Trident: Efficient 4PC Framework for Privacy Preserving Machine Learning

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Outline

- Privacy Preserving Machine Learning (PPML)
- Secure Multi-party Computation (MPC)
- Overview of Trident Protocol
- Benchmarking Results

Machine Learning (ML) Prediction – An Abstraction

Jasmine
(Model Owner)

Model
Parameters

ML Algorithm

Query

Result

Aladdin
(Client)

Privacy ??

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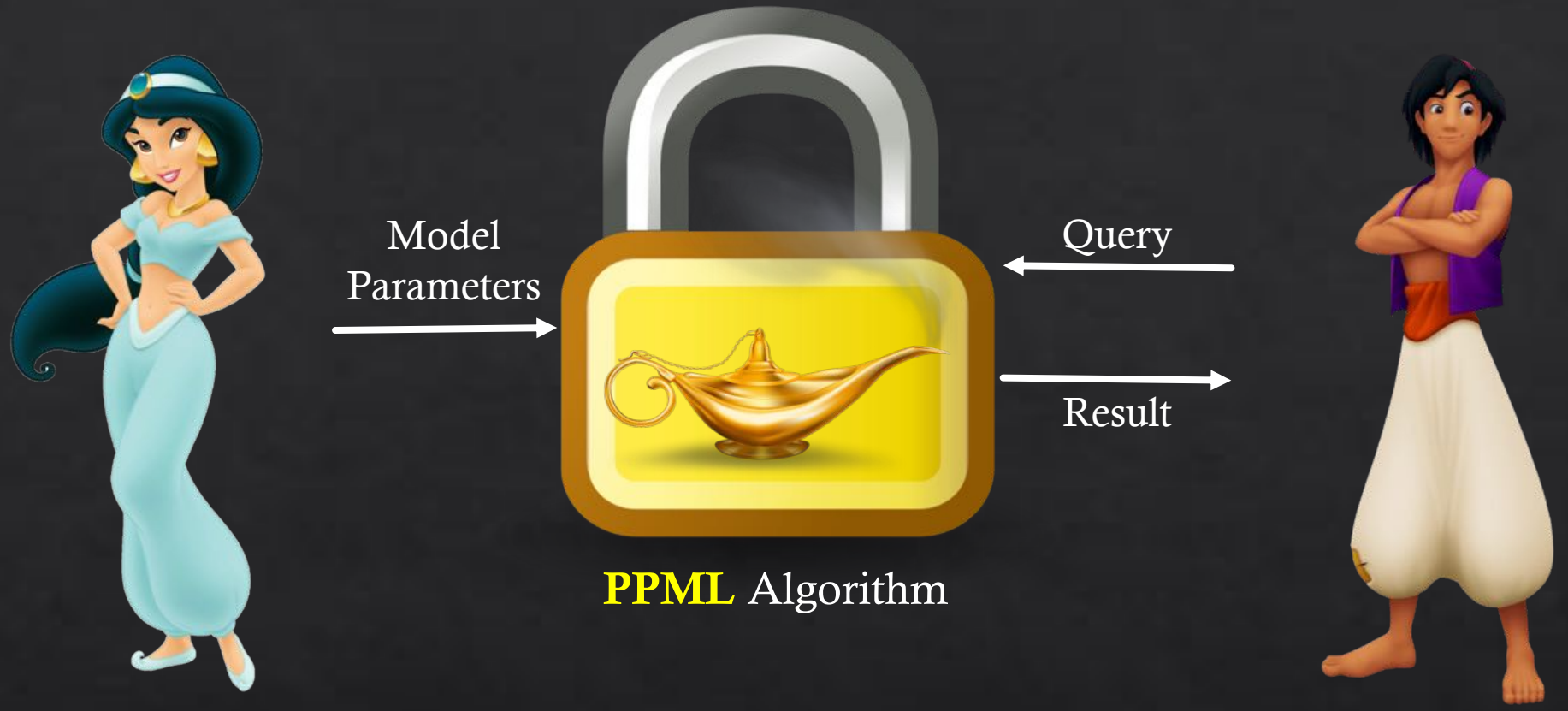
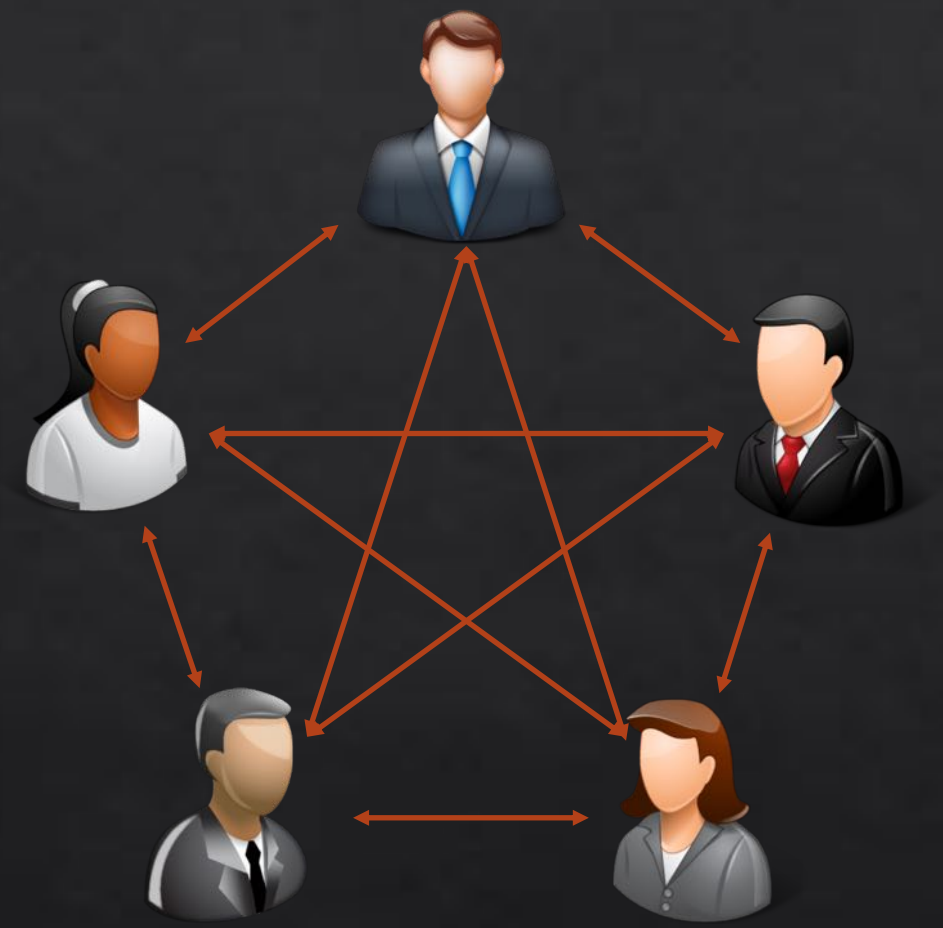
Privacy Preserving Machine Learning (PPML)



