

Blockchain and Federated Learning: Protecting and Securing Information

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- Rapid advancement of digital Information
- Data explosion
- Data collaboration
- Weakness of Traditional Data Sharing and Storage



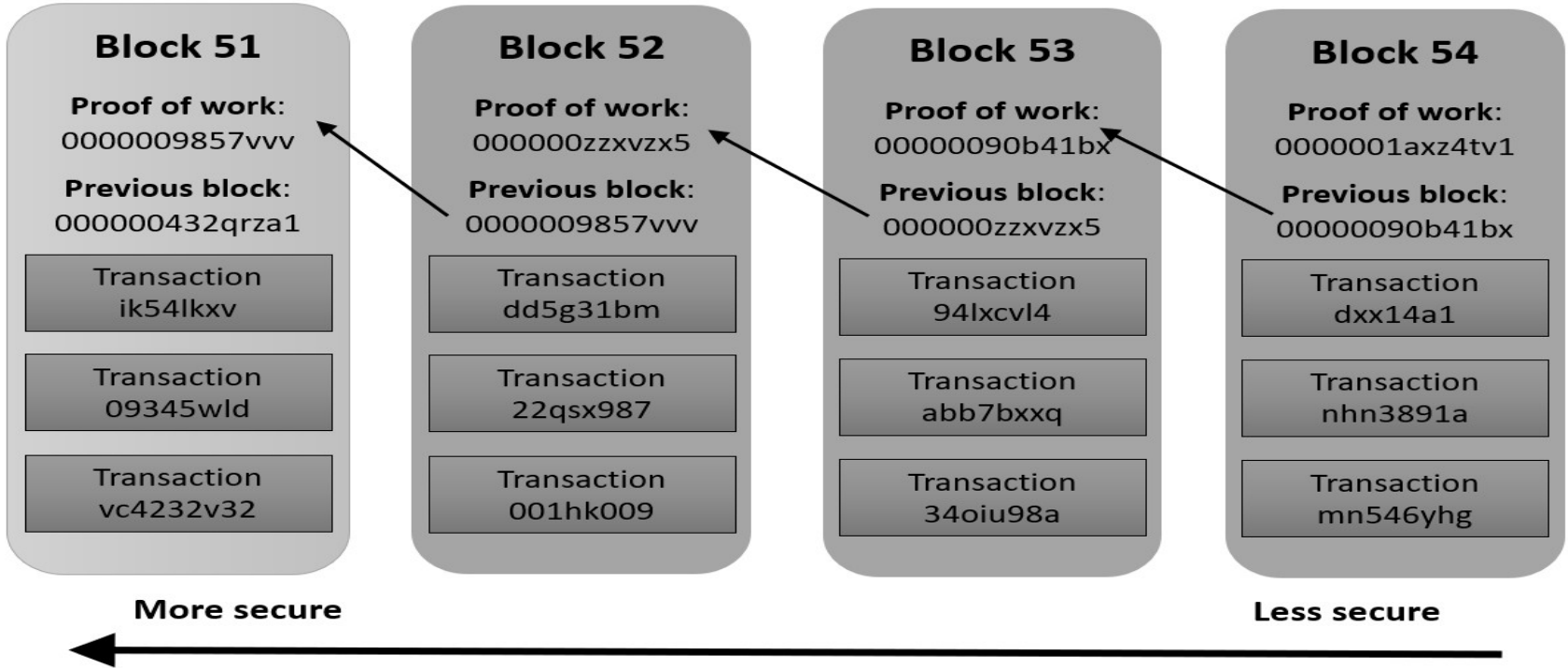
- Using blockchain to create an immutable audit trail for federated models
- Enhancing encryption between nodes and federated model to maintain better privacy-preservation



- Blockchain solves a challenging problem in **Data Science** of exchanging reliable information over unreliable network on which some of the participants cannot be **trusted**.



