

# How to Implement Anything in MPC

*Marcel Keller*   Peter Scholl   Nigel Smart

University of Bristol

19 February 2015

# Overview

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1. Compiler for MPC circuits
2. Implement ORAM as MPC circuit  
⇒ Oblivious array / memory
3. Execute a set of instructions (small circuits) at every step  
⇒ Oblivious execution

Oblivious machine

- ▶ 40 Hz with 1000-entry memory
- ▶ 2 Hz with  $10^6$ -entry memory

Problem:

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Problem: “The idea of implementing ORAM within the MPC is not new, and figuring out the details while cumbersome, doesn’t seem to require any new ideas or techniques. I was not able to pin down any such ideas or any surprises when implementing ORAM in MPC.”

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# How to Implement Anything in MPC

## Malicious-for-free OT Extension and Beyond

### All You Can OT

Tore Frederiksen<sup>1</sup>

*Marcel Keller*<sup>2</sup>

Emmanuela Orsini<sup>2</sup>

Peter Scholl<sup>2</sup>

<sup>1</sup>Aarhus University

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# Overview

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Malicious-for-free OT Extension and Beyond  
All You Can OT

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1. New correlation check for OT extension
2.  $\mathbb{F}_{2^n}$  SPDZ triples from OT



# OT Extension

1. New correlation check for OT extension
2.  $\mathbb{F}_2$ , SPDZ triples from OT

1.  $\kappa$  base OTs
2. Extend length with pseudorandom functions
3. Introduce correlation
4. Transpose
5. Hash to break correlation



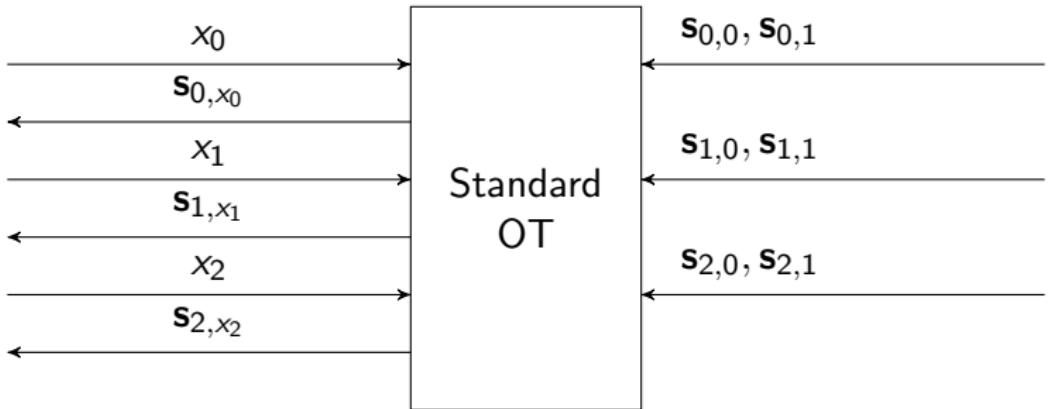
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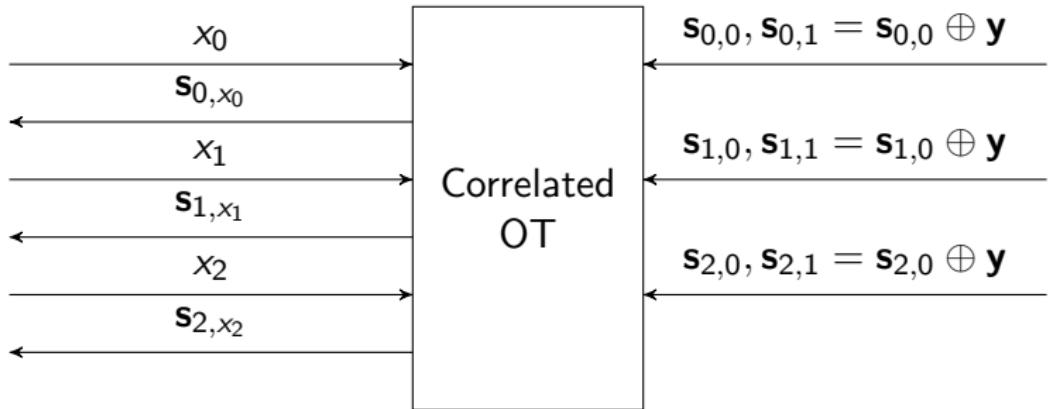


