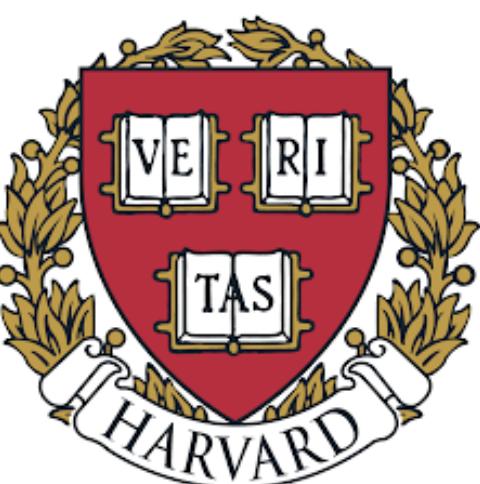


Robust Aggregation for Federated Learning

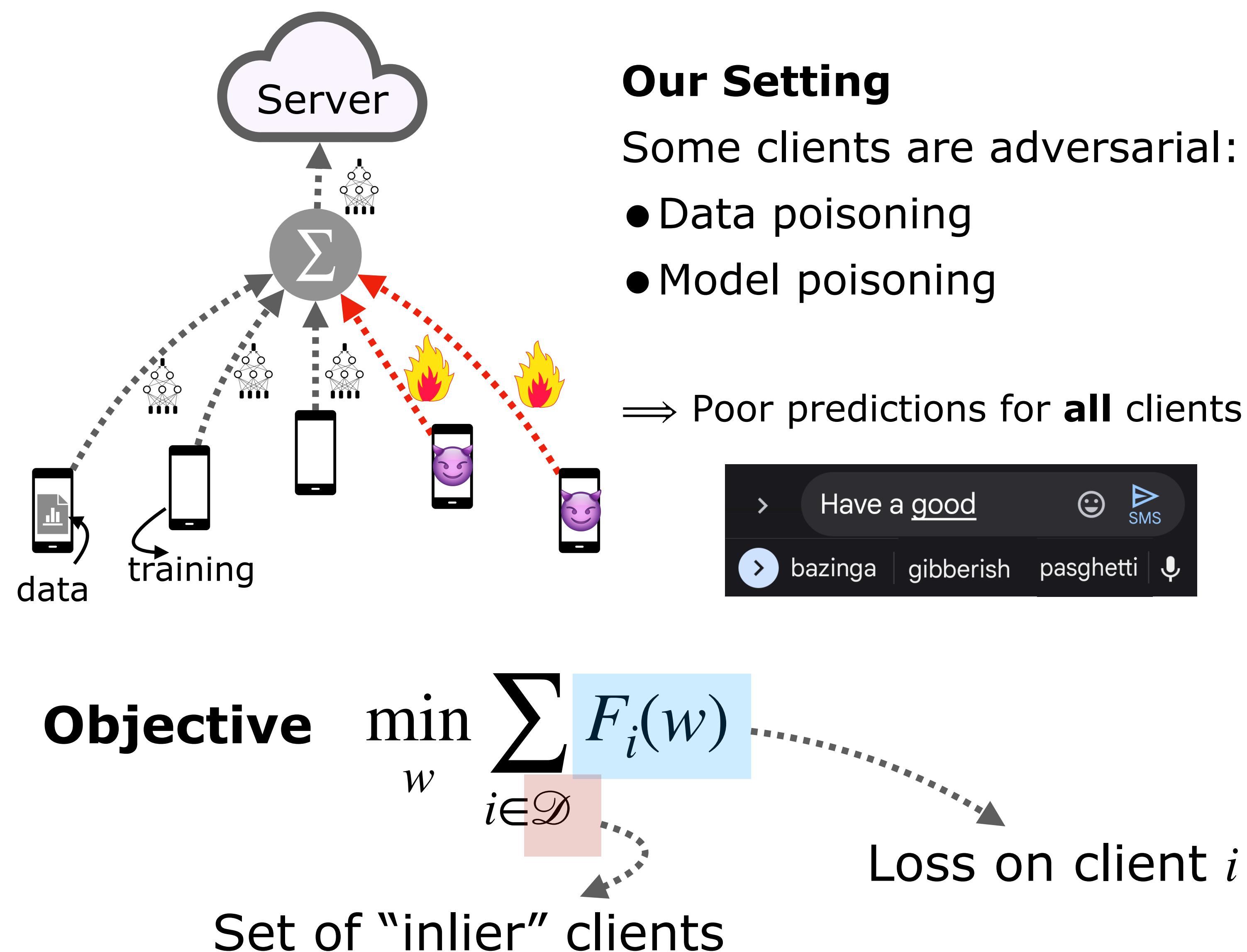
UNIVERSITY of
WASHINGTON



Google Research

Krishna Pillutla, Sham Kakade, Zaid Harchaoui

