

Outline

Federated Learning

- ▶ Data transfer, data privacy

A more complex slide

This slide illustrates the use of Beamer blocks. The following text, with its own headline, is displayed in a block:

Theorem (Org mode increases productivity)

- ▶ *org mode means not having to remember \LaTeX commands.*
- ▶ *it is based on ascii text which is inherently portable.*
- ▶ *Emacs!*



Hedonic Game (Game theoretic approach)

- ▶ different distributions
- ▶ join only for benefit

Model of federated learning

- ▶ Error model for
 - ▶ Uniform federation
 - ▶ Coarse-grained federation
 - ▶ Fine-grained federation
- ▶ Explain these different types of federation

Stability

Uniform Federation

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

Coarse-grained federation

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

Fine-grained federation

- ▶ this is fine grained company
- ▶ this is fine grained company
- ▶ this is fine grained company

Example (Fine grained image)



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
- ▶ Error model taken from the paper "Model sharing games" by the same author
- ▶ Swapping
- ▶ Monotonicity of joining
- ▶ Monotonicity of leaving
- ▶ Merging
- ▶ Model

$$\hat{\theta}_C = \frac{1}{\sum_{i \in C} n_i} \cdot \sum_{i \in C} n_i \cdot \hat{\theta}_i$$

$$err_j(C) = \frac{\mu_e}{\sum_{i \in C} n_i} + \sigma^2 \cdot \frac{\sum_{i \in C, i \neq j} n_i^2 + \left(\sum_{i \in C, i \neq j} n_i \right)^2}{\left(\sum_{i \in C} n_i \right)^2}$$

Price of Anarchy

- ▶ for $n_i \geq \frac{\mu_e}{\sigma^2}, \forall i$, the grand coalition π_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
- ▶ Π_M is maximum cost IS partition, then $err_i(\Pi_M) \leq \frac{\mu_e}{n_i}$ for all players i
- ▶ Error lower bound when a player j joins coalition C

$$err_j(C \cup \{n_j\}) \geq \begin{cases} \frac{1}{2} \cdot \frac{\mu_e}{n_j}, & n_j \geq \frac{\mu_e + \sigma^2}{2\sigma^2} \\ \sigma^2, & \text{otherwise} \end{cases}$$

- ▶ 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

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

- ▶ $n_i \leq \frac{\mu_e}{3\sigma^2}, \forall i$, with atleast 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

Limitations

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

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

My extension

- ▶ Nothing as of now

Conclusion

Thank You