$x_i$ : selection bit

$s_{i,0}, s_{i,1}$ : strings

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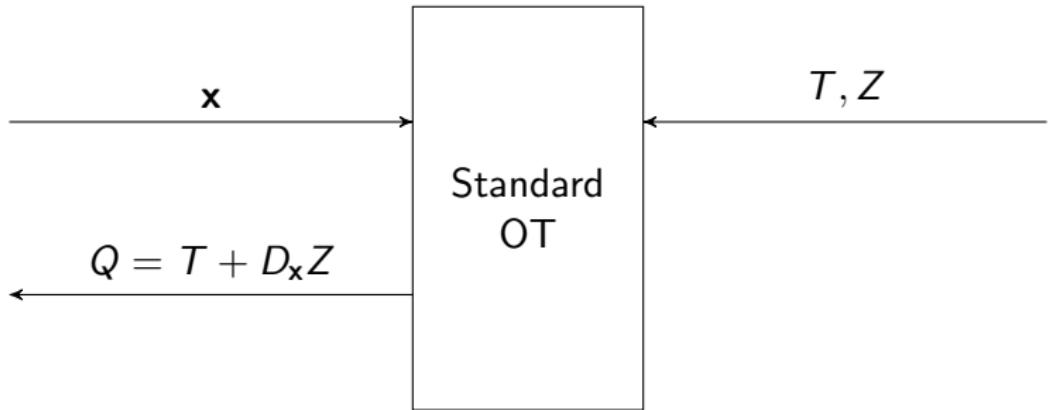


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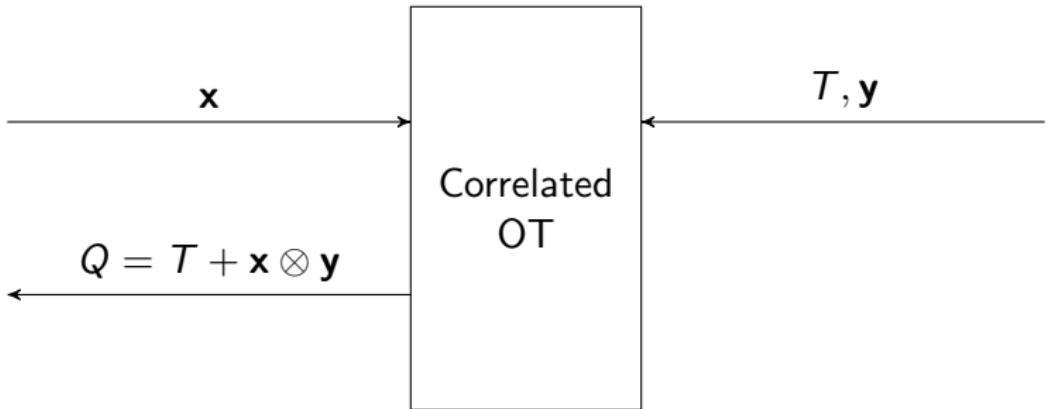
$x$ : string / vector in  $(\mathbb{F}_2)^\kappa$