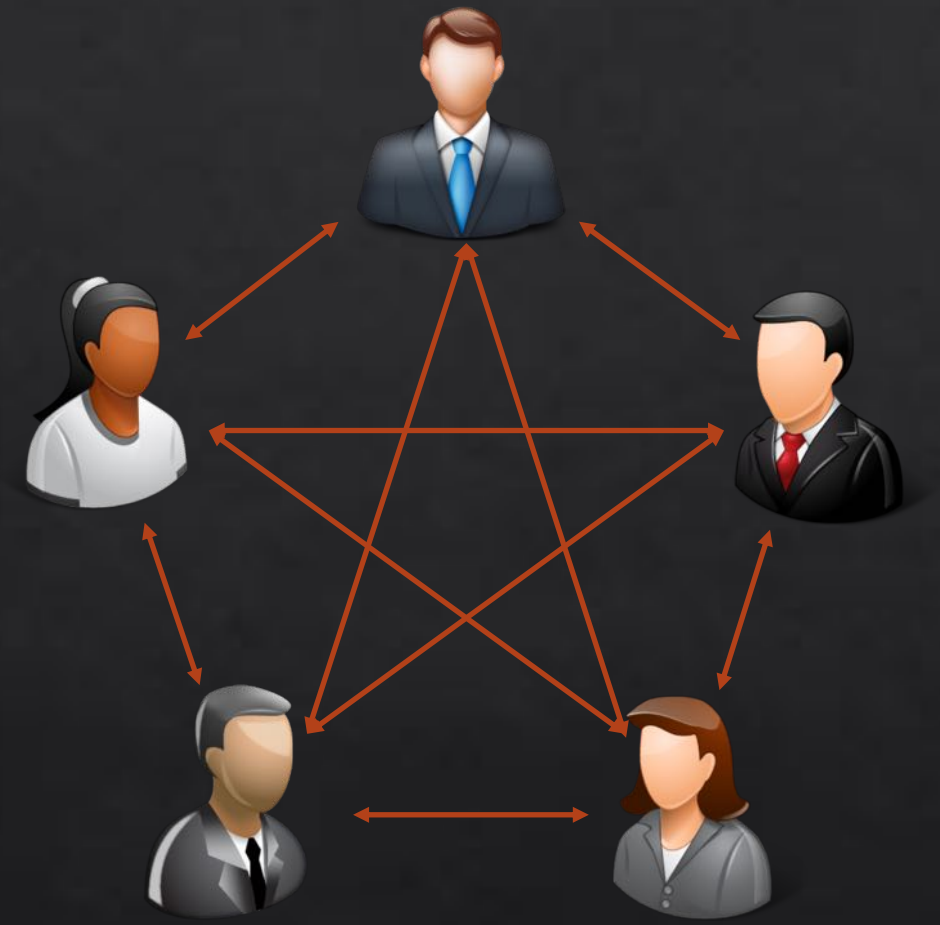
MPC meets ML

Secure Multi-party Computation (MPC) [Yao'82]

