

Lecture 14: 11 October 2023

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14.1 Correlated Quantization

We consider stochastic rounding in the case of n users. Let a random permutation π of $\{0, 1, \dots, n-1\}$ be shared among the users. User i is given the number π_i (here the users are 0-indexed). Then, each user generates $\gamma_i \sim \text{Unif}[0, \frac{1}{m}]$. Note that γ_i is *private randomness*. Define

$$U_i \triangleq \frac{\pi_i}{m} + \gamma_i. \quad (14.1)$$

The CDF of U_i , for $u \in [0, 1]$, is

$$\Pr(U_i \leq u) = \Pr\left(\frac{\pi_i}{m} + \gamma_i \leq u\right) \quad (14.2)$$

$$= \Pr(\pi_i + m\gamma_i \leq mu) \quad (14.3)$$

$$= \Pr(\pi_i < \lfloor mu \rfloor) + \Pr(\pi_i = \lfloor mu \rfloor, m\gamma_i \leq \mu - \lfloor mu \rfloor) \quad (14.4)$$

$$= \frac{\lfloor mu \rfloor}{m} + \frac{mu - \lfloor mu \rfloor}{m} = u \quad (14.5)$$

and hence $U_i \sim \text{Unif}[0, 1]$.

Now, each user i transmits

$$Y_i \triangleq \mathbb{1}_{\{U_i < x_i\}}. \quad (14.6)$$

Clearly, this scheme is unbiased because $\mathbb{E}[Y_i] = x_i$. We now present a claim for the MSE of this scheme.

Claim 14.1. *For the above scheme, the MSE is upper bounded by $\frac{3}{m}\sigma_{md} + \frac{12}{m^2}$, where*

$$\sigma_{md} \triangleq \frac{1}{m} \sum_{i=1}^m |x_i - \bar{x}| \leq \sqrt{\frac{1}{m} \sum_{i=1}^m (x_i - \bar{x})^2}. \quad (14.7)$$

Proof. Define

$$z_i \triangleq \frac{\lfloor mx_i \rfloor}{m} \quad (14.8)$$

$$Y_i \triangleq Q_i(x_i) \quad (14.9)$$

$$Q_i(t) \triangleq \mathbb{1}_{\{U_i < t\}} \quad (14.10)$$

The MSE is given by

$$\text{MSE} = \mathbb{E} \left[\left(\frac{1}{m} \sum_{i=1}^m Y_i - \frac{1}{m} \sum_{i=1}^m x_i \right)^2 \right] \quad (14.11)$$

$$= \frac{1}{m^2} \mathbb{E} \left[\left(\sum_{i=1}^m (Y_i - x_i) \right)^2 \right] \quad (14.12)$$

$$= \frac{1}{m^2} \mathbb{E} \left[\left(\sum_{i=1}^m ((x_i - z_i) + (z_i - Q_i(z_i)) + (Q_i(z_i) - Q_i(x_i))) \right)^2 \right] \quad (14.13)$$

$$\leq \frac{3}{m^2} \left\{ \mathbb{E} \left[\left(\sum_{i=1}^m (x_i - z_i) \right)^2 \right] + \mathbb{E} \left[\left(\sum_{i=1}^m (z_i - Q_i(z_i)) \right)^2 \right] + \mathbb{E} \left[\left(\sum_{i=1}^m (Q_i(z_i) - Q_i(x_i)) \right)^2 \right] \right\} \quad (14.14)$$

where we use in (14.13) the inequality

$$(\alpha + \beta + \gamma)^2 \leq 3(\alpha^2 + \beta^2 + \gamma^2). \quad (14.15)$$

for reals α, β, γ .

