



Privacy-friendly machine learning algorithms for intrusion detection systems

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- 1 Introduction
- 2 Intrusion detection systems
- 3 Multiparty computation
- 4 Methodology
- 5 Conclusion

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Privacy-friendly data pooling for enhancing intrusion detection systems



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Different types, based on where the intrusion takes place

- Network Intrusion Detection System (NIDS)
- Host Intrusion Detection System (HIDS)
- Hybrid Intrusion Detection System

Different detection methods

- Signature based
 - Advantages: accuracy and time
 - Disadvantages: only known intrusion types are detected
- Anomaly based
 - Advantages: new intrusion types can be detected
 - Disadvantages: malicious activity disguised as normal traffic can pass through
- Machine learning (classification)

Different types, based on where the intrusion takes place

- Network Intrusion Detection System (NIDS)
 - Advantages: detects attack before it occurs
 - Disadvantages: needs to be implemented on the network
- Host Intrusion Detection System (HIDS)
 - Advantages: collects broader data type
 - Disadvantages: needs to be implemented on each machine and only detects after the intrusion
- Hybrid Intrusion Detection System
 - Advantages: much more effective
 - Disadvantages: huge implementation necessary, not privacy-friendly

Privacy-friendly data pooling for machine learning network intrusion detection system



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Addition over \mathbb{Z}_2

- i players have each a secret number n_i
- they want to know if the sum of their numbers is even or uneven. $\sum n_i \mod 2 = 0$ or 1 ?
- they don't want anybody except them to know their number

- each player divides its number n_i into j $m_{i,j}$ parts. $\sum m_{i,j} = n_i$
- j players each receive the $m_{i,j}$ -part of each i players, sums it up and say if it is even or not. $\sum_i m_{i,j} \mod 2 = 0$ or 1 .
- the results of all j players is then summed up and is even if $\sum_i \mod 2 = 0$ and uneven otherwise. The problem is resolved.

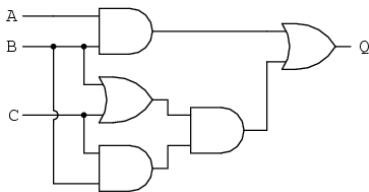
Solution

Garbled

- Boolean circuit
- Constant number of rounds
- Example on blackboard

Arithmetic

- Based on somewhat homomorphic encryption (SHE)
- Allows for addition and multiplication



Different leads

- Fairplay (boolean, n-party)
- BMR (boolean, n-party)
- Sharemind (3-party, proprietary)
- VIFF (obsolete)
- ABY (2-party)
- **SPDZ (arithmetic, n-party)**

Privacy-friendly collaborative network intrusion detection system

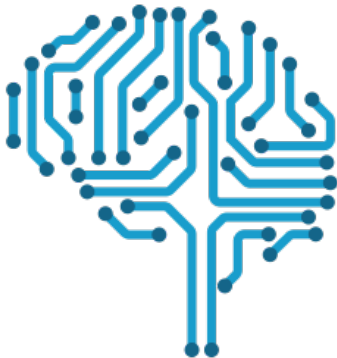


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Datasets

- KDD CUP '99
- AWID
- PCAP
- UNB ISCX
- CSIC 2010 HTTP Dataset
- West Point - NSA DataSet





Current research algorithms

- Semi-supervised learning algorithms with fuzziness
- K-Nearest Neighbors
- Support Vector Machines
- Bayes Classifier
- EM-Clustering
- Genetic algorithms
- Classification Tree

- Build on existing application (e.g. Bro Network Security Monitor, OpenNMS,...), in the form of a plug-in.
- Feed with constant new data, provided from other analysis tools
- User-friendly
- Off-line





- Begin March: Benchmark and designing working machine-learning algorithm
- Begin April: Design of final algorithm including MPC
- Begin May: Prototyping as a plug-in
- Begin June: Poster and report

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Further optimizations

- Hybrid Intrusion Detection Systems
- Speed and complexity optimizations (research on HE)



