

# **Chapter 3: Approaches for Private Computation**

**Lecture PETs4DS:  
Privacy Enhancing Technologies for Data Science**

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# The SPDZ Protocol for SMPC with Active Security

## SPDZ vs. CEPS

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- What does CEPS do?
  - Provides passive Security
  - But what if parties change the result ?
  - No active Security
- Design Goals of SPDZ
  - Efficient
  - Small Communication Overhead
  - Robust
  - Easy to set up
  - Secure against Malicious parties

# SPDZ – An Active Secure Protocol

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## SPDZ-Overview

- Commonly used SMPC protocol
- Divided in Offline and Online Phase
- Offline Phase
  - Parties share needed information for the function evaluation
  - Multiplication Triples for Multiplication Operations
  - MAC values for the correctness check
  - Parties communicate with each other, so this phase is “still online”, but not part of the actual calculation
- Online Phase
  - Evaluation of the function
- Security guarantee holds even if malicious majority (everybody malicious except for one player)

## Offline Phase

- Generate Multiplication Triples
  - Also called “Beaver Triples”
  - Consist of three numbers
  - $\langle a \rangle, \langle b \rangle, \langle c \rangle \Rightarrow a * b = c$
  
- Generate MAC Key
  - “Message Authentication Code”
  - Connected to the performed operations
  - Every party holds a part of the MAC Key
  - After the evaluation some properties still have to be satisfied for correctness



## Additive Shares

Given  $n$  parties  $P_1, \dots, P_n$  and a random value  $a$ , the value  $a$  is called additive shared between the parties when the following holds:

1. Every party holds a shared value  $a_i$
2.  $a = a_1 + \dots + a_n$
3. Party  $P_i$  has no information about  $a_j$ , when  $j \neq i$

- The shares are assumed to be random and evenly distributed.
- The protocol for additive sharing of values, is outside the scope of the lecture.

# SPDZ – An Active Secure Protocol

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## MAC Key

Consider  $n$  parties  $P_1, \dots, P_n$  and a random number  $\alpha$  NOT known to any party, called the MAC Key ("Message Authentication Code" Key).

The MAC Key is additive shared along the parties such that the following holds:

$$\alpha = \alpha_1 + \dots + \alpha_n$$

Note that every party  $P_i$  only holds  $\alpha_i$ , and does not know  $\alpha$ .

## MAC Value

Given  $n$  parties  $P_1, \dots, P_n$  and a fixed MAC Key  $\alpha$ .

Then the MAC value  $\gamma(a)$  can be additively shared in a finite field  $\mathbb{F}_p$  along the parties such that the following holds:

1.  $\gamma(a) = a * \alpha = \gamma(a)_1 + \dots + \gamma(a)_n$
2. Party  $P_i$  has no information about  $\gamma(a)$  and  $\gamma(a)_j$ , when  $j \neq i$
3. For a *public constant*  $c$  party  $P_i$  can easily compute a MAC value  $\gamma(c)_i = \alpha_i * c$ 
  - o note: party  $P_i$  can reconstruct  $c$

**Intuition about MAC values:** We will use this property of  $\gamma(a)$  to verify that no party has modified the results of its local computations. This enables active security.

## Notations

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### Representation of Values and MAC's

- In the **online** phase each shared value  $a \in \mathbb{F}_p$  is represented as follows

$$\langle a \rangle = ((a_1, \dots, a_n), (\gamma(a)_1, \dots, \gamma(a)_n))$$

where,  $a = a_1 + \dots + a_n$  and  $\gamma(a)_1 + \dots + \gamma(a)_n = \alpha * a$

- Every Party  $P_i$  holds  $(a_i, \gamma(a)_i)$  and a share of the MAC value  $\alpha_i$ .
- The interpretation is that  $\gamma(a) = \gamma(a)_1 + \dots + \gamma(a)_n$  is the MAC authenticating  $a$  under the global key  $\alpha$ .
- Both the shares  $a_i$ , and the corresponding MAC value  $\alpha_i$  have been pre-computed in the **offline** phase!

## Offline Phase-Generation of Multiplication Triples

- MAC Key
  - Generation of the MAC key is the first step during execution
  - Parties agree on a MAC key they will use to MAC their data
  - No individual party will actually know the full MAC key
  - The consistency of the result is checked using the MAC key
  - Any manipulated intermediate result can be determined that way
- Generate Multiplication triples using homomorphic encryption (HE)
  - Parties can perform passive secure MPC using HE
  - Malicious behavior possible
  - Idea to mitigate malicious behavior: “Sacrifice” Multiplication triple to check correctness of another triple (explained on the next slides)
    - Doubles the amount of multiplication triples required

## Online Phase – Addition

- Every party holds
  - additive share of  $\langle a \rangle = (a_i, \gamma(a)_i)$
  - additive share of  $\langle b \rangle = (b_i, \gamma(b)_i)$
- Every party computes
  - $a_i + b_i = c_i$
  - $\gamma(a)_i + \gamma(b)_i = \gamma(c)_i$
- Note this is a local operation and that we end up with
  - $c = \sum c_i = \sum(a_i + b_i) = (\sum a_i) + (\sum b_i) = a + b$
  - $\alpha * c = \sum \gamma_i(c) = \sum(\gamma_i(a) + \gamma_i(b)) = \alpha * a + \alpha * b = \alpha * (a + b)$

## Online Phase – Multiplication by Scalar

- Every party holds
  - additive share of  $\langle a \rangle = (a_i, \gamma(a)_i)$
  - Public constant  $c$
- Every party computes
  - $a_i * c = c_i$
  - $\gamma(a_i) * c = \gamma(a_i * c)$

