

# Secure and Efficient Computing on Private Data

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14/04/2023



Universiteit  
Leiden

Bij ons leer je de wereld kennen

# Media Player Classic – MPC



# An Information-Driven Society

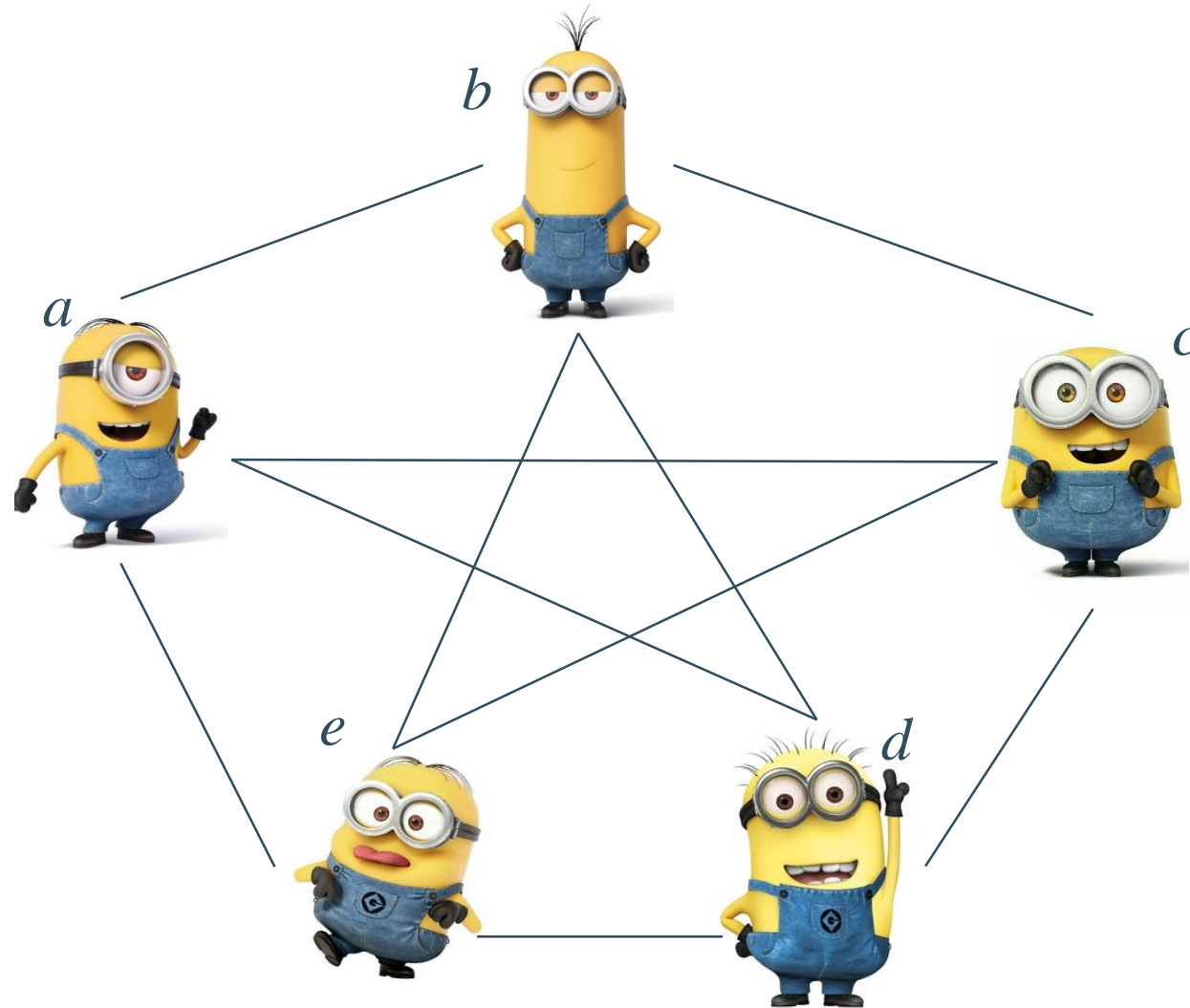


A close-up photograph of two hands, one on the left and one on the right, holding a blackboard. The hands are positioned at the top, with the fingers slightly curled, framing the word 'PRIVACY' which is written in white chalk on the black surface. The lighting is soft, highlighting the texture of the skin and the grain of the chalk. The background is a solid, dark color, making the white text stand out prominently.

PRIVACY

*“How to allow the collection and purposeful processing of private data, without compromising individual privacy?”*

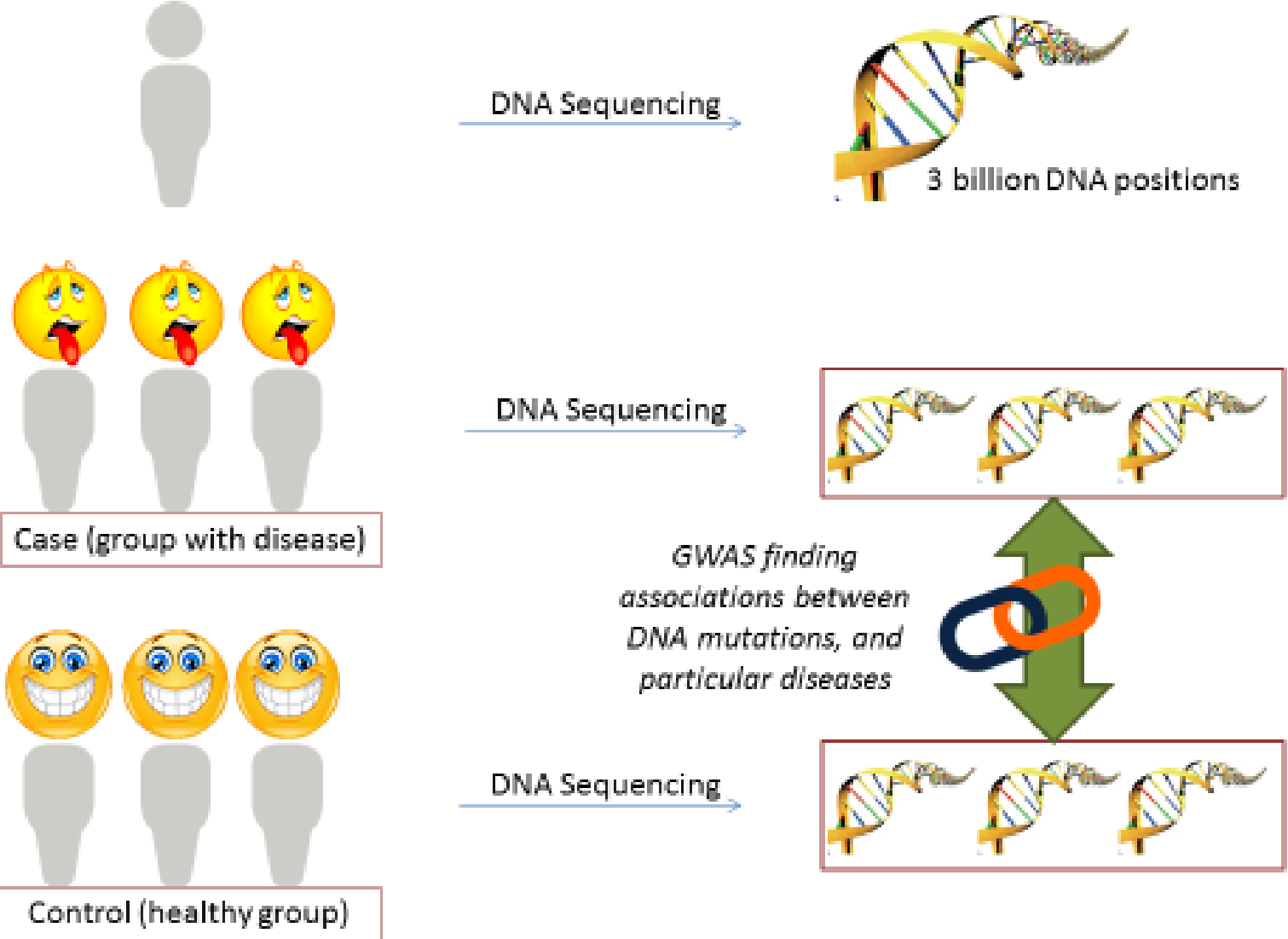
# Multiparty Computation – MPC



→ Securely compute  $f(a, b, c, d, e)$ .

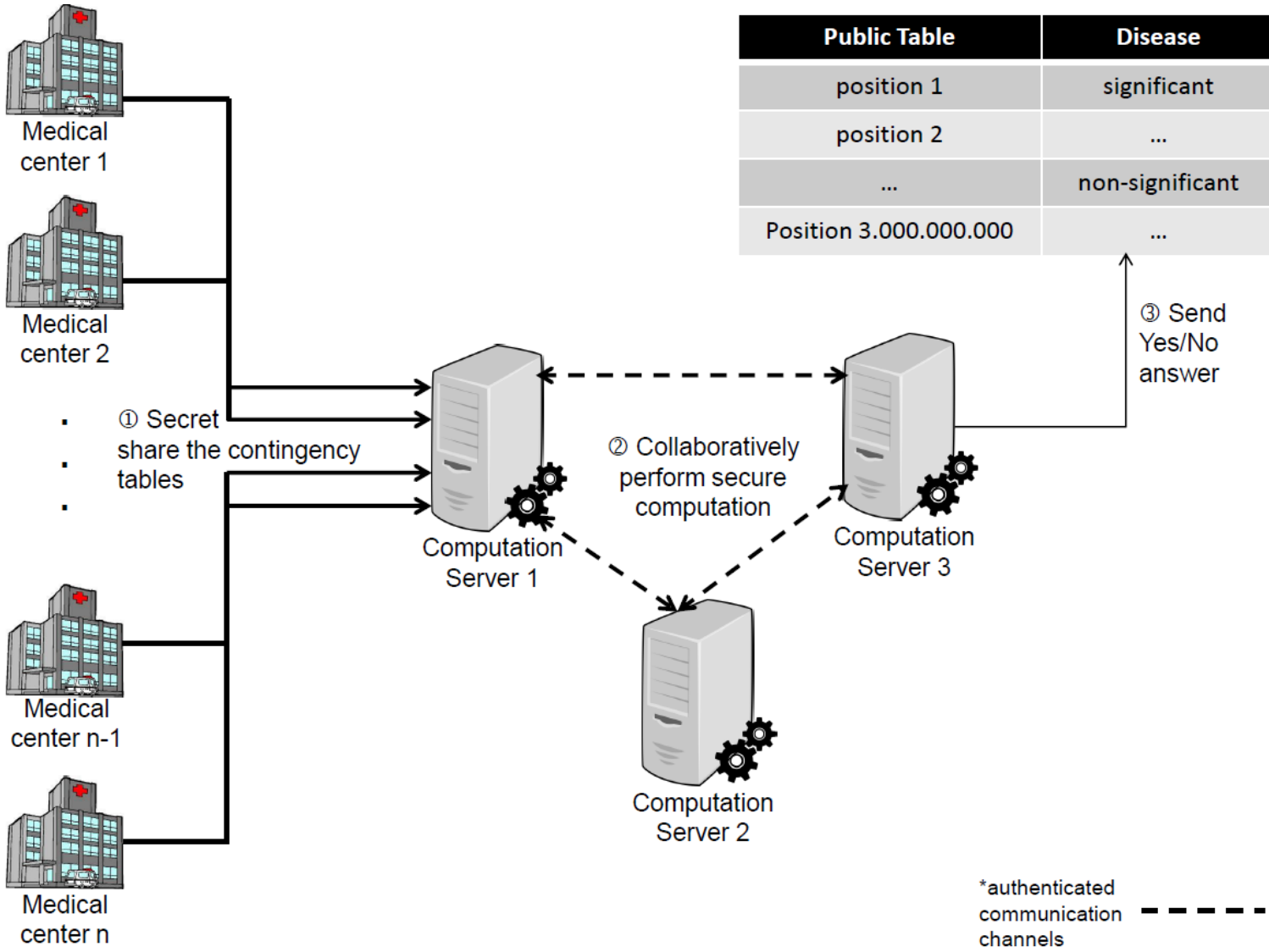
Application Scenario 1:  
Privacy-Preserving Genome-Wide Association Studies

# Privacy-Preserving GWAS

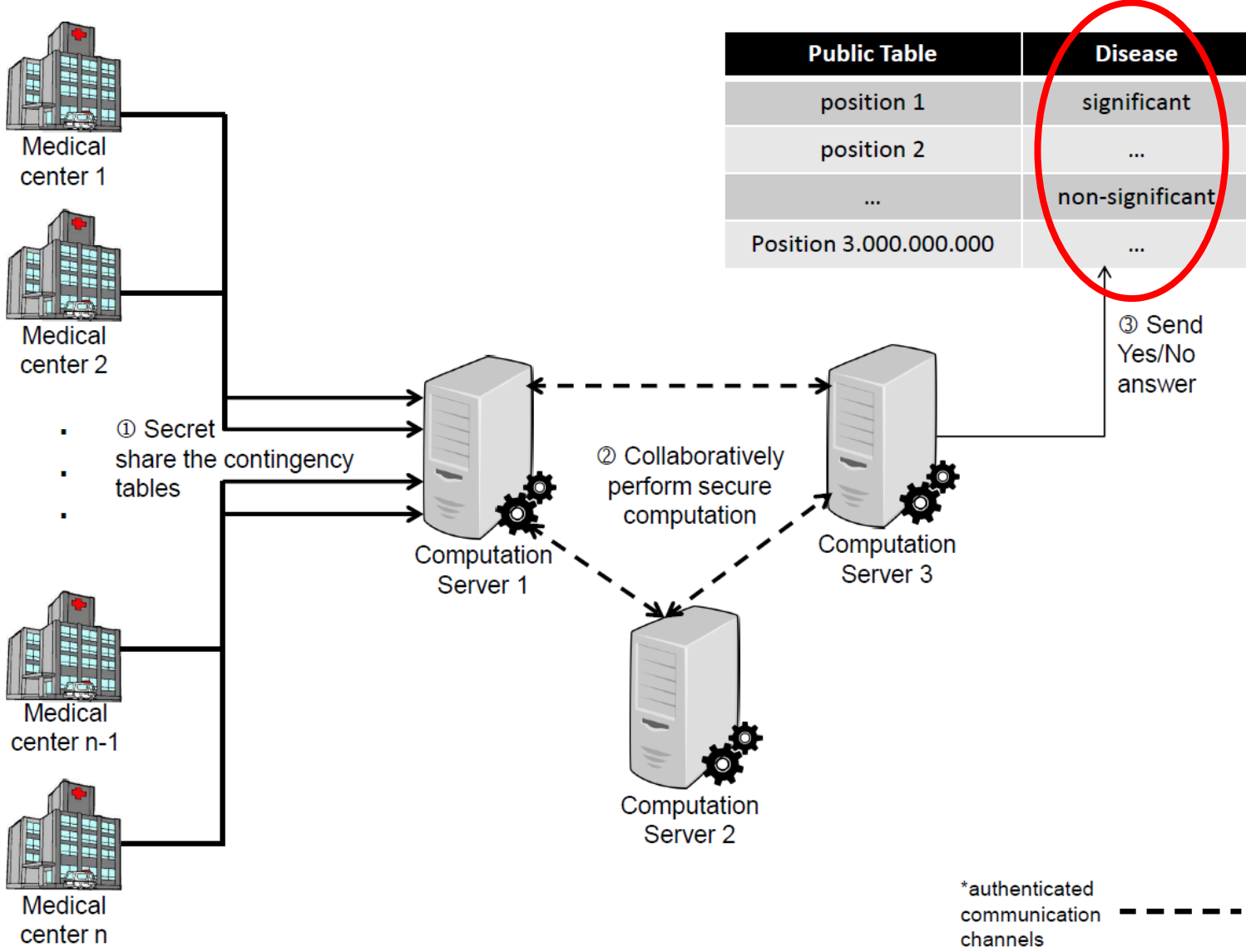




# MPC-based Solution



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# Problem of re-identification