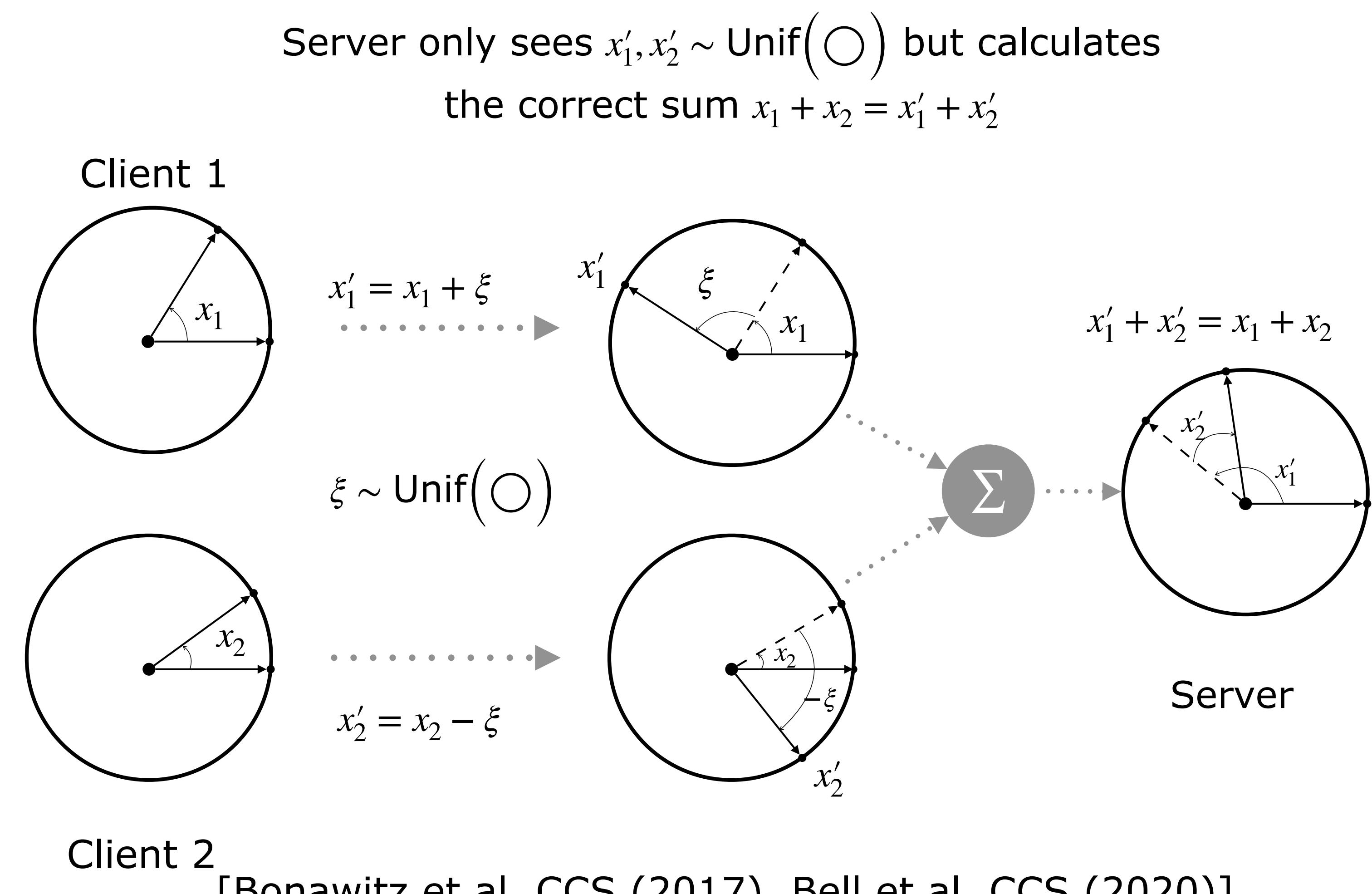
Robust federated learning



Goals of the work

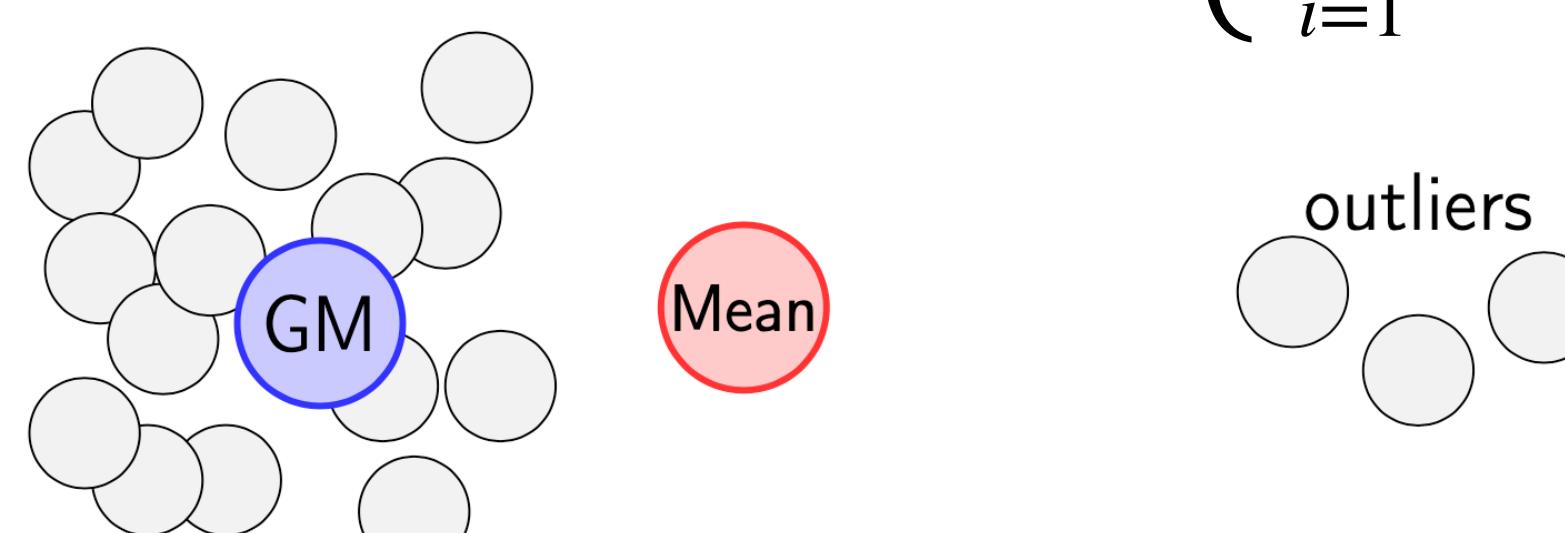
A robust aggregation approach that is

1. Communication-efficient
