

Lecture 14: 11 October 2023

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Note: The discussion started in Lecture 14 continues on in Lecture 15. This document contains the content of both lectures clubbed together for the sake of continuity.

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,

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

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

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

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

□