## Online Phase – Addition by Constant

- Every party holds
  - Additive share of  $\alpha = \alpha_i$
  - additive share of  $\langle a \rangle = (a_i, \gamma(a)_i)$
  - public constant  $c$
- Add public constant  $c$  to  $\langle a \rangle$
- $\langle a \rangle + c = ((\textcolor{red}{a_1 + c}), a_2 \dots, a_n), (\gamma(a)_1 + c * \alpha_1, \dots, \gamma(a)_n + c * \alpha_n)$
- Note:
  - $a + c = (a_1 + \dots + a_n) + c$
  - $\gamma(a + c) = \gamma(a)_1 + \dots + \gamma(a)_n + c * \alpha = \gamma(a)_1 + \dots + \gamma(a)_n + c * \alpha_1 + \dots + c * \alpha_n$   
 $= (\gamma(a)_1 + c * \alpha_1) + \dots + (\gamma(a)_n + c * \alpha_n)$

## Online Phase – Multiplication

- Every party holds
  - additive share of  $\langle a \rangle = (a_i, \gamma(a)_i)$
  - additive share of  $\langle b \rangle = (b_i, \gamma(b)_i)$
- To **open** a value  $\langle a \rangle$  means: Every party sends their share  $a_i$  of  $\langle a \rangle$  to all other parties, so that every party can recover the actual value of  $a$ .  
This does include sending  $\gamma(a)_i$
- To **partial open** a value  $\langle a \rangle$  means:  
Every Party sends their share  $a_i$  of  $\langle a \rangle$  to  $P_i$  who computes  $a = a_1 + \dots + a_n$   
This does NOT include sending the respective  $\gamma(a)_i$  as well!  
After that  $P_i$  broadcasts  $a$  to all other Parties

## Online Phase – Multiplication

- Use multiplication triple generated in the offline phase

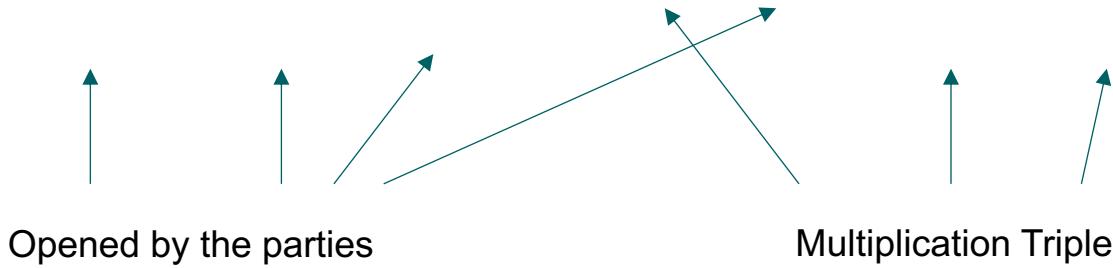
$$\langle x \rangle, \langle y \rangle, \langle z \rangle \Rightarrow x * y = z$$

- For each multiplication a new triple is needed
- We want to compute  $\langle a * b \rangle$  of the shared values  $\langle a \rangle, \langle b \rangle$ 
  - Every party “**partially opens**” **masked values of a and b.**
    - $\epsilon = a - x$
    - $\delta = b - y$
  - After this,  $\langle a * b \rangle$  can then be computed via the following local operation:

$$\langle a * b \rangle = \langle z \rangle + \langle y \rangle * \epsilon + \langle x \rangle * \delta + \epsilon * \delta$$

## Online Phase – Multiplication

$$\begin{aligned}\langle a * b \rangle &= \langle (a - x + x) * (b - y + y) \rangle \\ &= (a - x) * (b - y) + (b - y) * \langle x \rangle + (a - x) * \langle y \rangle + \langle x * y \rangle\end{aligned}$$



$$= \epsilon * \delta + \delta * \langle x \rangle + \epsilon * \langle y \rangle + \langle z \rangle$$

Please keep in mind that  $\epsilon$  and  $\delta$  refer to the masked values of  $a$  and  $b$  in this formula.

## How to share the Input of Party $P_i$

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- Problem: Party  $P_i$  wants to additively share its input  $a$
  - This means we want to compute  $\langle a \rangle = ((a_1, \dots, a_n), (\gamma(a)_1, \dots, \gamma(a)_n))$  such that:
    - $a = a_1 + \dots + a_n$
    - $\gamma(a) = \gamma(a)_1 + \dots + \gamma(a)_n$
  - Assume offline phase provides a random  $\langle r \rangle$ .
    - $\langle r \rangle$  is opened to  $P_i$ : This means only  $P_i$  knows  $r$
  - $P_i$  broadcasts  $\Delta = a - r$ 
    - now  $a$  is masked with a value only  $P_i$  knows
  - Then all players compute  $\Delta + \langle r \rangle = \langle a \rangle$ 
    - Now all players hold a share  $(a_j, \gamma(a)_j)$  without knowing the actual value of  $a$
    - This is an Addition with a public constant (see the relevant slide for details)
  - **Note:** This method shares  $a$  additively in a secure way and computes MAC values for  $a$  based on the MAC values of  $r$ ! In this way, pre-computed  $r_i$  and  $\gamma(r)_j$  from the offline phase are used to share  $a$  in the on-line phase!
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## Verifying Correctness at the end of the circuit evaluation – Part 1

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- Each party  $i$  has an agreed set of  $t$  partially opened values

$$\Omega_j, \quad 1 \leq j \leq t$$

- and each one has a sharing of the associated MAC value

$$\gamma(\Omega_j)_i, \quad 1 \leq j \leq t$$

- Each party  $i$  also has a share of the MAC key

$$\alpha_i.$$



## Verifying Correctness at the end of the circuit evaluation – Part 1

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- Each party  $i$  computes

$$\Omega = \sum_{j=1}^t \Omega_i$$

- They also compute their share of the MAC on  $\Omega$

$$\Gamma_i = \sum_{j=1}^t \gamma(\Omega_j)_i$$

- And then

$$\sigma_i = \Gamma_i - \alpha_i * \Omega$$

- Note, if all is correct then  $\sigma_i$  is a sharing of zero
  - Then all check whether  $\sigma_1 + \dots + \sigma_n = 0$  (by broadcasting  $\sigma_i$ )



