Optimal Compression of Locally Differentially Private Mechanisms

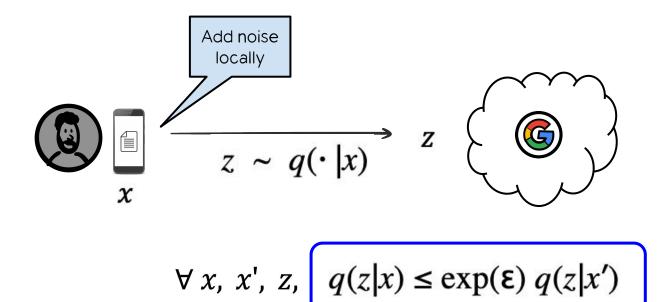
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Joint work with Wei-Ning Chen, Johannes Balle, Peter Kairouz, Lucas Theis

Private and efficient distributed learning

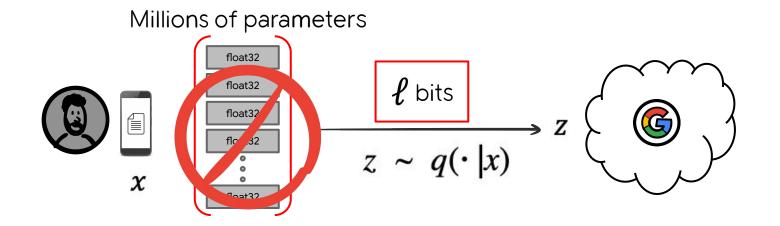
- 1. Preserving the **privacy** of the user's local data
- 2. **Communicating** the privatized data efficiently to a central server.
- 3. Achieving high **accuracy** on a task (e.g., mean estimation or frequency estimation)

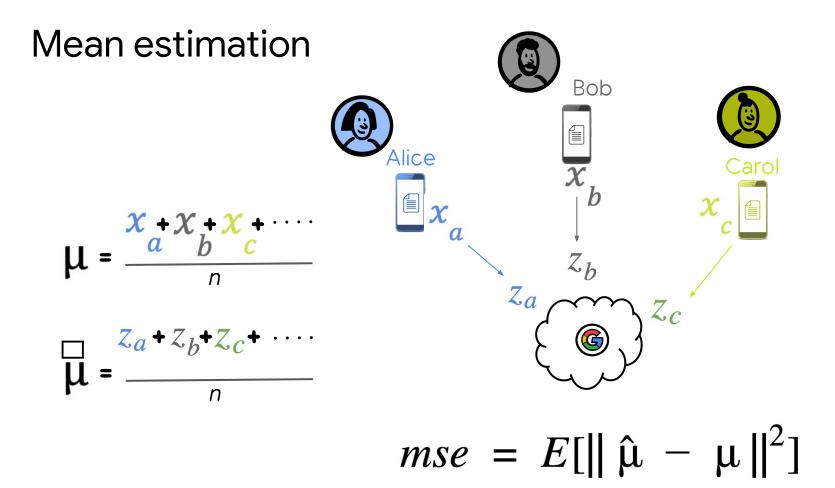
Local Differential Privacy (LDP)



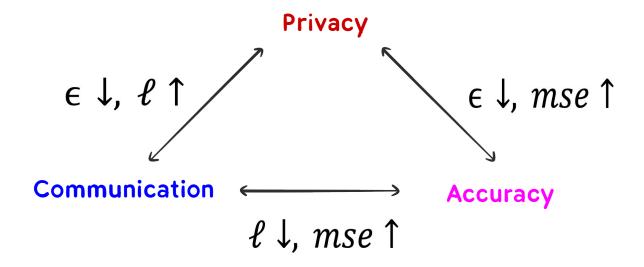
Smaller $\epsilon \longrightarrow$ larger privacy

Communication cost



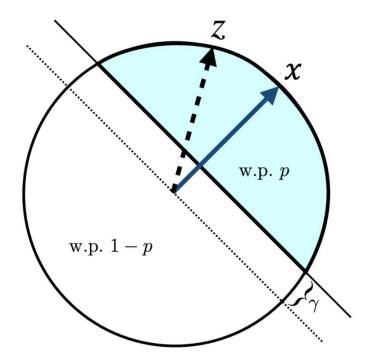


Privacy-Accuracy-Communication tradeoffs



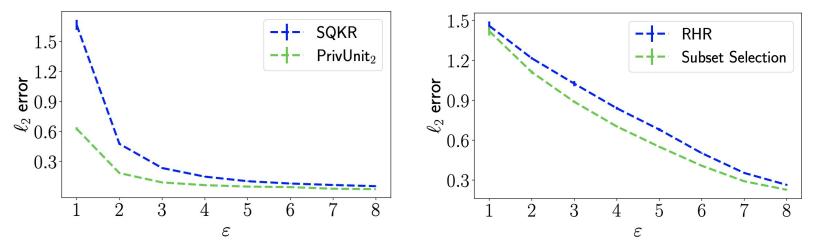
Best-known Privacy-Accuracy Tradeoff

- PrivUnit and Subset Selection are the ε-LDP schemes that provide the best-known accuracy for mean estimation and frequency estimation.
- However, their communication cost scales as *O(d)*.



SQKR and RHR

- Chen et al. (2020) presented minimax order-optimal mechanisms for mean estimation (SQKR) and frequency estimation (RHR) that required only ε bits, by using shared randomness.
- However, SQKR and RHR are not competitive in terms of accuracy with PrivUnit and Subset Selection.



Main Question

Can we attain the best known accuracy under ϵ -LDP for mean estimation and frequency estimation while only using on the order of ϵ bits of communication?

Yes! We leverage a technique based on importance sampling called Minimal Random Coding (MRC)

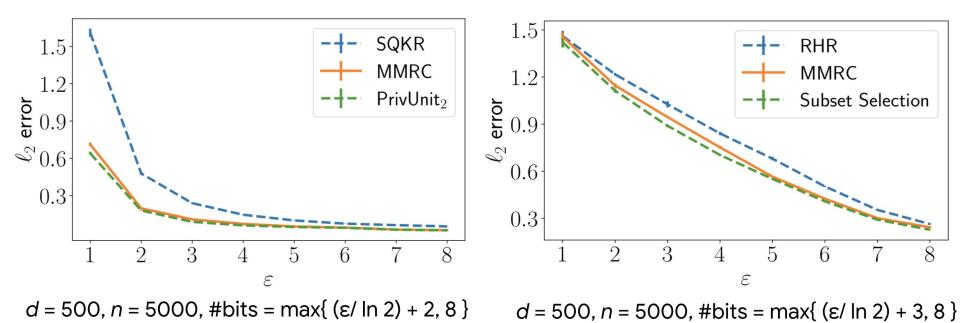
Pathway / Contributions

- MRC can compress any ϵ -LDP mechanism in a near-lossless fashion using only on the order of ϵ bits of communication. The resulting compressed mechanism is 2ϵ -LDP.
- A modified version, MMRC, can compress a large class of ε-LDP mechanisms in a near-lossless fashion using only on the order of ε bits of communication. The resulting compressed mechanism is ε-LDP.
- The class of LDP mechanisms MMRC can simulate includes the best-known schemes for mean and frequency estimation.

Empirical comparison

Mean estimation

Frequency estimation



Thank you! Please visit our poster!