Revisiting Ensembling in One-Shot Federated Learning



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Motivation

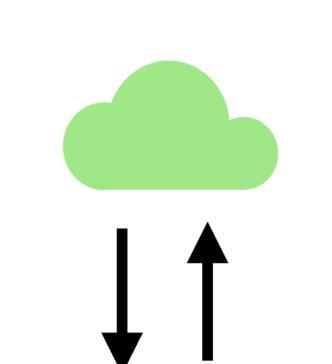
Federated Learning (FL)

1. Iterative training of models



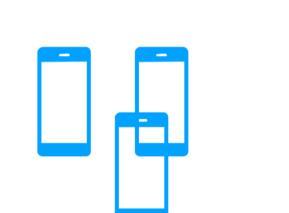
One-Shot Federated Learning (OFL)

1. One round of transfers



COPENHAGEN

- 2. Generally, achieves good accuracy
- 3. Significant communication costs



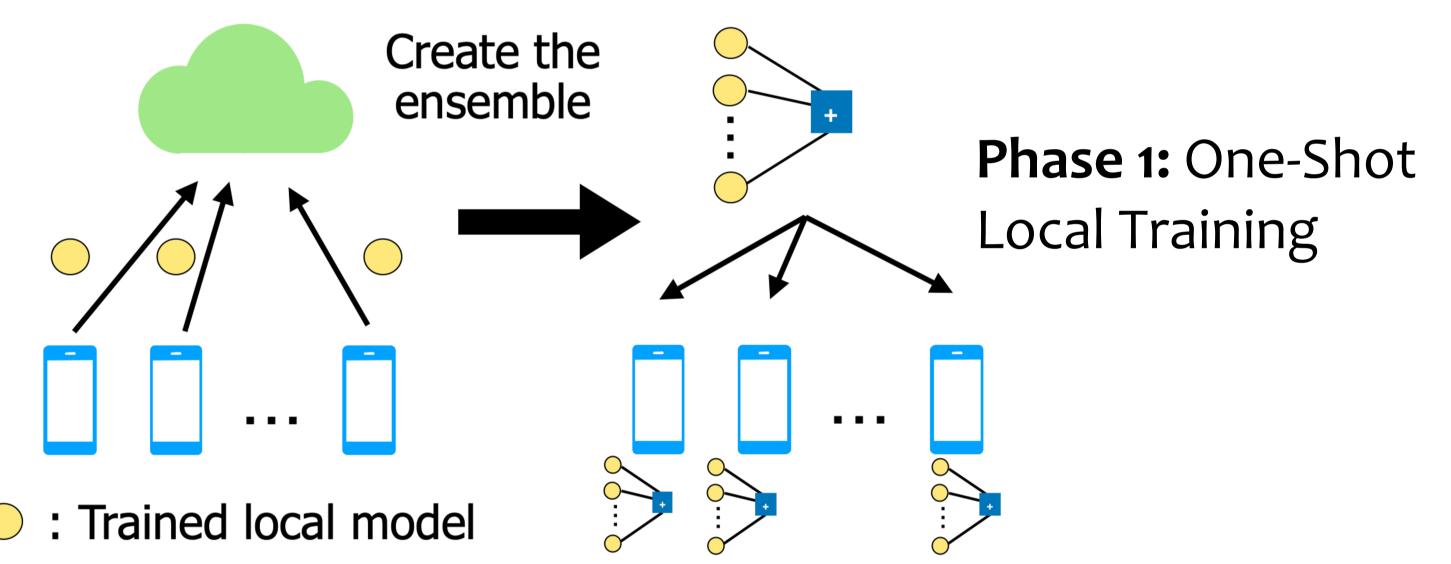
- 2. Low communication costs
- 3. Accuracy drop compared to FL



How do we neatly combine the two to achieve best of the both?

Fens: Hybrid of FL and OFL

Training in Two Phases



Fens: Key Properties

Fens stacks an aggregator model (a neural network) atop client local models in the ensemble

- 1. MLP suffices as the aggregator
- 1. Stacked model is trained on two separate partitions of local data

• : Trained local model

Phase 2: Iterative Aggregator Training

- Aggregation model (for logits/predictions)
- 2. Size of aggregator << Size of client model
- Phase 2 induces very low additional communication costs

Low comm. costs

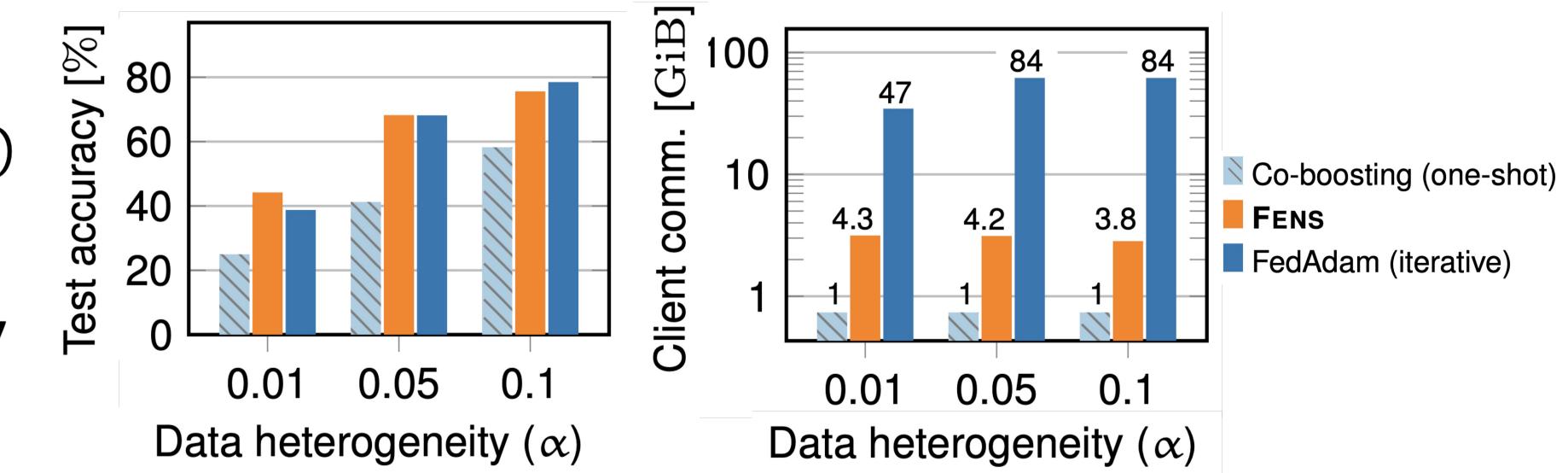
- 2. Aggregator training effectively resolves biases of client local models^[1]
- 3. Stacking significantly improves generalization

Improved generalization

Evaluation

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- 20 clients, ResNet-8
- 2. CIFAR-10 Non-IID Partitioning
 - Dirichlet Distribution ($\alpha = 0.01, 0.05, 0.1$) ullet
- 3. Co-Boosting^[2] (OFL) and FedAdam^[3] (FL)
 - State-of-the-art algorithms in each family



- 4. Fens (almost) achieves FedAdam's accuracy
 - Communication cost ~ 4x that of OFL, \bullet FL is 47-84x that of OFL

[1] Wolpert, D.H., 1992. Stacked generalization. *Neural networks*, 5(2), pp.241-259.

[2] Dai, Rong, et al. "Enhancing One-Shot Federated Learning Through Data and Ensemble Co-Boosting." In ICLR 2024.

[3] Reddi, Sashank, et al. "Adaptive federated optimization." In ICLR 2021.



