

# **Revisiting Ensembling in One-Shot Federated Learning**

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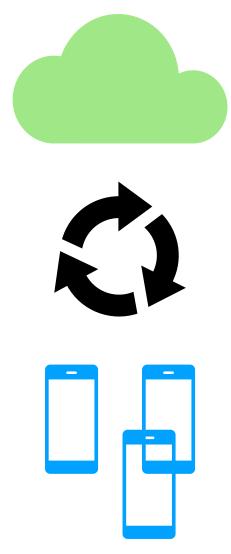


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# Introduction

#### (Iterative) Federated Learning (FL)

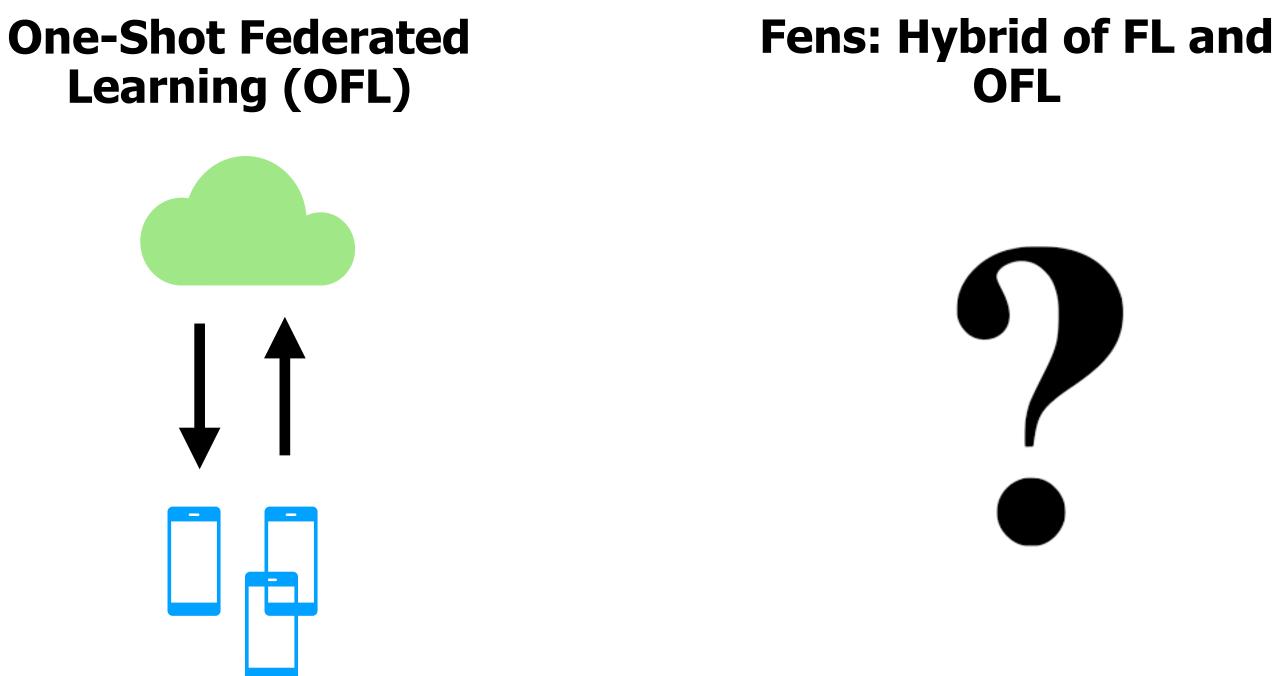


Huge communication costs

Low communication costs

Good accuracy

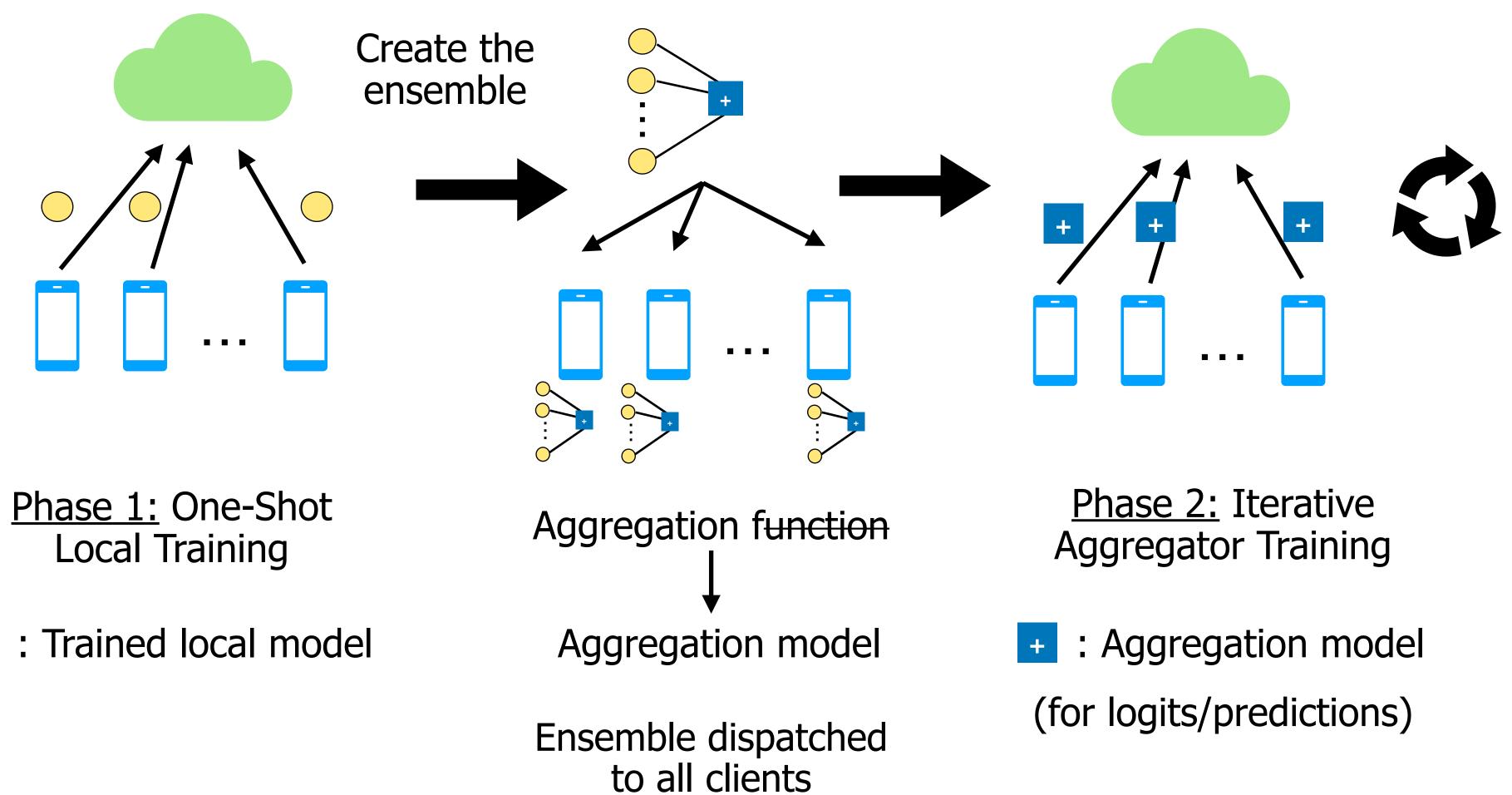
Not so good accuracy



Low communication costs Good accuracy



### Fens: Two Phase Learning







# Fens: Important Characteristics



Key observation: a MLP suffices as the aggregator model

