(X > t) dt - \int_{-\infty}^0 \Pr(X < t) dt \quad (21.3)$$

$$= \int_0^{\infty} \mathbb{E}[\mathbb{1}_{\{X > t\}}] dt - \int_{-\infty}^0 \mathbb{E}[\mathbb{1}_{\{X < t\}}] dt \quad (21.4)$$

$$\approx \sum_{j=0}^{\alpha} \mathbb{E}[\mathbb{1}_{\{X > j\Delta\}}] \Delta - \sum_{j=-\beta}^0 \mathbb{E}[\mathbb{1}_{\{X < j\Delta\}}] \Delta \quad (21.5)$$

Thus, we can create a potential scheme in which the  $j^{\text{th}}$  user transmits  $\mathbb{1}_{\{X_j < j\Delta\}}$ .

## 21.2 Threshold Based Schemes

In threshold-based schemes, we have  $\mathcal{A}_i = (-\infty, \theta_i]$ , and  $Y_i = \mathbb{1}_{\{X_i \leq \theta_i\}}$ . We state the following claim.

**Theorem 21.1.** *For any threshold-based scheme and symmetric log-concave  $f_X$ , we have*

$$\sqrt{m}(\hat{\mu} - \mu) \xrightarrow{d} \frac{1}{\kappa(\mu)} \quad (21.6)$$

as  $m \rightarrow \infty$ . Further,

$$\lim_{m \rightarrow \infty} \text{MSE} \rightarrow \frac{1}{\kappa(\mu)} \quad (21.7)$$

$$\kappa(\mu) \triangleq \int_{-\infty}^{\infty} \frac{f_X^2(t - \mu)}{F_X(t - \mu) F_X(\mu - t)} \lambda(t) dt \quad (21.8)$$

$$\Lambda_m(\tau) \triangleq \frac{1}{m} \sum_{i=1}^m \mathbb{1}_{\theta_i \leq \tau} \quad (21.9)$$

$$\lambda(\tau) \triangleq \frac{d}{d\tau} \lim_{m \rightarrow \infty} \Lambda_m(\tau) = \frac{d\Lambda(\tau)}{d\tau}. \quad (21.10)$$

Note that  $\Lambda_m(z)$  is the approximation of a CDF. Further, if  $\theta_i$  is uniform over  $[-\beta, \alpha]$ , then

$$\kappa(\mu) = \int_{-\beta}^{\alpha} \frac{f_X^2(t - \mu)}{F_X(t - \mu) F_X(\mu - t)} \frac{1}{\alpha + \beta} dt. \quad (21.11)$$