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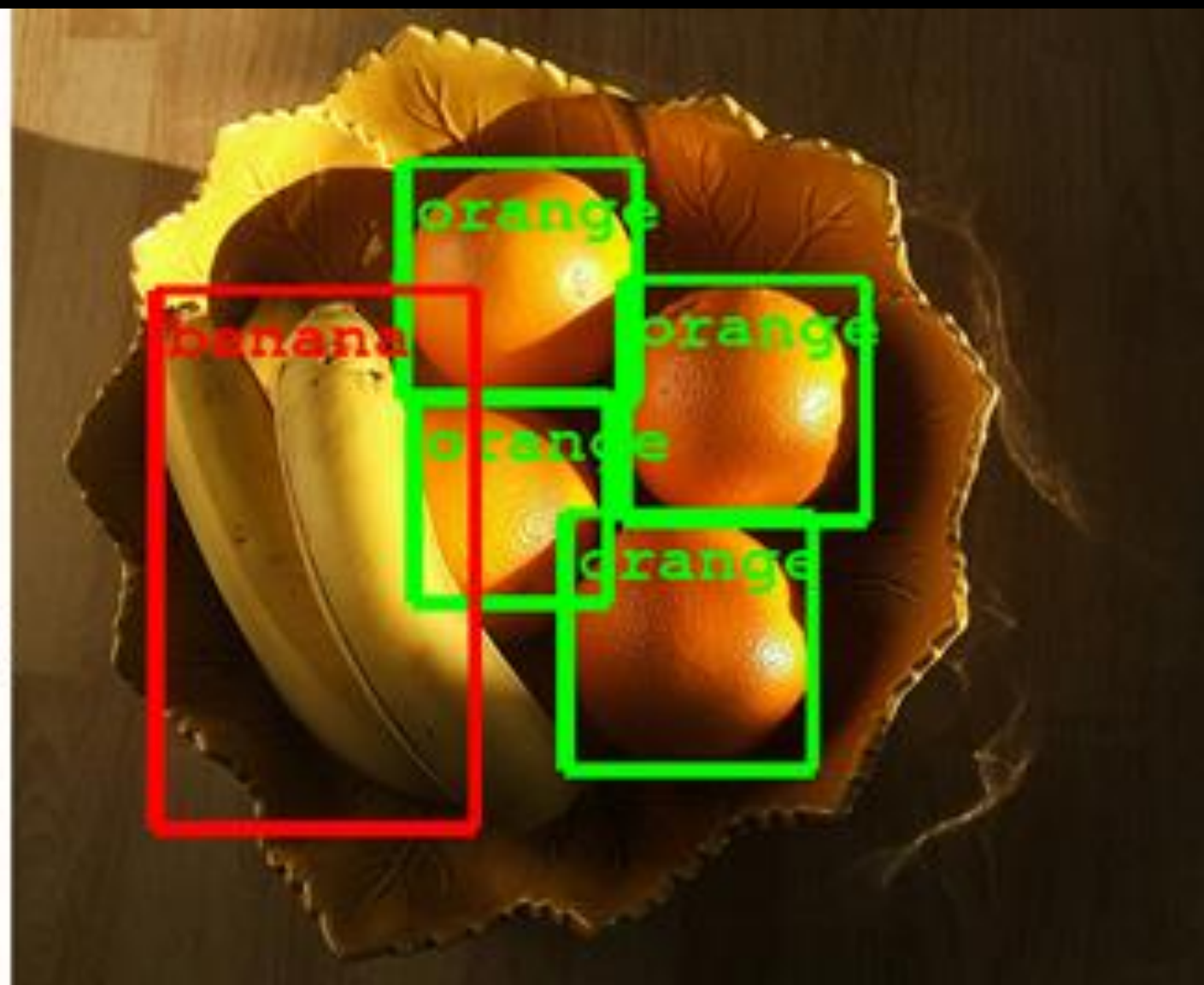
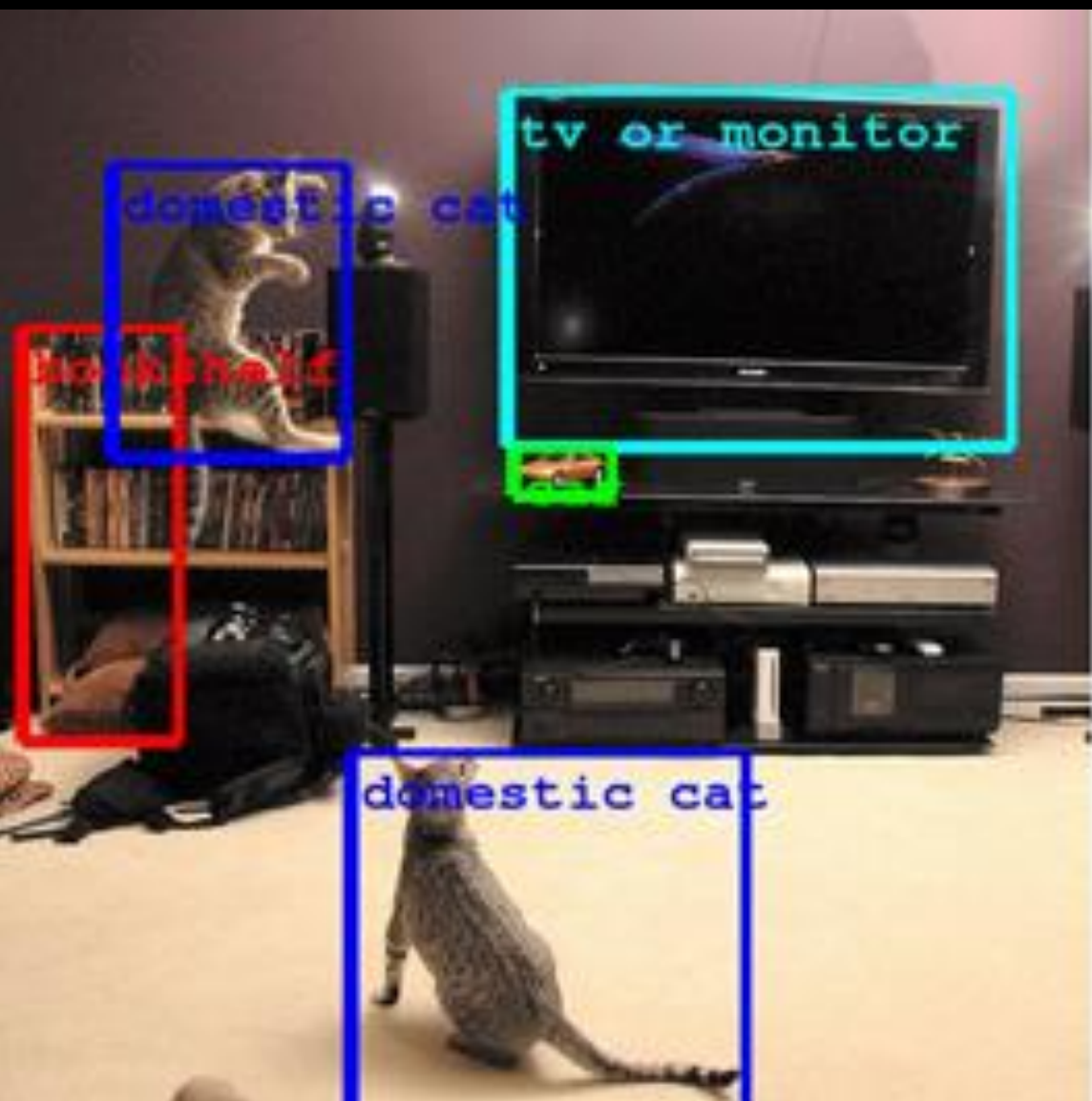
# MPC-based GWAS: Performance

Medical Centers	Number of Patients	CPU Time (Server 1)	Data Sent (Server 1)	CPU Time (Server 2, and Server 3)
20	200000	2.2ms	12.7KB	1.9ms
40	400000	2.3ms	17.8KB	2.0ms
60	600000	2.3ms	23KB	2.0ms
80	800000	2.5ms	28.1KB	2.2ms
100	1000000	2.4ms	33.2KB	2.1ms

# MPC-based GWAS vs. HE-based GWAS

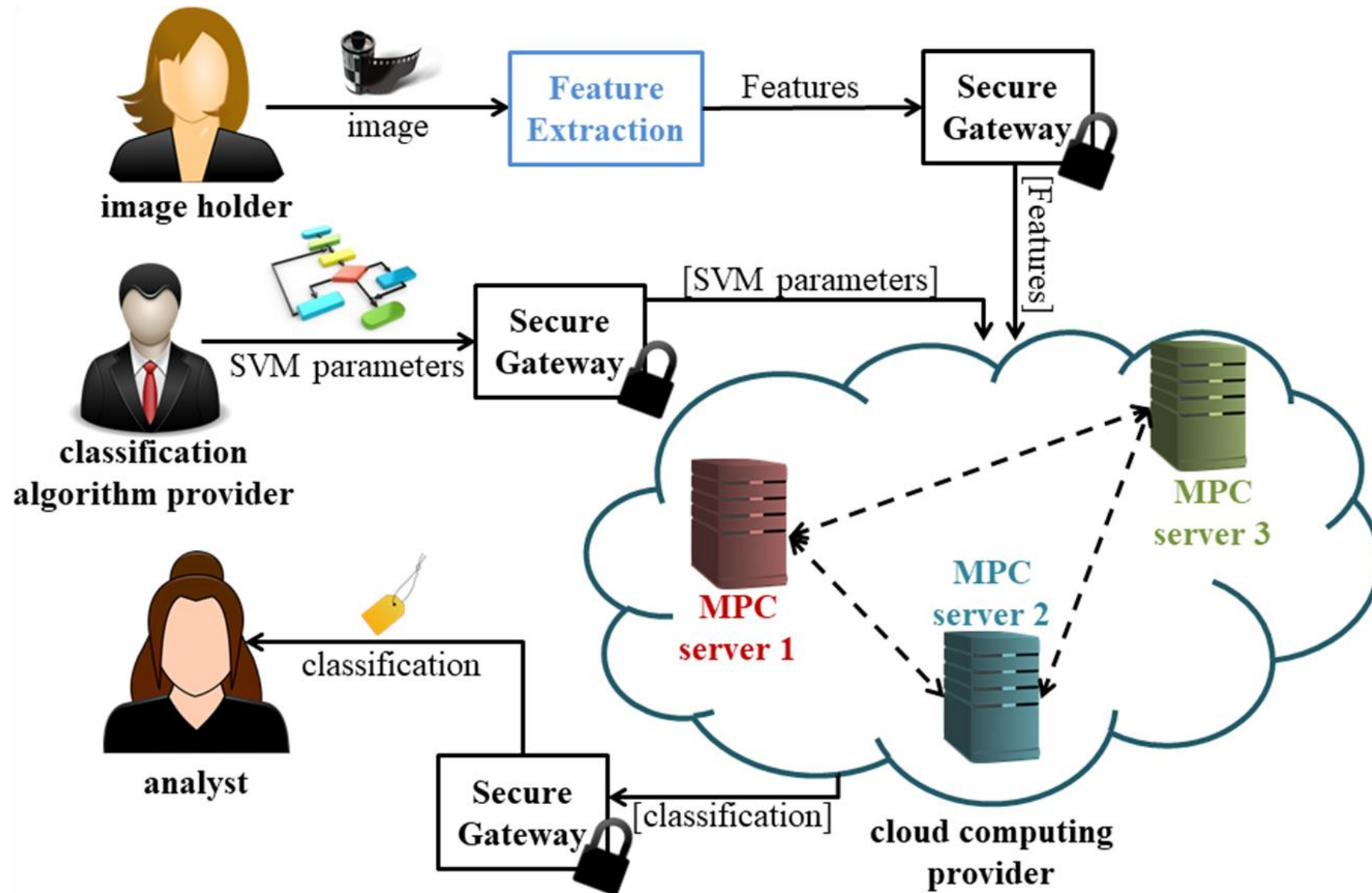
- CPU time in the range of milliseconds
  - Total communication cost in the range of KB
  - Multiple parties ( $\geq 2$ ) required to perform the computation
  - Active Security Guarantees
- CPU time in the range of seconds
  - Total communication cost in the range of MB
  - One party performs the computation
  - Semi-honest Decryptor assumption

