SecureMLlib: Privacy-**Preserving Distributed Machine** Learning

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TECHNOLOGY AND SCIENCE

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- The exponential growth of digital information is raising novel and tougher challenges for large-scale data analytics;
 - Outsourcing of computation to AI-based infrastructures;



Big Data & Al

- Legislations, such as GDPR or HIPAA, are blocking how data could be leveraged by new Al algorithms;
- Private data should be kept private;



Privacy-Preserving Distributed Machine Learning

Several threats and attacks end jeopardising the normal functioning of the machine learning pipeline.



Model Extraction

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How can we prevent data leakage in distributed machine learning frameworks?



Challenges

How can we guarantee that our models do not remember the training data and how can we prevent data leakage?



Challenges

How can we protect the intellectual property of our models?





Trained Model



The solution?

SecureMLlib

- By relying on Apache Spark, we offer:
 - Scalability;
 - High-Availability;
 - Different APIs for different purposes.

SecureMLlib

- Data encryption with different cryptographic primitives;
- Modification of MLlib algorithms to compute inside Intel's SGX.





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