

Linear Codes for Secure Computation

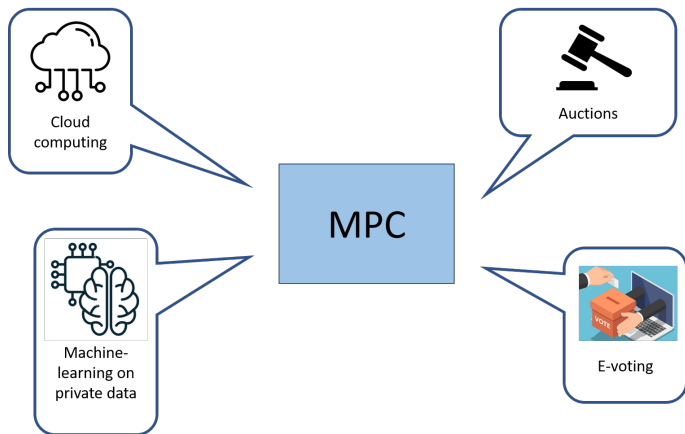
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Secure Computation : what for ?

Classical cryptography goal : protecting communications. But data can be used in computations.



What is Secure Multiparty Computation (MPC)?

Goal : Consider n players, each one owning a secret value x_i . Each player wants to compute the result of a function f_i on the entries $(x_j)_{1 \leq j \leq n}$

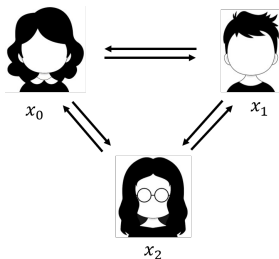
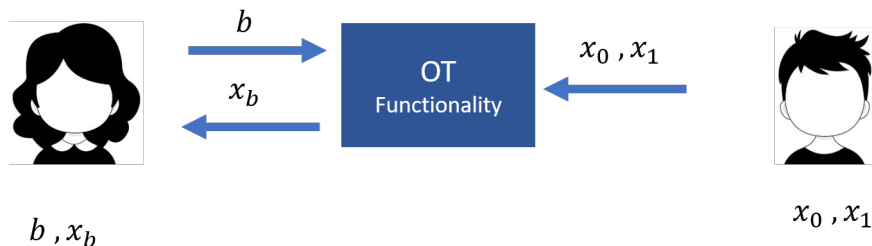


Figure: 3 players MPC

- *Correctness* \rightarrow Each player should get the correct result.
- *Security* \rightarrow Any group of players who ally themselves must not learn more than is already implied by their secret entries and their function.

Example 1 : Oblivious Transfer



OT requires **public-key cryptography**

Useful correlation for efficiently computing Boolean circuit.

Example 2 : OLE and Vector OLE

OLE : Oblivious Linear Evaluation



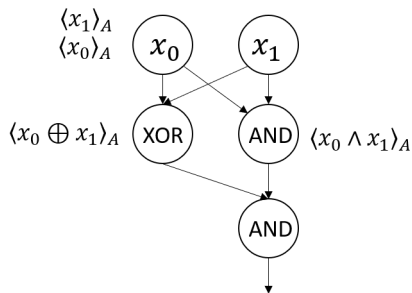
Useful correlation for efficiently computing arithmetic circuit.



How to achieve MPC? [GMW87]

Secret sharing of each inputs !

$$\langle x_0 \rangle_A \oplus \langle x_0 \rangle_B = x_0$$



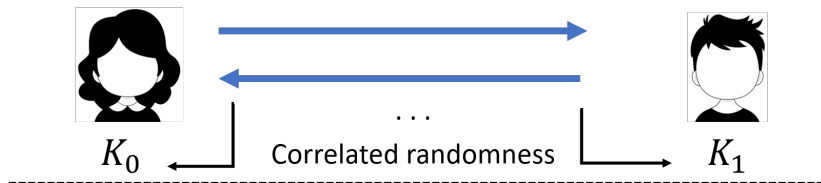
$$\langle x_0 \oplus x_1 \rangle_A = \langle x_0 \rangle_A \oplus \langle x_1 \rangle_A$$

$$\langle x_0 \wedge x_1 \rangle_A = ? \text{ Requires 2 OT}$$

How to achieve MPC?