$Q, T$ : matrices in  $(\mathbb{F}_2)^{\kappa \times \ell}$

$D_x$ : matrix with diagonal  $x$

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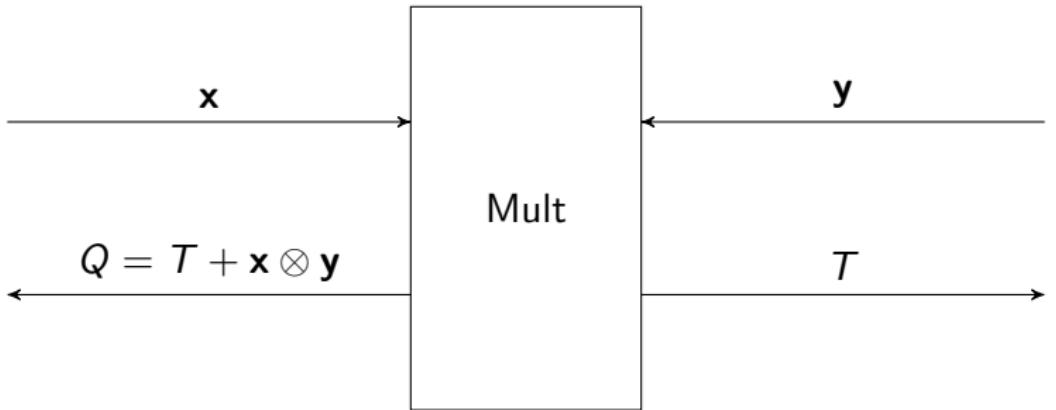
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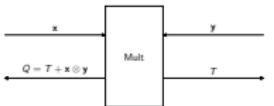


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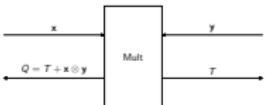
 $x, y$ : strings / vectors in  $(\mathbb{F}_2)^n$  $Q, T$ : matrices in  $(\mathbb{F}_2)^{n \times \ell}$  $D_x$ : matrix with diagonal  $x$  $x \otimes y$ : tensor product, matrix of all possible products

# Errors in Correlation

Honest receiver:

$$Q + T = D_x Z = \mathbf{x} \otimes \mathbf{y} =$$

$$\begin{pmatrix} x_1y_1 & x_1y_2 & x_1y_3 & x_1y_4 & x_1y_5 & x_1y_6 & x_1y_7 \\ x_2y_1 & x_2y_2 & x_2y_3 & x_2y_4 & x_2y_5 & x_2y_6 & x_2y_7 \\ x_3y_1 & x_3y_2 & x_3y_3 & x_3y_4 & x_3y_5 & x_3y_6 & x_3y_7 \\ x_4y_1 & x_4y_2 & x_4y_3 & x_4y_4 & x_4y_5 & x_4y_6 & x_4y_7 \\ x_5y_1 & x_5y_2 & x_5y_3 & x_5y_4 & x_5y_5 & x_5y_6 & x_5y_7 \\ x_6y_1 & x_6y_2 & x_6y_3 & x_6y_4 & x_6y_5 & x_6y_6 & x_6y_7 \\ x_7y_1 & x_7y_2 & x_7y_3 & x_7y_4 & x_7y_5 & x_7y_6 & x_7y_7 \end{pmatrix}$$



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# Errors in Correlation

Dishonest receiver:

$$Q + T = D_x Z = \mathbf{x} \otimes \mathbf{y} + D_x E =$$

$$\left( \begin{array}{ccccccc} x_1 y_1 & x_1 y_2 & x_1 y_3 & x_1 y_4 & x_1 y_5 & x_1 y_6 & x_1 y_7 \\ x_2 y_1 & x_2 y_2 & x_2 y_3 & x_2 y_4 & x_2 y_5 & x_2 y_6 & x_2 y_7 \\ x_3 y_1 & x_3 y_2 & x_3 y_3 & x_3 y_4 & x_3 \poop & x_3 y_6 & x_3 y_7 \\ x_4 y_1 & x_4 y_2 & x_4 y_3 & x_4 y_4 & x_4 y_5 & x_4 y_6 & x_4 y_7 \\ x_5 y_1 & x_5 \poop & x_5 y_3 & x_5 y_4 & x_5 y_5 & x_5 y_6 & x_5 y_7 \\ x_6 \poop & x_6 y_2 & x_6 \poop & x_6 \poop & x_6 y_5 & x_6 y_6 & x_6 y_7 \\ x_7 y_1 & x_7 y_2 & x_7 y_3 & x_7 y_4 & x_7 y_5 & x_7 y_6 & x_7 y_7 \end{array} \right)$$