Note that

$$x_i - z_i = x_i - \frac{\lfloor mx_i \rfloor}{m} = \frac{mx_i - \lfloor mx_i \rfloor}{m} \leq \frac{1}{m}, \quad (14.16)$$

hence

$$\mathbb{E} \left[\left(\sum_{i=1}^m (x_i - z_i) \right)^2 \right] \leq 1. \quad (14.17)$$

Now,

$$\mathbb{E} \left[\left(\sum_{i=1}^m z_i - Q_i(z_i) \right)^2 \right] = \sum_{i=1}^m \mathbb{E} \left[(z_i - Q_i(z_i))^2 \right] + \sum_{i=1}^m \sum_{\substack{j=1 \\ j \neq i}}^m \mathbb{E} [(z_i - Q_i(z_i))(z_j - Q_j(z_j))] \quad (14.18)$$

$$= \sum_{i=1}^m \mathbb{E} \left[(z_i - Q_i(z_i))^2 \right] + \sum_{i=1}^m \sum_{\substack{j=1 \\ j \neq m}}^m (\mathbb{E} [Q_i(z_i) Q_j(z_j)] - z_i z_j) \quad (14.19)$$

Note that

$$Q_i(z_i) = \mathbb{1}_{\{\pi_i + m\gamma_i < mz_i\}} \quad (14.20)$$

$$= \mathbb{1}_{\{\pi_i < mz_i\}} \quad (14.21)$$

and

$$\mathbb{E}[Q_i(z_i) Q_j(z_j)] = \Pr(\pi_i < mz_i, \pi_j < mz_j) \quad (14.22)$$

$$= \min\{z_i, z_j\} \frac{m \max\{z_i, z_j\} - 1}{m - 1} \quad (14.23)$$

$$= \frac{mz_i z_j - \min\{z_i, z_j\}}{m - 1} \quad (14.24)$$

$$= \frac{mz_i z_j}{m - 1} - \frac{1}{m - 1} \left(\frac{z_i + z_j}{2} - \frac{|z_i - z_j|}{2} \right). \quad (14.25)$$

Using (14.25) in (14.19),

$$\mathbb{E} \left[\left(\sum_{i=1}^m z_i - Q_i(z_i) \right)^2 \right] = \sum_{i=1}^m z_i (1 - z_i) + \sum_{i=1}^m \sum_{\substack{j=1 \\ j \neq i}}^m \left(\frac{z_i z_j}{m - 1} - \frac{z_i + z_j}{2(m - 1)} + \frac{|z_i - z_j|}{2(m - 1)} \right) \quad (14.26)$$

$$= \sum_{i=1}^m z_i (1 - z_i) + \frac{(\sum_i z_i)^2 - \sum_i z_i^2}{m - 1} - \sum_i z_i + \sum_{i=1}^m \sum_{\substack{j=1 \\ j \neq i}}^m \frac{|z_i - z_j|}{2(m - 1)} \quad (14.27)$$

$$\leq \frac{(\sum_i z_i)^2}{m} - \sum_i z_i^2 + \sum_{i=1}^m \sum_{\substack{j=1 \\ j \neq i}}^m \frac{|z_i - z_j|}{2(m - 1)} \quad (14.28)$$

$$\leq \sum_{\substack{j=1 \\ j \neq i}}^m \frac{|z_i - z_j|}{2(m - 1)} \quad (14.29)$$

$$\leq \sum_{\substack{j=1 \\ j \neq i}}^m \frac{|x_i - x_j| + \frac{1}{m}}{2(m - 1)} \quad (14.30)$$

$$\leq \sum_{i=1}^m \sum_{\substack{j=1 \\ j \neq i}}^m \frac{|x_i - x_j|}{2(m - 1)} + \frac{1}{2}. \quad (14.31)$$

Finally, note that

$$Q_i(z_i) = 1 \implies U_i < z_i \leq x_i \implies Q_i(x_i) = 1 \quad (14.32)$$

and thus

$$\mathbb{E} \left[\left(\sum_{i=1}^m (Q_i(z_i) - Q_i(x_i)) \right)^2 \right] \leq 2. \quad (14.33)$$

Putting (14.17), (14.31) and (14.33) together,

$$\text{MSE} \leq \frac{3}{m^2} \left(4 + \sum_{i=1}^m \sum_{\substack{j=1 \\ j \neq i}}^m \frac{|x_i - x_j|}{2(m-1)} \right) \quad (14.34)$$

$$\leq \frac{12}{m^2} + \frac{3}{m} \sigma_{md} \quad (14.35)$$

as desired. It is left as an exercise to show that

$$\sum_{i=1}^m \sum_{\substack{j=1 \\ j \neq i}}^m \frac{|x_i - x_j|}{2m(m-1)} \leq \sigma_{md} = \frac{1}{m} \sum_{i=1}^m |x_i - \bar{x}|. \quad (14.36)$$

□