Research leads to split the protocol in two phases [Bea95, IKNP03]

First Phase : Preprocessing

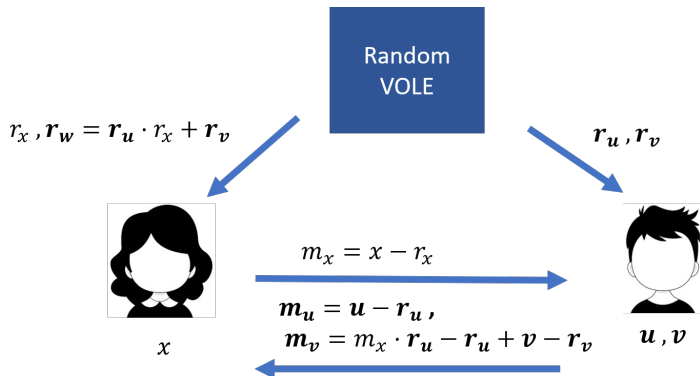


Second Phase



The first phase is input-independent, and can be done ahead of time

From random VOLE to Pseudorandom Vector OLE

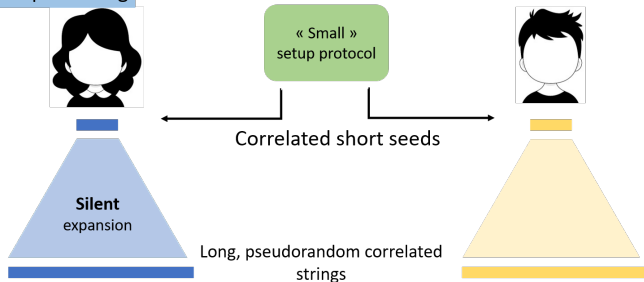


$$w = m_u x + m_v + r_w = u \cdot x + v$$

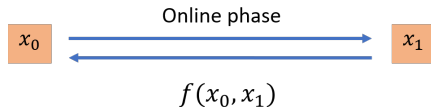
Question: how to generate many random OTs ? (or others correlations).

Correlated Randomness generation [BCGI18, BCG⁺19]

First Phase : Preprocessing



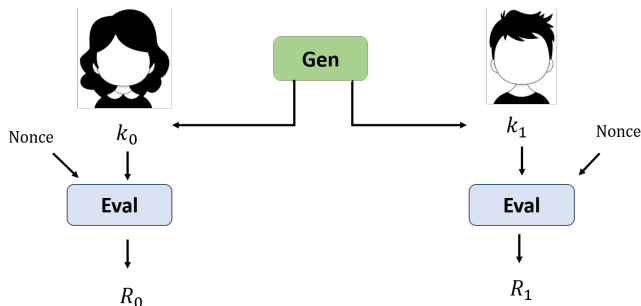
Second Phase



- Very fast online phase
- Few communication to compute
- Downside : we have to do again all the computation when it is done.

Pseudorandom Correlated Functions(PCF) [BCG⁺20]

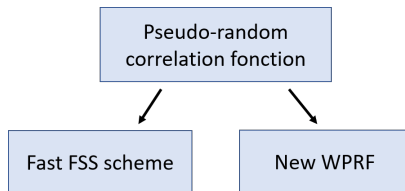
PRF : functions that cannot be distinguished from truly random functions
Equivalent for correlation ?



Correctness : $(R_0, R_1) \approx$ fresh sample of correlation

Security : against insiders

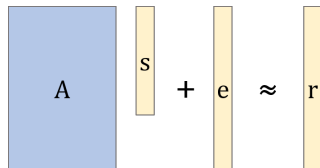
How to construct a PCF ?



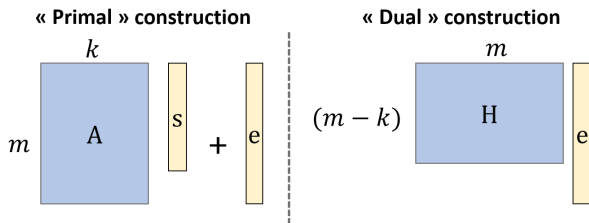
- A Weak Pseudo-Random Function = PRF but the adversary can't choose where to evaluate the functions.

The LPN assumption

Let $A \in \mathbb{F}_q^{m \times k}$, $s \in \mathbb{F}_q^k$, $e \in \mathbb{F}_q^m$, $r \in \mathbb{F}_q^m$, with $\mathcal{HW}(e)$ small.