# Application Scenario 2: Private Image Classification

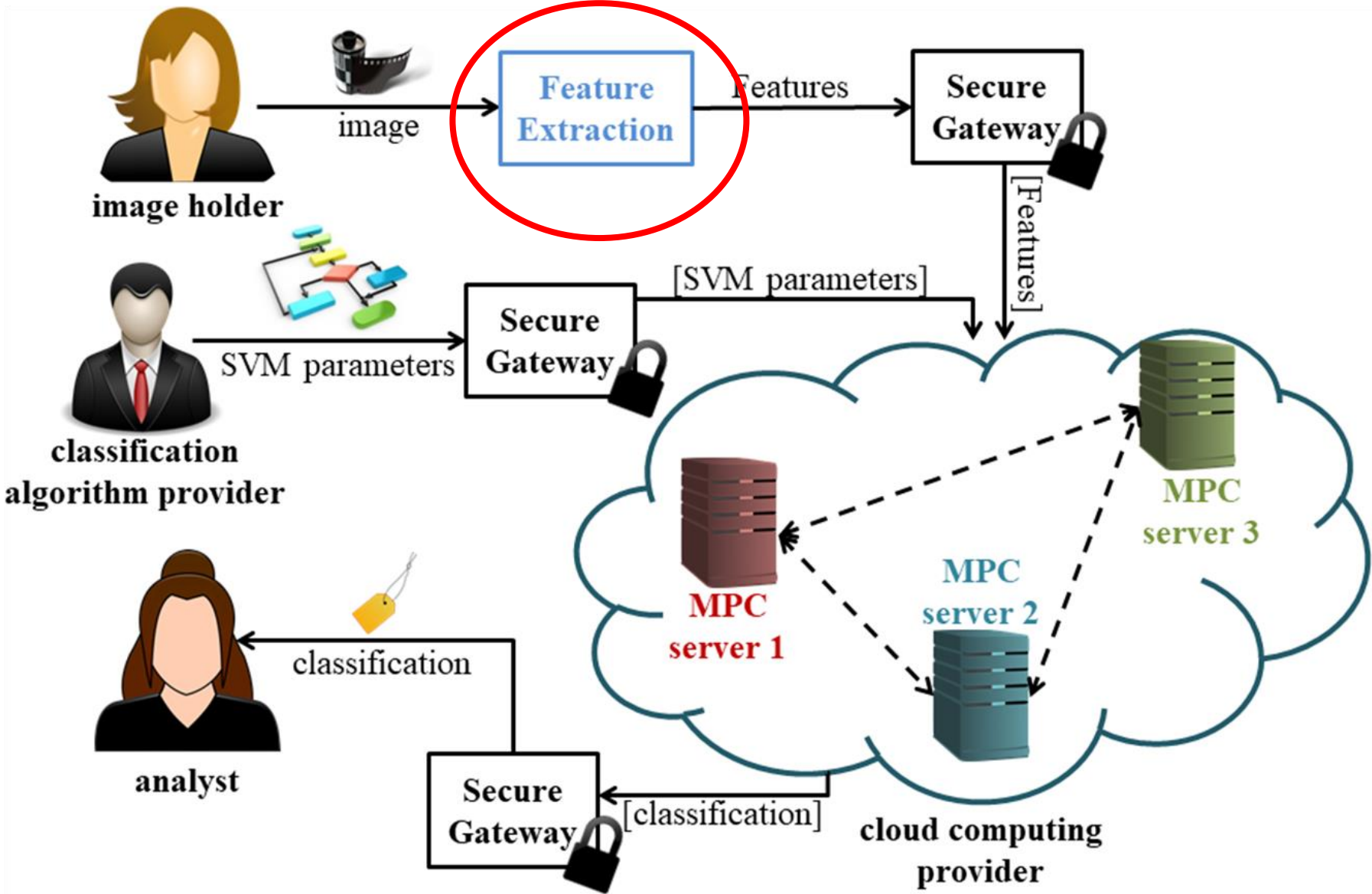




# EPIC: Efficient Private Image Classification



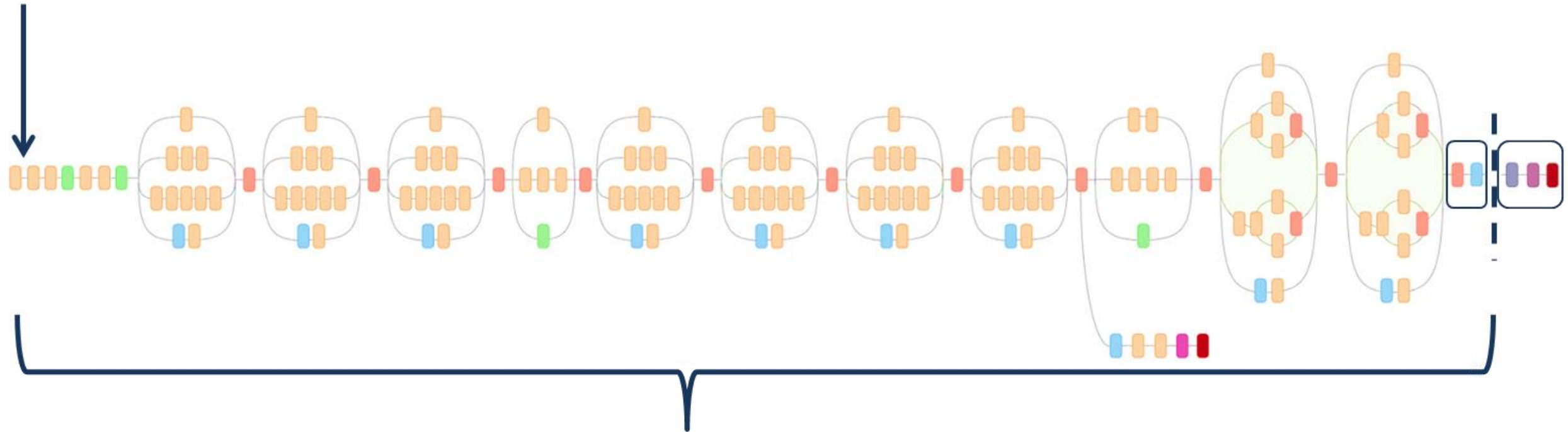
# EPIC: Efficient Private Image Classification



# Transfer Learning Feature Extraction (or: Learning from the Masters)

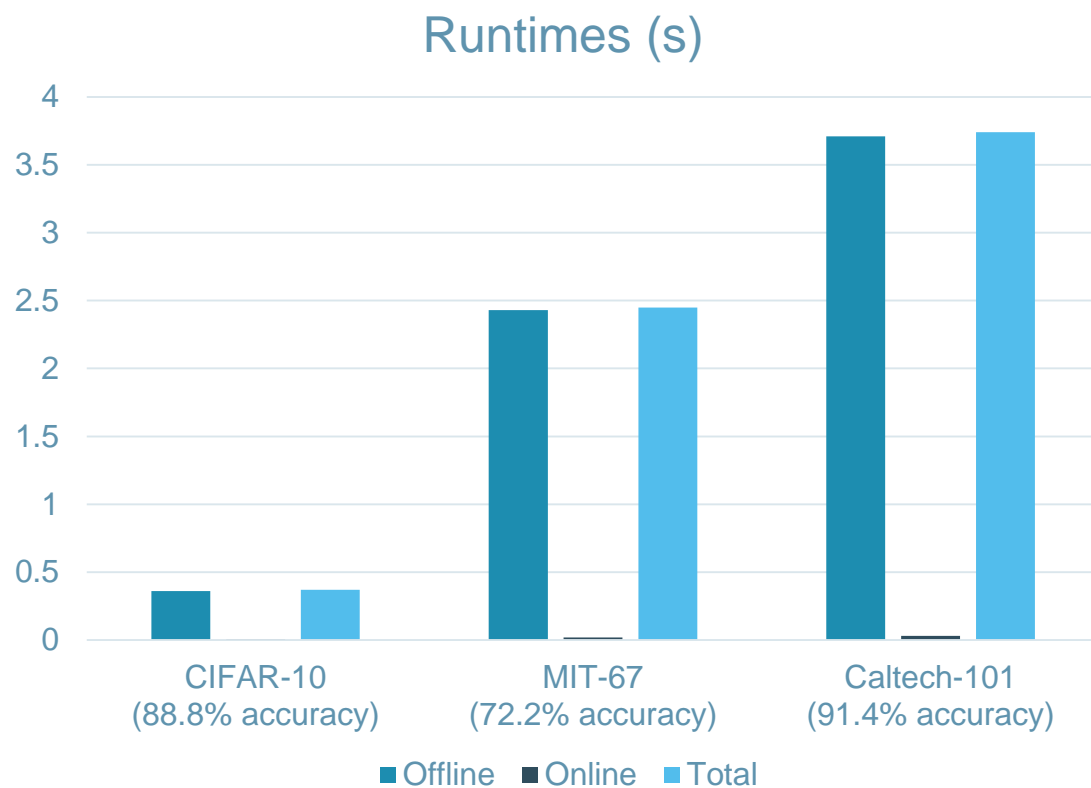


Plaintext (non-sensitive) images

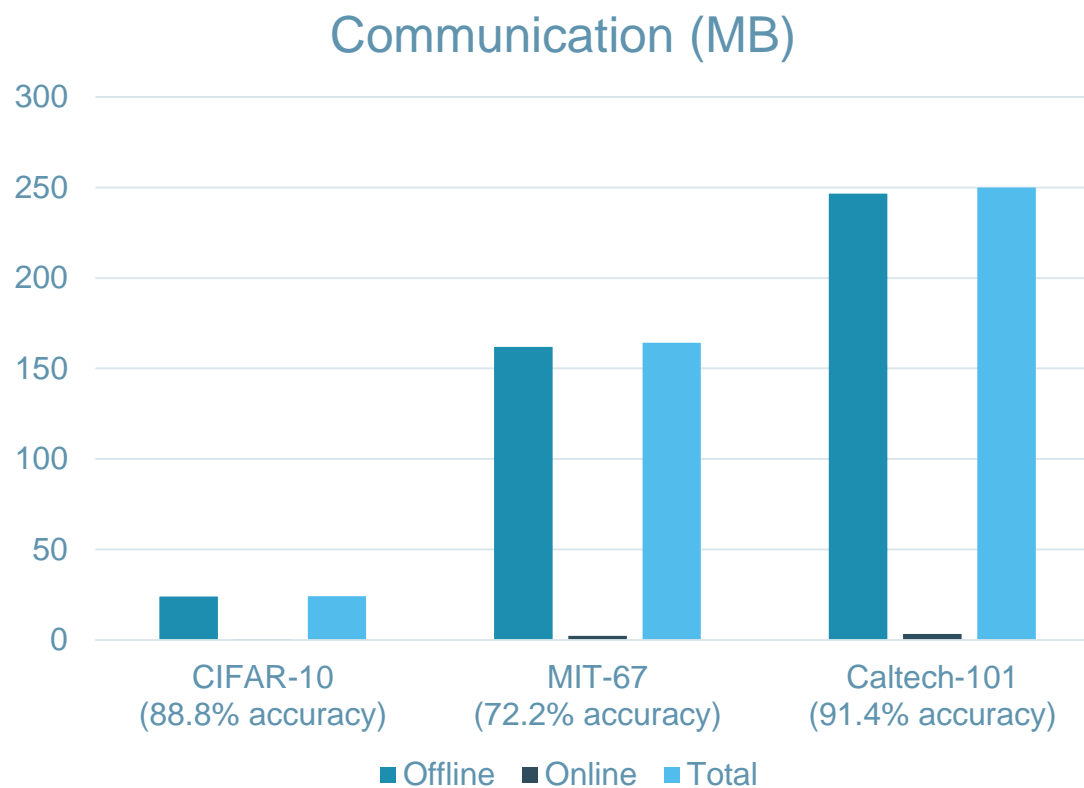


# EPIC Performance – Simple Variant

## Computation Cost

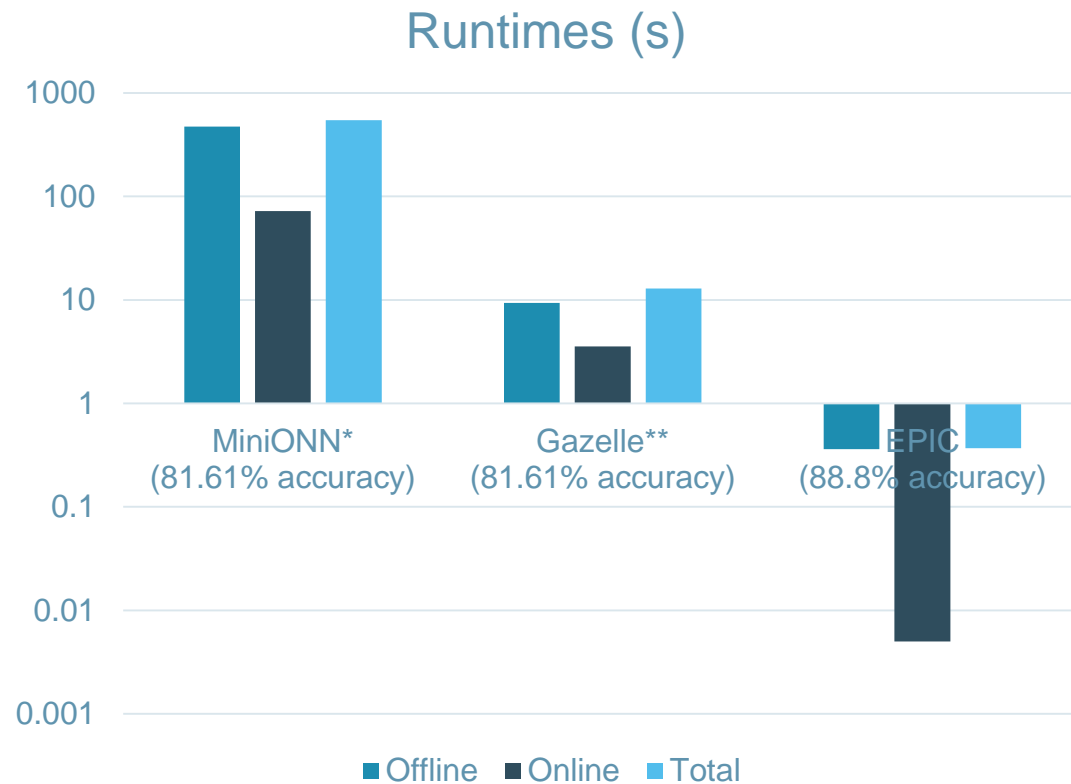


## Communication Cost

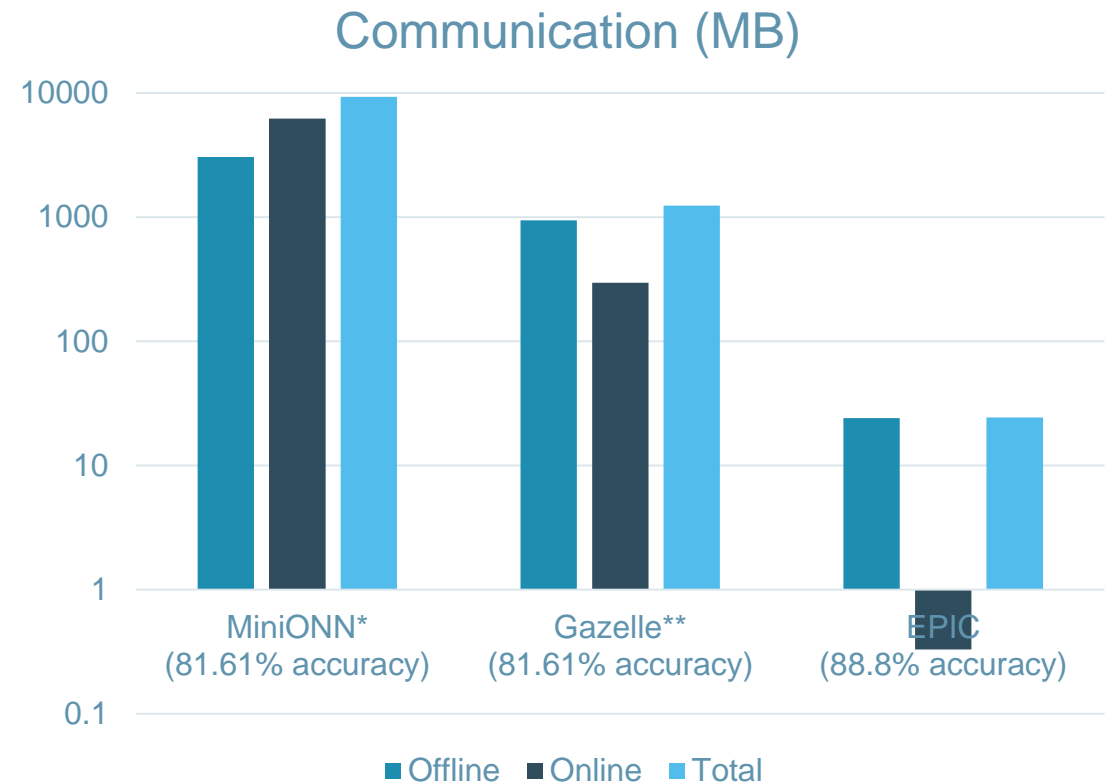


# Performance of the state-of-the-art private image classification

## Computation Cost



## Communication Cost



\* Jian Liu, Mika Juuti, Yao Lu, N. Asokan. **Oblivious Neural Network Predictions via MiniONN Transformations**. In *Proceedings of the 2017 ACM SIGSAC Conference on Computer and Communications Security* (pp. 619-631). ACM.