2. Implementable via **secure summation**

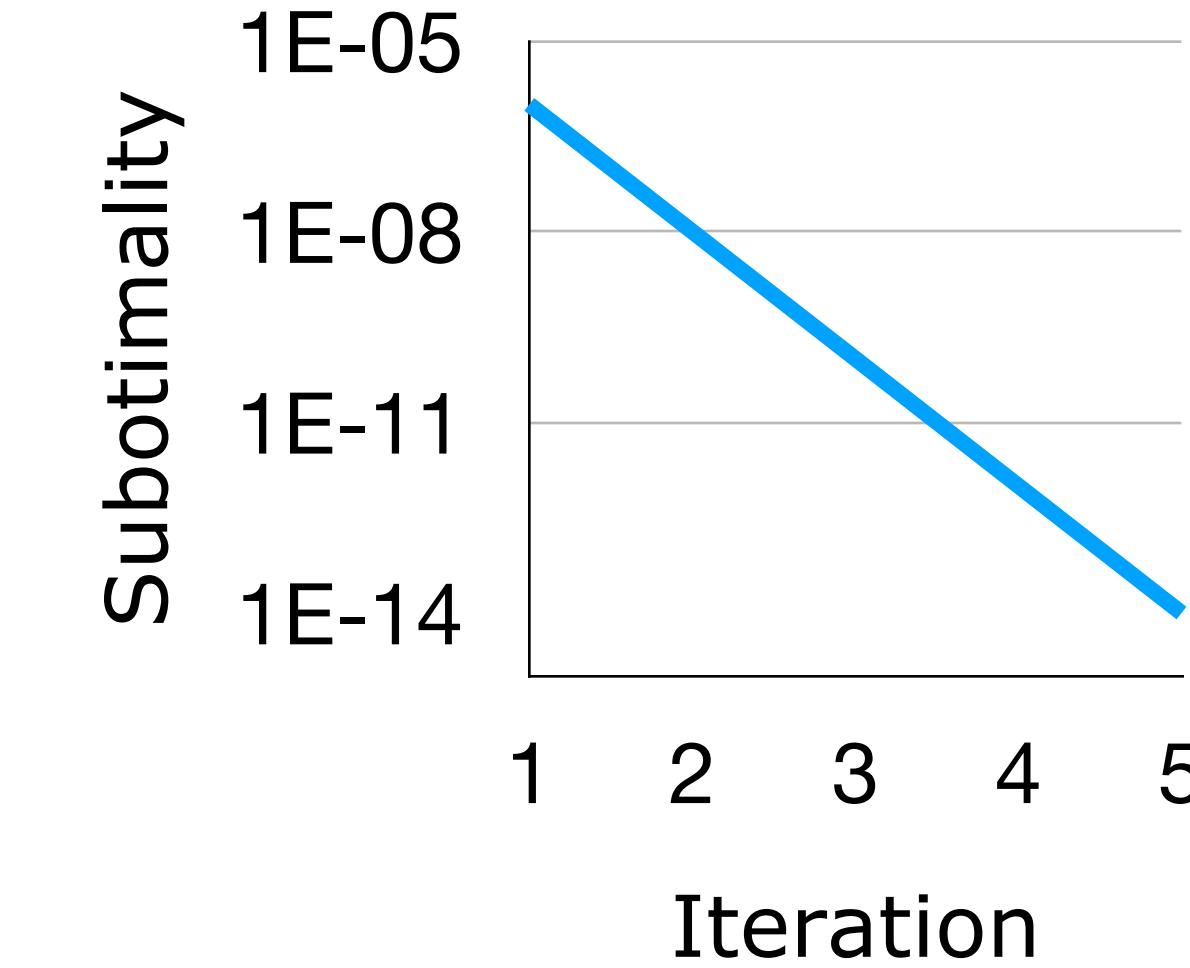


RFA: Geometric Median Aggregation

$$\text{GM}(w_1, \dots, w_m) = \arg \min_z \left\{ \sum_{i=1}^m \|z - w_i\|_2 \right\}$$



Robustness:
Breakdown point = 1/2



Weiszfeld's Algorithm:

$$\beta_{i,t} = 1/\max\{\|z_t - w_i\|_2, \nu\}$$

$$z_{t+1} = \frac{\sum_i \beta_{i,t} w_i}{\sum_i \beta_{i,t}}$$

Theory (Least Squares)