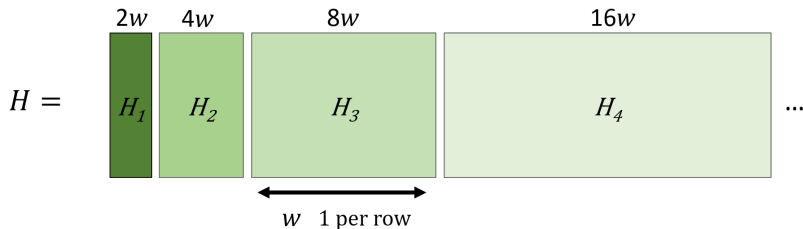

$$A s + e \approx r$$

We mostly focus on the dual version of this assumption, which is equivalent

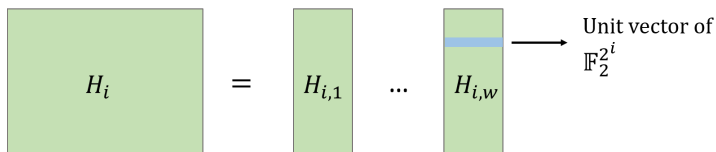


The LPN and the VDLPN assumption

The matrix H and the noise e have some structure !



Exponentially decreasing density. The noise have the shape of one line of H .



Linear attacks examples

Attacks	Types of attacks
Gaussian elimination	Linear
Statistical decoding	Linear
Information set decoding	Linear
BKW	Linear
Algebraic	Non-linear
Statistical Query Algorithm	Non-linear

Linear attacks for our variants

Definition (Bias of a distribution)

Given a distribution \mathcal{D} over \mathbb{F}_2^n , a vector $\mathbf{u} \in \mathbb{F}_2^n$:

$$\text{bias}_u(\mathcal{D}) = \left| \frac{1}{2} - \Pr_{\mathbf{v} \leftarrow \mathcal{D}} [\mathbf{u}^\top \cdot \mathbf{v} = 1] \right|$$

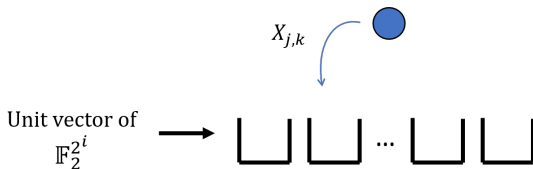
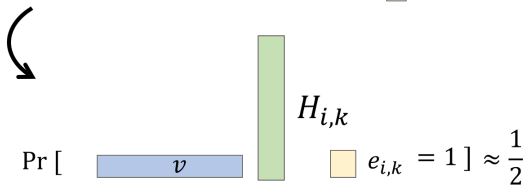
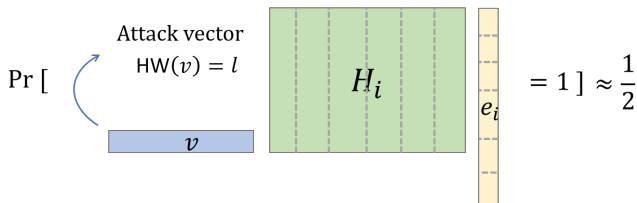
Definition (Resistance against linear attacks)

We obtain the resistance against linear attacks when

$$\Pr_{x^{(1)}, \dots, x^{(N(\lambda))} \leftarrow \mathbb{F}_2^{n(\lambda)}} [\text{bias}(\mathcal{D}(x)) > \epsilon(\lambda)] < \delta(\lambda)$$

where ϵ and δ are small depending on the security parameter λ .

Analysis of security



Analysis of security

We define $R_{i,l,k} = (\mathbf{v}^\top \cdot H_{i,k}) = \left(\bigoplus_{j=1}^l X_{j,k} \right)$ and $Z_{i,l,k}$ as $Z_{i,l,k} = |2^{i-1} - R_{l,k}|$.

Definition (δ -Bad Matrices)

Let $M \in \mathbb{F}_2^{N \times 2^i}$. We say that $M \in \text{Bad}_{\delta, \mathbf{v}}$ with respect to a vector $\mathbf{v} \in \mathbb{F}_{2^N}$ if

$$(\mathbf{v}^\top \cdot M) = Z_{l,k} \in [(1/2 - \delta) \cdot 2^i, 2^{i-1}].$$

Given vector \mathbf{v} , we denote $B_{\delta, \mathbf{v}} = \#\text{Bad}_{\delta, \mathbf{v}}$.