\*\* Chiraag Juvekar, Vinod Vaikuntanathan, and Anantha Chandrakasan. **GAZELLE: A low latency framework for secure neural network inference**. In *27th USENIX Security Symposium (USENIX Security '18)*, Baltimore, MD, 2018. USENIX Association.

# Application Scenario 3: Secure RSA Modulus Generation

# RSA Modulus

- A biprime  $N$ , with two secret prime factors,  $p$  and  $q$ .
- Heart of the first public key cryptosystem; security based on factoring hardness assumption.

# Why RSA Moduli?

- Signatures and Encryption
  - [RSA-77], [Paillier-99].
- Cryptographic accumulators
  - [Benaloh-deMare-93], [Camenisch-Lysyanskaya-02], [Li-Li-Xue-07], [Boneh-Bünz-Fisch-19],
- VDF and Timelock puzzles
  - [Rivest-Shamir-Wagner-99], Boneh-Bonneau-Bünz-Fisch-18], [Wesolowski-19], [Pietrzak-19], [Ephraim-Freitag-Komargodski-Pass-19].
- Efficient zk-SNARKs
  - [Bünz-Fisch-Szepieniec-19], [Lai-Malavolta-19]
- And others...



# Why *distributed* RSA Moduli?

- Threshold Cryptography

## Call 2021a for Feedback on Criteria for Threshold Schemes

### NIST Multi-party Threshold Cryptography

2021-July-02: <https://csrc.nist.gov/projects/threshold-cryptography>

Please send comments to [threshold-MP-call-2021a@nist.gov](mailto:threshold-MP-call-2021a@nist.gov) by September 13, 2021.

**1. Scope of proposals.** The future call for proposals will be intended to gather expert submissions of concrete threshold schemes for primitives that are *interchangeable* (in the sense of IR 8214A, Section 2.4) with<sup>2</sup> ECDSA, EdDSA, **RSA signing/decryption, RSA keygen**, AES, and ECC-based key agreement.<sup>3</sup> After an evaluation period, and possibly various stages for tweaks,

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# Why *distributed* RSA Moduli?

- *Companies or foundations*



# Connections with related work

CCD+20

OT based

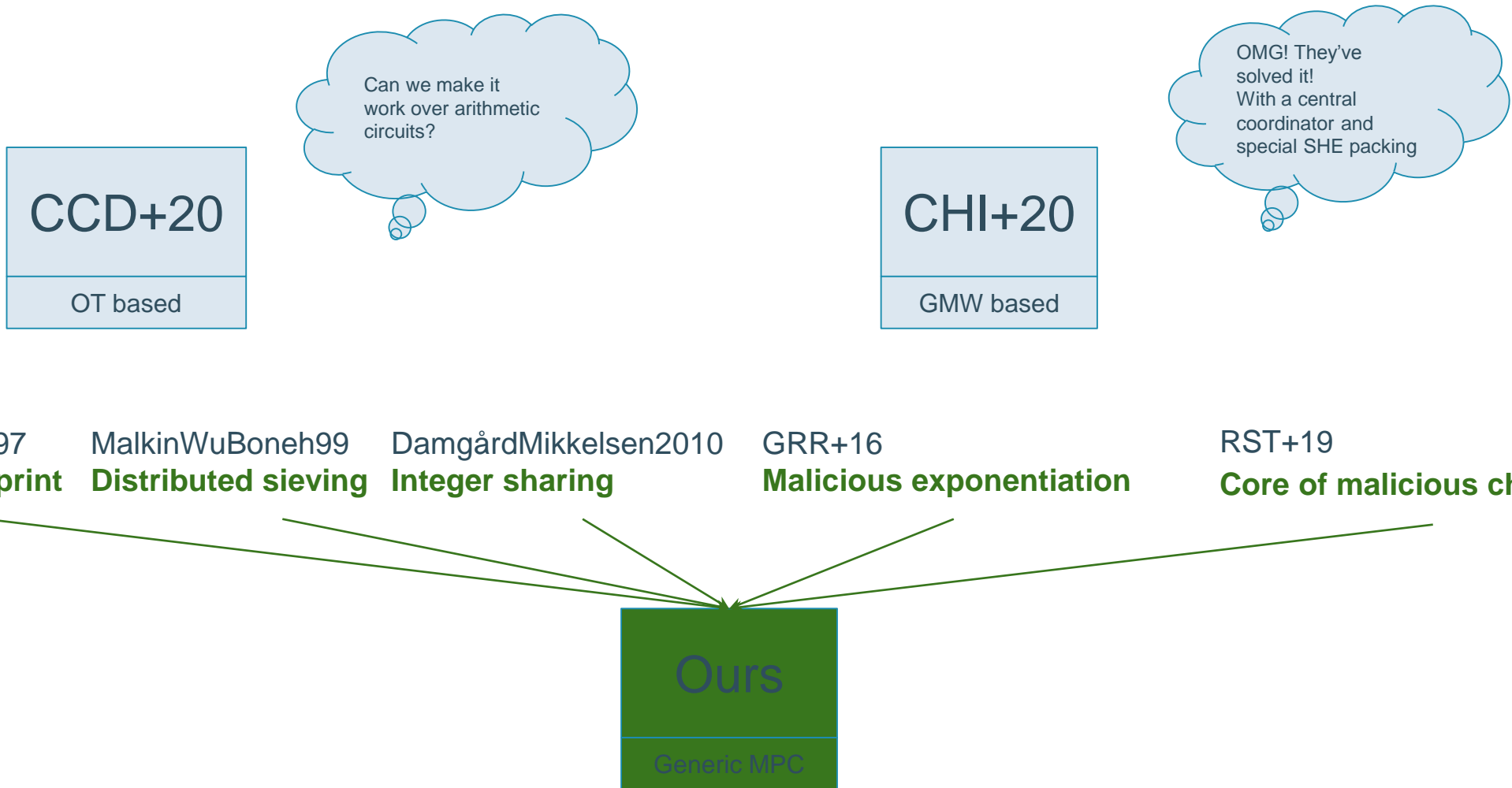
Can we make it  
work over arithmetic  
circuits?

CHI+20

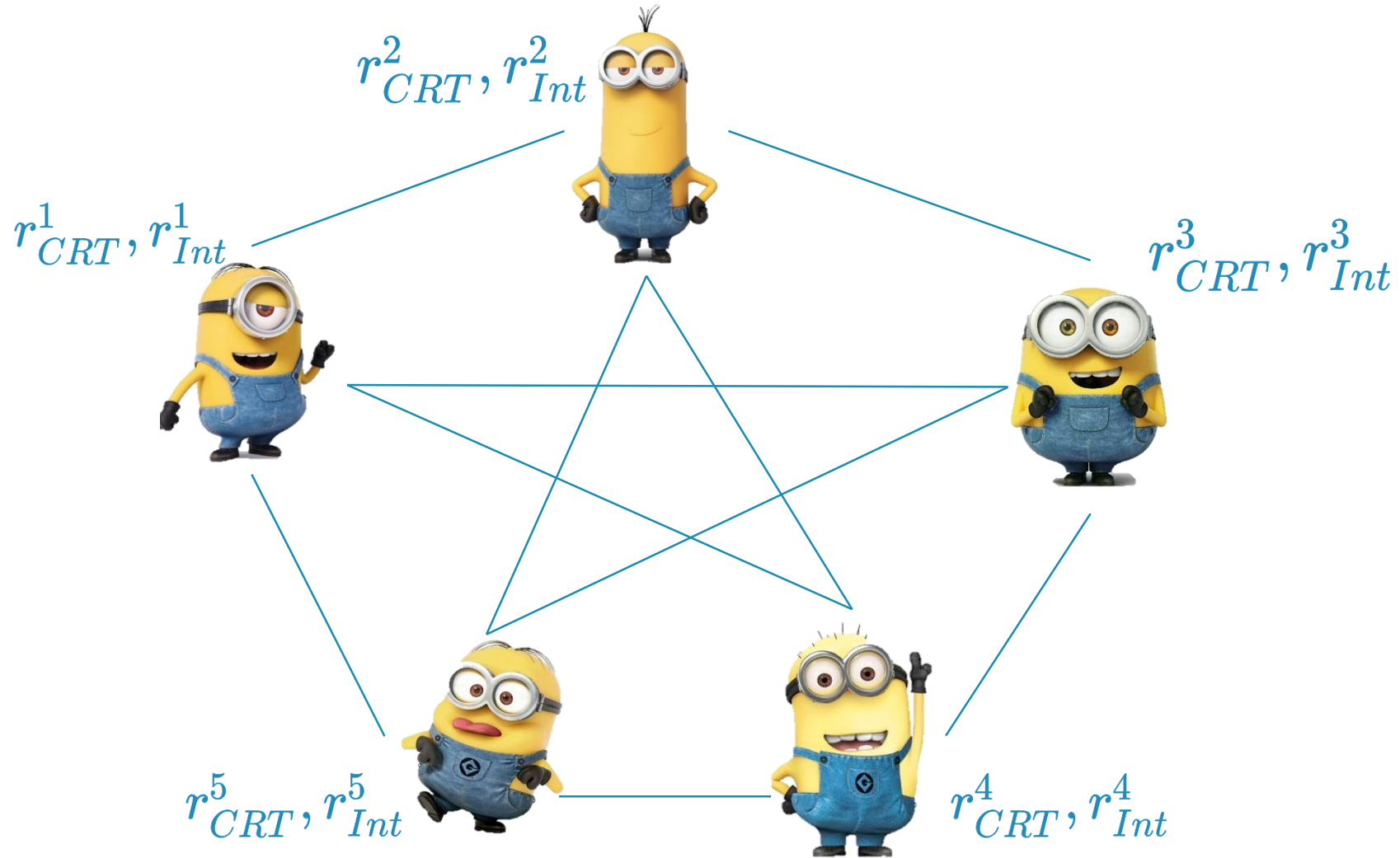
GMW based

OMG! They've  
solved it!  
With a central  
coordinator and  
special SHE packing