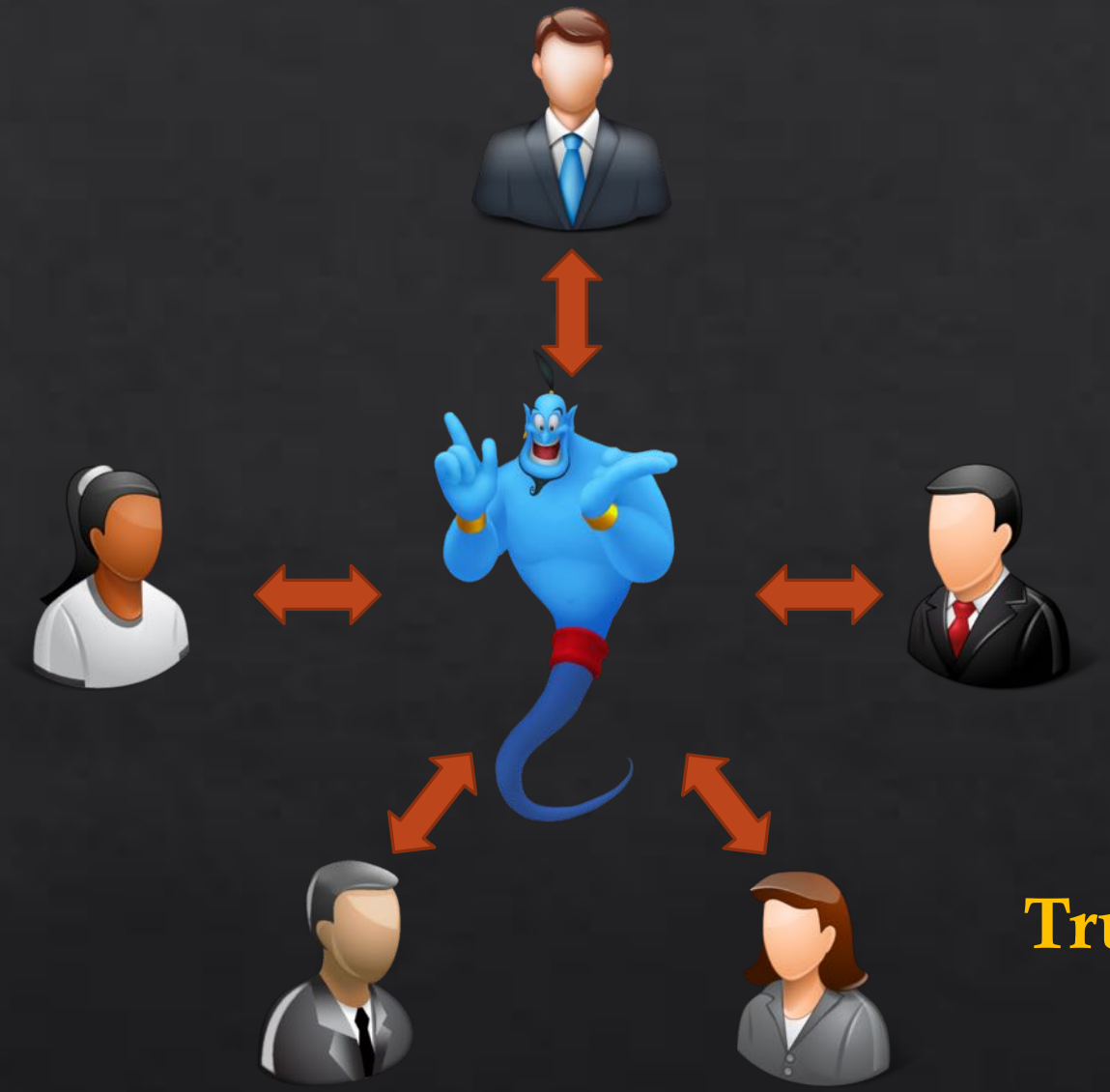


- ✓ A set of parties with private inputs wish to compute some joint function of their inputs.
- ✓ Goals of MPC:
  - **Correctness** – Parties should correctly evaluate the function output.
  - **Privacy** – Nothing more than the function output should be revealed

Secure Multi-party Computation (MPC) [Yao'82]



Trusted Third Party (TTP)



Trusted Third Party (TTP)

MPC

MPC emulates TTP

TRIDENT PROTOCOL

AJITH SURESH | CRYPTOGRAPHY AND INFORMATION SECURITY LAB ,CSA, IISC

26-02-2020

Trident protocol

- A new 4PC protocol over ring in the pre-processing model

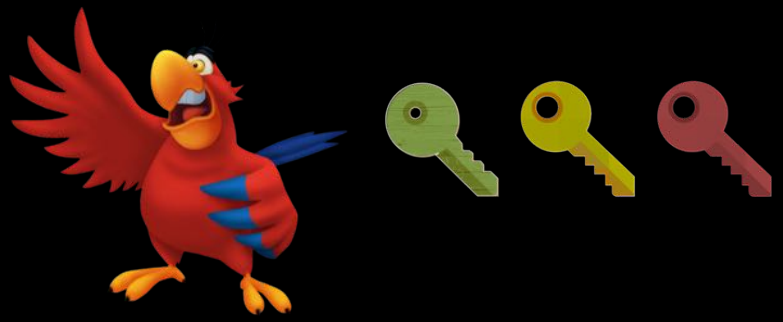
Trident protocol

- A new **4PC protocol** over ring in the pre-processing model

Trident protocol

- A new 4PC protocol over ring in the **pre-processing** model

Sharing Semantics



Pre-processing








Pre-processing

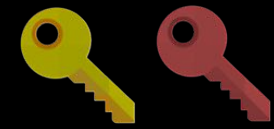
Online





Shares of 




Trident protocol

- A new 4PC protocol over ring in the pre-processing model
- Malicious security with guarantee of fairness

Trident protocol

- A new 4PC protocol over ring in the pre-processing model
- **Malicious** security with guarantee of fairness

Trident protocol

- A new 4PC protocol over ring in the pre-processing model
- Malicious security with guarantee of **fairness**

Multiplication $(x.y)$

Ref	Pre-processing (#elements)	Online (#elements)	Security
Araki et al'17 (3PC)	12	9	Abort

Multiplication $(x.y)$

Ref	Pre-processing (#elements)	Online (#elements)	Security
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ASTRA (3PC)	21	4	Fair

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Gordon et al.'18 (4PC)	2	4	Abort

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Araki et al'17 (3PC)	12	9	Abort
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Gordon et al.'18 (4PC)	2	4	Abort
Trident	**3**	**3**	**Fair**

