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Optimized homomorphic evaluation of Boolean functions

Nicolas Bon

Joint work with David Pointcheval and Matthieu Rivain

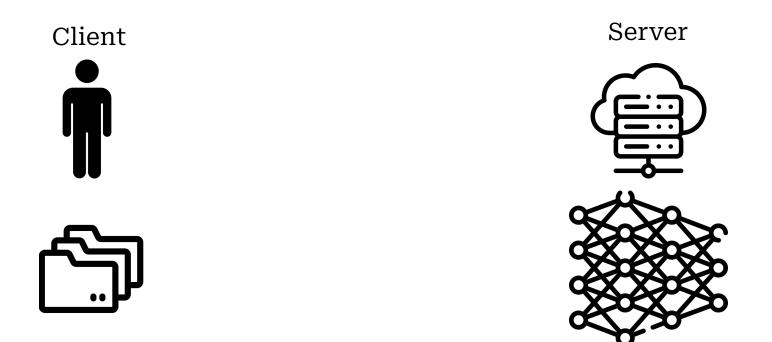
Outline

- 1. Introduction to Fully Homomorphic Encryption
- 2. Introduction to the TFHE cryptosystem
- 3. Our contributions : a framework for fast evaluation of Boolean functions

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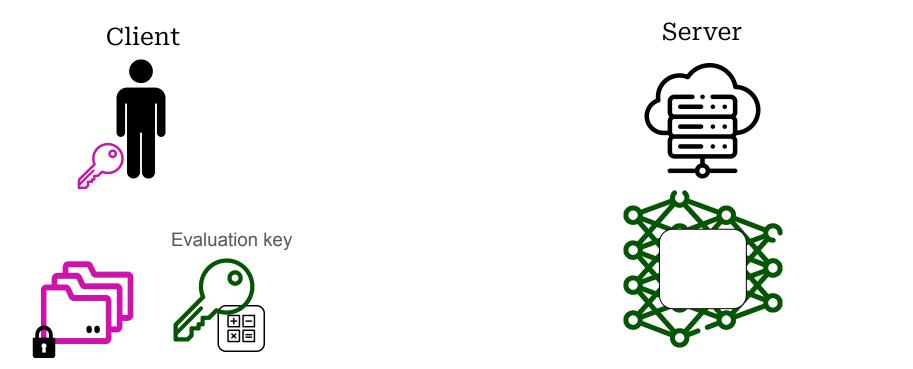
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What is FHE ?

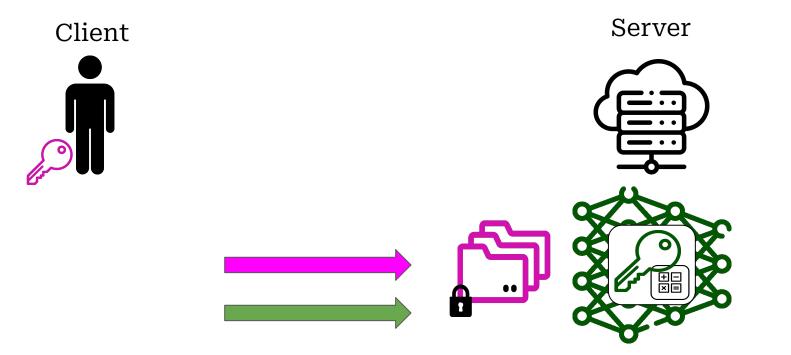


The client wants to use the neural network on its data.

- The data is encrypted using a homomorphic scheme
- The server runs a homomorphized version of the neural network
- All computations can be performed without any decryption or information leak.



The client encrypts its data and crafts an evaluation key that will be used in the homomorphized neural network.



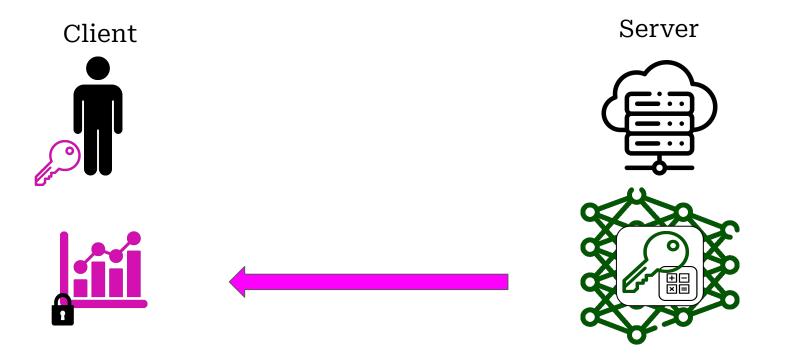
The server gets the encrypted data and the evaluation key.