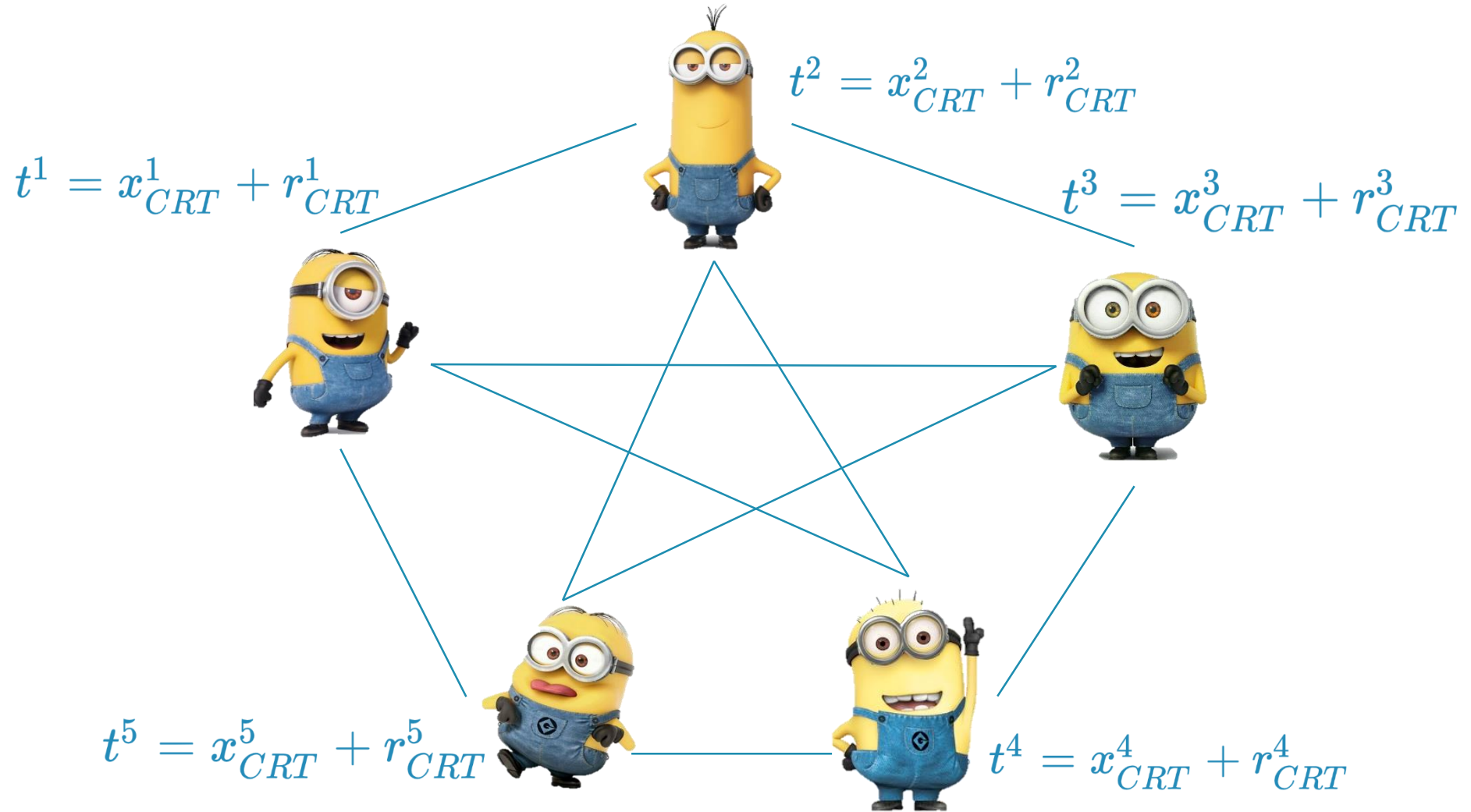
# Connections with related work



# ConvInt Protocol

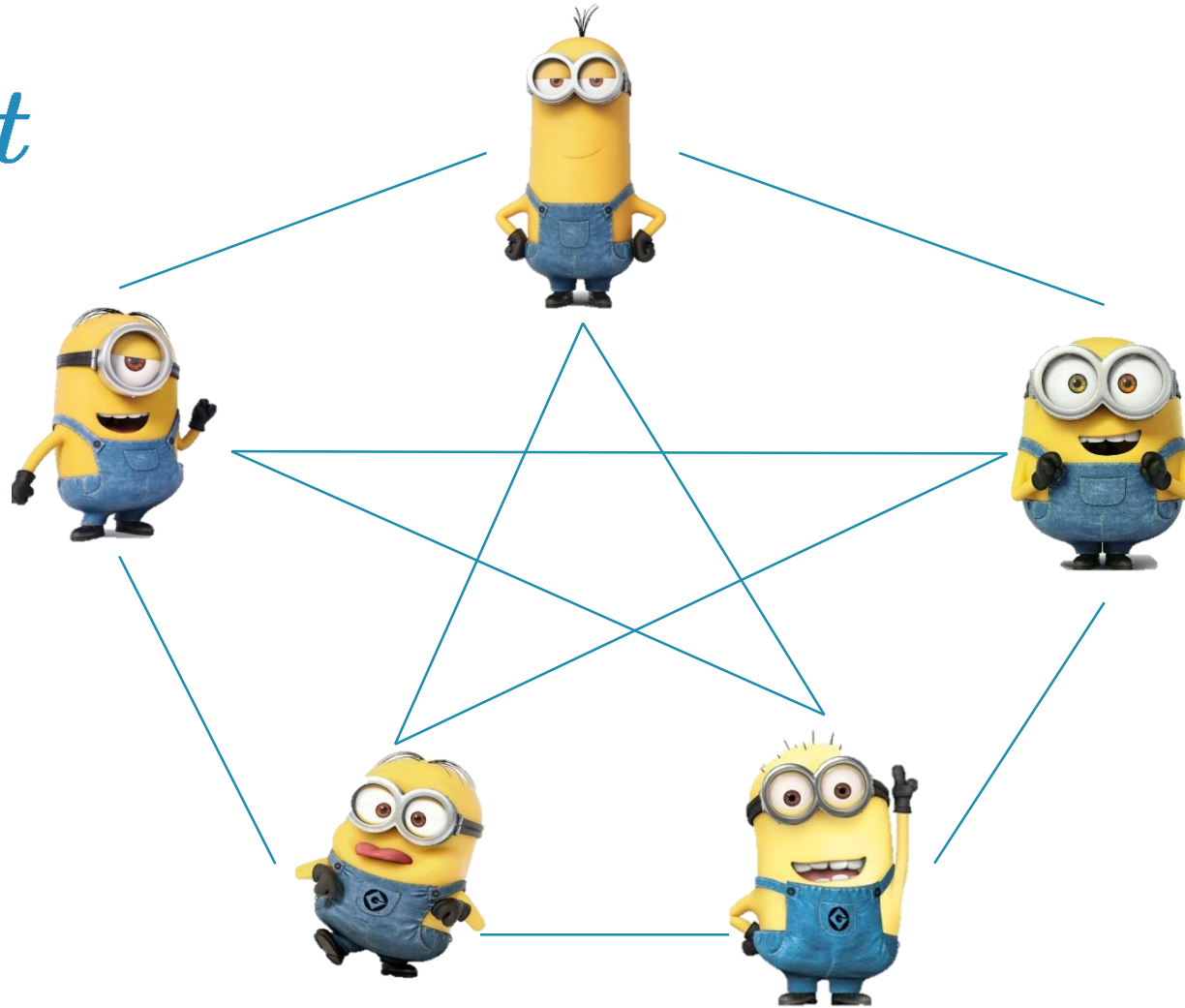


# ConvInt Protocol

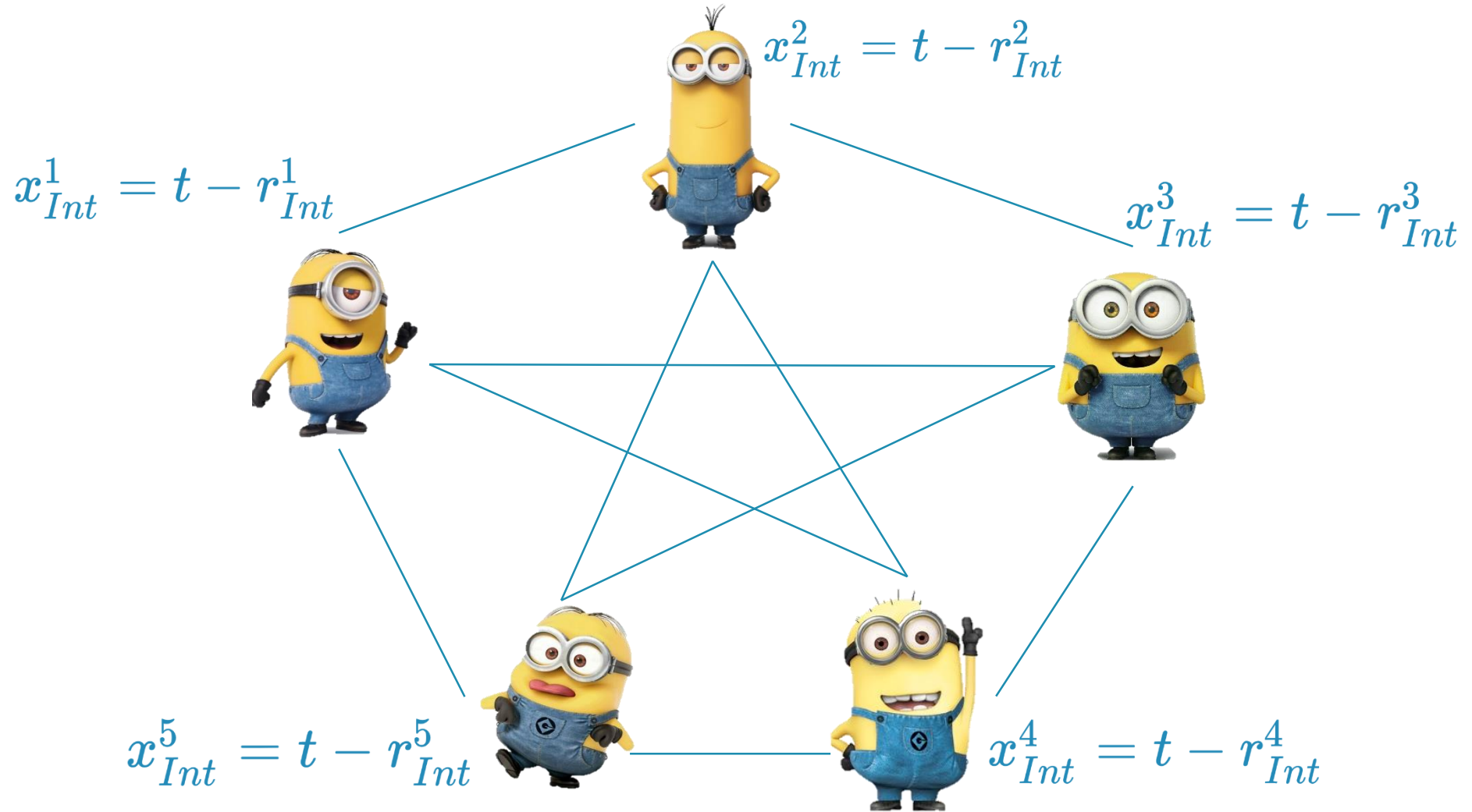


# ConvInt Protocol

Open  $t$



# ConvInt Protocol



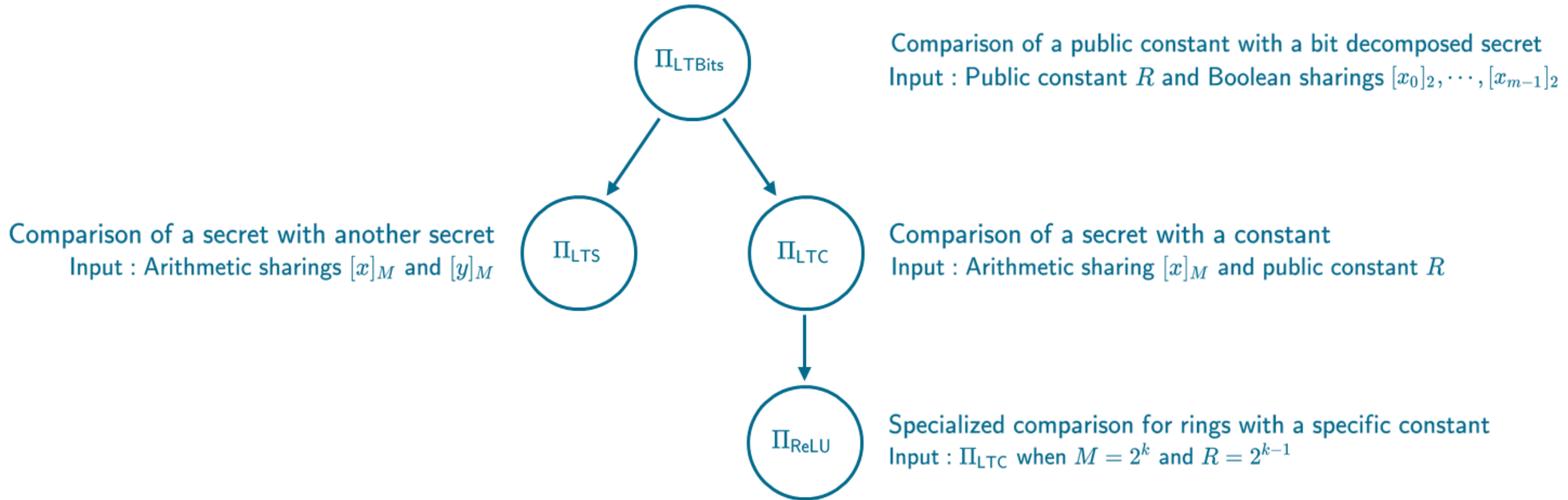


# Contributions in Secure RSA modulus Generation

- RSA modulus generation protocol with generic MPC.
- Exploit *Distributed Sieving techniques* and *public knowledge* to perform parts of the protocol semi-honestly without degrading security.
- Convert to Integer protocol, of independent interest.
- Up to 37x better communication cost compared to CCD+20.

# Improving MPC Primitives: The Case of Comparisons

# Rabbit: Comparison Protocols Collection



# Rabbit Intuition: Observation 1

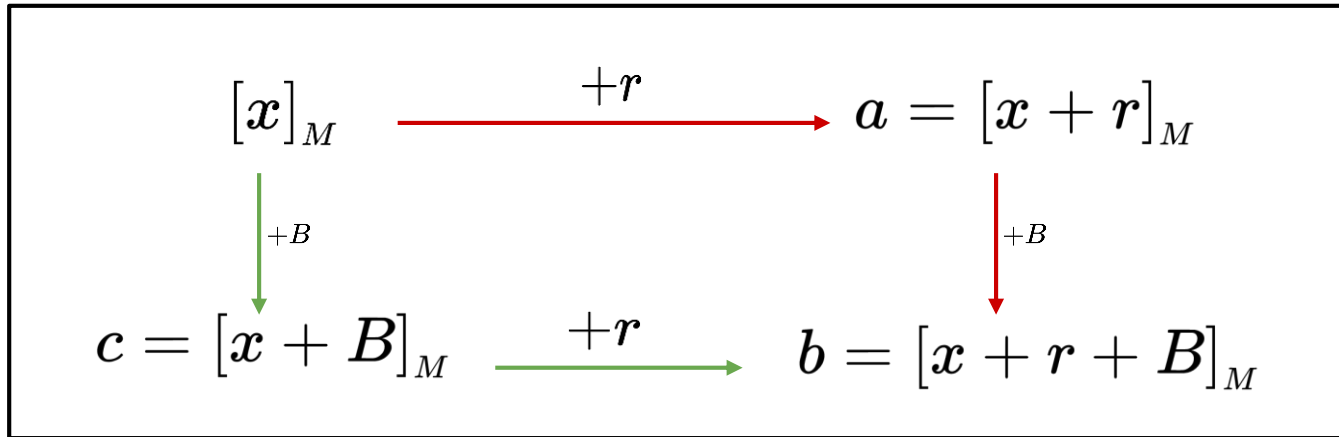
**GOAL:** *Detect when a sum over a particular modulus wraps around and correct for it.*

- Given a function:  $\text{LT}(\cdot, \cdot) : \mathbb{Z} \times \mathbb{Z} \rightarrow \{0, 1\} \subseteq \mathbb{Z} : \begin{cases} \text{LT}(x, y) = 1 & \text{if } (x < y); \\ \text{LT}(x, y) = 0 & \text{otherwise,} \end{cases}$
- We can compute a modular sum by performing computations over the integers:  
$$x + y \bmod M = x + y - M \cdot \text{LT}(x + y \bmod M, x) = x + y - M \cdot \text{LT}(x + y \bmod M, y)$$

# Rabbit Intuition: Observation 2

Note that we are  
after:  $x \stackrel{?}{<} M - B$

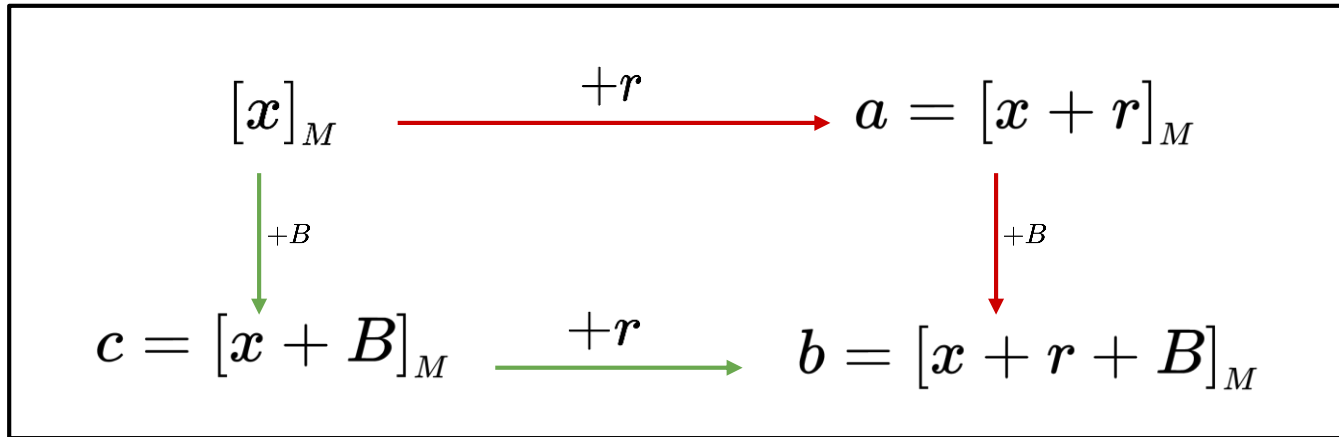
- Exploit the commutativity of addition:



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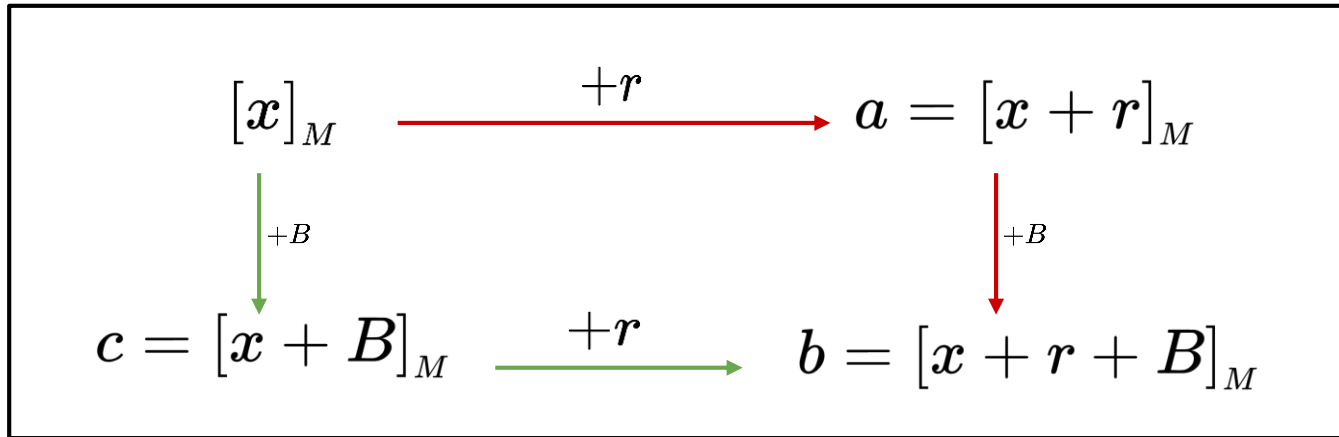


- Combine with Observation 1 (performing modular sums over the integers):

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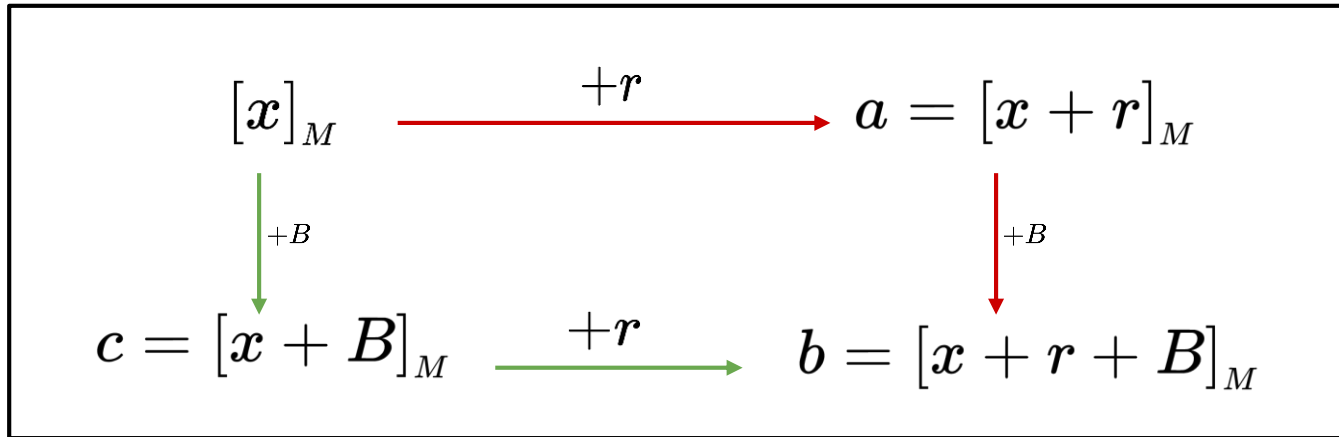
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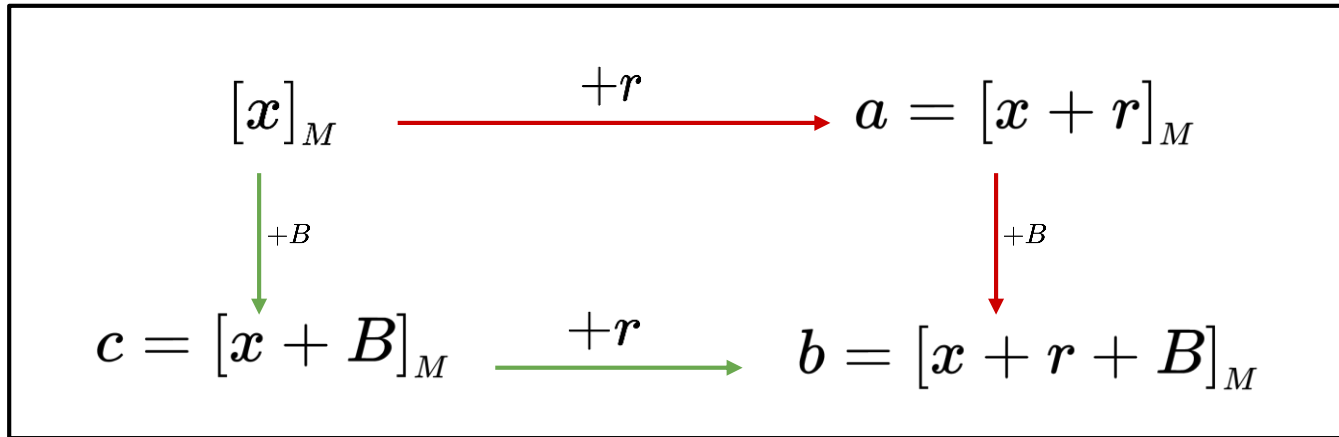
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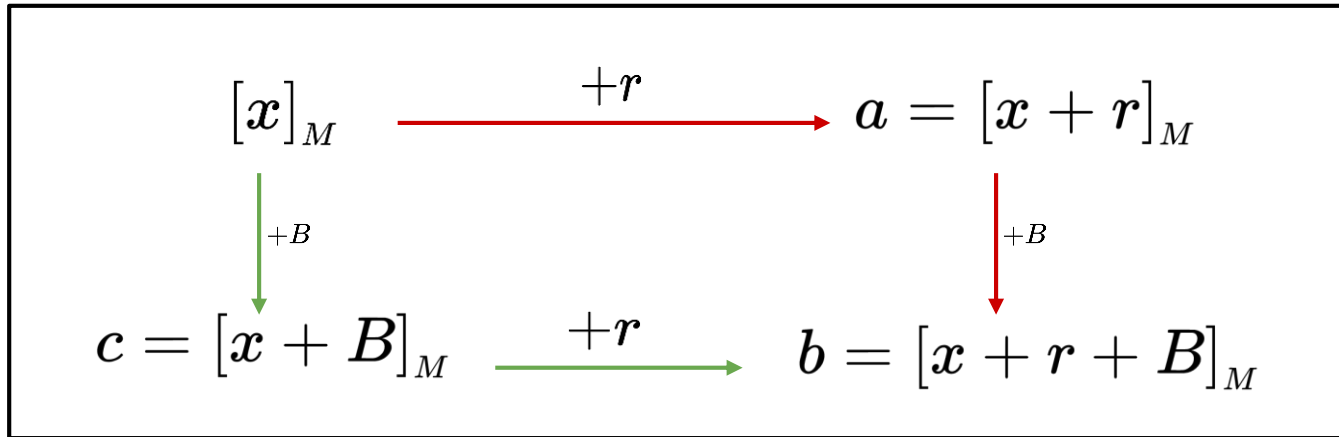
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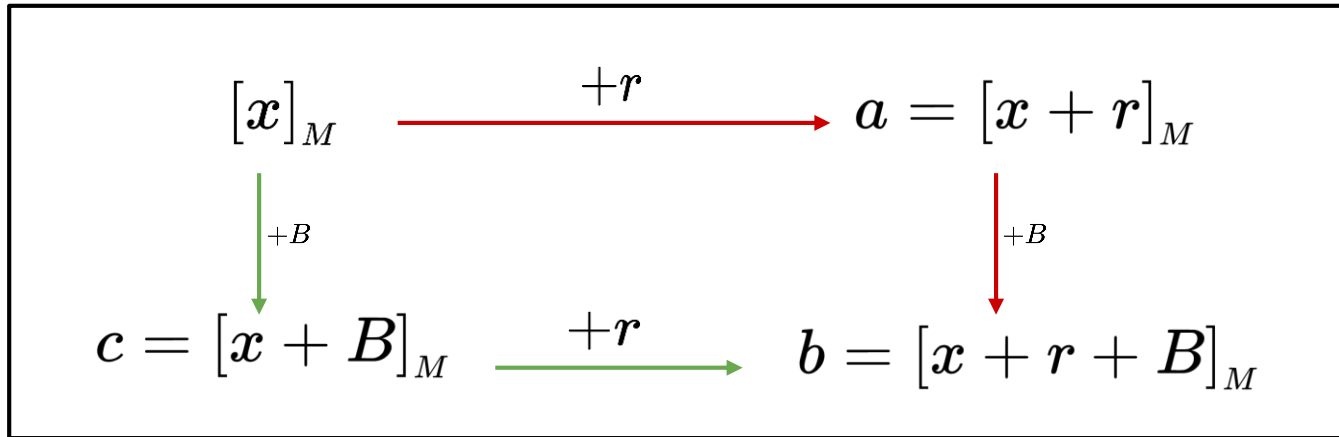
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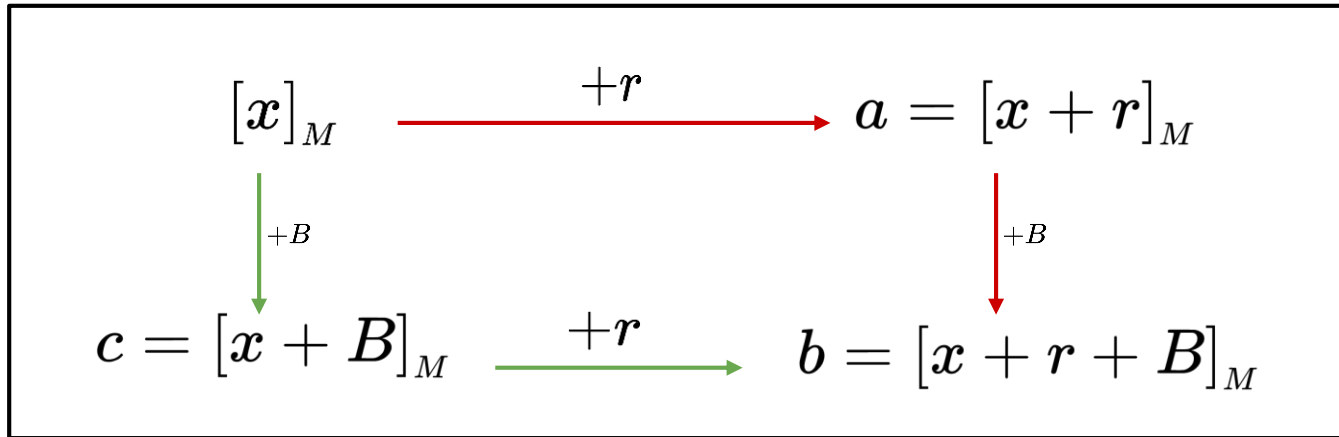
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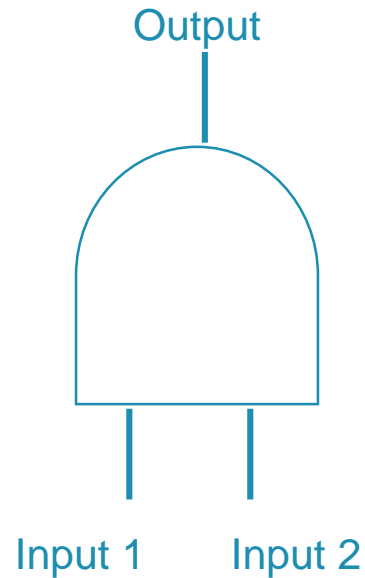
# Rabbit's conclusions

- Rabbit comparisons are **more efficient** than edaBit<sup>1</sup> comparisons with:
  - ~1.5x better throughput in most adversarial settings
  - Over 2.3x better throughput in the passive, honest majority setting
  - Lower communication cost
  - Lower memory footprint for the HE-based preprocessing
  
- Rabbit eliminates the need for “slack”
  
  
- The core of the Rabbit comparison algorithms is unconditionally secure













<sup>1</sup> Daniel Escudero, Satrajit Ghosh, Marcel Keller, Rahul Rachuri, and Peter Scholl. **Improved primitives for MPC over mixed arithmetic-binary circuits**. In *Annual International Cryptology Conference*, pp. 823-852. Springer, Cham, 2020.

