## Optimality and Stability in Federated Learning

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## 1. Federated Learning

• Data transfer, data privacy

# 2. Hedonic Game (Game theoretic approach)

- different distributions
- join only for benefit

## 3. Model of federated learning

- Error model for
  - Uniform federation
  - Coarse-grained federation
  - Fine-grained federation

## 4. Stability

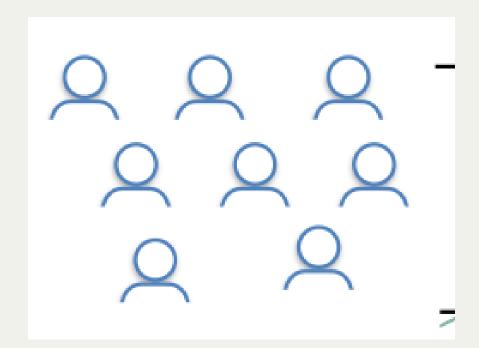
#### 4.1. Uniform Federation

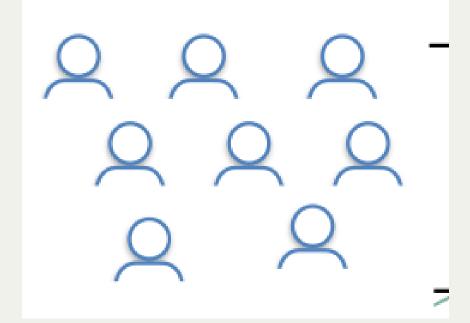
- Same number of samples
- Small/large number of samples

#### 4.2. Coarse-grained federation

- Same number of samples
- Small/large number of samples
- Same number of samples
- Small/large number of samples

#### 4.3. Fine-grained federation





## 5. Optimality

- Optimal: minimizes weighted sum of errors across all agents
- Algorithm:
  - Start with every agent doing local learning
  - Group the agents together in ascending order of size, stopping when the first agent would increase its error by joining the coalition
- Equivalence of player preference and reducing cost

- Swapping
- Monotonicity of joining
- Monotonicity of leaving
- Merging

Model

$$\hat{ heta}_C = rac{1}{\sum_{i \in C} n_i}.\sum_{i \in C} n_i.\hat{ heta}_i$$

$$err_j(C) = rac{\mu_e}{\sum_{i \in C} n_i} + \sigma^2. \, rac{\sum_{i \in C, i 
eq j} n_i^2 + \left(\sum_{i \in C, i 
eq j} n_i
ight)^2}{\left(\sum_{i \in C} n_i
ight)^2}$$

### 6. Price of Anarchy

- ullet for  $n_i \geq rac{\mu_e}{\sigma^2}, orall i$ , the grand coalition  $\pi_g$  is always core stable
- for  $n_i \leq \frac{\mu_e}{\sigma^2}$ ,  $\forall i$ , the individually stable or core stable is also optimal
- $\Pi_M$  is maximum cost IS partition, then  $err_i(\Pi_M) \leq \frac{\mu_e}{n_i}$  for all players i

ullet Error lower bound when a player j joins coalition C

$$err_j(C \cup \{n_j\}) \geq egin{cases} rac{1}{2}.rac{\mu_e}{n_j}, n_j \geq rac{\mu_e + \sigma^2}{2\sigma^2} \ \sigma^2, ext{otherwise} \end{cases}$$

• Error upper bound when a player j joins coalition C if total number of samples  $N_C \geq \frac{\mu_e}{3\sigma^2}$ , then

$$err_j(C \cup \{n_j\} \leq 7.25.\sigma^2$$

- $n_i \leq \frac{\mu_e}{3\sigma^2}$ ,  $\forall i$ , with at least one player in a coalition with mass of its partners no more than  $\frac{\mu_e}{3\sigma^2}$ , then the only stable arrangement of these players is to have all of them federating together
- Price of Anarchy

$$PoA = rac{f_w(\Pi_M)}{f_w(\Pi_{opt})} \leq 9$$

#### 7. Limitations

- Theoretical study, results might be different in practice
- Optimality bound has some assumptions, this may lead to different bounds

#### 8. Related Work

- Donahue and Kleinberg studied models of fairness
- Hu et al. 2023 models clients behaviour in network
- Cui et al 2021 tries to find collaboration equilibrium
- Le et al. 2021 analyzes incentives for agents to contribute computational resources while using an auction approach

## 9. My extension

Nothing as of now

### 10. Conclusion

Thank You