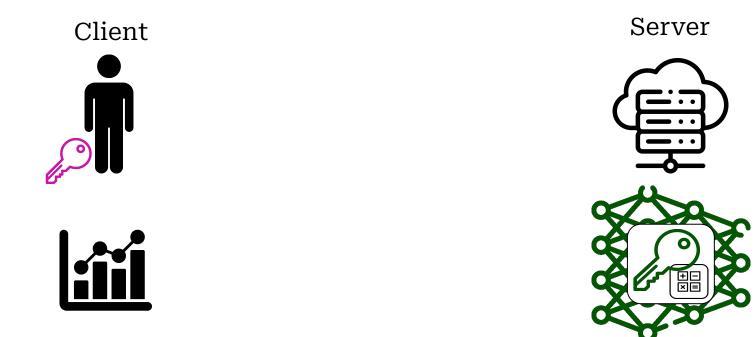
Server



Thanks to the evaluation key, the server evaluates the neural network on the data without decryption and gets a result in an encrypted form



The server sends back the encrypted result to the client.



The client can then decrypt the result !

Main challenges of FHE





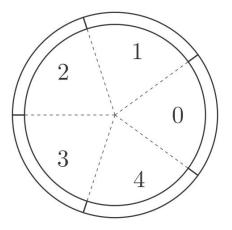


Limited set of supported homomorphic operations

Outline

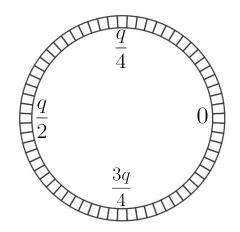
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Clear space: \mathbb{T}_p



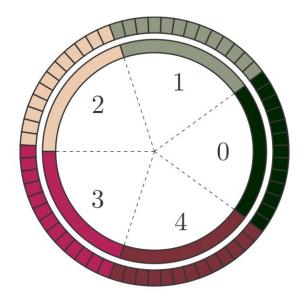
p has a size of few bits.

Encrypted space: \mathbb{T}_q

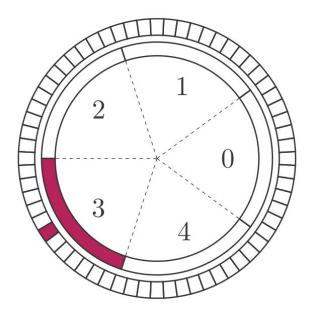


 $q = 2^{32}$ or 2^{64}

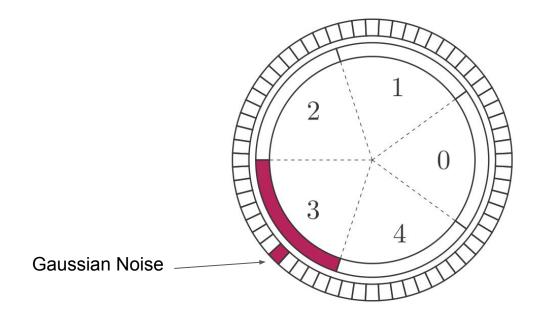
Natural embedding of
$$\, \mathbb{T}_p \,$$
 in $\, \mathbb{T}_q \,$



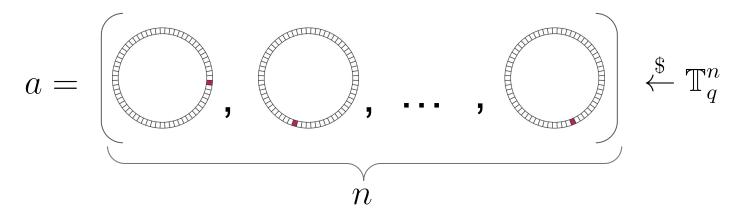
Encoding of a message $m \in \mathbb{T}_p$