## Summary of online phase for SPDZ protocol

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1. **Initialisation:** The parties agree on MAC key  $\alpha$ , a set of multiplication triples and a set of random values to use. All of these values have been pre-computed and pre-shared in the offline phase, and no party knows the actual values corresponding to these shared values.
  2. **Input:** Each party shares its inputs to all other parties as described earlier.
  3. **Computation:** The circuit is evaluated using addition and multiplication.
  4. **Output:** The output of the circuit is recovered and all parties verify the correctness of the result and all partially opened values.
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- **The details of the offline phase of SPDZ are not covered in the lecture**

# Complexity of SPDZ

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- Offline Phase
  - Expensive part of the calculation
  - Generation of the Multiplication Triples has to be done with HE
  - Correctness check of triples consumes triple
  - Check if  $\langle x \rangle * \langle y \rangle = \langle z \rangle$  holds using triple and MAC
  - $O(n^3)$
  
- Online Phase
  - All Operations can be calculated in  $O(n)$
  - Produces Overhead of  $O(n)$  + complexity of the arbitrary function

# **Improvements of SPDZ: MASCOT and SPDZ 2.0**

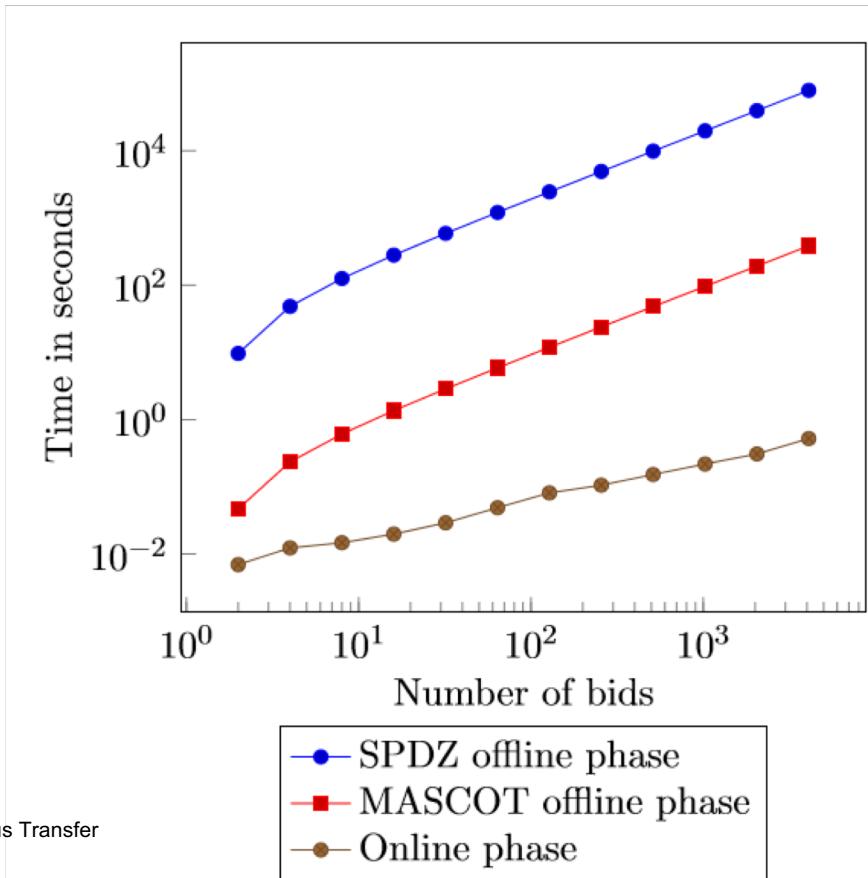
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- The offline phase is a bottleneck of the SPDZ protocol
- FHE for the multiplication triple generation is expensive
  - $O(n^3)$
- MASCOT tries to improve offline phase
  - Oblivious Transfer for Multiplication Triple Generation is much faster
  - FHE isn't needed anymore
  - Parallelism of generation possible
- **Online phase stays the same**

## SPDZ vs MASCOT

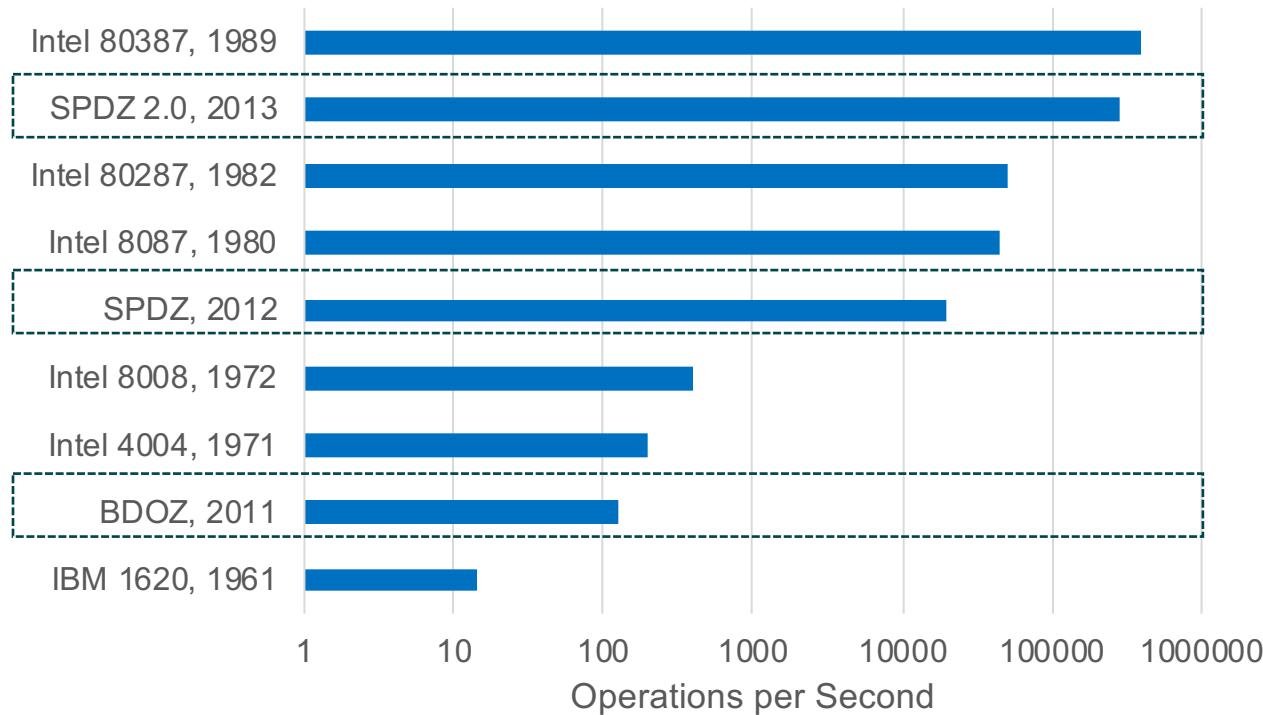
- Implementation of Vickrey auction
  - Every party sends bid privately
  - Party with highest bid wins, but only has to pay second highest bid
- MASCOT offline phase is ~ 200 times faster