Lemma

For any $\mathbf{v} \in S_{i,N}$, there is a constant C such that

$$\Pr \left[B_{\delta, \mathbf{v}} > \alpha \cdot w \right] \leq 2^{-C \cdot 2^i \cdot w}$$

Analysis of security

We introduce a function Φ

$$\Phi(X_{1,1}, \dots, X_{l,w}) = 2^{i-1} \cdot w - \sum_{k=1}^w Z_{l,k}.$$

$$\Pr \left[B_{\delta, \mathbf{v}} \geq \alpha \cdot w \right] \leq \Pr \left[\Phi(X_{1,1}, \dots, X_{l,w}) < \gamma \cdot w \cdot 2^i \right],$$

Φ is 2-Lipschitz : we use the Bounded Difference Inequality.

Proposition (Bounded Difference Inequality)

Let $\Phi : [n]^m \rightarrow \mathbb{R}$ be a function satisfying the Lipschitz property with constant d , and let (X_1, \dots, X_m) be independent random variables over $[n]$

$$\Pr[\Phi(X_1, \dots, X_m) < \mathbb{E}[\Phi(X_1, \dots, X_m)] - t] \leq \exp\left(-\frac{2t^2}{m \cdot d^2}\right)$$

Analysis of security

Remains to find an upper bound of $\mathbb{E}[\Phi] \rightarrow$ find an upper bound of $\mathbb{E}[Z_{l,k}]$.
There was an error in the proof [BCG⁺20]!

Correction

$$\mathbb{E}[Z_{l,k}] = \sum_{j=0}^{2^{i-1}-1} \Pr(R_{l,k} \geq j+1+2^{i-1}) + \sum_{j=0}^{2^{i-1}-1} \Pr(R_{l,k} \leq 2^{i-1}-j-1)$$

- 1 We bound the shares that we can with the Generalized Chernoff Inequality.
- 2 For that we had to prove that the distribution of the $R_{l,k}$ shows some kind of independence.
- 3 We bound the remaining shares with a trivial bound.

We have to remember the union bound !

This corrects the proof but is highly unpractical, with $w \approx 10^6$.

Analysis of security - Generalized Chernoff Inequality

Proposition

Let $n \in \mathbb{N}$ an integer, and let (Y_1, \dots, Y_n) be independent boolean random variables such that, for some $\eta \in [0, 1]$ it holds that for every subset $S \in [n]$, $\Pr \left[\bigwedge_{q \in S} Y_q \right] \leq \eta^{|S|}$. Then for any $\kappa \in [\eta, 1]$,

$$\Pr \left[\sum_{q=1}^n Y_q \geq \kappa n \right] \leq \exp(-n \cdot D_{KL}(\kappa || \eta)),$$

where $D_{KL}(\kappa || \eta)$ denotes the relative entropy function, defined as

$$D_{KL}(\kappa || \eta) = \kappa \cdot \ln \frac{\kappa}{\eta} + (1 - \kappa) \ln \left(\frac{1 - \kappa}{1 - \eta} \right).$$

Analysis of security - New approach

A new proof :

- Simulation to prove that $\mathbb{E}[Z] < \beta^i$ with better β
- Erasing the corner cases.
- A new idea to bound the bias :

$$\Pr[\text{bias}_{\mathbf{v}}(\mathcal{O}_{\text{par}}^i) > B] = \Pr \left[\prod_{k=1}^w Z_{i,l,k} > 2^{(i-1)w} \times (2B) \right].$$

- The *sum* $\sum_k Z_{i,l,k}$ is minimized when all the terms in the product are equal.

$$\Pr[\text{bias}_{\mathbf{v}}(\mathcal{O}_{\text{par}}^i) > B] \leq \Pr \left[\sum_{k=1}^w Z_{i,l,k} > w \cdot 2^{(i-1)} \cdot c \right],$$

- With this sum we can apply again our results with the function Φ .

- We obtain with this new proof $w \approx 350$.

Estimation of the **concrete cost** of the PCFs.

- Seed size: 2.55MB
- PCF evaluation time: ≈ 500 PCF evaluations per second on a single 3GHz processor.

Another work of [BCG⁺20] also suggested an improved all prefix variant. No proof of the security for this variant yet, but very promising values.

- Seed size: 0.34MB.
- PCF evaluation time: around 3500 evaluations per second on a single 3GHz processor.

Conclusion and open questions

- Pseudo-random Function achieves very promising parameters.

Open Problems and ongoing works :

- All prefix Variant
- Variable density matrix shapes.
- Ring LPN and variants

Questions ?



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