





# **Upgrade Federated Learning with MPC-based Secure Aggregation**

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# Motivation **3**

Learning from data improves, nowadays, virtually all areas in our life: e.g., next-word predictions on virtual keyboards, (premature) tumor analysis on MRI images, or enhancing autonomous driving. Federated Learning (FL), introduced by Google in 2016, enables Machine Learning locally on participants' devices. A prominent example is the virtualkeyboard application Gboard, which learns/t on millions of people's device to improve the global ML model via FL. Though, "plain FL" is vulnerable to data-reconstruction attacks. Thus, techniques such as (MPC-based) Secure Aggregation (SecAgg), which reveals only the final sum of all participants, have been added to FL. In recent years, several flavors of SecAgg protocols have been created.

- Your Mission, should you choose to accept it, is to enter the realm of MPC-based SecAgg-enhanced FL and accomplish the project's goals
- Interested to get to know more info? Please feel free to contact me 🙃

# Goals 🎯

- 📮 Get to know
- Secure Multi-Party Computation (MPC)
- Tederated Learning (FL)
- 📒 Familiarize with MPC-based SecAgg protocols in FL
- $\P$  Dig Deeper into  $\geq 1$  Protocol
- X Implement & Evaluate the Protocol(s)

#### Literature

> K. Bonawitz et al.

Practical Secure Aggregation for **Privacy-Preserving Machine Learning** CCS 2017

https://dl.acm.org/doi/10.1145/3133956. 3133982

#### **Courses & Deliverables**

- Introduction to Scientific Working Short report on background Short presentation
- **☑** Bachelor Project Project code and documentation
- ☑ Bachelor's Thesis Project code Thesis Final presentation

## Recommended if you're studying

**™**CS ☑ICE ☑SEM

### **Prerequisites**

- > Basic crypto background
- Motivation to dig into the realm of privacy-preserving computations

#### **Advisor Contact**

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