At least one error in  $i$ -th row  $\Rightarrow$  selective failure attack on  $x_i$



# Correlation Check

Dishonest receiver:

$$Q + T = D_x Z = \mathbf{x} \otimes \mathbf{y} + D_x E = \begin{pmatrix} x_1 y_1 & x_2 y_2 & x_3 y_3 & x_4 y_4 & x_5 y_5 & x_6 y_6 & x_7 y_7 \\ x_2 y_1 & x_2 y_2 & x_2 y_3 & x_2 y_4 & x_2 y_5 & x_2 y_6 & x_2 y_7 \\ x_3 y_1 & x_3 y_2 & x_3 y_3 & x_3 y_4 & x_3 y_5 & x_3 y_6 & x_3 y_7 \\ x_4 y_1 & x_4 y_2 & x_4 y_3 & x_4 y_4 & x_4 y_5 & x_4 y_6 & x_4 y_7 \\ x_5 y_1 & x_5 y_2 & x_5 y_3 & x_5 y_4 & x_5 y_5 & x_5 y_6 & x_5 y_7 \\ x_6 y_1 & x_6 y_2 & x_6 y_3 & x_6 y_4 & x_6 y_5 & x_6 y_6 & x_6 y_7 \\ x_7 y_1 & x_7 y_2 & x_7 y_3 & x_7 y_4 & x_7 y_5 & x_7 y_6 & x_7 y_7 \end{pmatrix}$$

At least one error in  $i$ -th row  $\Rightarrow$  selective failure attack on  $x_i$ 

Honest receiver:  $Q + T = \mathbf{x} \otimes \mathbf{y}$   
 $\Leftrightarrow \mathbf{q}_j + \mathbf{t}_j = y_j \cdot \mathbf{x}$  for all columns of  $Q, T$

## Tricks

- ▶ Check random linear combination for  $j = 1, \dots, \ell$
- ▶ Confuse  $(\mathbb{F}_2)^\kappa$  and  $\mathbb{F}_{2^\kappa}$

## Protocol

1. Sample  $\chi_1, \dots, \chi_\ell$  securely
2. Receiver computes and sends  $\mathbf{v} = \sum_j \chi_j \cdot \mathbf{t}_j$  and  $\mathbf{w} = \sum_j \chi_j \cdot \mathbf{y}_j$
3. Sender checks whether  $\sum_j \chi_j \cdot \mathbf{q}_j + \mathbf{v} = \mathbf{w} \cdot \mathbf{x}$



Honest receiver:  $Q + T = \mathbf{x} \odot \mathbf{y}$   
 $\Leftrightarrow \mathbf{q}_j + \mathbf{t}_j = y_j \cdot \mathbf{x}$  for all columns of  $Q, T$

## Tricks

- ▶ Check random linear combination for  $j = 1, \dots, \ell$
- ▶ Confuse  $(\mathbb{F}_2)^\ell$  and  $\mathbb{F}_{2^k}$ .