Trident protocol

- A new 4PC protocol over ring in the pre-processing model
- Malicious security with guarantee of fairness
- Efficient Mixed World Conversions

Mixed World Conversions

Boolean World

- Comparison, Bit Extraction …

Mixed World Conversions

Boolean World

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Arithmetic World

- Addition, Multiplication …

Mixed World Conversions

Boolean World

- Comparison, Bit Extraction …

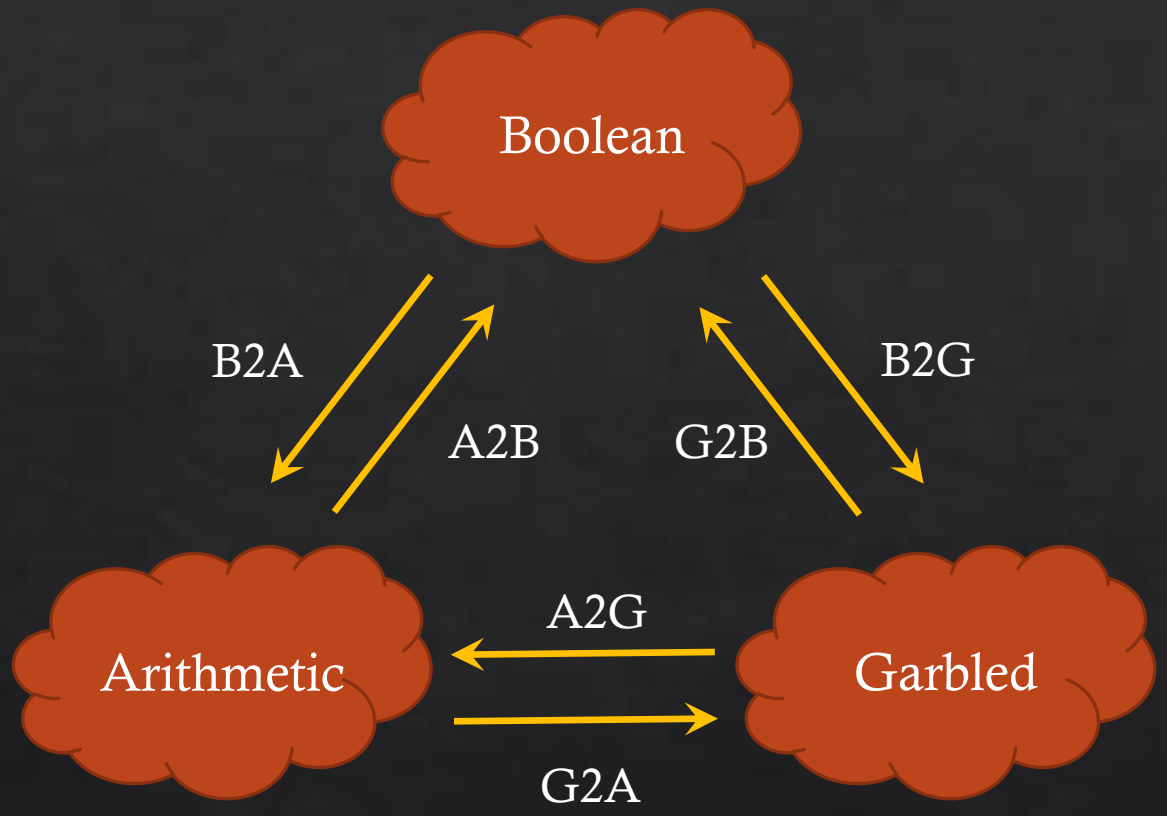
Arithmetic World

- Addition, Multiplication …

Garbled World

- Division over rings …