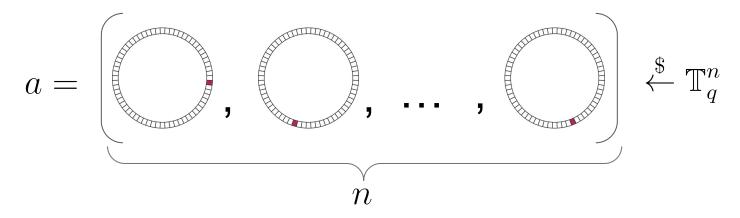
Encoding of a message $m \in \mathbb{T}_p$



Sampling of a mask :



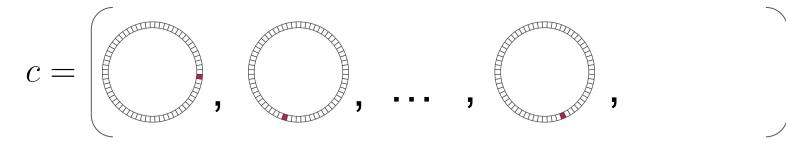
Sampling of a mask :



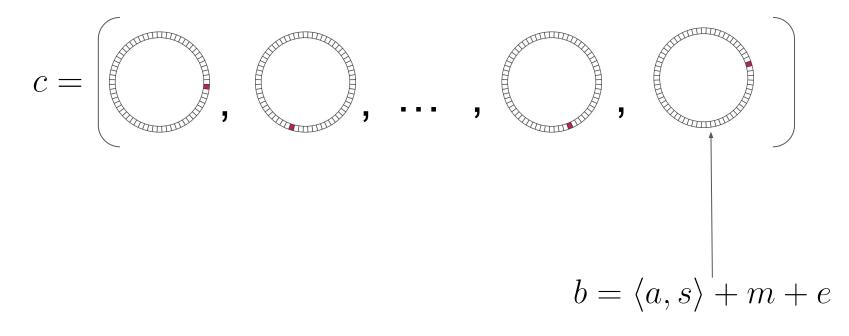
Secret key:

$$s_k = (1, 0, \dots, 1) \xleftarrow{\$} \mathbb{B}^n$$

Construction of ciphertexts:



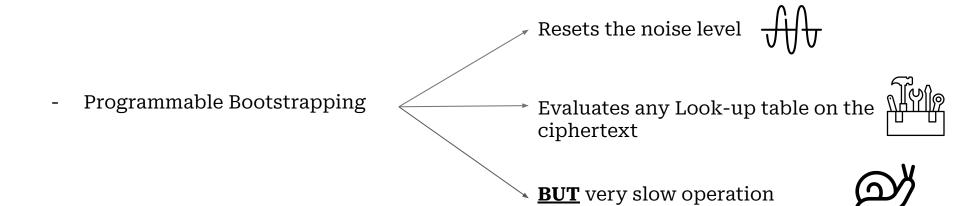
Construction of ciphertexts:



TFHE: available operations

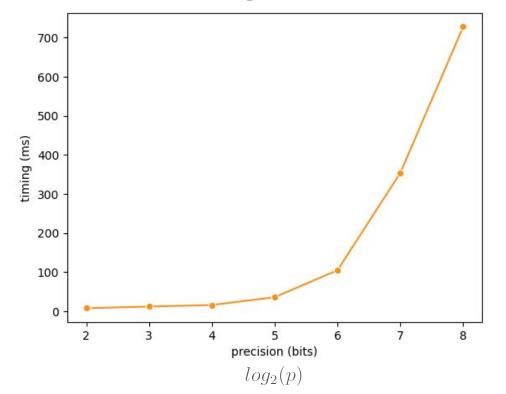
- Sum on \mathbb{T}_p
- External product on \mathbb{T}_p by a clear constant



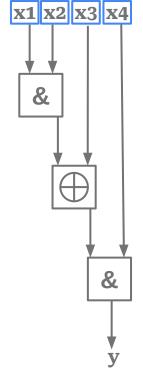


TFHE only allows small precisions

Timing of a PBS

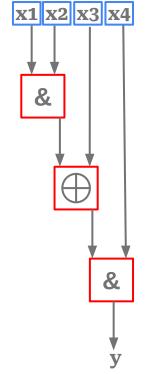


Natural approach of Boolean function evaluation: gate bootstrapping



- See Boolean functions as Boolean circuits
- Each bit is a ciphertext
- Each gate is a 2-input Look-up table