MASCOT: Faster Malicious Arithmetic Secure Computation with Oblivious Transfer  
<https://eprint.iacr.org/2016/505.pdf>



- Reduction of shares is done by a *program dependency graph*  $G = (V, E)$ 
  - Vertices relate to instructions
  - Edges represent flow between instructions
  - Merge nodes which don't have any preconditions
  - Merged instructions can be done in one open sharing procedure
- Off-line phase not just for multiplication triples but also for pre-sharing of secrets for other operations:
  - Square  $b = a^2$
  - Inverse  $a^{-1} = a$
- Rearrange Operations
  - $a \oplus b = (a - b)^2$  is expensive normally
  - Through bit decomposition only one open sharing is needed, because  $a \oplus b = a + b - 2 * a * b$  in binary
- Polynomial approximation of  $\sin(x)$  can be reduced from 5874 OS to 759 OS

## Performance of the SPDZ Protocols



Based on : <https://www.cs.bris.ac.uk/home/ps7830/spdz2.pdf>

# Summary of SPDZ and Active Security for SMPC

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## Discussion

- Advantages

- Secure Multiparty Computations works fine
- Open source implementations available e.g. on github
- Detection of malicious behavior even if  $n-1$  parties are malicious
- Fixed settings might be practicable since the offline phase can be run in background

- Disadvantages

- Bad Scaling
- Computational power of the 80s
- How to reuse results ? If one party is malicious the whole calculation gets worthless
- Changing settings require new preprocessing

⇒ Usage of SMPC in practical applications is currently limited due to the overheads in performance and scalability, but future potential is huge!

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# Secure Multi-party Computation: past thesis topics 1/2

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Master thesis:

Jonas Nagy-Kuhlen

Interactive Private Multi-Party Calendar Scheduling

## Motivation

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- Popularity of calendar scheduling applications
  - Appointments can be arranged easily
- Increased productivity for companies and individuals



	Okt 18 MI	Okt 19 DO	Okt 21 SA	Okt 24 DI	Okt 28 SA	Nov 2 DO
Thomas	✓	✓		✓		
Marie		✓	✓			
Hans	✓	✓	✓	✓	✓	✓
Julia		✓				

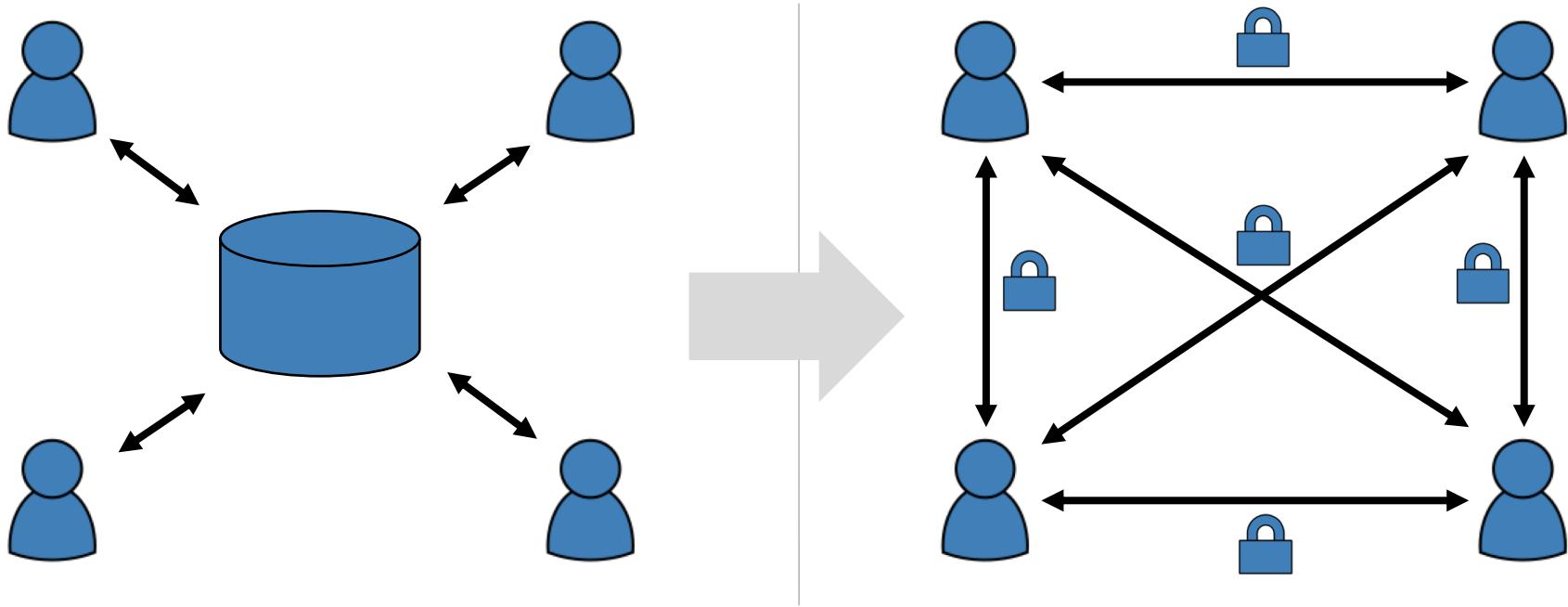
<https://doodle.com>



## Motivation

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Can we do better?



