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MobNet: Online Learning at the Edge via Staleness Awareness and Performance Prediction

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The individual privacy

- Conflict of interests between service providers and users
 - Service providers need users' data to improve their services
 - Many users value their individual privacy



>\$100 billion was knocked off





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Private Machine Learning is the future

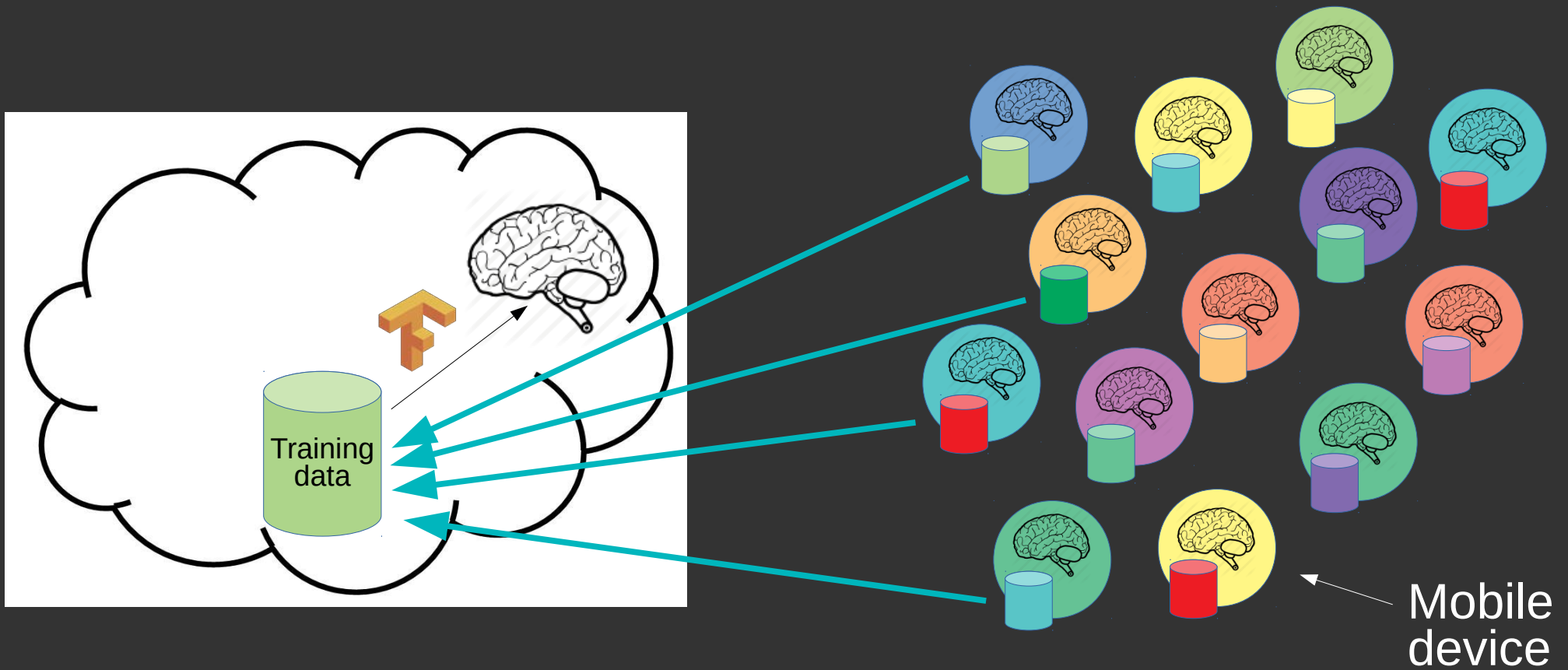


Private Machine Learning

- Secure cloud-based Machine Learning
 - Secure processors (e.g. SGX)
 - Homomorphic Encryption, Multi-party Computation
- Federated Learning