# Improving MPC Primitives: The Case of Multiparty Arithmetic Garbling

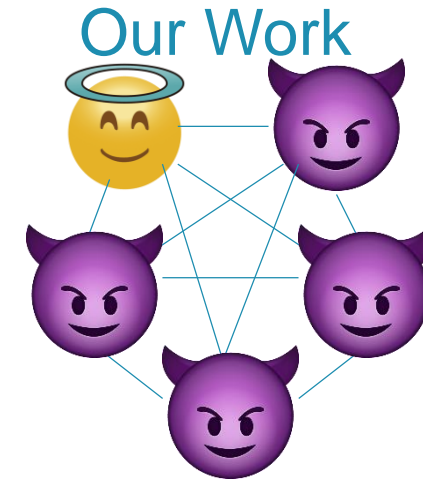
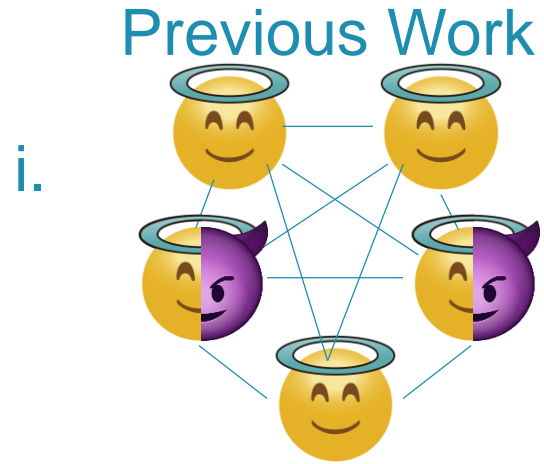
# Full-Threshold Actively-Secure Multiparty Arithmetic Circuit Garbling



Garbling →

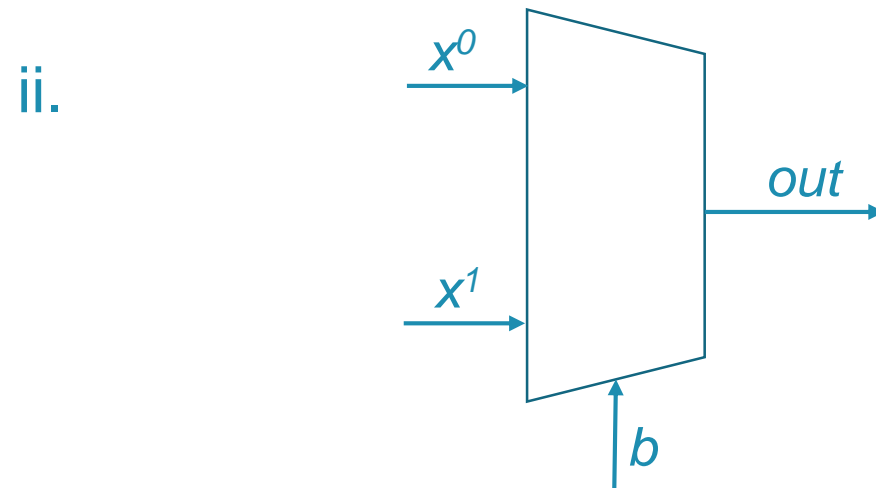
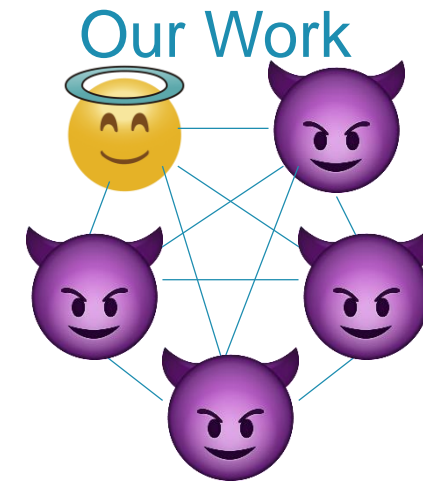
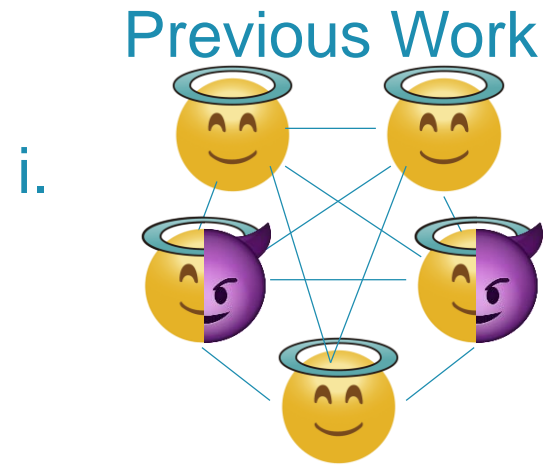
Input 1	Input 2	Output
		(0) 
		(0) 
		(0) 
		(1) 

# *Full-Threshold Actively-Secure* Multiparty Arithmetic Circuit Garbling: Contributions

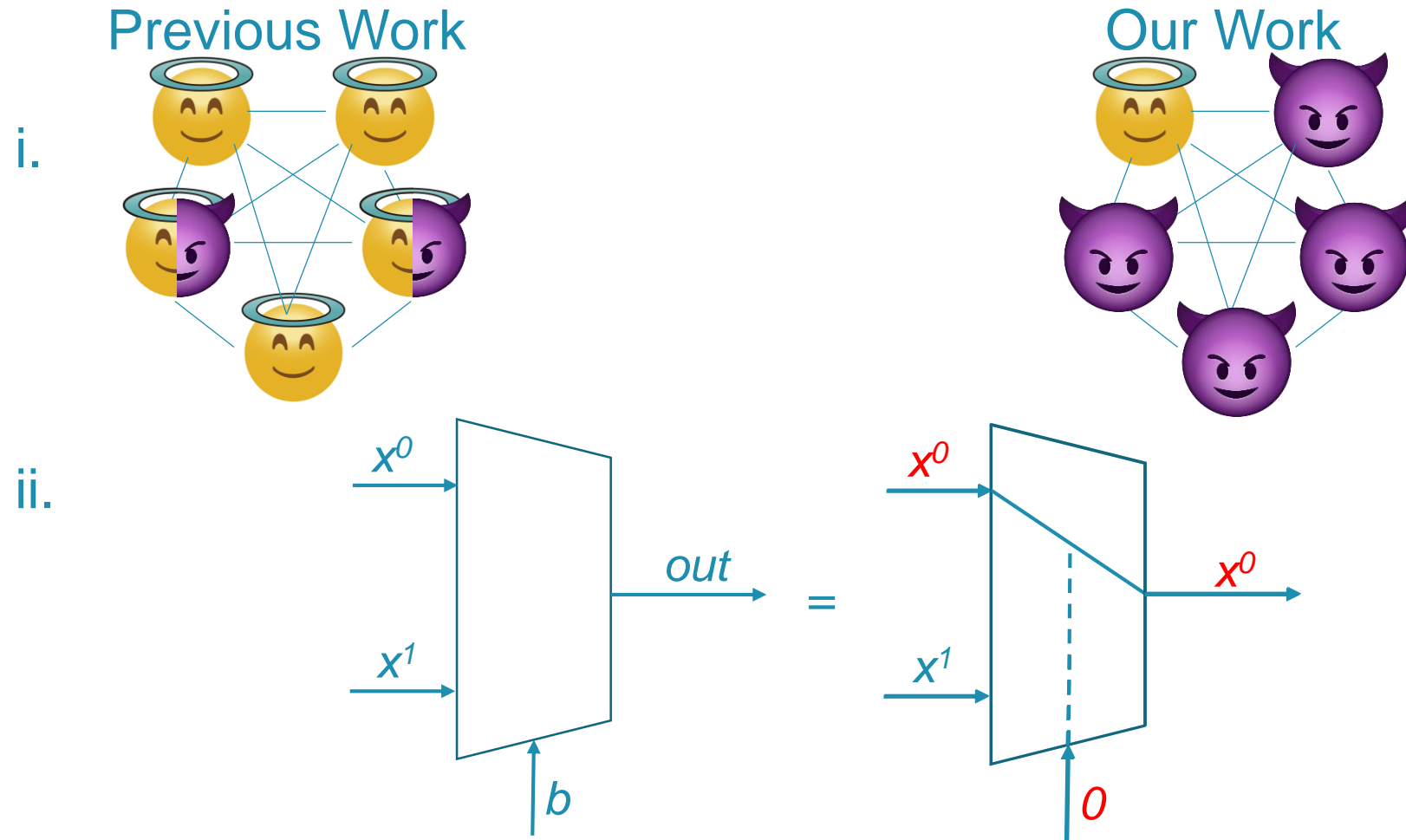




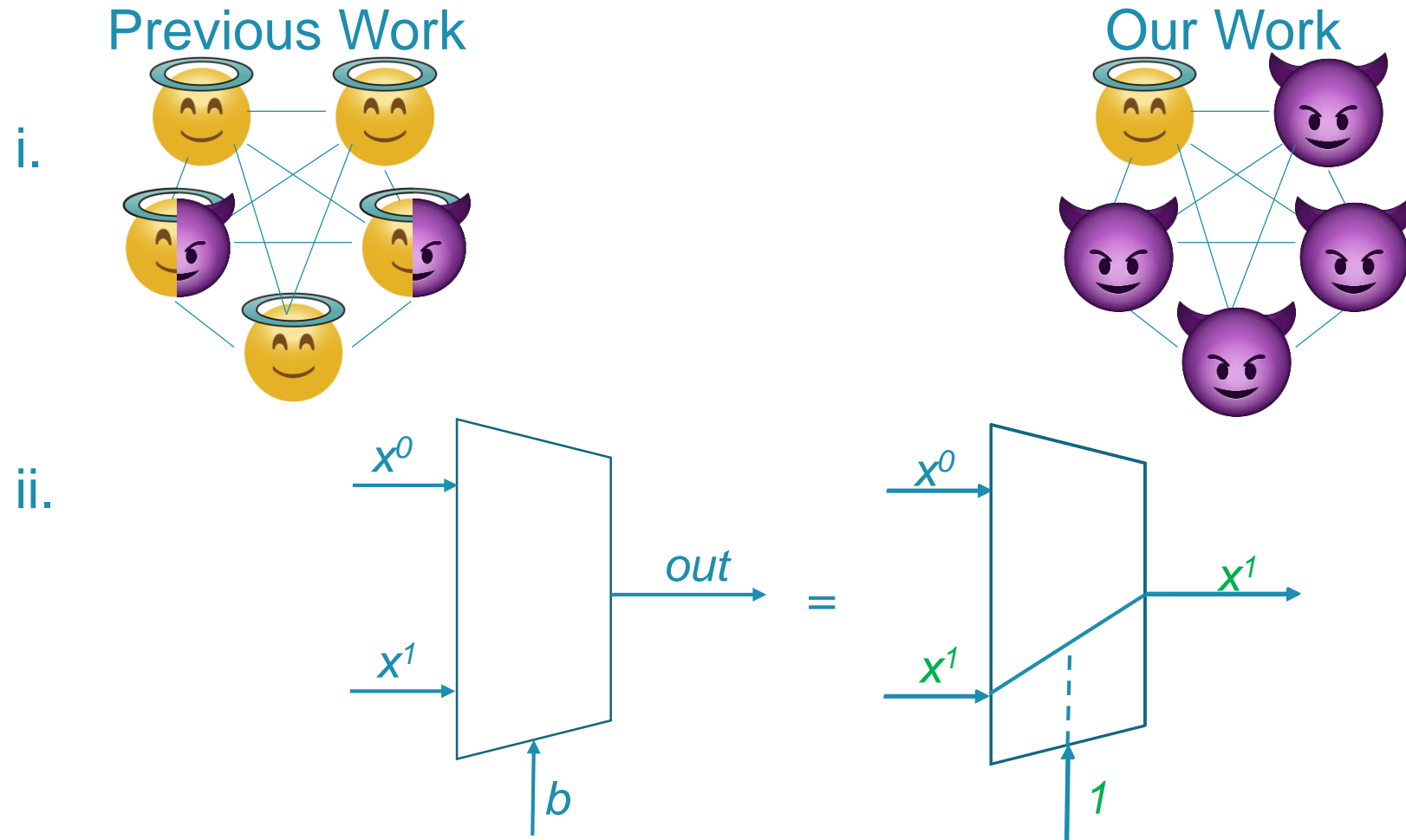
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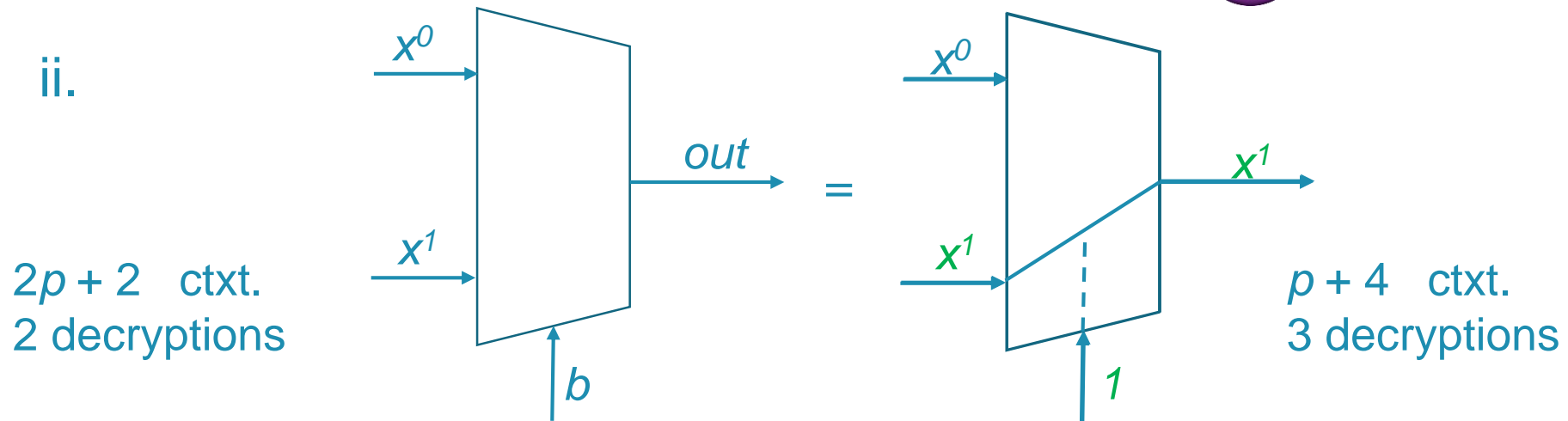
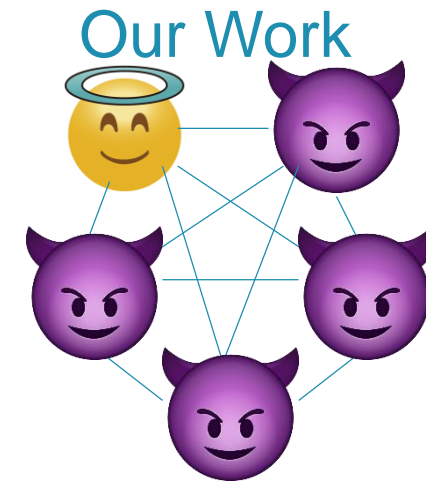
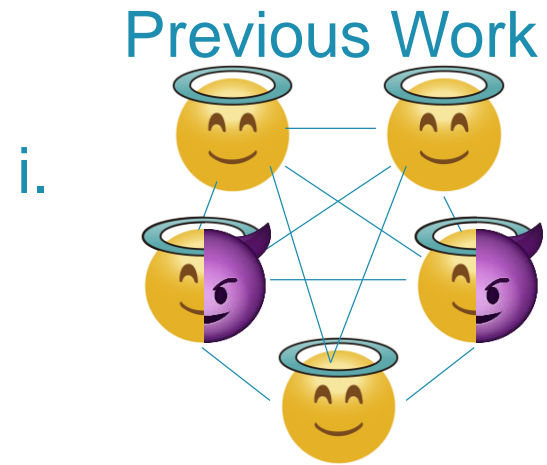
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# Closing Remarks

