

Random Orthogonalization for Federated Learning in Massive MIMO Systems

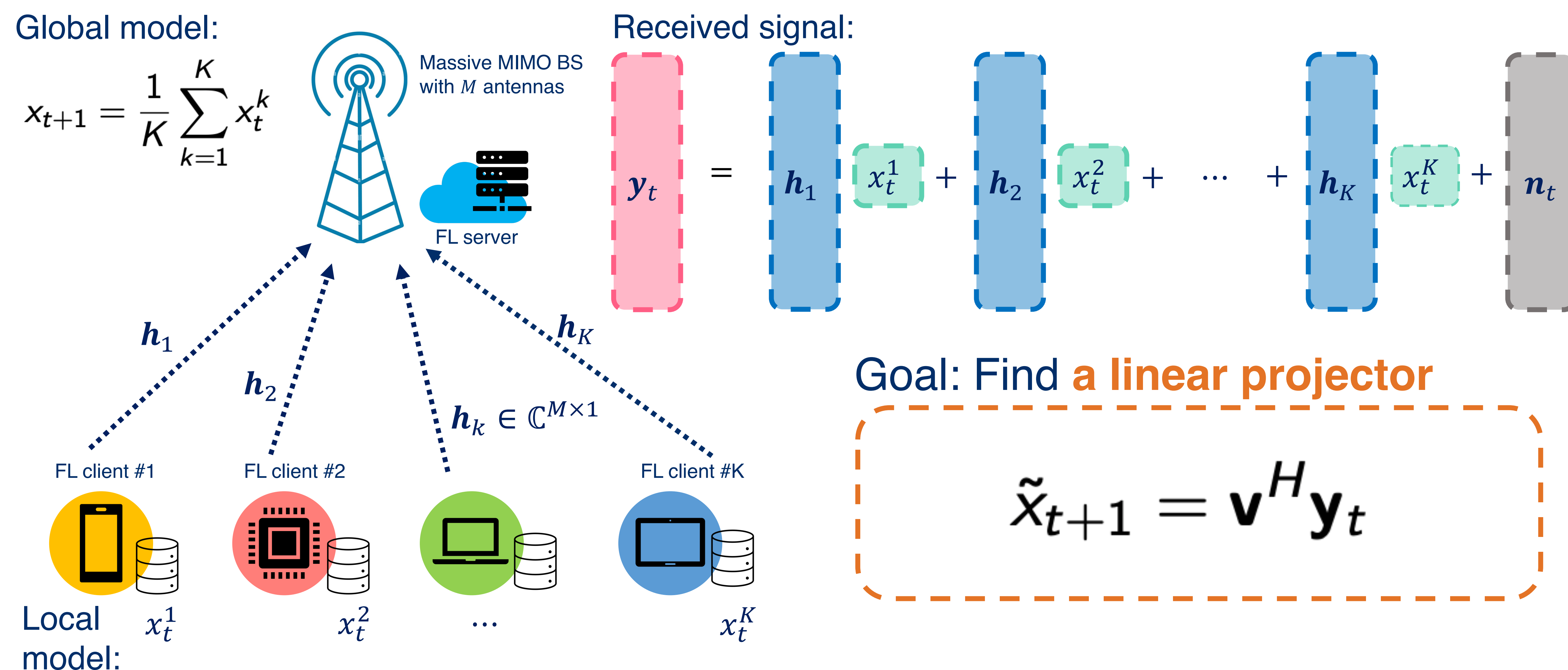


SCHOOL of ENGINEERING & APPLIED SCIENCE

Charles L. Brown Department of Electrical and Computer Engineering

Xizixiang Wei, Cong Shen, Jing Yang, H. Vincent Poor

Goal: AirComp federated learning without CSIT



Solution: random orthogonalization

Channel hardening: $\mathbf{h}_k^H \mathbf{h}_k \rightarrow 1, \text{ as } M \rightarrow \infty, \forall k$

Favorable propagation: $\mathbf{h}_k^H \mathbf{h}_j \rightarrow 0, \text{ as } M \rightarrow \infty, \forall k \neq j$