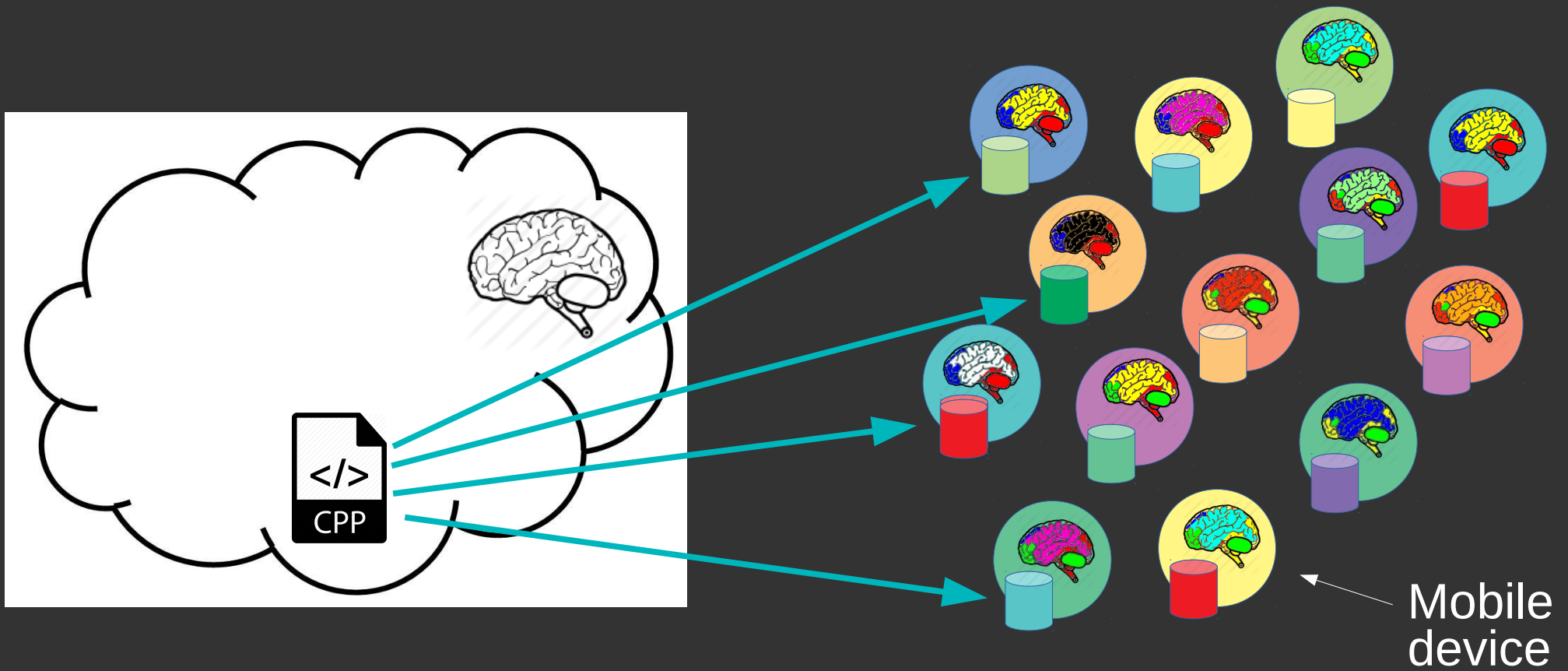
Cloud-based Machine Learning

- Instead of centralizing the users' data...

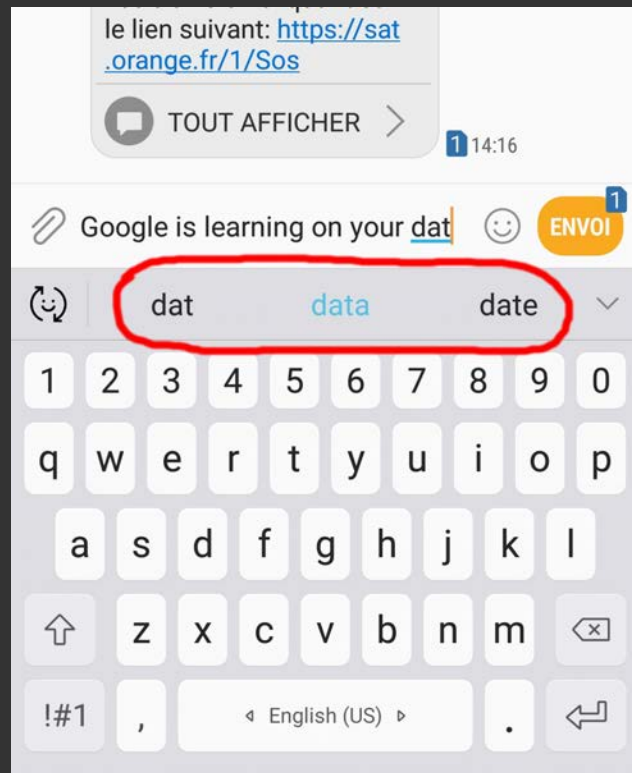


Federated Learning

- Train models right on the device.