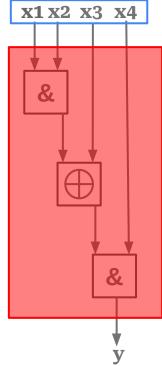
Natural approach of Boolean function evaluation: gate bootstrapping



- See Boolean functions as Boolean circuits
- Each bit is a ciphertext
- Each gate is a 2-input Look-up table

Problem: each gate costs 1 Programmable Bootstrapping

Other approach of Boolean function evaluation: large LUT



- See Boolean Function as a LUT on $\ell\,$ bits
- Pack all bits in a single ciphertext of order p^ℓ
- Evaluate the function with a very large PBS.

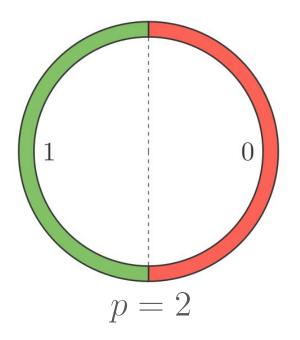
Problem: the PBS becomes very slow and not practical.

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How to represents bit in TFHE?

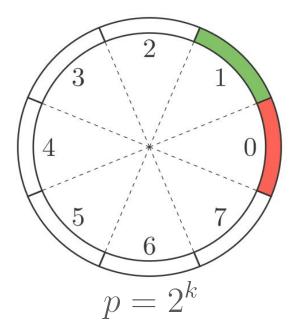
Obvious (and terrible) solution:



- No 2-input LUT
- Only XOR can be evaluated

How to represents bit in TFHE?

Slightly better solution of the literature:



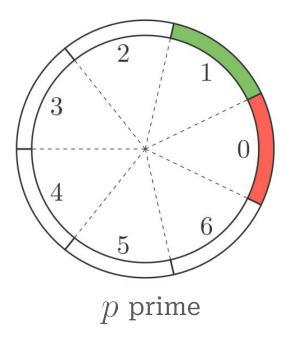
Negacyclity problem causes one of the following:

-> Not every functions are evaluable

-> Sum is no longer homomorphic

How to represent bit in TFHE?

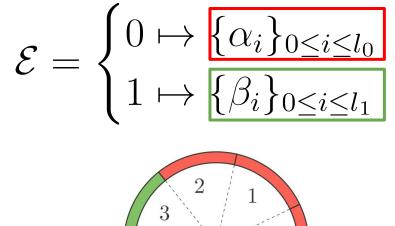
Our approach:

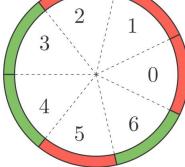


Negacyclicity problem vanishes!

Enter the p-encodings

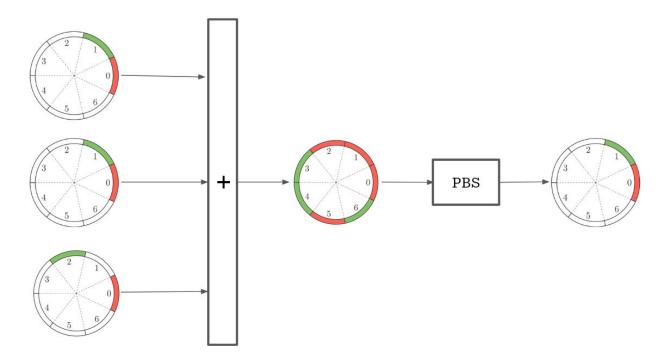
A p-encoding is a function $\mathcal{E}:\mathbb{B}\mapsto 2^{\mathbb{Z}_p}$