Proposed projector: sum channel

$$\mathbf{v}_s = \mathbf{h}_s = \frac{1}{K} \sum_{k=1}^K \mathbf{h}_k$$

Linear projection: an unbiased estimator

$$\tilde{x}_{t+1} = \frac{1}{K} \mathbf{h}_s^H \mathbf{y}_t$$

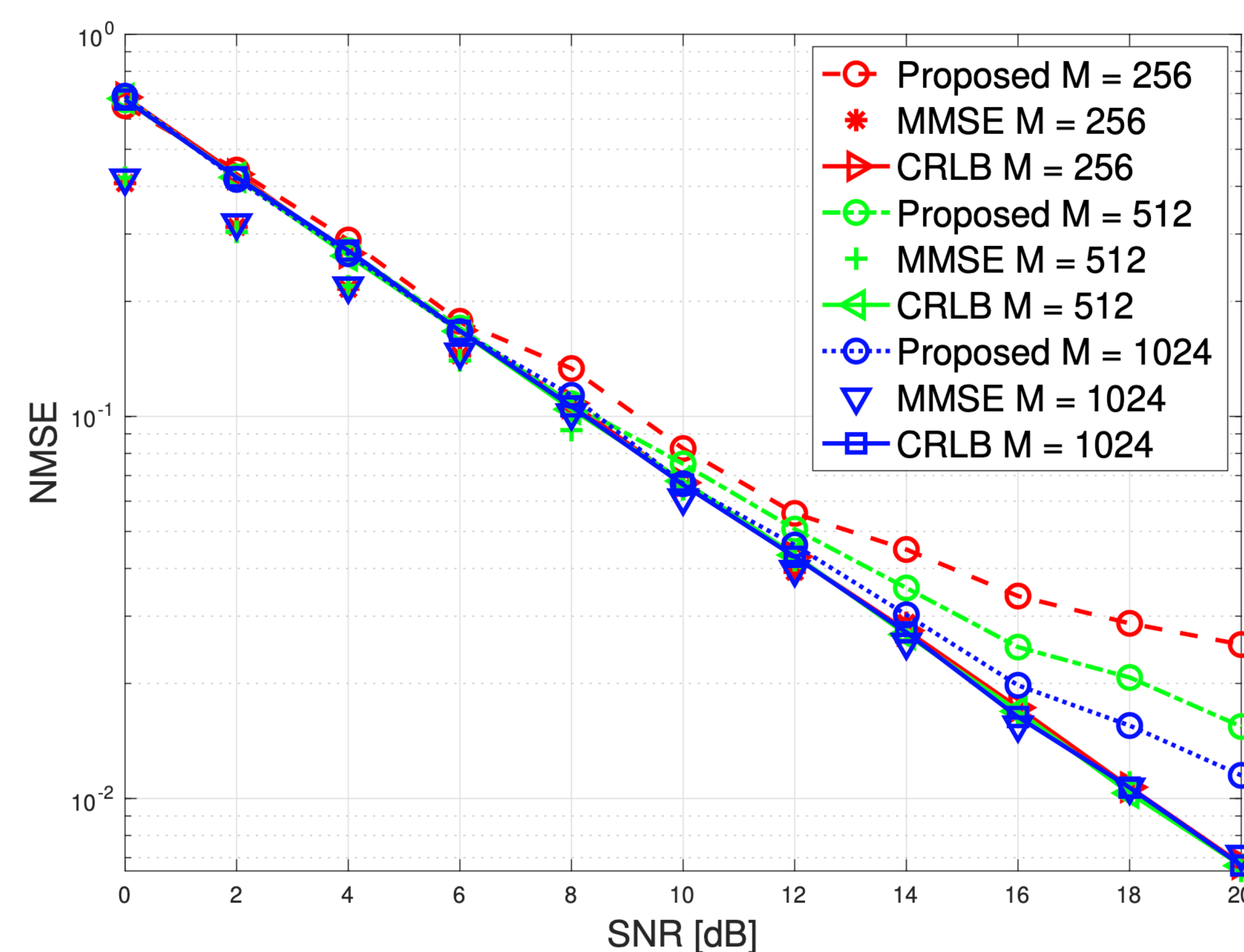
Motivation

- Traditional over-the-air computation (AirComp) methods for FL are efficient but require CSIT.
- Design a low-complexity AirComp method that removes the CSIT requirement.

Random Orthogonalization Solution

- Leveraging channel hardening and favorable propagation in massive MIMO channels.
- A low-complexity 3-step procedure:
 - Step#1: uplink sum channel estimation;
 - Step#2: uplink local model transmission;
 - Step#3: linear projection to estimate global model.

Experiment results



- Estimates yielded by random orthogonalization become closer CRLB as M goes larger

Summary

Random orthogonalization preserves all the advantages of AirComp and enjoys:

- Low communication overhead: partial CSIR
- Low complexity receiver: linear projection
- Asymptotically efficient: estimates achieves CRLBs as M goes infinity

Reference

X. Wei, C. Shen, J. Yang and H. Vincent Poor, "Random Orthogonalization for Federated Learning in Massive MIMO Systems," in *IEEE Transactions on Wireless Communications*.