Suppose $Y_i = X_i^\top w^* + \xi_i$ where $\xi_i \sim \mathcal{N}(0, \sigma^2)$

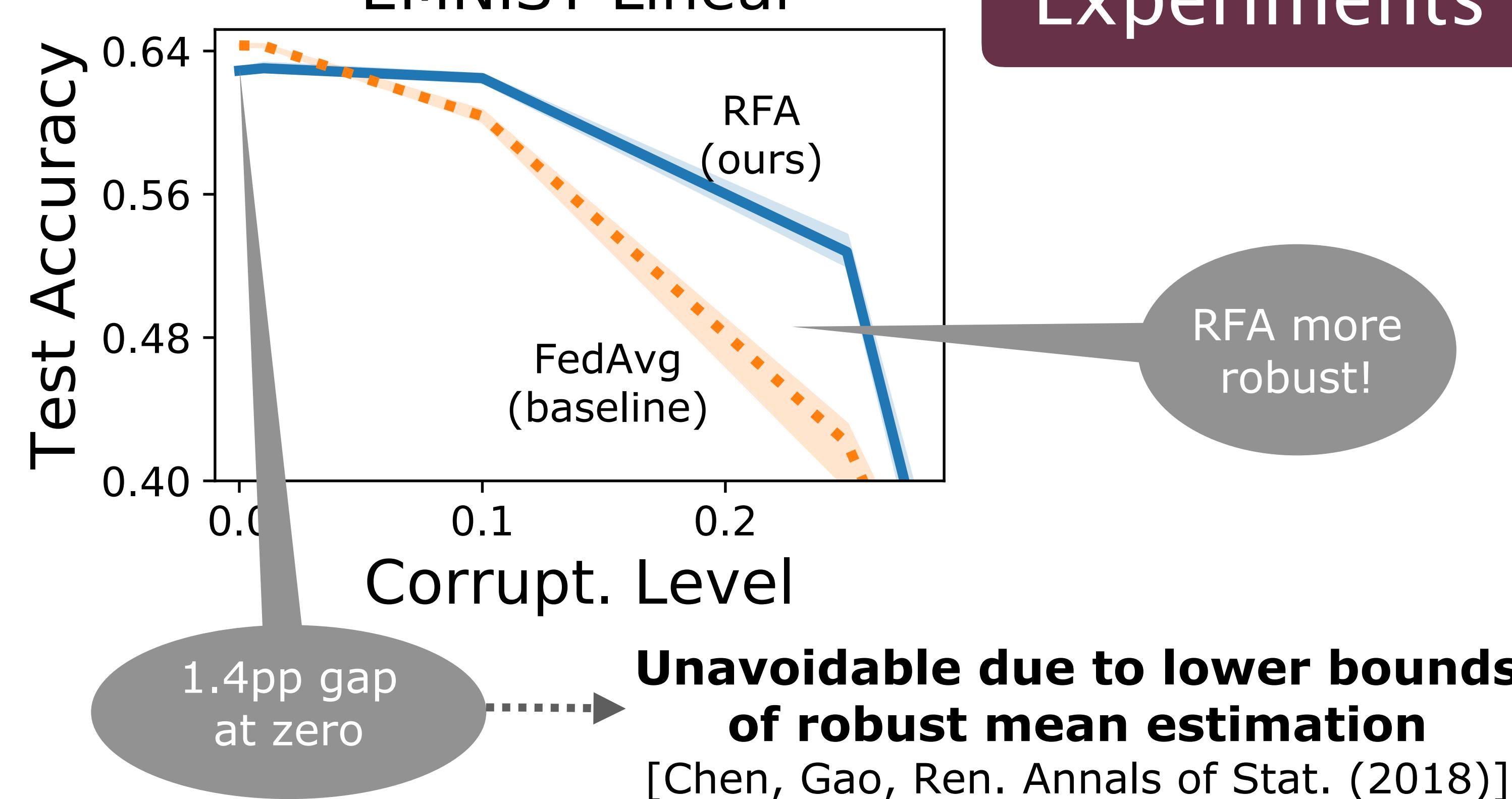
Theorem: Assume that $F(w)$ is strongly convex, $\|X\| \leq 1$ and # local steps $\propto 2^t$. Let \mathcal{E} denote the event that $\geq 1/2 + c/2$ non-corrupted devices are chosen (out of the $1/2 + c$ available) in each round.

