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# HOMWORK 8. MIXED-PRECISION TRAINING

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November 30, 2023

**Attention:** Turn in your homework at the beginning of our lecture on Dec. 5, 2023

## 1 SGD with mixed-precision

Consider the stochastic optimization problem:

$$\min_{x \in \mathbb{R}^d} f(x) = \mathbb{E}_{\xi \sim \mathcal{D}}[F(x; \xi)] \quad (1)$$

SGD with mixed-precision training can be approximated by

$$g_k = \nabla F(x_k; \xi_k) \quad (2)$$

$$x_{k+1} = x_k - \gamma Q(g_k) \quad (3)$$

where operator  $Q(\cdot)$  quantizes  $g_k$  with fewer bits. Assume  $f(x)$  is  $L$ -smooth, each stochastic gradient  $\nabla F(x; \xi)$  is unbiased and has bounded variance  $\sigma^2$ . Furthermore, we make the following assumptions on quantization operator  $Q(\cdot)$ .

**Assumption 1.1.** The (probably randomized) quantization operator  $Q(\cdot)$  satisfies

$$\mathbb{E}[Q(g)] = g, \quad \forall g \quad (4)$$

$$\mathbb{E}[\|Q(g) - g\|^2] \leq \zeta^2, \quad \forall g \quad (5)$$

and the random quantization operator  $Q(\cdot)$  is independent of the random sample  $\xi$ .

Please prove the convergence rate of algorithm (2)–(3).