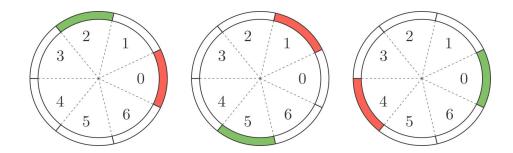
New function evaluation algorithm

Function evaluation is made by a sum followed by a PBS

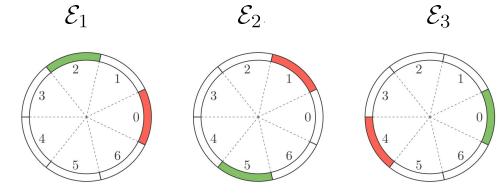


Let f be a Boolean function: $f:\mathbb{B}^3\mapsto\mathbb{B}$

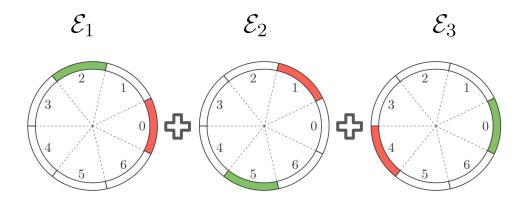
Let $\mathcal{E}_1, \mathcal{E}_2, \mathcal{E}_3$ be three p-encodings.

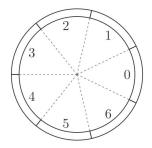


b1	b2	b3	f(b1, b2, b3)
0	0	0	
1	0	0	
0	1	0	

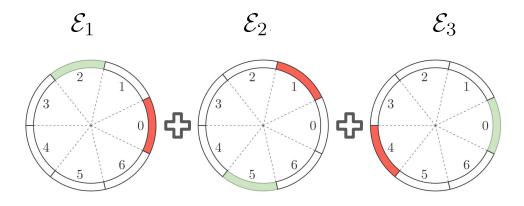


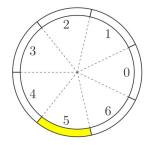
b1	b2	b3	f(b1, b2, b3)
0	0	0	
1	0	0	
0	1	0	



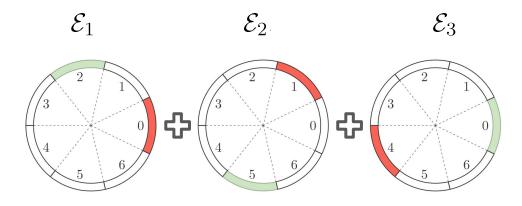


b1	b2	b3	f(b1, b2, b3)
0	0	0	
1	0	0	
0	1	0	

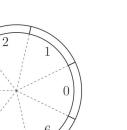




b1	b2	b3	f(b1, b2, b3)
0	0	0	1
1	0	0	
0	1	0	



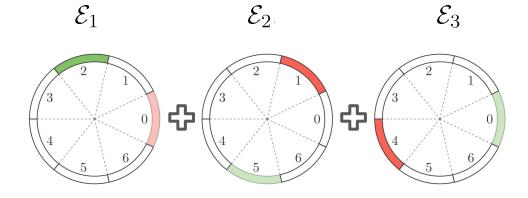




Toy Example

Truth table of f:

b1	b2	b3	f(b1, b2, b3)
0	0	0	1
1	0	0	
0	1	0	

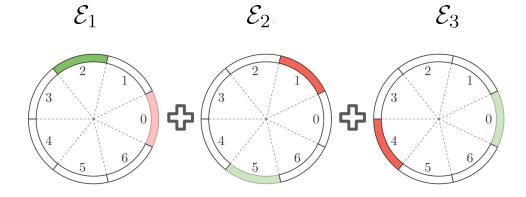


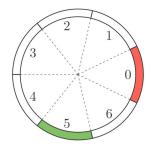


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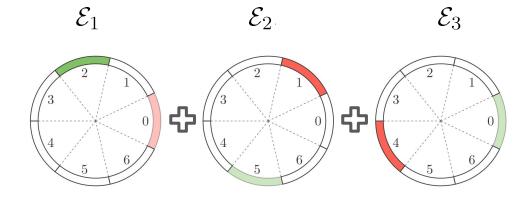




Toy Example

Truth table of f:

b1	b2	b3	f(b1, b2, b3)
0	0	0	1
1	0	0	0
0	1	0	



The encodings are valid if the red and the green parts do not overlap after all the lines have been treated.



Advantages of the method

- One single bootstrapping to evaluate the whole function
- No need to extend the plaintext space to fit more inputs
- Scales much better that the conventional gate approach

Boolean function on p-encodings

Problem : for a given function, how to find a valid set of p-encodings (and the best p) ?