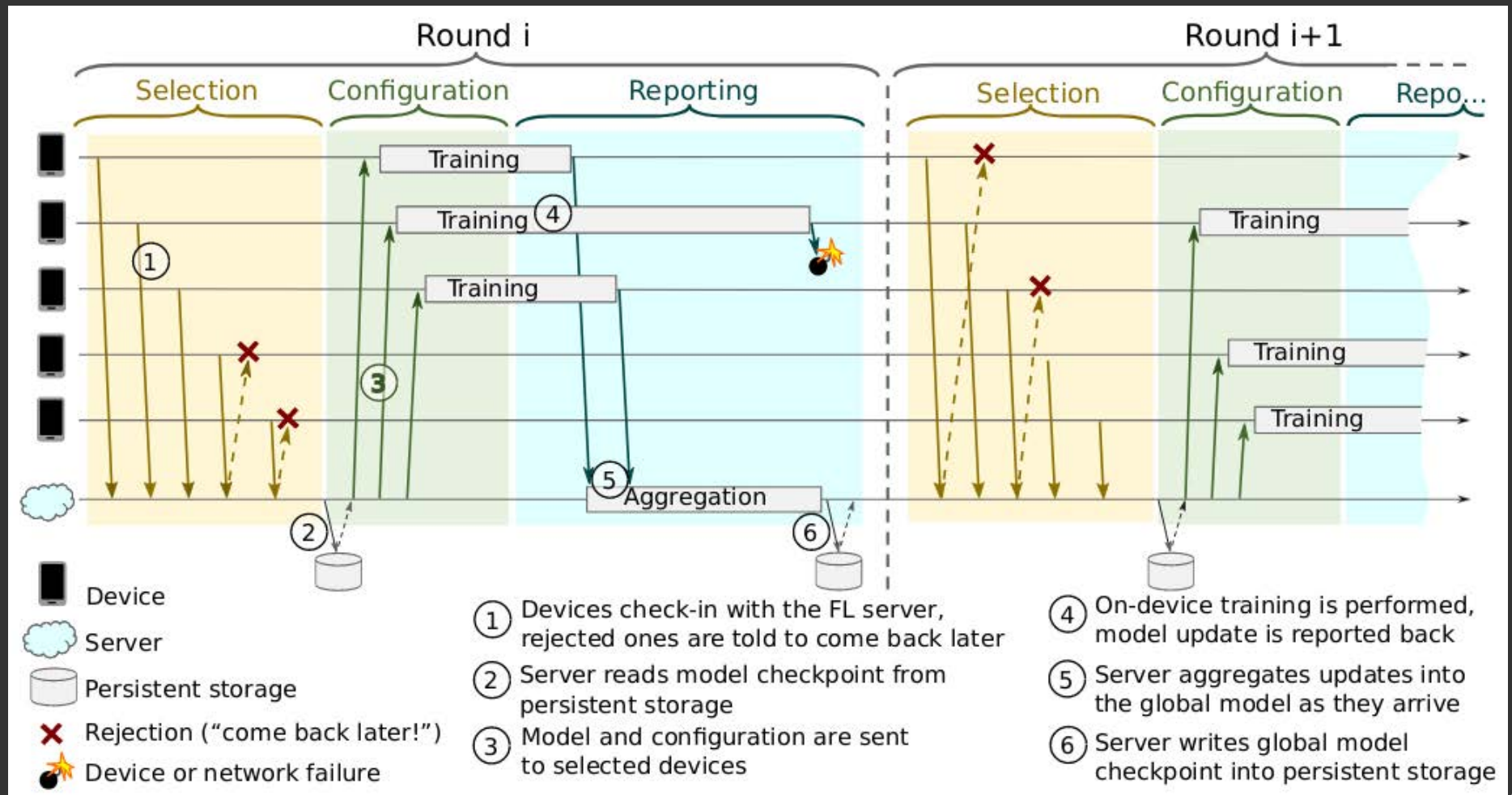
Federated Learning



*Bonawitz, Keith, et al. "Towards Federated Learning at Scale: System Design." SysML 2019.