Then, RFA with ϵ -approximate GM satisfies

$$\mathbb{E} [\|w_t - w^*\|^2 | \mathcal{E}] \lesssim \frac{\|w_0 - w^*\|^2}{2^t} + \frac{1}{c^2} \left(d\sigma^2 \frac{t}{2^t} + \frac{\epsilon^2}{m^2} + \Omega_X^2 \Omega_Y^2 \right)$$

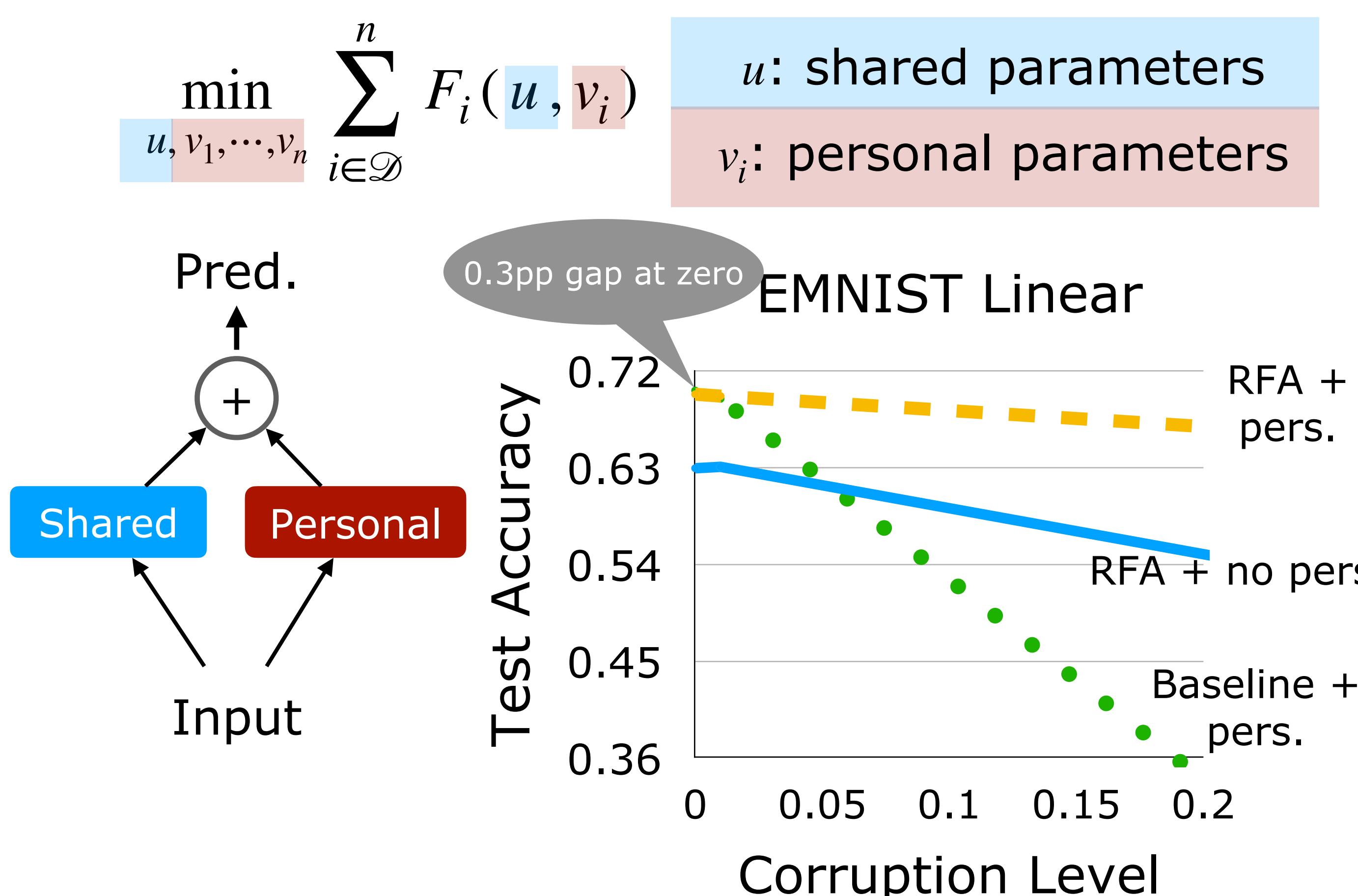
Optimization error
Statistical error
GM Approx. Error
Heterogeneity Error