Mixed World Conversions



Online Rounds	Online Communication
Up to 7x	2x - 67x

Range of improvement over ABY3

Mixed World Conversions – An Example

$$\frac{\min(x_1 + x_2, x_3)}{x_4}$$

Mixed World Conversions – An Example

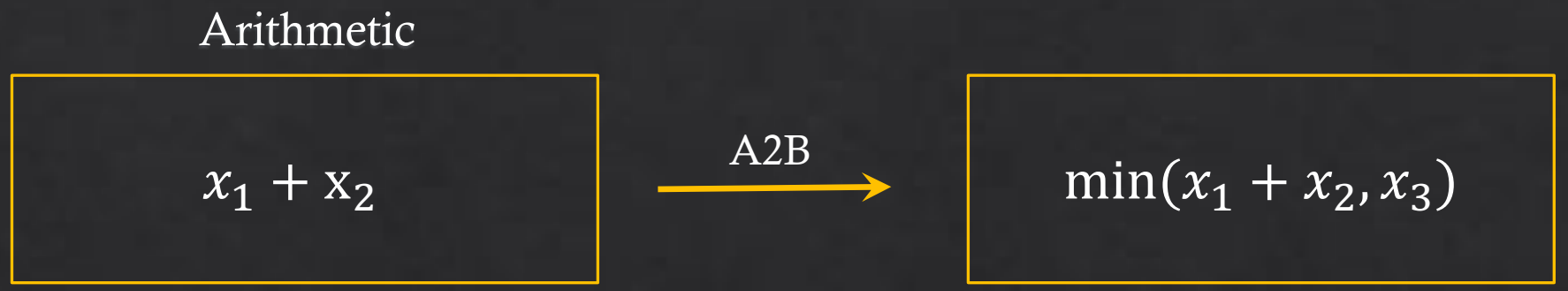
$x_1 + x_2$

Mixed World Conversions – An Example

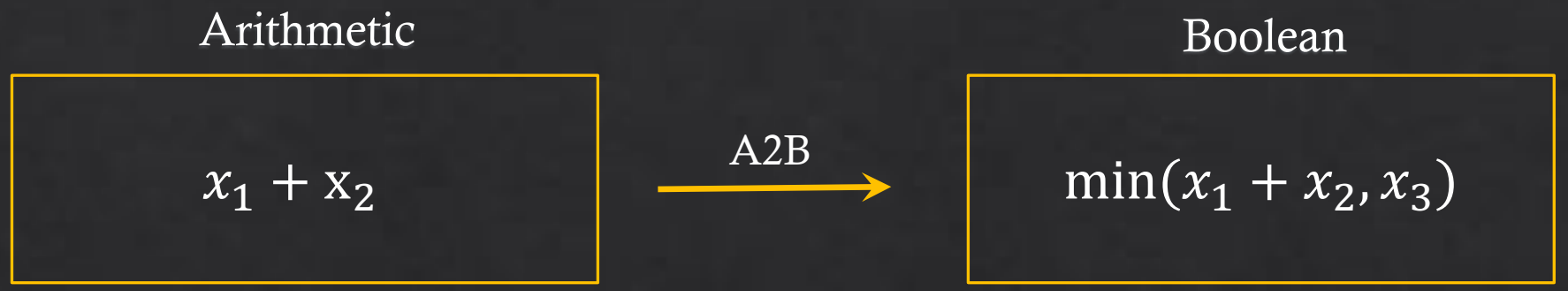
Arithmetic

$x_1 + x_2$

Mixed World Conversions – An Example



Mixed World Conversions – An Example



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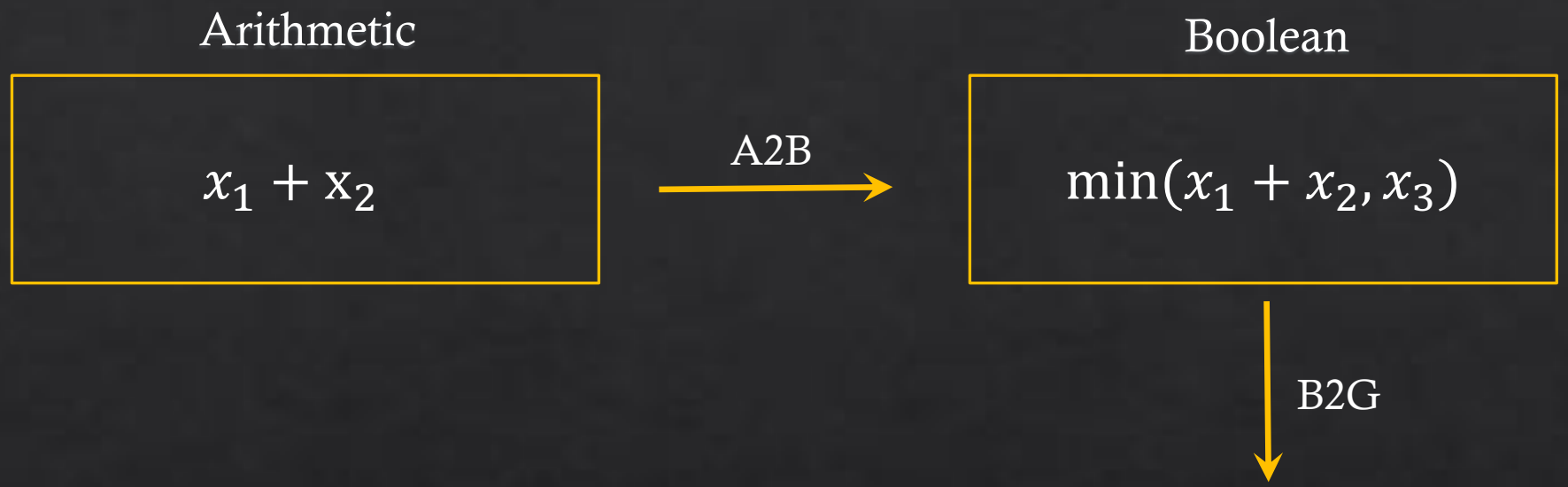