=> Exhaustive search. But some restrictions have to be made in the search space.

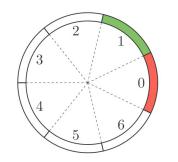
=> We restrict the search to p-encodings with form:

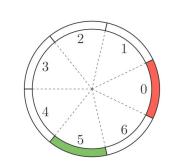
$$\mathcal{E}_{i} = \begin{cases} 0 \mapsto \{0\} \\ 1 \mapsto \{d_{i}\} \end{cases} \quad \text{with:} \mathcal{E}_{1} = \begin{cases} 0 \mapsto \{0\} \\ 1 \mapsto \{1\} \end{cases}$$

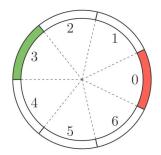
with no loss of generality

Truth table of f:

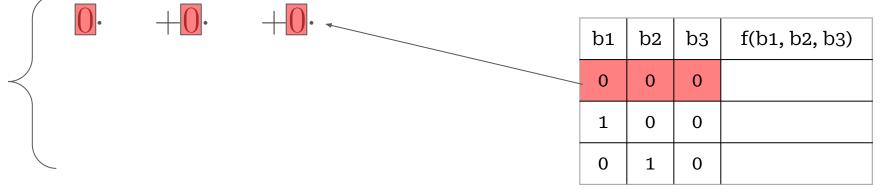
b1	b2	b3	f(b1, b2, b3)
0	0	0	
1	0	0	
0	1	0	

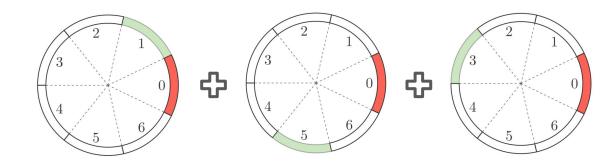


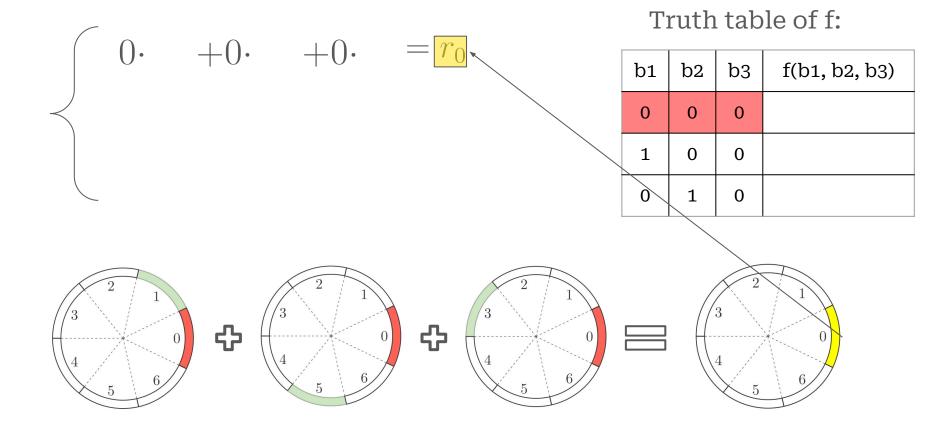


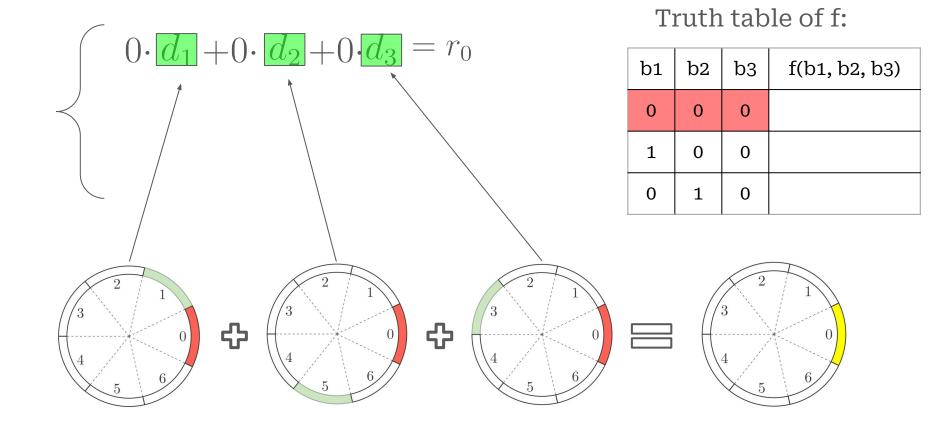


Truth table of f:

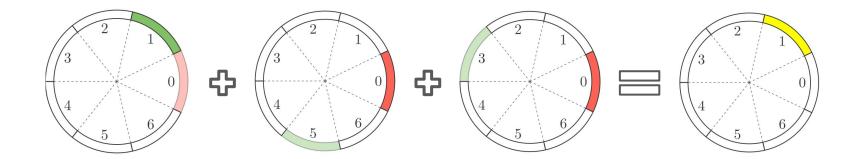




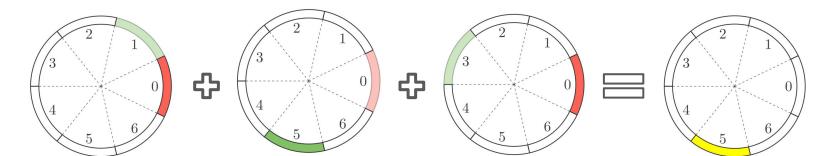




 $\left(\begin{array}{cccc} 0 \cdot d_1 + 0 \cdot d_2 + 0 \cdot d_3 = r_0 \\ 1 \cdot d_1 + 0 \cdot d_2 + 0 \cdot d_3 = r_1 \end{array} \right) \left(\begin{array}{cccc} b_1 & b_2 & b_3 & f(b_1, b_2, b_3) \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{array} \right)$



. . . .

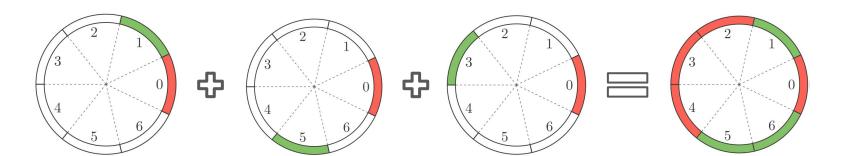


 $0 \cdot d_1 + 0 \cdot d_2 + 0 \cdot d_3 = r_0 \mod p$ $1 \cdot d_1 + 0 \cdot d_2 + 0 \cdot d_3 = r_1 \mod p$ $0 \cdot d_1 + 1 \cdot d_2 + 1 \cdot d_3 = \underbrace{r_2 \mod p}_{\text{mod } p}$

Truth table of f:

b1	b2	b3	f(b1, b2, b3)
0	0	0	0
1	0	0	1
0	1	0	1

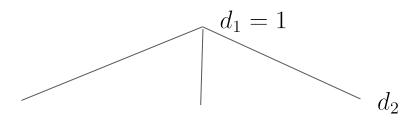
. . .

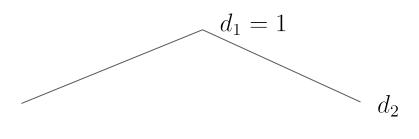


By writing all the inequations

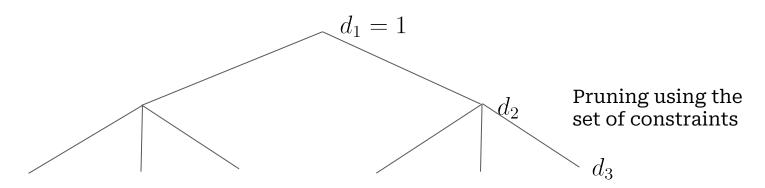


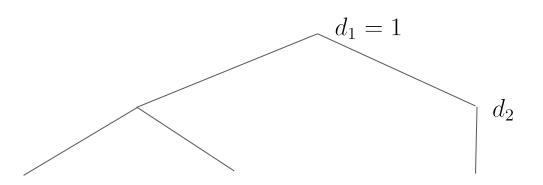
$$\begin{cases} c_1^{(1)} \cdot d_1 + \dots + c_l^{(1)} \cdot d_l \neq 0 \mod p \\ c_1^{(2)} \cdot d_1 + \dots + c_l^{(2)} \cdot d_l \neq 0 \mod p & \text{with } c_i^{(j)} \in \{0, \pm 1\} \\ \vdots & \end{cases}$$