## Protocol

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# Correlation Check II

Dishonest receiver:  $Q + T = D_{\mathbf{x}} Z$

- ▶  $\mathbf{z}_1, \dots, \mathbf{z}_\ell$  columns of  $Z$ , not all  $(0, \dots, 0)$  or  $(1, \dots, 1)$
- ▶  $\mathbf{x} * \mathbf{z}_1, \dots, \mathbf{x} * \mathbf{z}_\ell$  columns of  $D_{\mathbf{x}} Z$

Receiver needs to compute  $\mathbf{v}, \mathbf{w}$  such that

$$\begin{aligned}\mathbf{v} + \mathbf{w} \cdot \mathbf{x} &= \sum_j \chi_j \cdot \mathbf{q}_j \\ &= \sum_j \chi_j \cdot (\mathbf{t}_j + \mathbf{z}_j * \mathbf{x})\end{aligned}$$

Intuition: Impossible without some guessing about  $\mathbf{x}$   
if  $\mathbf{z}_1, \dots, \mathbf{z}_\ell$  not all  $(0, \dots, 0)$  or  $(1, \dots, 1)$

Proof: Not straightforward

- Dishonest receiver:  $Q + T = D_k Z$
- ▶  $z_1, \dots, z_l$ : columns of  $Z$ , not all  $(0, \dots, 0)$  or  $(1, \dots, 1)$
  - ▶  $x + z_1, \dots, x + z_l$ : columns of  $D_k Z$

Receiver needs to compute  $v, w$  such that

$$\begin{aligned} v + w \cdot x &= \sum_j \chi_j \cdot q_j \\ &= \sum_j \chi_j \cdot (t_j + z_j + x) \end{aligned}$$

Intuition: Impossible without some guessing about  $x$   
if  $z_1, \dots, z_l$  not all  $(0, \dots, 0)$  or  $(1, \dots, 1)$

Proof: Not straightforward

# Correlation Check III

Receiver needs to compute  $w$  such that

$$v + w \cdot x = \sum_j \chi_j \cdot (t_j + z_j * x)$$

## Example

$z_j = (1, 0, 1, 0, \dots, 1, 0)$  for all  $j$ . If  $x_1 = x_2, x_3 = x_4, \dots,$

$\Rightarrow x * z_j = x / (1 + X)$  for all  $j$

$\Rightarrow v = \sum_j \chi_j \cdot t_j, w = (1 + X)^{-1} \cdot \sum_j \chi_j$

## Theorem (Almost sufficient)

*The receiver can learn whether  $x$  is in some affine  $(\kappa - m)$ -dimensional space with success probability  $2^{-m}$ .*



# Correlation Check – The Other Side

Receiver needs to compute  $w$  such that

$$\mathbf{v} + \mathbf{w} \cdot \mathbf{x} = \sum_j \chi_j \cdot (\mathbf{t}_j + \mathbf{z}_j \cdot \mathbf{x})$$

## Example

$\mathbf{z}_j = (1, 0, 1, 0, \dots, 1, 0)$  for all  $j$ . If  $x_1 = s_2, x_3 = s_4, \dots$

$\Rightarrow \mathbf{x} \cdot \mathbf{z}_j = \mathbf{x}_j \cdot (1 + X)$  for all  $j$

$\Rightarrow \mathbf{v} = \sum_j \chi_j \cdot \mathbf{t}_j, \mathbf{w} = (1 + X)^{-1} \cdot \sum_j \chi_j$

## Theorem (Almost sufficient)

The receiver can learn whether  $\mathbf{x}$  is in some affine  $(n - m)$ -dimensional space with success probability  $2^{-m}$ .

## Leakage

$\sum_j \chi_j \cdot y_j$  gives information about  $\mathbf{y}$

## Solution

Discard enough bits of  $\mathbf{y}$



# Experiments

## Leakage

$\sum_i \chi_i \cdot x_i$  gives information about  $y$

## Solution

Discard enough bits of  $y$

- ▶ 10 million OTs
- ▶ 8 threads

	LAN	WAN
Passive security	3.3258 s	13.1510 s
Active security	3.3516 s	13.4157 s

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## Part II

# Beyond – The Road to SPDZ

- ▶ 10 million OTs
- ▶ 8 threads

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# Amplified Correlated OT

TinyOT: Check correlated OT, then amplify

SPDZ: Check in sacrificing, amplify first?