# Conclusions

MPC is practical

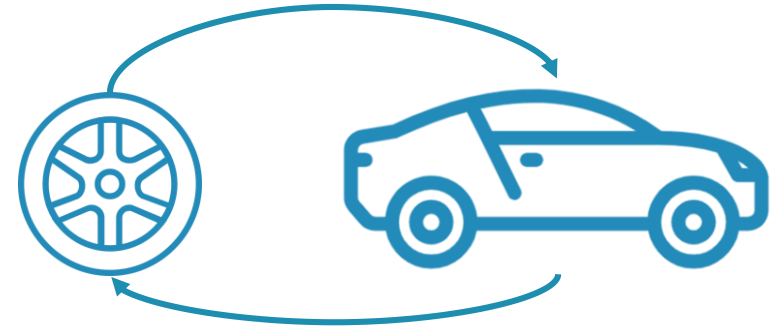


Theory - Practice

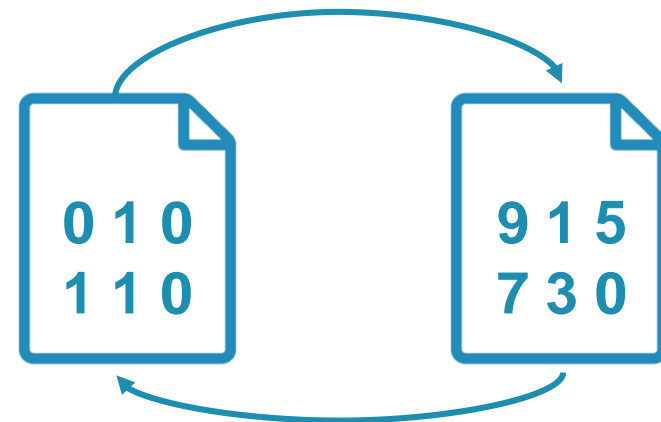
Exploit interdisciplinary research



Primitives and Application Scenarios

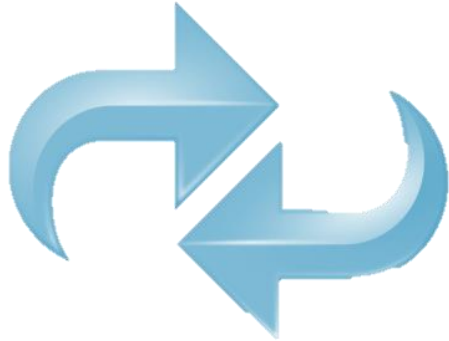


Use each data representation for what is best



# Future Work

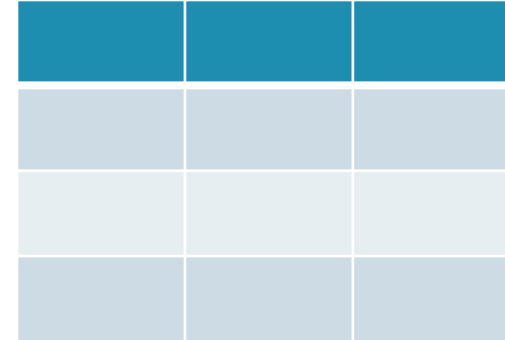
Switching Protocols



Generalized  
Beaver Tuples



Special Preprocessing



MPC for Machine  
Learning

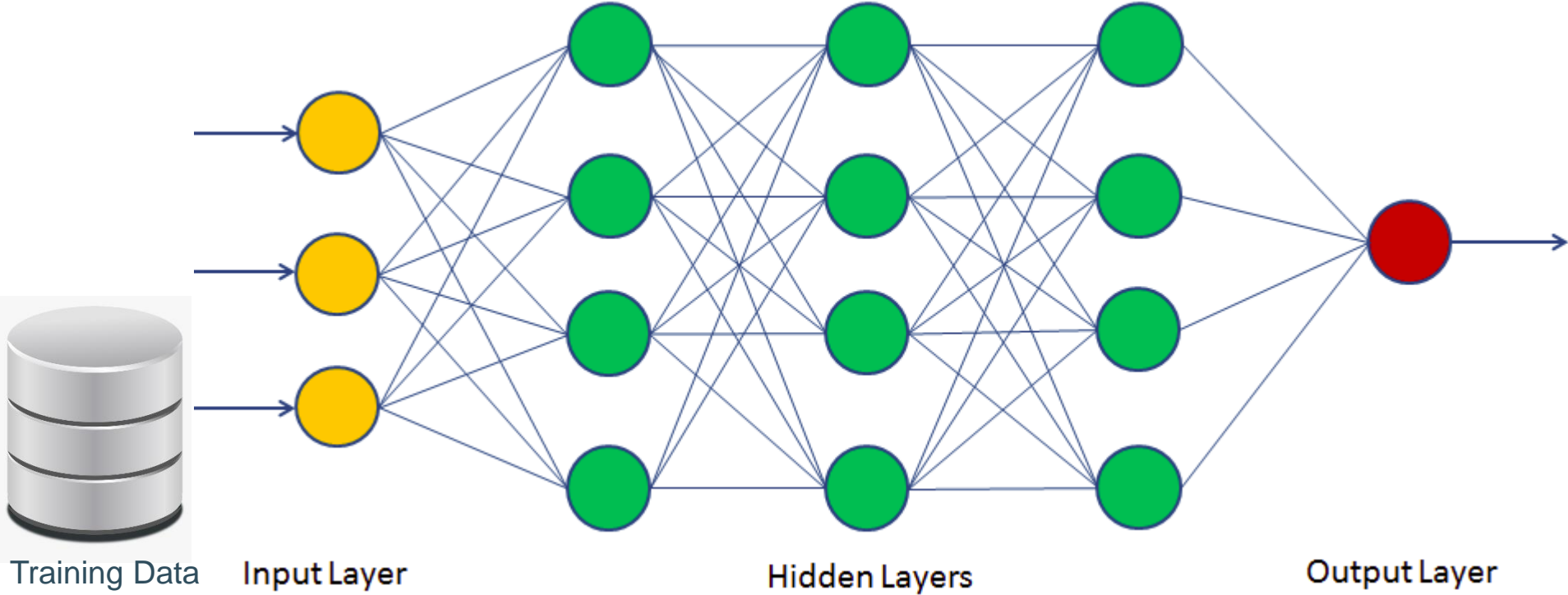


# Future Research

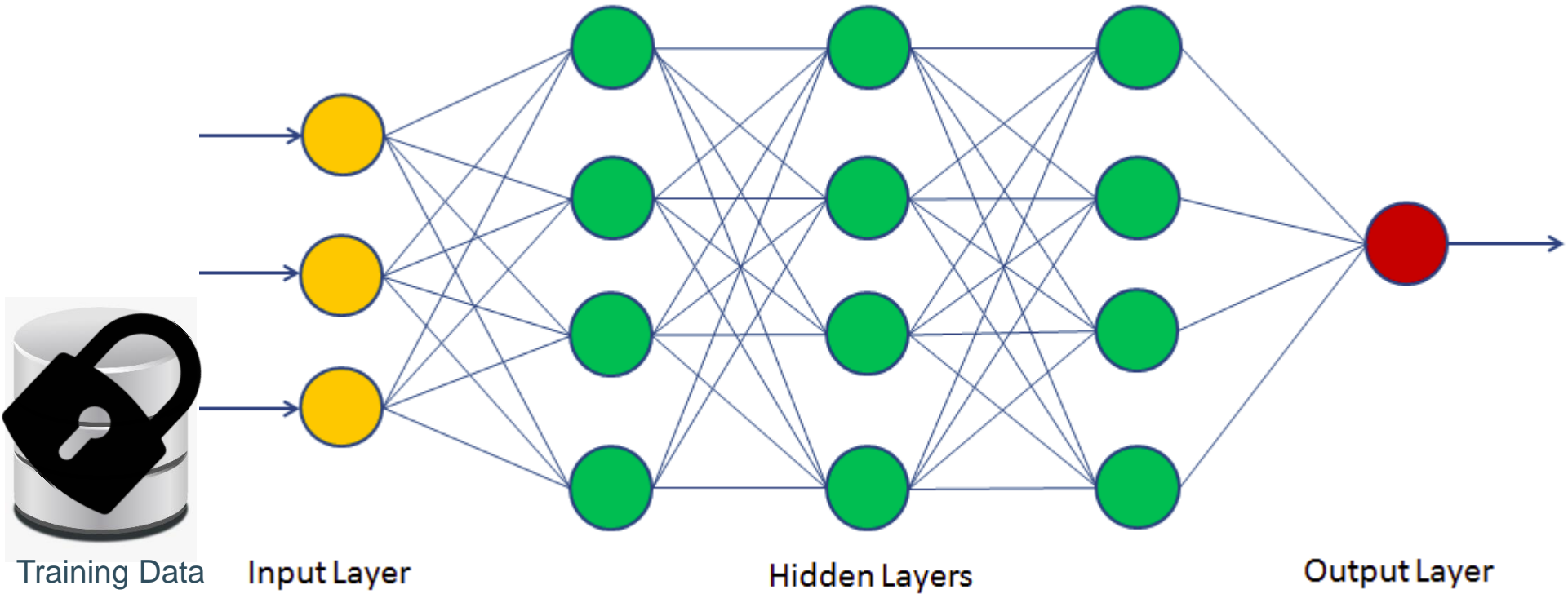




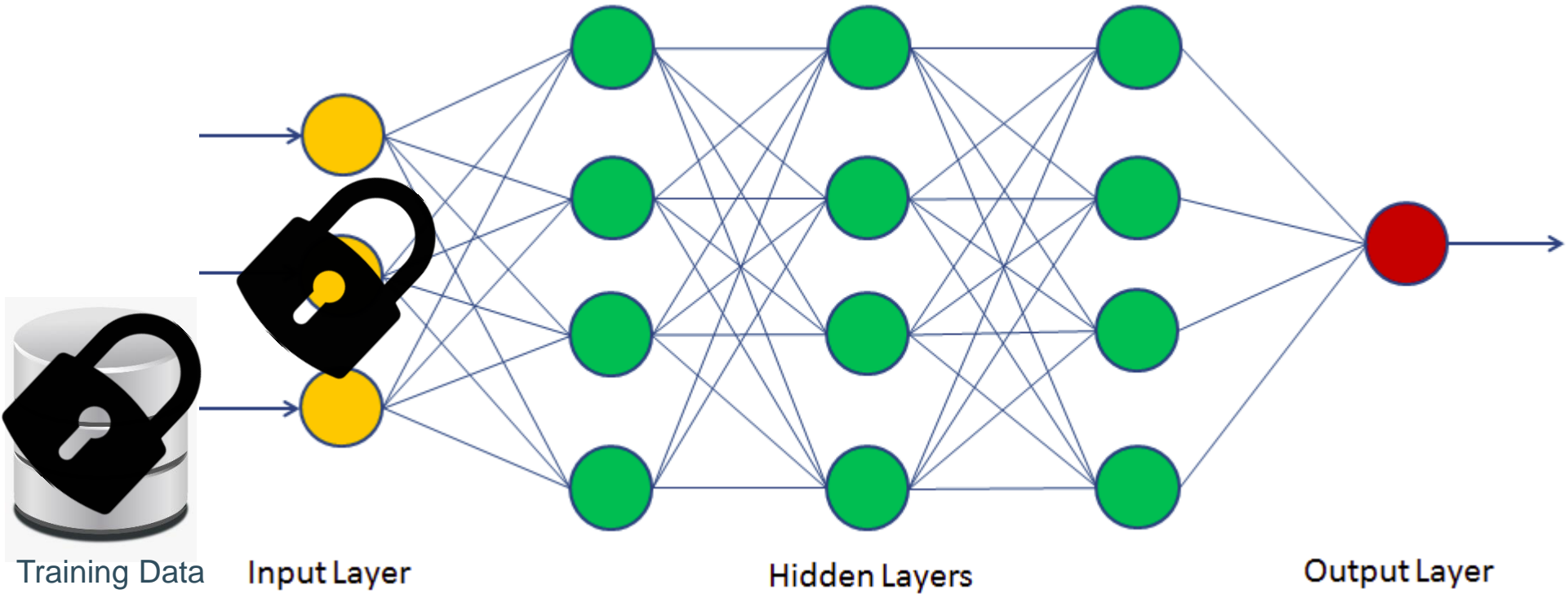
# Privacy-Preserving Machine Learning



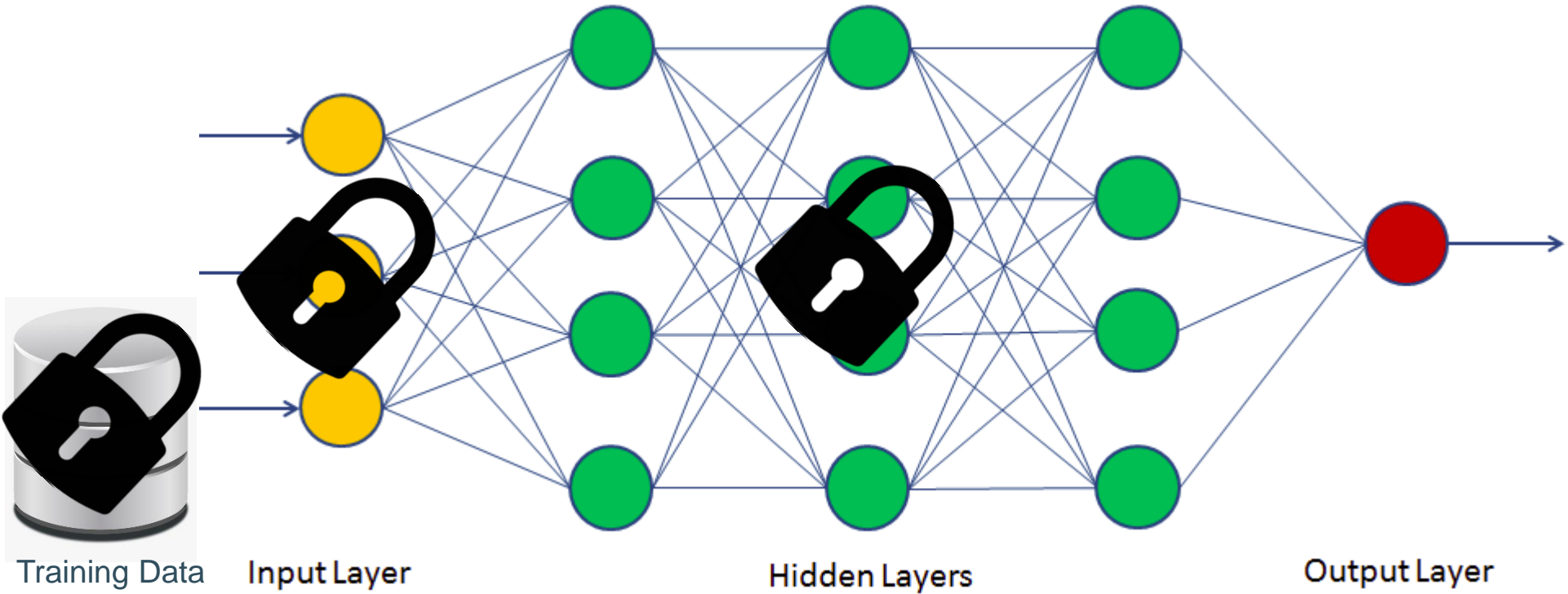
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