- A Database encompassing a physical chain of fixed length blocks that can include 1 to N transactions
- Each transaction added to a new block is validated
- When block is completed it is added to the end of existing chain of blocks



Blocks are “more secure” as you go further back in the chain

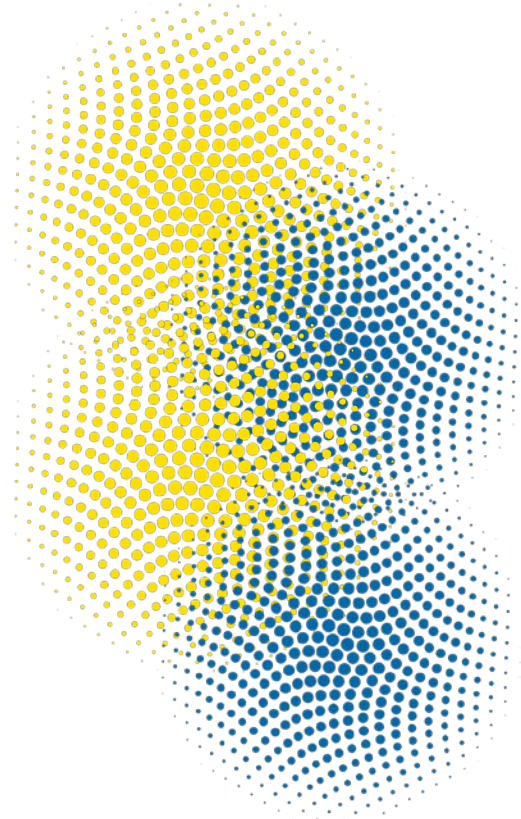


- Add new and undeletable transactions and organize them into blocks
- Cryptographically verify each transaction in the block
- Append the new block to the end of the existing immutable blockchain



Examples: Smart Contracts

- Finance
- Agriculture
- Health
- Supply Chain
- Smart Cities





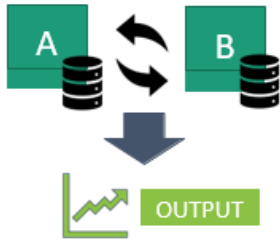
Data Collaboration– utilizing data across several institutions for the purposes of creating knowledge or providing services without sharing the data.



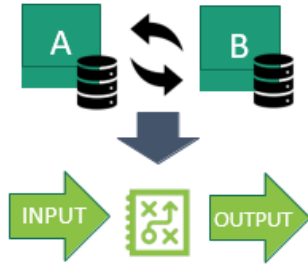


Data Collaboration Taxonomy

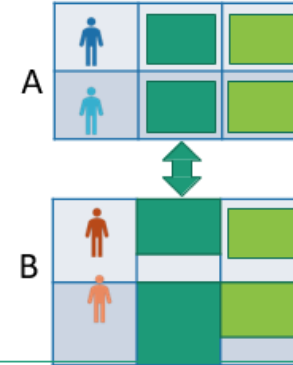
Data -Mining



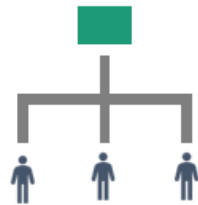
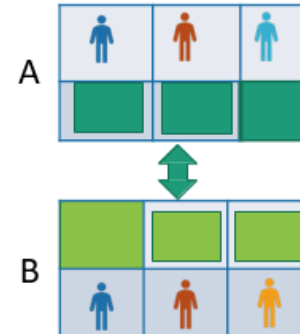
Machine-Learning



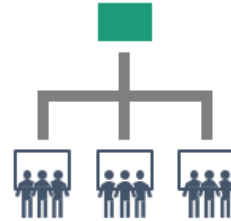
Horizontal Partition



Vertical Partition



On-device learning
(1 partition – 1 sample)