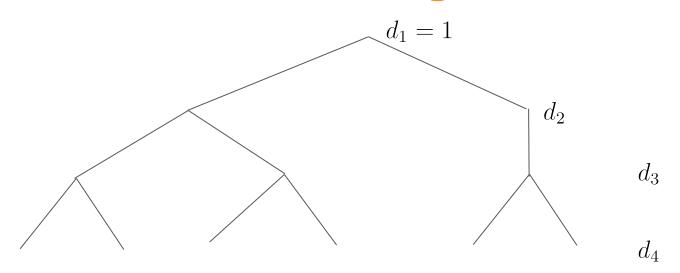
Pruning using the set of constraints







Pruning using the set of constraints



... until we find a path of length ℓ

- The search algorithm finds an **optimal** solution for a given p.
- To identify relevant values for $p\,$ we developed an ${\bf heuristic}$ method that finds an upper bound on the optimal $p\,$

Experimental results

Primitive	Section or Other work	Performances
	Gate Bootstrapping	174 s
One full run of SIMON	$[BSS^+23] \dagger$ Our work (Section 7.1)	128 s 10 s
One warm-up phase of Trivium (*)	Gate Bootstrapping [BOS23] (estimation on our machine) Our work (Section 7.2)	1498 s 53 s 32.8 s
One Full Keccak permutation (*)	Gate Bootstrapping Our work (Section 7.3)	30.7 min 8.8 min
One Ascon hashing (*)	Gate Bootstrapping Our work (Section 7.4)	200s 92 s

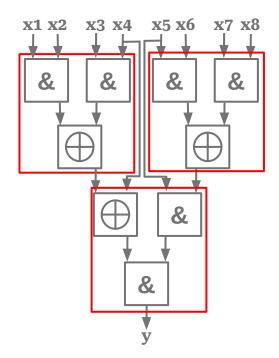
Use-cases includes transciphering, OPRF, ...

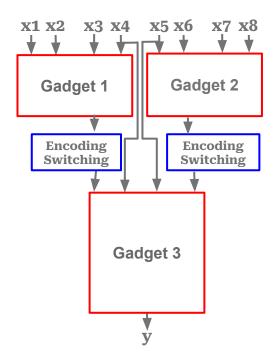
Experimental results

Primitive	Section or Other work	Performances
	Gate Bootstrapping	174 s
One full run of SIMON	$[BSS^+23]$ †	128 s
p = 9	$\bigcirc Our work (Section 7.1)$	10 s
	Gate Bootstrapping	1498 s
One warm-up phase of Trivium (*)	[BOS23] (estimation on our machine)	53 s
p = 9	Our work (Section 7.2)	32.8 s
One Full Keccak permutation (*)	Gate Bootstrapping	30.7 min
p = 3	Our work (Section 7.3)	8.8 min
One A <u>scon hash</u> ing (*)	Gate Bootstrapping	200s
p = 17	Our work (Section 7.4)	92 s

Use-cases includes transciphering, OPRF, ...

Extension to bigger circuits





AES: performances

	[GHS12] †	$18 \min$
One full evaluation of AES-128	[CLT14] †	$5 \min$
$(\epsilon = 2^{-23})$ on one thread	[TCBS23]	270 s
	Our work (Section 7.5)	103 s
One full evaluation of AES-128 ($\epsilon = 2^{-40}$) on one thread	Gate Bootstrapping	234 s
	Our work (Real implementation)	$135 \mathrm{s}$
	Our work (Theoretical timing with two keys)	$105 \mathrm{~s}$



Conclusion



- New Framework to evaluate Boolean functions in TFHE
- One bootstrapping per function, with any number of input. Fixed size.
- Optimal algorithm to find a solution for a given function
- Heuristic to split bigger circuits into evaluable functions
- Adaptation of the bootstrapping to remove the padding bit

Thank you !

https://eprint.iacr.org/2023/1589 For more details

Slides by Nicolas Bon

Mathematical figures built with Manim (https://github.com/ManimCommunity/manim)

Icons from Flaticons