## EE6367: Topics in Data Storage and Communications

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## 13.1 Stochastic Rounding

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

$$U_i \triangleq \frac{\pi_i}{m} + \gamma_i. \tag{13.1}$$

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

$$\Pr\left(U_i \leqslant u\right) = \Pr\left(\frac{\pi_i}{m} + \gamma_i \leqslant u\right) \tag{13.2}$$

$$=\Pr\left(\pi_i + m\gamma_i \leqslant mu\right) \tag{13.3}$$

$$= \Pr\left(\pi_i < \lfloor mu \rfloor\right) + \Pr\left(\pi_i = \lfloor mu \rfloor, \ m\gamma_i \leqslant \mu - \lfloor mu \rfloor\right)$$
(13.4)

$$=\frac{\lfloor mu\rfloor}{m} + \frac{mu - \lfloor mu\rfloor}{m} = u \tag{13.5}$$

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

Now, each user i transmits

$$Y_i \triangleq \mathbb{1}_{\{U_i \leqslant x_i\}}.\tag{13.6}$$

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

Claim 13.1. For the above scheme, the MSE is upper bounded by  $\frac{3}{m}\sigma_{md} + \frac{12}{m^2}$ , where

$$\sigma_{md} \triangleq \frac{1}{m} \sum_{i=1}^{m} |x_i - \bar{x}| \le \sqrt{\frac{1}{m} \sum_{i=1}^{m} (x_i - \bar{x})^2}.$$
 (13.7)

Proof. Define

$$z_i \triangleq \frac{\lfloor mx_i \rfloor}{m} \tag{13.8}$$

$$Y_i \triangleq Q_i\left(x_i\right) \tag{13.9}$$

$$Q_i(t) \triangleq \mathbb{1}_{\{U_i < t\}} \tag{13.10}$$

The MSE is given by

$$MSE = \mathbb{E}\left[\left(\frac{1}{m}\sum_{i=1}^{m}Y_{i} - \frac{1}{m}\sum_{i=1}^{m}x_{i}\right)^{2}\right]$$
(13.11)

$$= \frac{1}{m^2} \mathbb{E}\left[\left(\sum_{i=1}^m (Y_i - x_i)\right)^2\right]$$
(13.12)

$$= \frac{1}{m^2} \mathbb{E} \left[ \left( \sum_{i=1}^m \left( (x_i - z_i) + (z_i - Q_i(z_i)) + (Q_i(z_i) - Q_i(x_i)) \right)^2 \right]$$
 (13.13)

$$\leq \frac{3}{m^{2}} \left\{ \mathbb{E} \left[ \left( \sum_{i=1}^{m} (x_{i} - z_{i}) \right)^{2} \right] + \mathbb{E} \left[ \left( \sum_{i=1}^{m} (z_{i} - Q_{i}(z_{i})) \right)^{2} \right] + \mathbb{E} \left[ \left( \sum_{i=1}^{m} (Q_{i}(z_{i}) - Q_{i}(x_{i})) \right)^{2} \right] \right\}$$
(13.14)
$$(13.15)$$

where we use in (13.13) the inequality

$$(\alpha + \beta + \gamma)^2 \leqslant 3(\alpha^2 + \beta^2 + \gamma^2). \tag{13.16}$$

for reals  $\alpha, \beta, \gamma$ .

Note that

$$x_i - z_i = x_i - \frac{\lfloor mx_i \rfloor}{m} = \frac{mx_i - \lfloor mx_i \rfloor}{m} \leqslant \frac{1}{m}, \tag{13.17}$$

hence

$$\mathbb{E}\left[\left(\sum_{i=1}^{m} (x_i - z_i)\right)^2\right] \leqslant 1. \tag{13.18}$$

Now,

$$\mathbb{E}\left[\left(\sum_{i=1}^{m} z_{i} - Q_{i}(z_{i})\right)^{2}\right] = \sum_{i=1}^{m} \mathbb{E}\left[\left(z_{i} - Q_{i}(z_{i})\right)^{2}\right]$$
(13.19)