Experiments



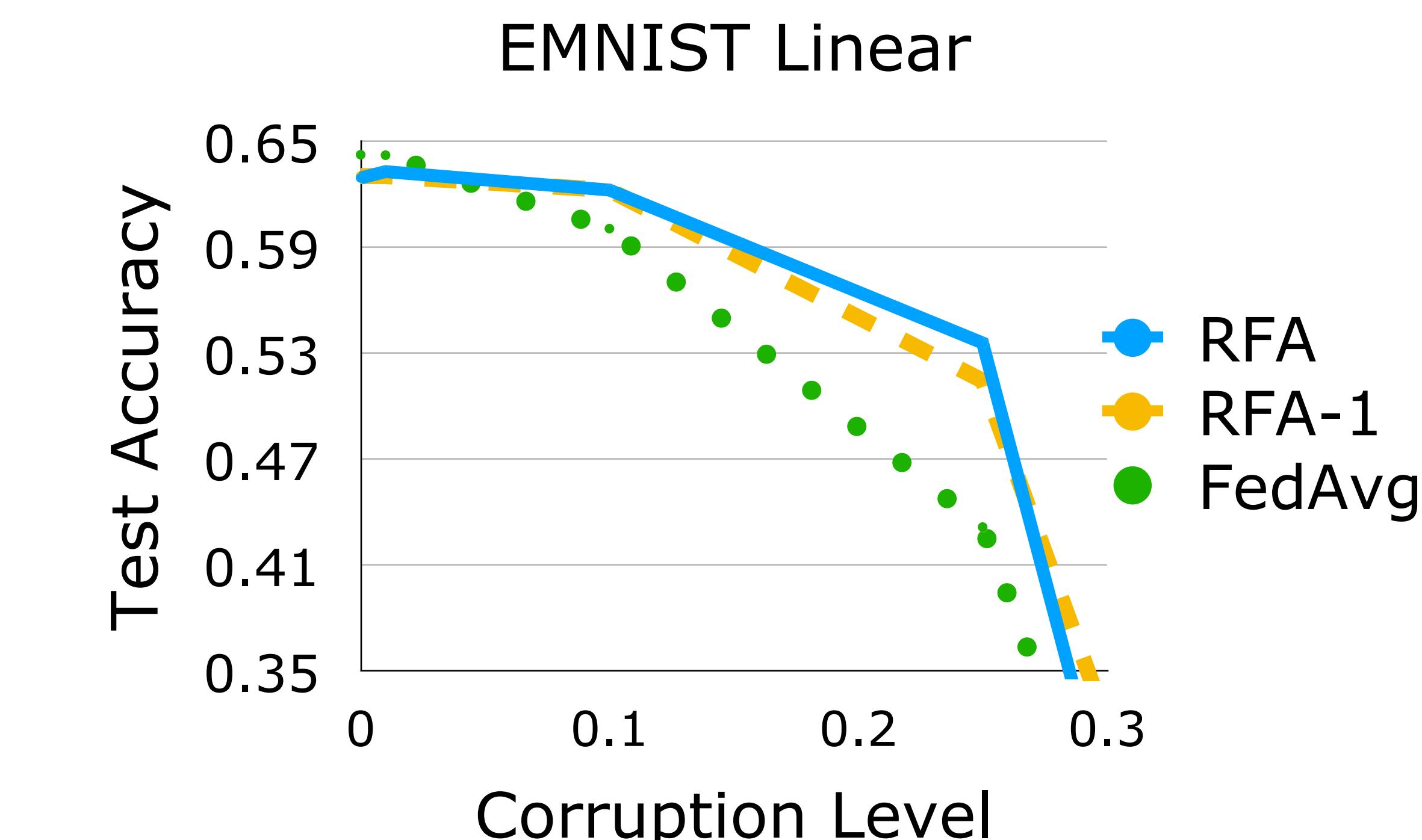
Handling Heterogeneity

Personalize parts of the model



Improving the Communication Cost

Single Weiszfeld iteration is also robust!



Software for the geometric median

Install: pip install geom-median

Documentation: github.com/krishnap25/geom_median



krishnap25



KrishnaPillutla



krishnap25.github.io

Code



SCAN ME