Size of Aggregator  $\ll$  Size of local model

Iterative training induces very low additional comm. cost

Low communication costs

Stacked Generalization [Wolpert, 1992]

Level 1 (L1) generalizers correct the biases of Level 0 (L0) generalizers

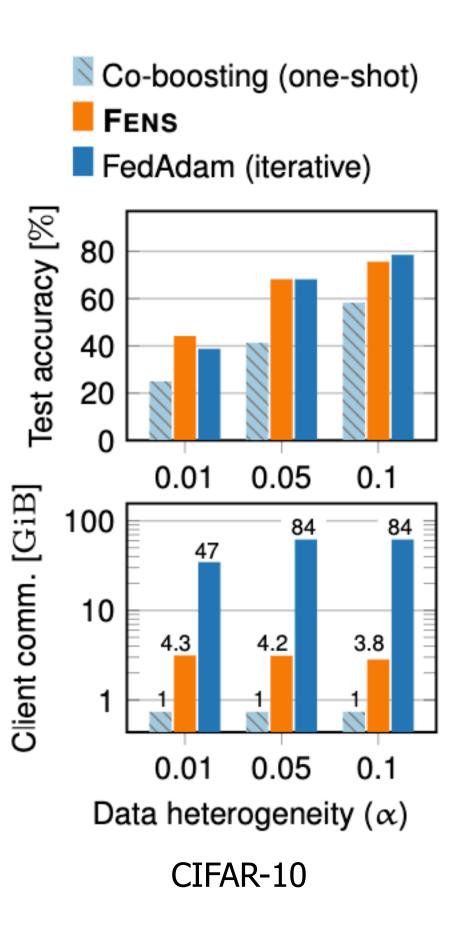
Higher accuracy than standard OFL



### Results

#### Fens vs OFL and FL

Accuracy properties of FL with communication properties of OFL



# Please checkout our paper for more experiments and results!

### THANK YOU !