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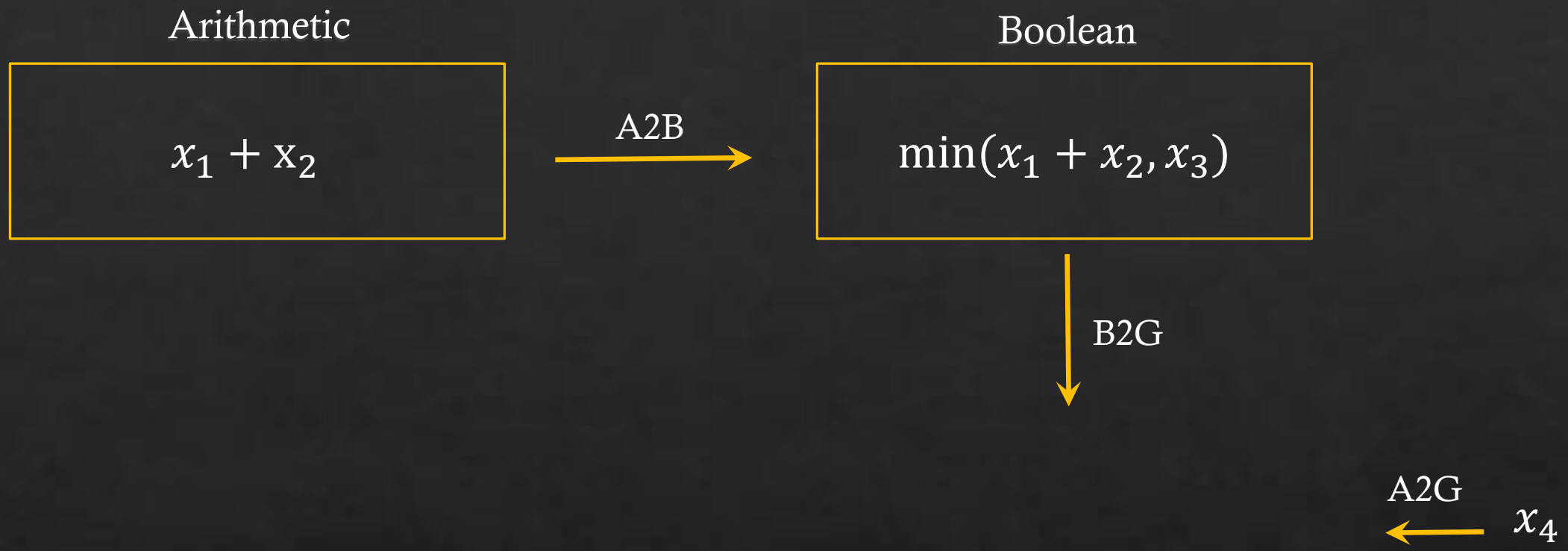

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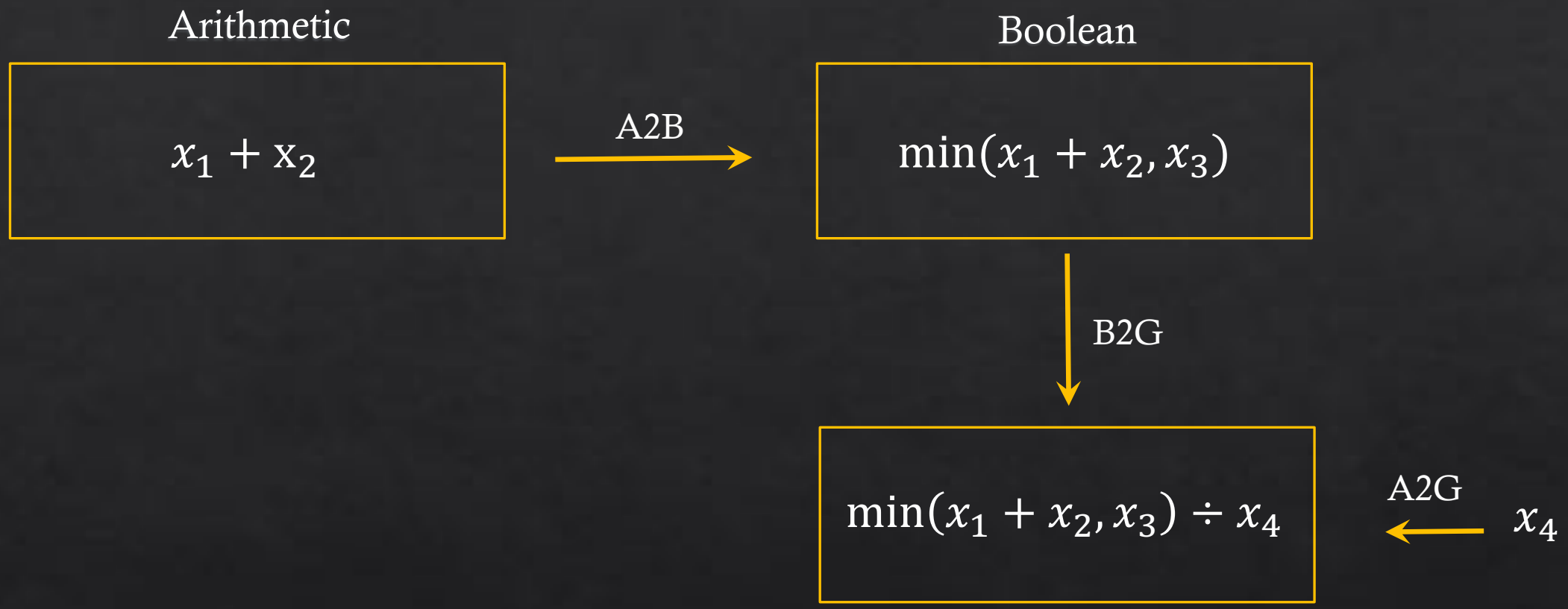