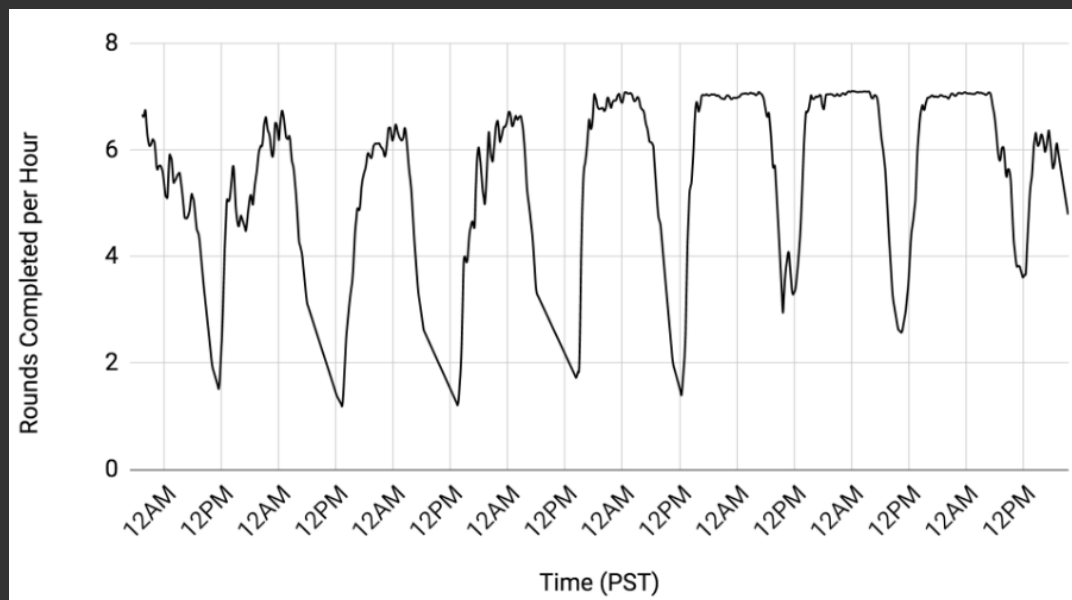
Federated Learning

➤ The network protocol:



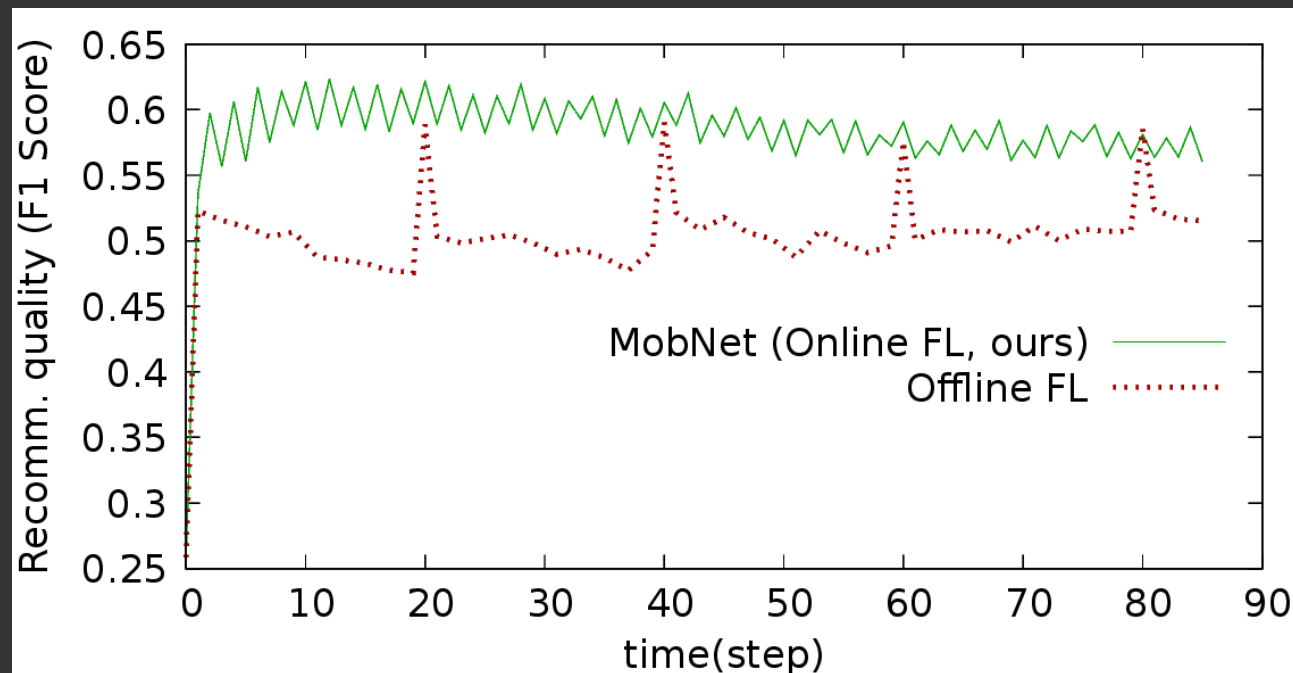
Federated Learning

- Designed for no impact on user experience
- Requirements for mobile devices
 - Idle, charging, connected to WiFi