# Goal: Development of a Secure Calendar Application

Private Calendar Scheduling

General Help Testing

PETs4DS Working Group Meeting [Best Match] X

A public discussion of new cryptographic techniques that allow to maintain privacy in data science.

✓ Alice	<input type="checkbox"/> Wednesday, February 28, 2018 from 12:00 to 13:00
✓ Bob	<input checked="" type="checkbox"/> Wednesday, February 28, 2018 from 14:00 to 15:00
	<input checked="" type="checkbox"/> Wednesday, February 28, 2018 from 16:00 to 17:00
	<input type="checkbox"/> Thursday, March 1, 2018 from 12:00 to 13:00
	<input type="checkbox"/> Friday, March 2, 2018 from 12:00 to 13:00
	<input checked="" type="checkbox"/> Friday, March 2, 2018 from 14:00 to 15:00
	<input checked="" type="checkbox"/> Friday, March 2, 2018 from 16:00 to 17:00

Ready      Import Calendar      Resend Invite

Participating as user Alice. [Change Name](#)

Private Calendar Scheduling

General Help Testing

PETs4DS Working Group Meeting [Best Match] X

A public discussion of new cryptographic techniques that allow to maintain privacy in data science.

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	<input checked="" type="checkbox"/> Friday, March 2, 2018 from 14:00 to 15:00
	<input checked="" type="checkbox"/> Friday, March 2, 2018 from 16:00 to 17:00

 Scheduling completed. The selected date is

Friday, March 2, 2018 from 14:00 to 15:00

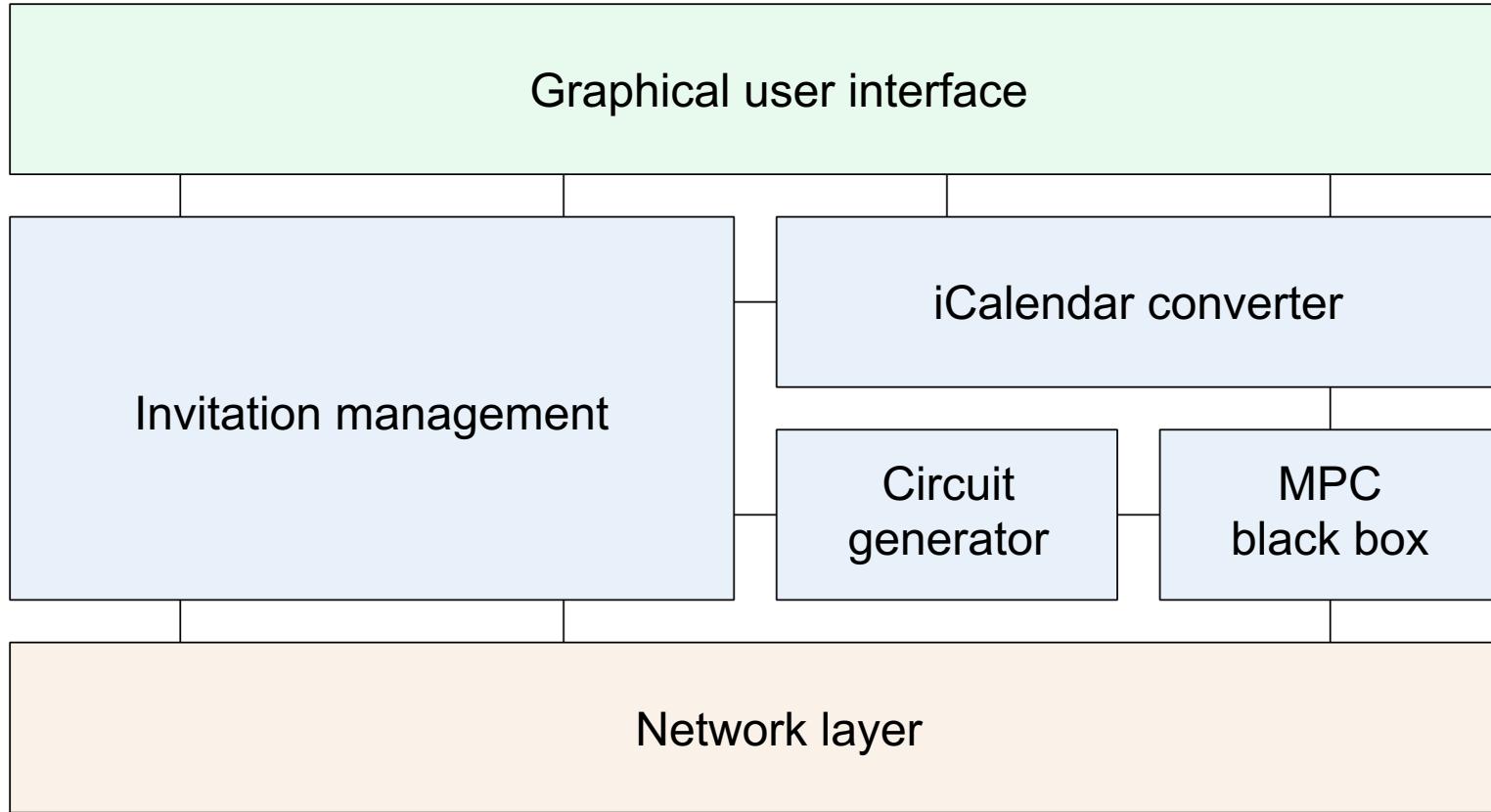
[Export appointment](#)

Ready      Import Calendar      Resend Invite

Participating as user Alice. [Change Name](#)