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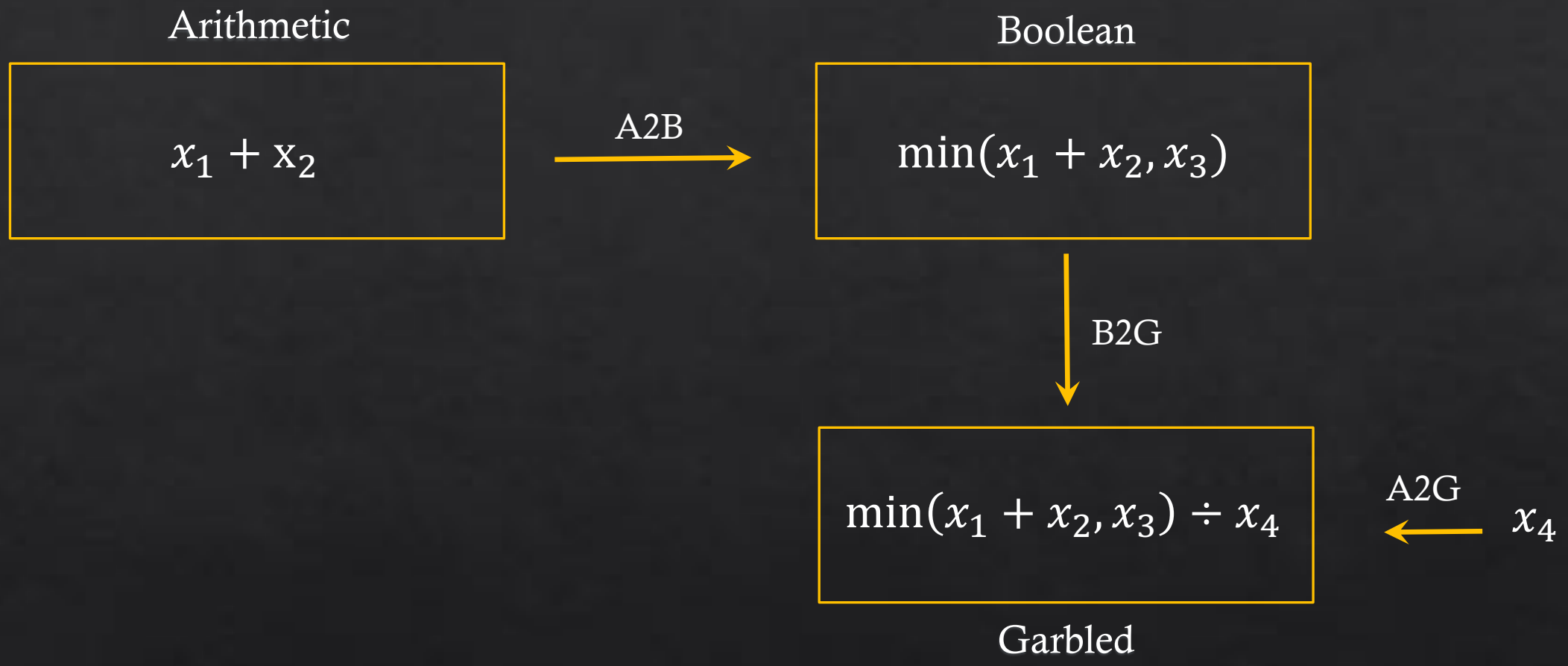

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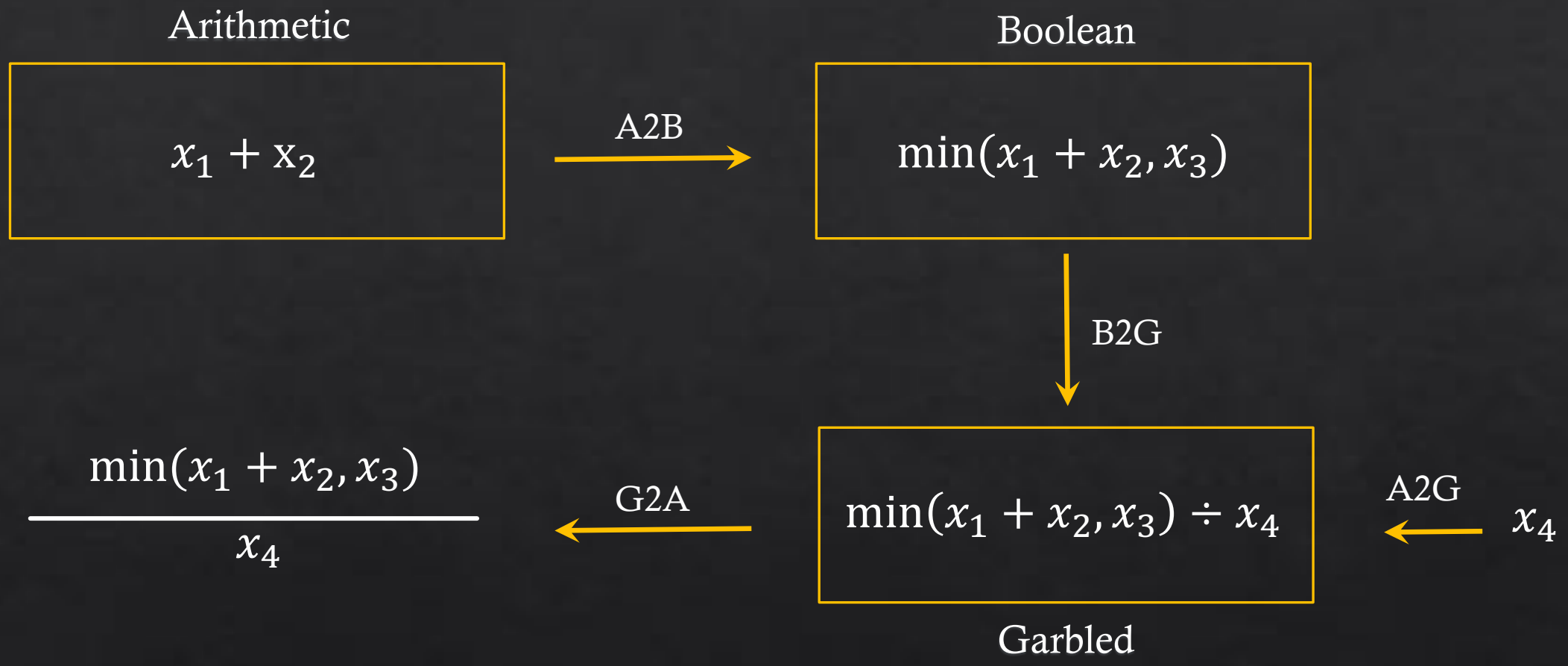

