EDEN:

Communication-Efficient and Robust Distributed Mean Estimation for Federated Learning

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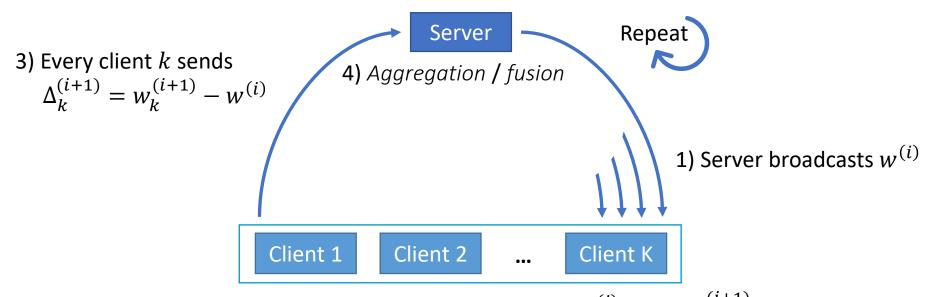
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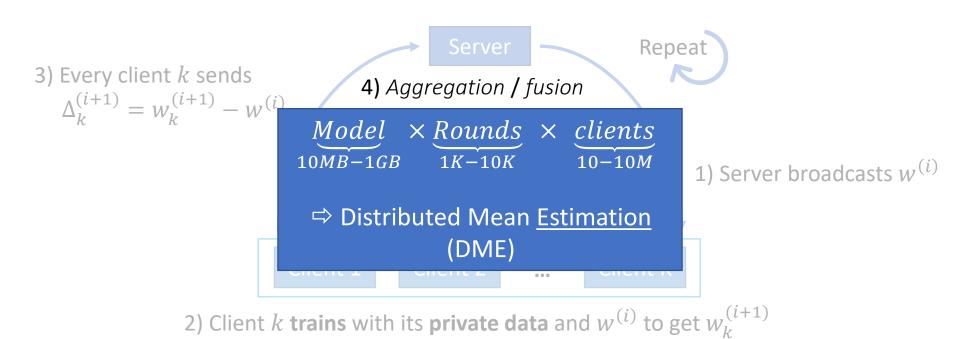
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Federated Optimization



2) Client k trains with its private data and $w^{(i)}$ to get $w_k^{(i+1)}$

Federated Optimization



1) First, we **randomly rotate** the input vector, $x \in \mathbb{R}^d$, after which each coordinate distribution is similar to a normal distribution:

$$\mathcal{N}\left(0, \frac{\|x\|_2^2}{d}\right)$$

- 2) We use this fact, multiply the post-rotation vector by $\frac{\sqrt{d}}{\|x\|_2}$, and apply a **deterministic quantization** over $\mathcal{N}(0,1)$ instead of a stochastic one.
- 3) With proper scaling, after applying the inverse rotation, the reconstructed vector, \hat{x} , is unbiased, and the eventual distributed mean estimation is more accurate.

We define a post-rotation quantization, Q, as follows:

Given a set of intervals, \mathcal{I} , that constitutes a symmetric partition of \mathbb{R} .

Each coordinate, $a \in I \in \mathcal{I}$, is (deterministically) quantized into I's center of mass:

$$Q(a) = \mathbb{E}[z \mid z \in I \text{ and } z \sim \mathcal{N}(0,1)]$$

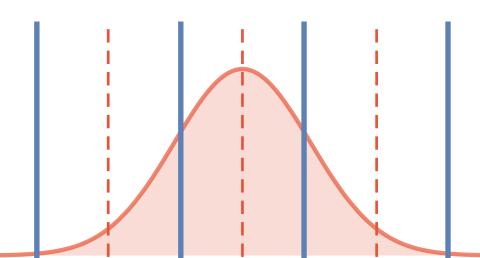
Key theoretical result

$$\frac{\mathbb{E}[\|x - \hat{x}\|_{2}^{2}]}{\|x\|_{2}^{2}} \le \frac{1}{1 - \mathbb{E}\left[\left(z - Q(z)\right)^{2}\right]} - 1$$

where $d \to \infty$ and $z \sim \mathcal{N}(0,1)$

Two principal quantization strategies:

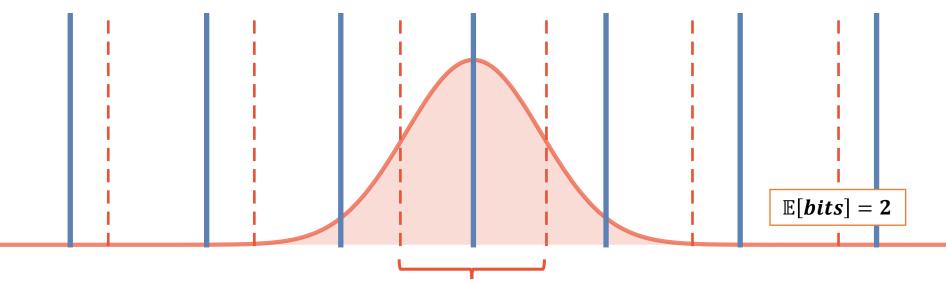
1) Lloyd-Max provides the **optimal** quantization for a **fixed** number of bits by minimizing $\mathbb{E}\left[\left(z-Q(z)\right)^2\right]$.



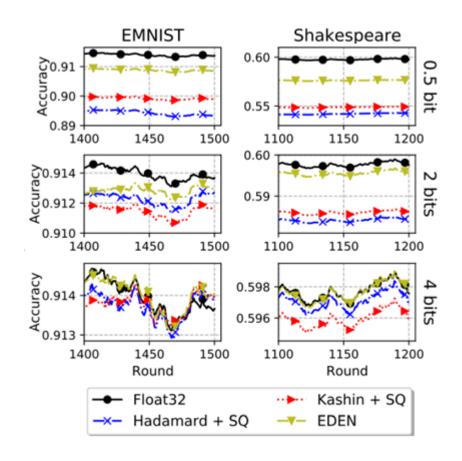
bits = 2

Two principal quantization strategies:

2) When considering variable length coding, we use Shannon's source coding theorem to determine the expected number of bits (i.e., the entropy).
Given a bit budget, b, we search for the smallest Δ that achieves entropy ≤ b.



Federated learning benchmarks



More in the paper:

- Heterogeneous sender communication budgets.
- Handling packet losses.
- Support non-integer number of bits per coordinate.
- The Hadamard transform as a rotation.
- Full source code is available at: https://github.com/amitport/EDEN-Distributed-Mean-Estimation.

Thank You!