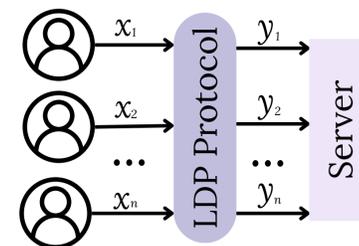




Motivation

- Local Differential Privacy (LDP) → allows to **privatize data before sharing**
- Highly adopted by **Big Tech** → large-scale **frequency monitoring systems**
- Real-world AI systems operate on **very large alphabets**
- Challenge: four-way trade-off**
 - Utility (↑), Communication cost (↓), Server runtime (↓), Robustness to reconstruction attacks (↑)
 - Existing protocols optimize only **subsets** of these constraints

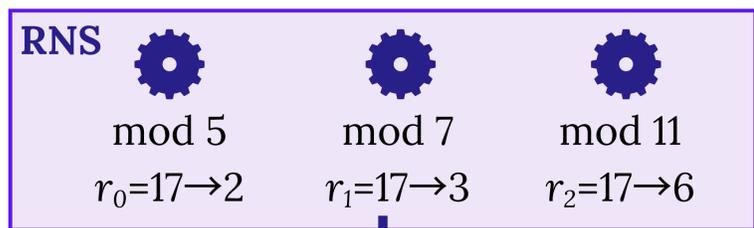


We need LDP protocols that are **accurate, lightweight, fast, and attack-resistant**

Our Contribution: Modular Subset Selection (MSS)

User: Report one noisy modular view

$x = 17$



Randomly select **one** modulus $j \in \{0,1,2\}$
Privatization via ϵ -LDP(r_j)

$y = (\text{modulus index, privatized residue})$

Example (“top-20 URL”):

- domain size = 20
- Moduli: {5, 7, 11}

Residue Number System (RNS): split a large alphabet into few small modular views

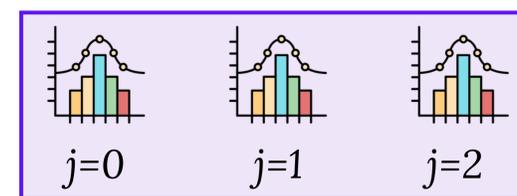
Randomized modular view (**single report**)

Server: Aggregation & decoding

Reports:

($j=0, Y$) ($j=1, Y$) ($j=2, Y$)

group by modulus j



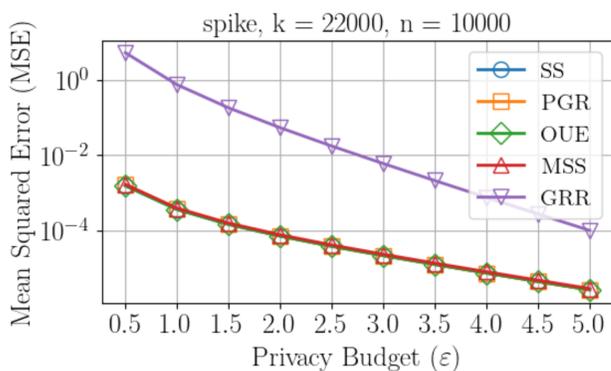
Weighted least squares (CRT-based)

Estimated **frequency vector**

Why MSS Works (Results & Takeaway)

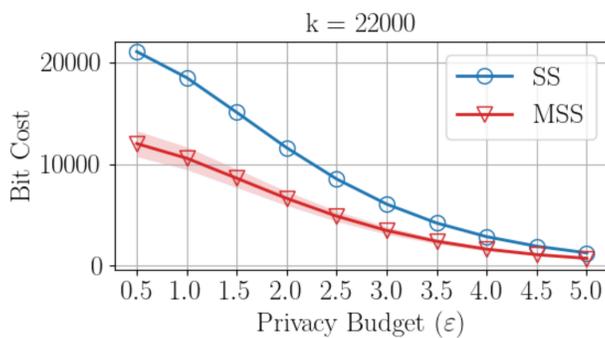
Near-optimal utility 🎯

Matches state-of-the-art
Stable across privacy budgets
No collapse for large alphabets



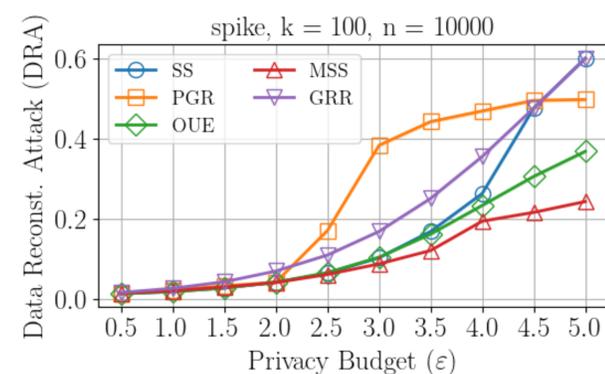
System efficiency ⚡

Short messages (few bits)
Fast server-side decoding
Scales to very large domains



Practical privacy 🛡️

Randomized modular view
Lowest reconstruction risk
Stronger protection in practice



MSS shows that large-alphabet LDP can be both practical and robust to reconstruction attacks

References

- [1] Kasiviswanathan, Shiva Prasad, et al. "What can we learn privately?." SIAM Journal on Computing 40.3 (2011): 793–826..
- [2] Szabo, Nicholas S., and Richard I. Tanaka. "Residue arithmetic and its applications to computer technology." (1967).
- [3] Ye, Min, and Alexander Barg. "Optimal schemes for discrete distribution estimation under locally differential privacy." IEEE Transactions on Information Theory 64.8 (2018): 5662–5676.