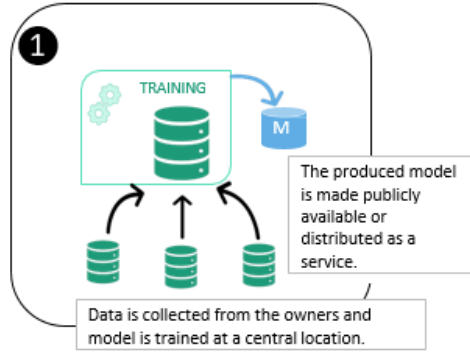


Institutional
Learning

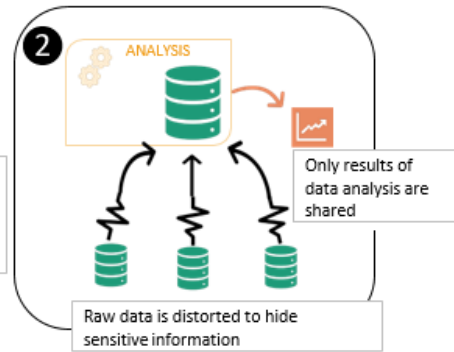


Data Collaboration Architectures

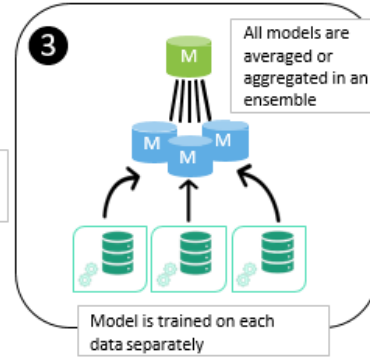
Conventional Machine Learning



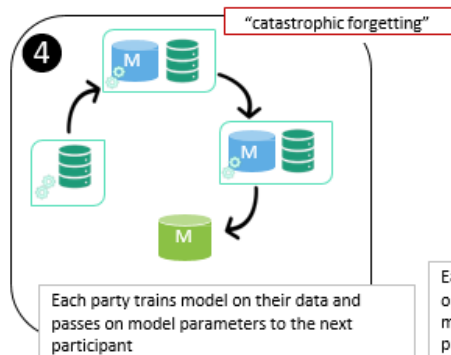
Privacy-Preserving Data Mining



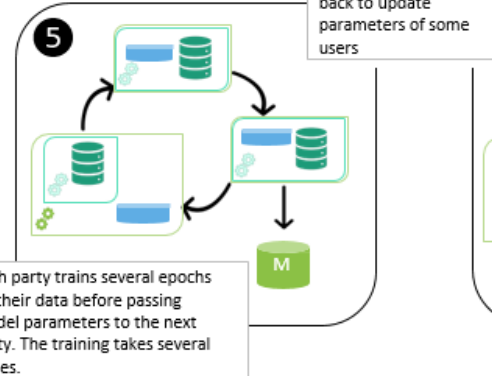
Model Ensembling



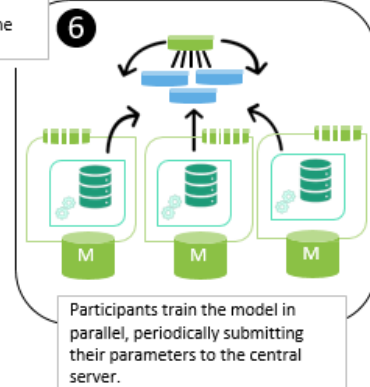
Incremental Learning



Cyclic Learning



Federated Learning





Privacy and Security of Federated Learning

Privacy Protection by design:

- Ephemeral
- Focused
- Aggregate

Data leakage to the central server

Individual model updates can be sensitive depending on the model (e.g. keyboard prediction model)

Inference attacks on the model

If the resulting model can be inspected or queried by the end user membership inference attacks are possible

Membership Inference Attacks Against Machine Learning Models. (Shokri et al. 2017)

Secure Aggregation:

SMC, homomorphic encryption

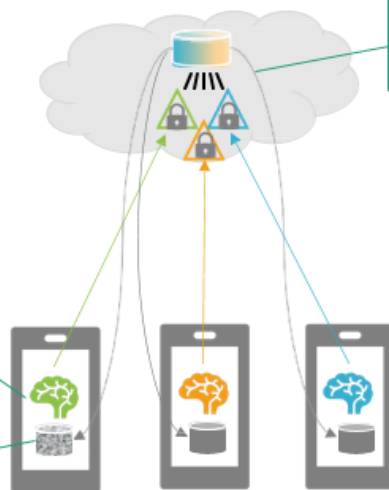
Practical Secure Aggregation for Privacy-Preserving Machine Learning. (Bonawitz et al. 2017)

Privacy Accountant

Deep Learning with Differential Privacy (Abadi et al. 2016)

Differential Privacy

Differentially Private Federated Learning: A Client Level Perspective. (Geyer et al. 2017).