Amplification with random matrix  $M \in \mathbb{F}_2^{\ell \times 3\ell}$  (honest receiver):

$$\begin{aligned} MQ + MT &= MD_x Z \\ &= M \cdot (\mathbf{x} \otimes \mathbf{y}) \\ &= (M\mathbf{x}) \otimes \mathbf{y} \end{aligned}$$

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# Amplification of Errors

Amplification with random matrix  $M \in \mathbb{F}_2^{\ell \times 3\ell}$ :

$$MD_x Z = (Mx) \otimes y + MD_x E =$$

$$M \cdot \left( \begin{array}{ccccccc} x_1 y_1 & x_1 y_2 & x_1 y_3 & x_1 y_4 & x_1 y_5 & x_1 y_6 & x_1 y_7 \\ x_2 y_1 & x_2 y_2 & x_2 y_3 & x_2 y_4 & x_2 y_5 & x_2 y_6 & x_2 y_7 \\ x_3 y_1 & x_3 y_2 & x_3 y_3 & x_3 y_4 & x_3 \poop & x_3 y_6 & x_3 y_7 \\ x_4 y_1 & x_4 y_2 & x_4 y_3 & x_4 y_4 & x_4 y_5 & x_4 y_6 & x_4 y_7 \\ x_5 y_1 & x_5 y_2 & x_5 y_3 & x_5 y_4 & x_5 y_5 & x_5 y_6 & x_5 y_7 \\ x_6 y_1 & x_6 y_2 & x_6 y_3 & x_6 y_4 & x_6 y_5 & x_6 y_6 & x_6 y_7 \\ x_7 y_1 & x_7 y_2 & x_7 y_3 & x_7 y_4 & x_7 y_5 & x_7 y_6 & x_7 y_7 \end{array} \right)$$

Few errors:  $Mx$  independent of errors with high probability

Many errors:  $MD_x E$  has high entropy

$\Rightarrow$  SPDZ sacrifice will fail with high probability



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Amplification with random matrix  $M \in \mathbb{R}^{d \times M}$ 

$$MD_x Z = (Mx) \otimes y + MD_x E =$$

$$M \cdot \begin{pmatrix} x_1 & x_1 y_2 & x_1 y_3 & x_1 y_4 & x_1 y_5 & x_1 y_6 & x_1 y_7 \\ x_2 y_1 & x_2 y_2 & x_2 & x_2 y_4 & x_2 y_5 & x_2 y_6 & x_2 y_7 \\ x_3 y_1 & x_3 y_2 & x_3 y_3 & x_3 y_4 & x_3 y_5 & x_3 y_6 & x_3 y_7 \\ x_4 y_1 & x_4 y_2 & x_4 y_3 & x_4 y_4 & x_4 y_5 & x_4 y_6 & x_4 y_7 \\ x_5 y_1 & x_5 y_2 & x_5 y_3 & x_5 y_4 & x_5 y_5 & x_5 y_6 & x_5 y_7 \\ x_6 y_1 & x_6 y_2 & x_6 & x_6 y_4 & x_6 y_5 & x_6 y_6 & x_6 y_7 \\ x_7 y_1 & x_7 y_2 & x_7 y_3 & x_7 y_4 & x_7 y_5 & x_7 y_6 & x_7 y_7 \end{pmatrix}$$

Few errors:  $Mx$  independent of errors with high probabilityMany errors:  $MD_x E$  has high entropy

→ SPDZ sacrifice will fail with high probability

# Generating $\mathbb{F}_{2^n}$ SPDZ Triples

Amplified correlated OT is a two-party tensor product box  
 ⇒ Use for every pair of parties

## Protocol

1. Pairwise amplified OT to get secret-shared triple  $[a], [b], [c = a \cdot b]$
2. Pairwise leaky correlated OT to get secret-shared MACs  $[a \cdot \Delta], [b \cdot \Delta], [c \cdot \Delta]$
3. Improved SPDZ sacrifice (error detection)