# High-Level Components

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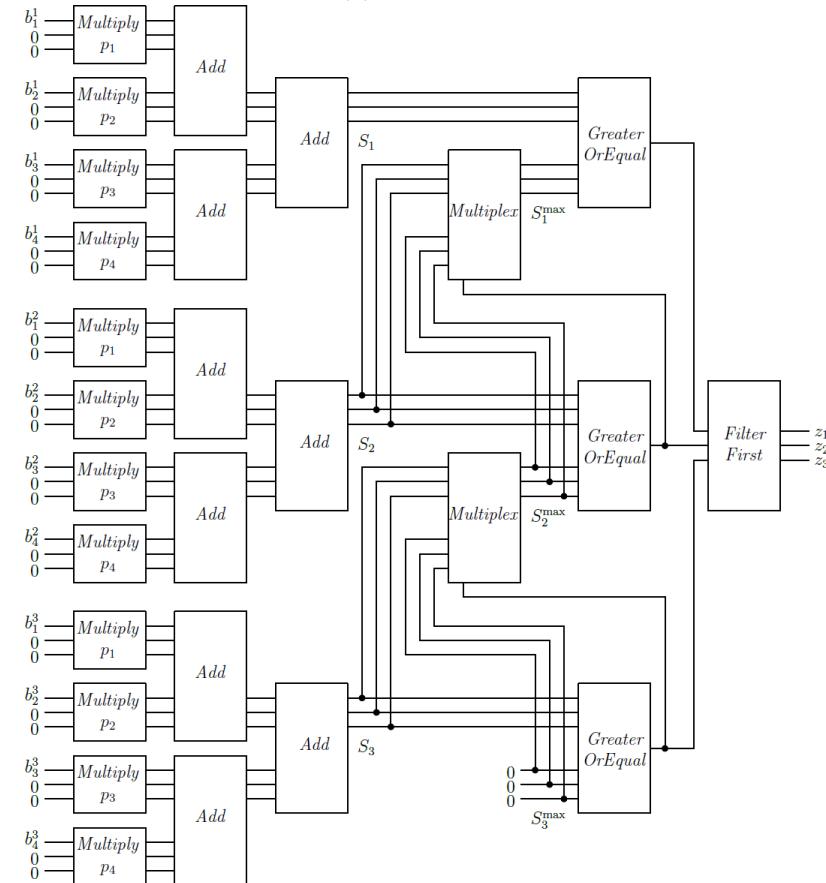


## Weighted best match

- Integer addition allows to accumulate scores for each time slot
- Comparison and multiplexing blocks allow to find index with highest score

Number of parties  
Number of time slots

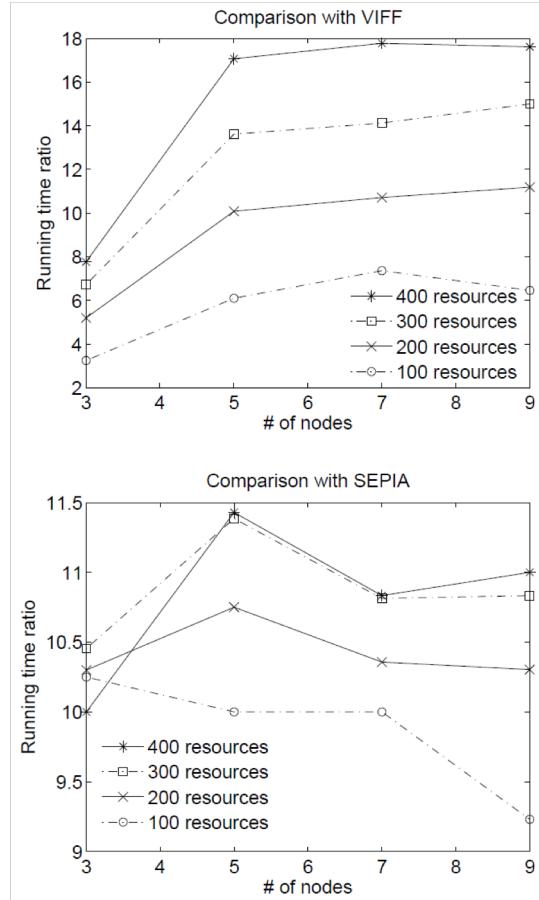
$n = 4$   
 $m = 3$



## Why use SCAPI as MPC library?

- GMW protocol is secure in semi-honest setting
  - Security against malicious adversaries comes with additional costs
- Boolean circuits considered more flexible than arithmetic circuits in this use case
  - Only small fixed-size integer logic required
- Well-documented and well-structured
  - Can be easily integrated into Java code

“Secure Multi-Party Computation of Boolean Circuits with Applications to Privacy in On-Line Marketplaces”, Choi et al., 2012



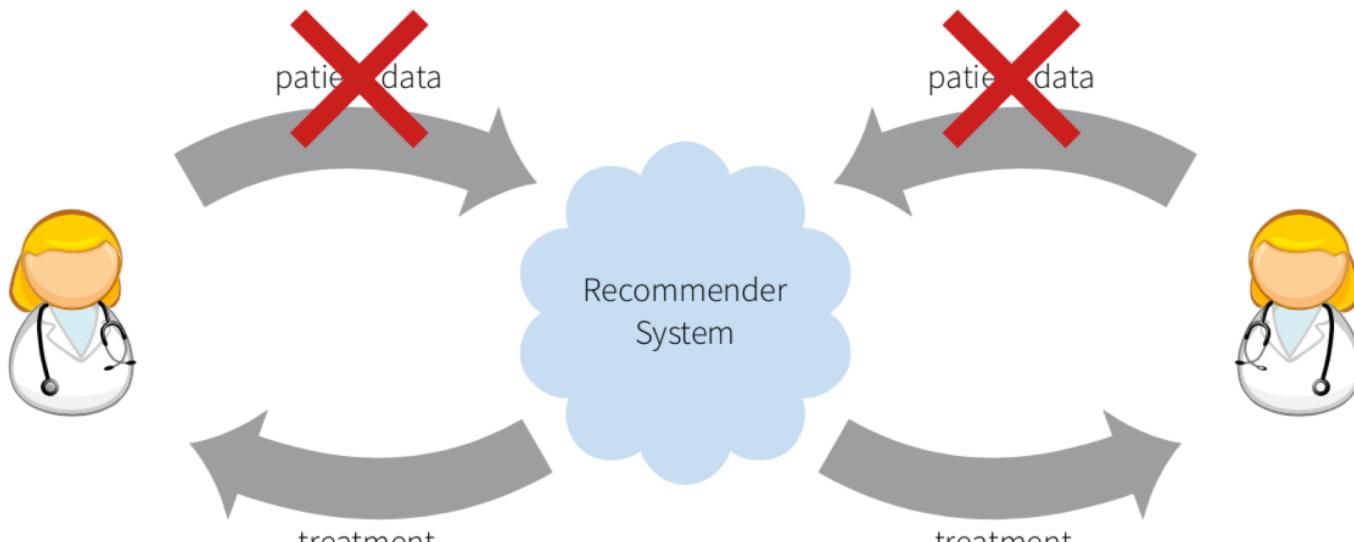
## Secure Multi-party Computation: past thesis topics 2/2