Model Poisoning

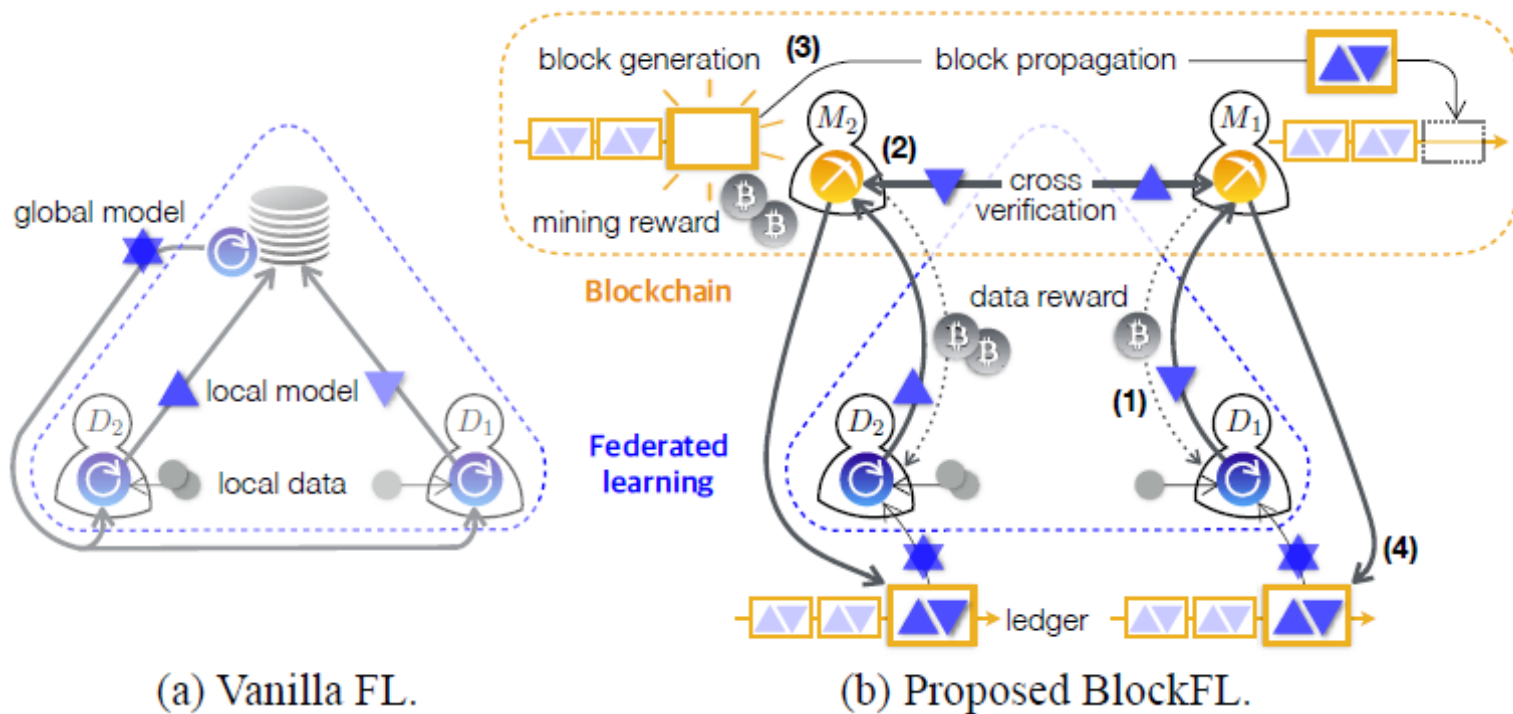
Vulnerable by design against a client-side adversary, because the central server is blind to user data

How To Backdoor Federated Learning. (Bagdasaryan et al. 2018)

Blockchain Technology

On-Device Federated Learning via Blockchain and its Latency Analysis. (Kim et al. 2018)

* None of the Privacy-Preserving Federated Learning Frameworks had been tested so far on real-life medical data





- Each node and uploads the local model update to its associated miner in the blockchain network
- Miners exchange and verify all the local model updates
- Miners generates a block where the verified local model updates are recorded
- The generated block storing the aggregate local model updates is added to a blockchain.
- Each device computes the global model update from the new block.

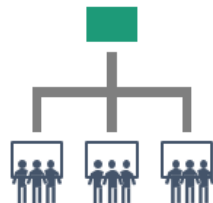


Medical Data Collaboration Use Cases

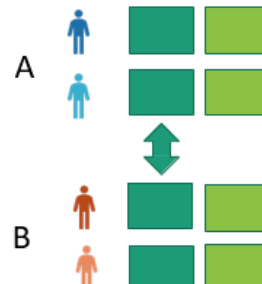
Clinical Decision Support System

- Computer-aided Interpretation of medical images
- COVID
- Design treatment plans
- Survival prediction

Multi-institutional



Horizontal





1. Blockchained On-Device Federated Learning Hyesung Kim et al.
2. Blockchain : Bambara, JJ and Allen, PR. McGraw Hill 2018.
3. Risk and Advantages of Federated Learning for Health Care Data Collaboration. ASCE Journal of Risk and Uncertainty Mangement. Bogdanova, A; Attoh-Okine, N; and Sakurai, T.



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