## Running Time

Estimate: 200 times faster than SPDZ for  $\mathbb{F}_{2^n}$

Amplification with random matrix  $M \in \mathbb{R}^{d \times M}$

$$MD_x Z = (Mx) \otimes y + MD_x E =$$

$$M \cdot \begin{pmatrix} x_1 \triangleup \\ x_2 \triangleup \\ x_3 \triangleup \\ x_4 \triangleup \\ x_5 \triangleup \\ x_6 \triangleup \\ x_7 \triangleup \end{pmatrix} \otimes \begin{pmatrix} x_1 y_2 & x_1 y_3 & x_1 y_5 & x_1 y_6 & x_1 y_7 \\ x_2 y_1 & x_2 y_2 & x_2 y_4 & x_2 y_5 & x_2 y_7 \\ x_3 y_1 & x_3 y_2 & x_3 y_3 & x_3 y_6 & x_3 y_7 \\ x_4 y_1 & x_4 y_2 & x_4 y_3 & x_4 y_5 & x_4 y_7 \\ x_5 y_1 & x_5 y_2 & x_5 y_3 & x_5 y_5 & x_5 y_7 \\ x_6 y_1 & x_6 y_2 & x_6 y_3 & x_6 y_5 & x_6 y_7 \\ x_7 y_1 & x_7 y_2 & x_7 y_3 & x_7 y_5 & x_7 y_7 \end{pmatrix}$$

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 Many errors:  $MD_x E$  has high entropy  
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# Generating $\mathbb{F}_{2^n}$ SPDZ Triples

## Protocol

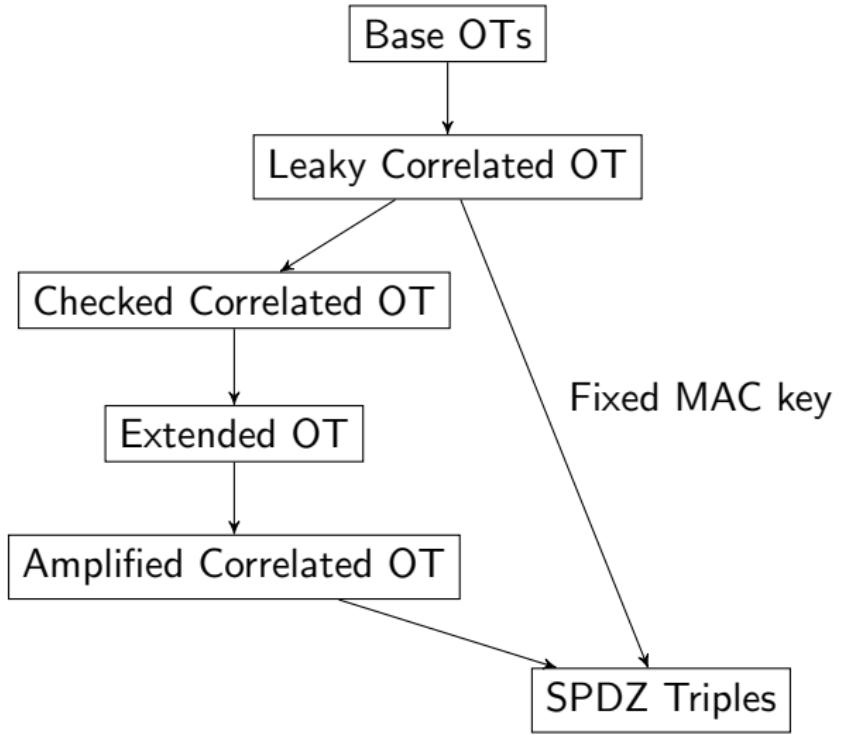
1. Pairwise amplified OT to get secret-shared triple
2. Pairwise leaky correlated OT to get secret-shared MACs
3. Improved SPDZ sacrifice (error detection)