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Master thesis:  
Thibaud Kehler  
Privacy-preserving collaborative filtering with SPDZ

## Use Case

### Medical Data



Patient data is confidential by law!

## Threat Analysis

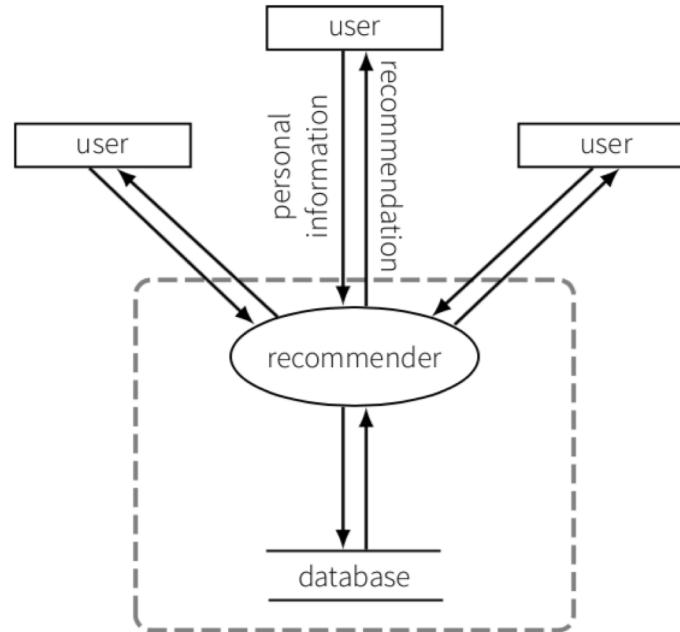
### Items of Interest (IOI)

- personal information
- recommendations

**Linkability** Are two IOIs linked?

**Identifiability** Who is the subject of an IOI?

**Disclosure of Information** IOI is disclosed to some unauthorized entity.



## Requirements

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**Accuracy** Recommendations/Predictions are valuable to the user.

### Algorithmic Complexity

- Responses in reasonable time
- Good scaling behaviour

### Privacy

- Unlikability
- Anonymity
- Confidentiality

**Robustness** A user of the recommender system has only minimal influence on the output.

**Does collaborative filtering with SPDZ meet these requirements?**

## Summary of our approach

### User-Based Model

1. Enter the rating matrix as private input
  - in plain representation.
  - in sparse representation.
2. Build a user-based cosine similarity model.
3. Predict the ratings securely.
4. Reveal predictions only to their users.

#### Build Model

Plain Ratings  $\Theta(n^2 \cdot m)$

Sparse Ratings  $\Theta(n^2 \cdot c^2)$

Approx. k-NN Predictions  $\Theta(f' \cdot n)$

### Item-Based Model

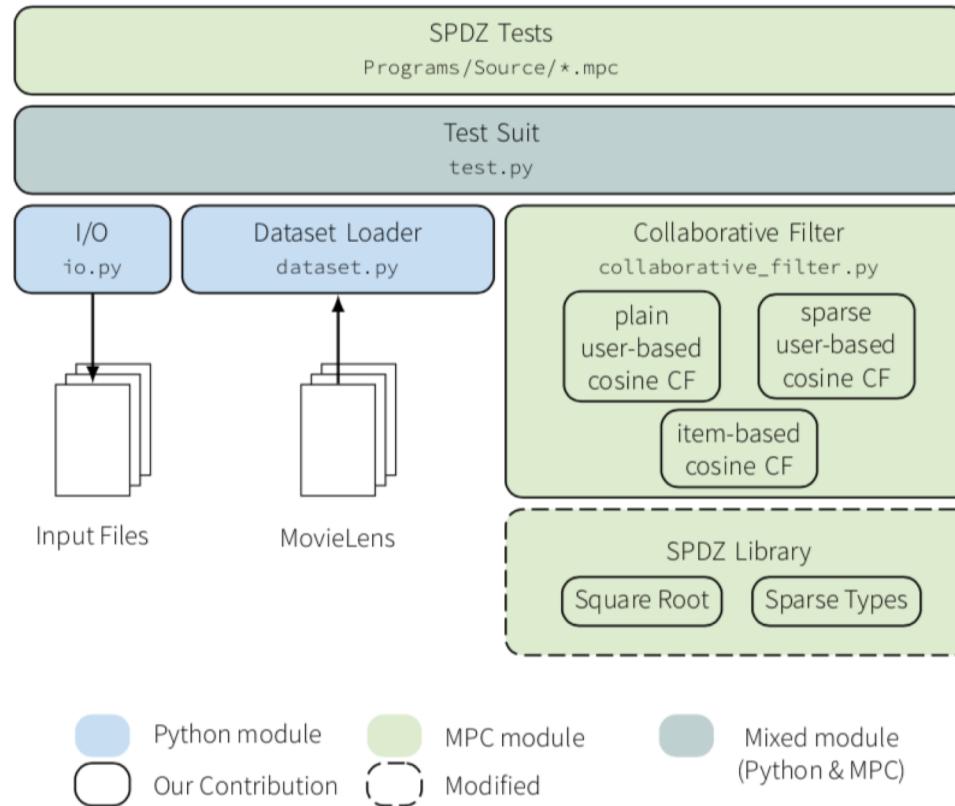
1. Enter the rating matrix as private input
  - in plain representation.
2. Build an item-based cosine similarity model
3. Reveal the similarity model.
4. Users predict their ratings locally.