Trident protocol

- A new 4PC protocol over ring in the pre-processing model
- Malicious security with guarantee of fairness
- Efficient Mixed World Conversions
- Special tools for PPML

Dot Product

$$X \blacksquare Y = \sum_{i=1}^{d} x_i \,.\, yi$$

Ref	Pre-processing (#elements)	Online (#elements)	Security
ABY3 (3PC)	12d	9d	Abort

d – #elements in each vector

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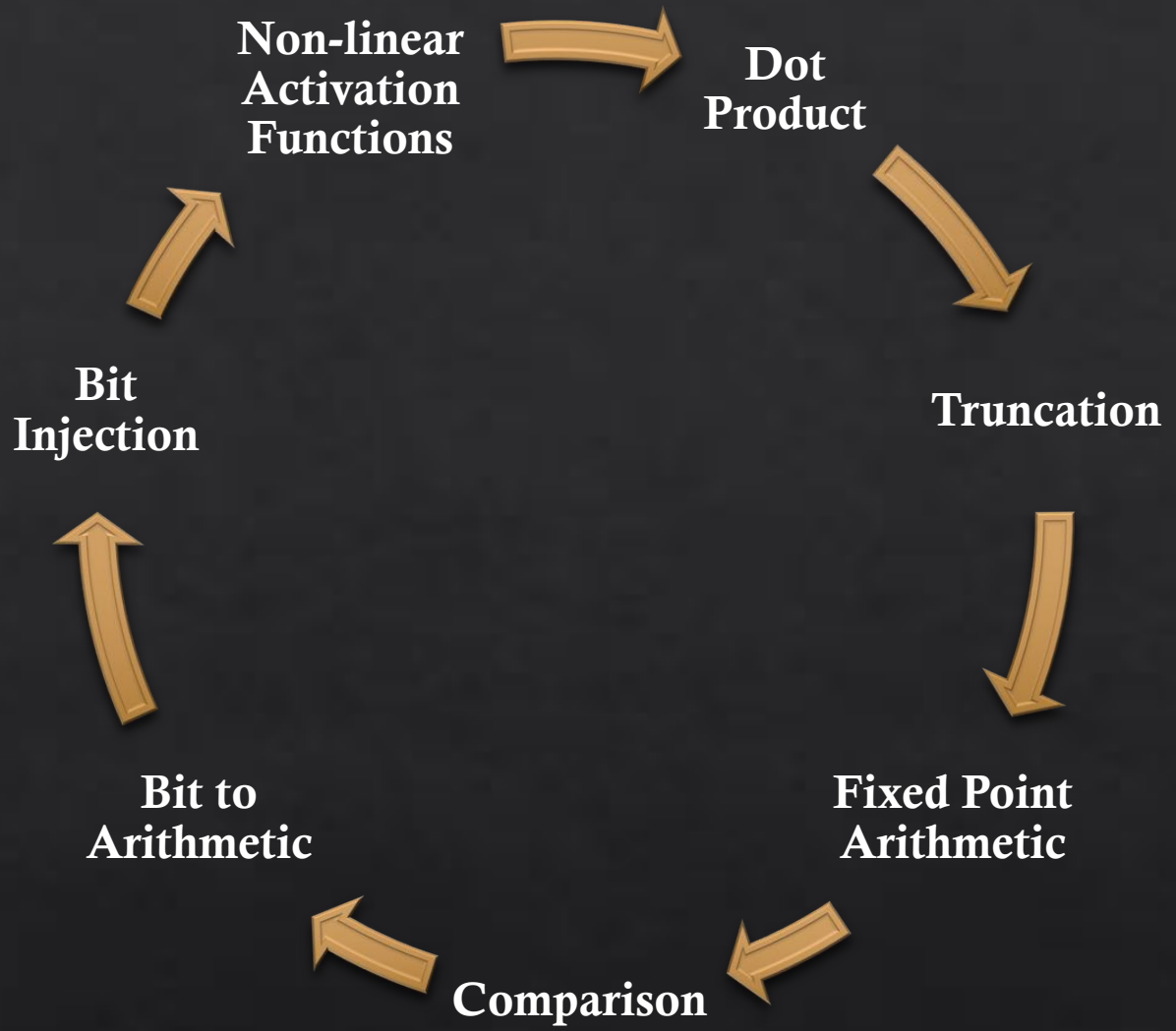

Tools for PPML

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- Efficient Mixed World Conversions
- Special tools for PPML
- Lower monetary cost in the outsourced setting

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Benchmarking

- Implemented both Trident and ABY3, using the ENCRYPTO library.
- Benchmarked the protocols over LAN (40 Mbps) and WAN (1 Gbps) with the Google Cloud Platform.
- Servers located in West Europe, East Australia, South Asia, and South East Asia.
- For benchmarking, we used batch sizes up to 512 and feature sizes up to 1000.

Summary of Our Benchmarking Results

ML Algorithm	Improvement in terms of Online Throughput over ABY3	
	Training	Prediction
Linear Regression	251.84x	145.81x
Logistic Regression	34.58x	149.63x
Neural Networks	63.71x	407.12x
Convolutional Neural Networks	42.81x	741.56x

*Throughput for Training - #iterations processed by servers / minute
*Throughput for Prediction - #queries processed by servers / minute

thank you!

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