Online vs. Offline FL

- Offline FL is not suitable for fast-data apps
 - e.g. social networks, recommenders
- Train 10 times more often → 23% better

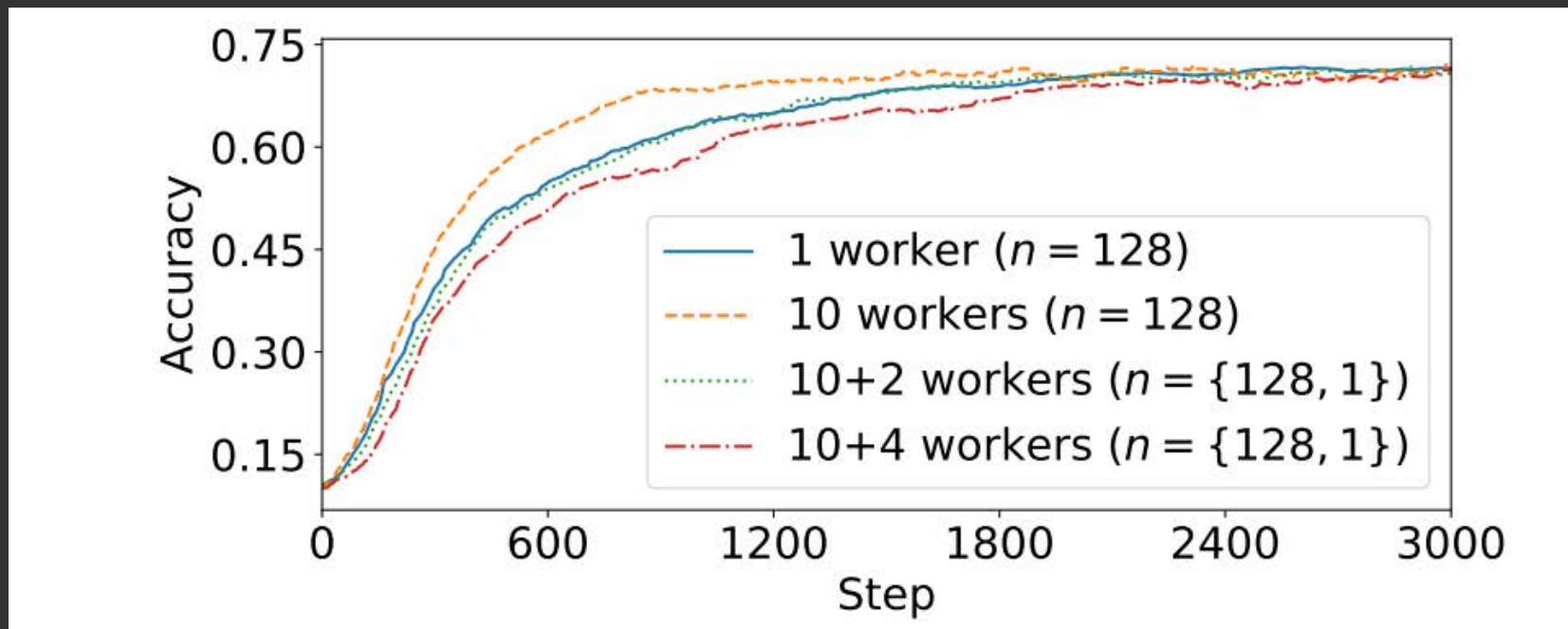


Drop the idle policy

- Gradient computation is compute-intensive
- Drop idle => gradient computation shares resources with the other applications
- AdaSGD: asynchronous rounds and shift decision-making to the mobile device

Drop the charging policy

- Drop charging => we need to predict and control the energy consumption
- Weak devices disrupt the training process



Conclusion

- The state-of-the-art federated learning is not suitable for fast-data applications
- We present MobNet:
 - AdaSGD – a staleness-aware learning algorithm
 - iProf – a regression-based profiler