#### Build Model

Plain Ratings  $\Theta(m^2 \cdot n)$

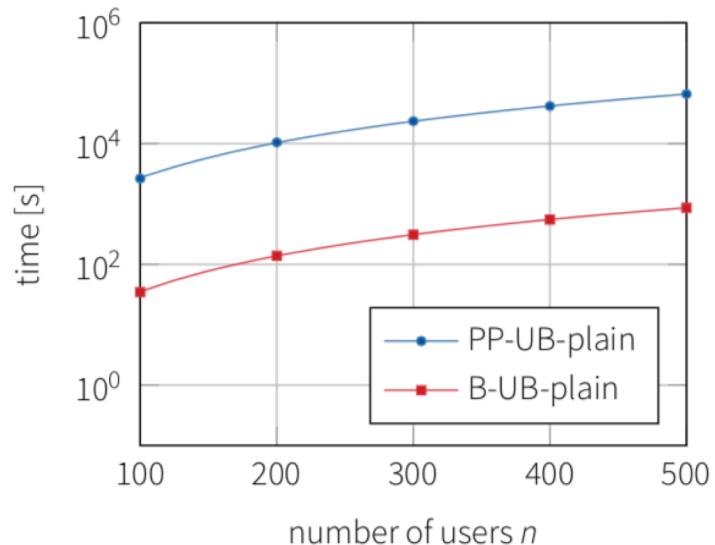
k-NN Predictions Computed locally

## Implementation



## Performance

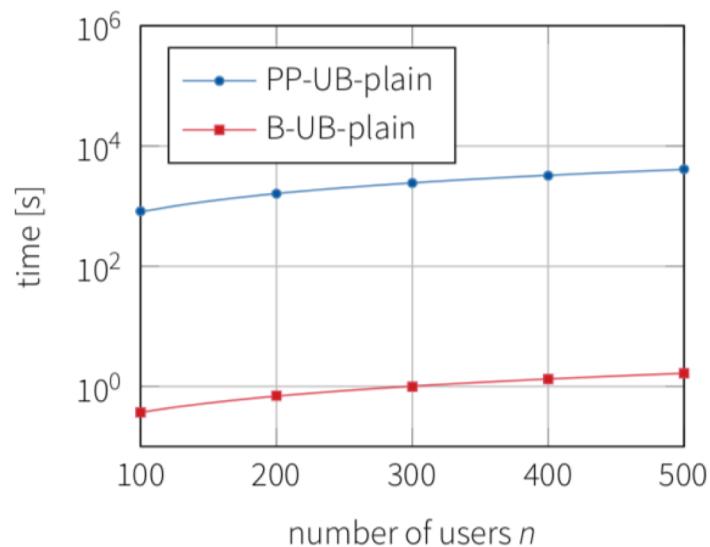
### User-Based Model



- 3000 items
- $\approx 78$  times slower

## Performance

### User-Based Predictions



- 3000 items
- $k = 11, f' = 4$
- $\approx 2560$  times slower

## Summary of Chapter 3

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# Comparison of Approaches for Private Computation

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- **Homomorphic Encryption (HE)**
  - Encrypt data before sending to cloud
- **Secure Multi-Party Computation (SMPC)**
  - Compute results together without sharing private data



# Comparison of Approaches for Private Computation

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Approach	Adversary Type	Confidentiality	Integrity	Requires Interaction
Homomorphic Encryption (HE)	Honest-but-curious (HBC)	YES	NO	NO
Secure Multi-Party Computation (SMPC)	Honest-but-curious (HBC) or Malicious	YES	YES	YES

# Homomorphic Encryption: Overview of Properties

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- **Confidentiality:** yes
  - Compute and storage nodes do not have access to the unencrypted data
- **Integrity:** no
  - Modification of data is possible
  - Faulty computation is possible
- **Adversary type addressed:** honest-but-curious adversaries
  - No party can learn anything from the other party
  - But no guarantee about malicious behavior of any party
- **Requires interaction:** no
  - HE is possible with only one party, with two parties and with three or more parties
  - However, the same key has to be used by all parties wishing to share results

# Secure Multi-Party Computation : Overview of Properties

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- **Confidentiality:** yes
  - No node learns anything new, besides the result of the computation.
- **Integrity:** yes
  - Intermediate results have to be shared, which allows verification of correctness.
- **Adversary type addressed:** malicious and HBC
  - Most SMPC schemes allow  $t < \frac{n}{2}$  HBC adversaries or  $t < \frac{n}{3}$  malicious adversaries, but higher bounds also possible.
- **Requires interaction:** yes
  - SMPC usually requires 3 nodes or more.
  - Confidentiality and integrity of SMPC are enforced through interaction between nodes.

# Business use case for computation on secret data

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## Show me the money!

- Important applications of Data Science:
  - Trends / forecasting
  - Clustering
  - Anomaly detection
- Common aspects:
  - Aggregation of data from multiple parties
  - Computation adds value to existing data
  - Computation requires sharing of data
  - Sensitive nature of input **and** output data
- Current approach: centralised computation
  - Requires trusting central party
  - Easy if everybody on the same team
  - Problem for multiple companies / jurisdictions / countries
- Whenever the participants do not trust each other, but the data analytics result is worth a lot of money, and processing time is unimportant, then SMPC and HE should be considered.



## Revisiting the goals of this chapter

**Two approaches for computation with secret data:**

- Homomorphic encryption (HE)
- Secure multi-party computation (SMPC)

**Multiple implementations for each approach:**

- HE: RSA, Paillier Cryptosystem
- SMPC: YGC, CEPS, SPDZ

**Differences:**

- Number of parties
- Complexity overhead
- Different guarantees (confidentiality, integrity, interaction)

**Challenges:** Computational overhead + architecture!