## Attacks

- ▶ Cheating in amplified OT:  
Erased by amplification or fail error detection
- ▶ Cheating in leaky correlated OT:  
Some bits of MAC key are leaked. Not an issue because complete leakage succeeds only with negligible probability
- ▶ Use different inputs for different parties: fail error detection



# Toolchain



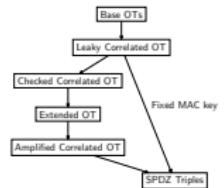
## Protocol

1. Pairwise amplified OT to get secret-shared triple
2. Pairwise leaky correlated OT to get secret-shared MACs
3. Improved SPDZ sacrifice (error detection)

## Attacks

- ▶ Cheating in amplified OT:
  - Erasé by amplification or fail error detection
- ▶ Cheating in leaky correlated OT:
  - Some bits of MAC key are leaked. Not an issue because complete leakage succeeds only with negligible probability
- ▶ Use different inputs for different parties: fail error detection

## Toolchain





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## Field Multiplication from Tensor Product

$$(1, X, X^2, \dots) \cdot (\mathbf{x} \otimes \mathbf{y}) \cdot (1, X, X^2, \dots)^\top =$$

$$(1 \ X \ X^2 \ \dots) \cdot \begin{pmatrix} x_1y_1 & x_1y_2 & x_1y_3 & \dots \\ x_2y_1 & x_2y_2 & x_2y_3 & \dots \\ x_3y_1 & x_3y_2 & x_3y_3 & \dots \\ \vdots & & & \ddots \end{pmatrix} \cdot \begin{pmatrix} 1 \\ X \\ X^2 \\ \vdots \end{pmatrix}$$

$$= x_1y_1 + (x_1y_2 + x_2y_1) \cdot X + (x_1y_3 + x_2y_2 + x_3y_1) \cdot X^2 + \dots$$

$$= \mathbf{x} \cdot \mathbf{y}$$

# Hashing Step

$$\begin{aligned}
 & (1, X, X^2, \dots) \cdot (\mathbf{x} \otimes \mathbf{y}) \cdot (1, X, X^2, \dots)^T = \\
 & \left( \begin{array}{cccc} 1 & X & X^2 & \dots \end{array} \right) \cdot \left( \begin{array}{cccc} x_1 y_1 & x_1 y_2 & x_1 y_3 & \dots \\ x_2 y_1 & x_2 y_2 & x_2 y_3 & \dots \\ x_3 y_1 & x_3 y_2 & x_3 y_3 & \dots \\ \vdots & \vdots & \vdots & \ddots \end{array} \right) \cdot \left( \begin{array}{cccc} 1 & X & X^2 & \dots \end{array} \right)^T \\
 & = x_1 y_1 + (x_1 y_2 + x_2 y_1) \cdot X + (x_1 y_3 + x_2 y_2 + x_3 y_1) \cdot X^2 + \dots \\
 & = \mathbf{x} \cdot \mathbf{y}
 \end{aligned}$$

Extended random OT:  $j$ -th input is  $H(\mathbf{q}_j)$  and  $H(\mathbf{q}_j + \mathbf{x})$

$$\begin{aligned}
 H(\mathbf{q}_j) &= H(\mathbf{t}_j + \mathbf{x} * \mathbf{z}_j) \\
 H(\mathbf{q}_j + \mathbf{x}) &= H(\mathbf{t}_j + \mathbf{x} * \overline{\mathbf{z}}_j)
 \end{aligned}$$

## Theorem (Sufficient)

*If the check passes with probability  $2^{-m}$ , then  $\kappa - m$  bits of either  $\mathbf{x} * \mathbf{z}_j$  or  $\mathbf{x} * \overline{\mathbf{z}}_j$  remain unknown to the sender of the base OT / receiver of the extended OT.*