Towards Scalable, Efficient and Privacy Preserving Machine Learning
Rania Talbi, Sara Bouchenak

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Objectives

- Minimize the computational costs incurred by privacy preservation.
- Provide an end-to-end privacy preserving outsourced data classification service.
- Enable a set of mutually untrusted data owners to have a global vision on the union of their data without breaching the privacy of each one of them.
- Enable dynamic data model updates when new training data samples are available.

Design principles

- Cryptographic based protection (data model, training data, classification queries and responses)
- Partial homomorphic encryption (PHE ) based building blocks
- Combine PHE with cryptographic blinding (DTPKC cryptosystem [6])
- We implemented the VFDT incremental decision tree learning algorithm [7]

(1) Blind inputs
(2) Partially decrypt blinded values
(3) Decrypt blinded values
(4) Run operation over blinded values

Preliminary results

- We have used a synthetic dataset for fraud detection in a B2B network.
- This dataset contains 1000 bank transactions with 9 attributes each.
- We compare our work to the Ciphermed framework [8].

References

- [3] Liu et al.: Privacy-Preserving Mining of Association Rule on Outsourced Data from Multiple Parties. ASONAM 2018: 412-413