

New Algorithms to Find Lightweight (AND,XOR) Implementations of S-boxes

Marie Bolzer, Sébastien Duval, Marine Minier

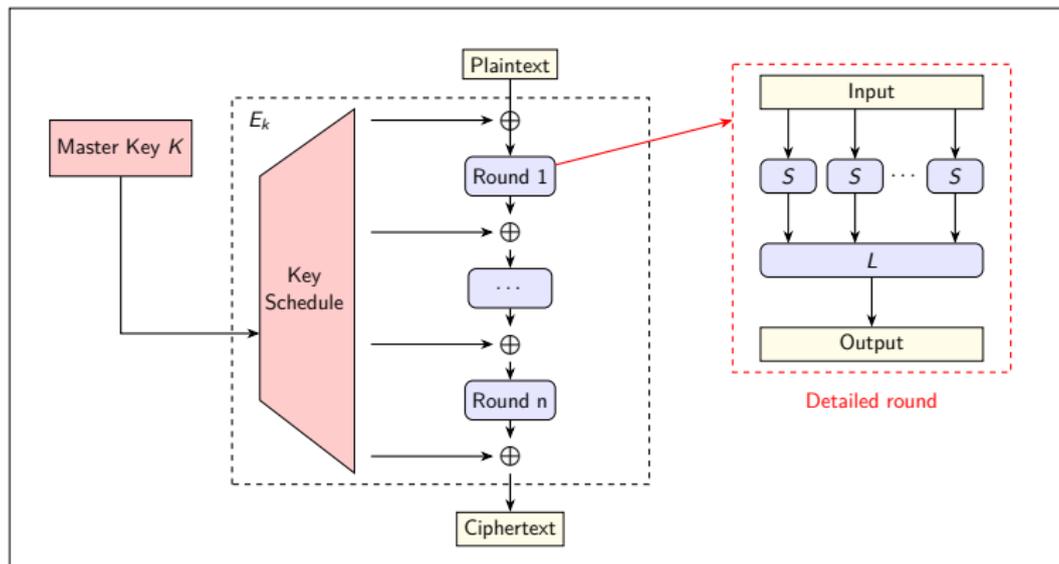
March 20, 2026

Séminaire de cryptographie de l'université de Rennes

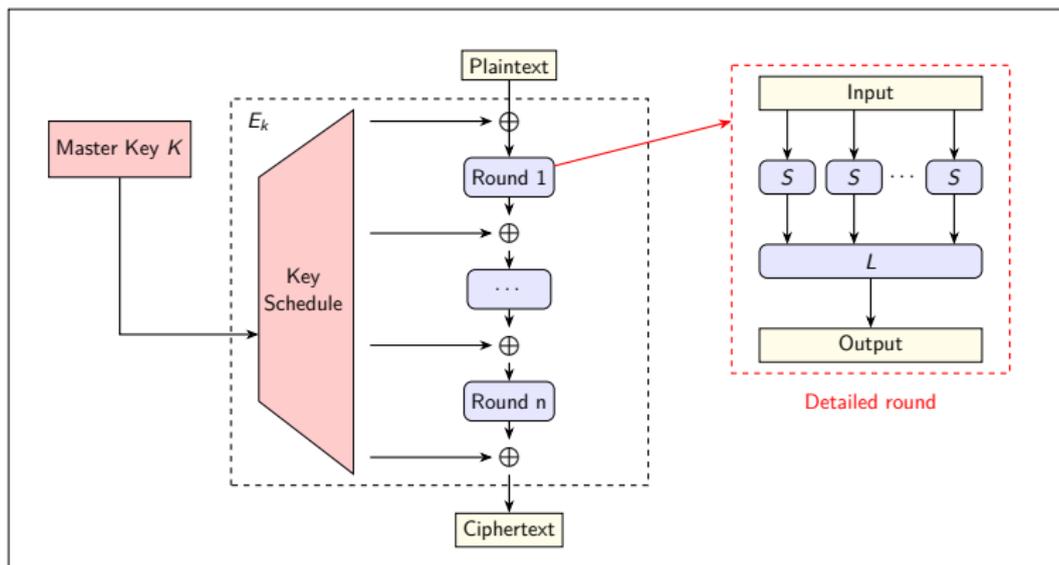


- 1 Context
- 2 Quadratic case
- 3 Higher-Degree cases
- 4 Perspectives and conclusion

Context: Substitution-Permutation Network

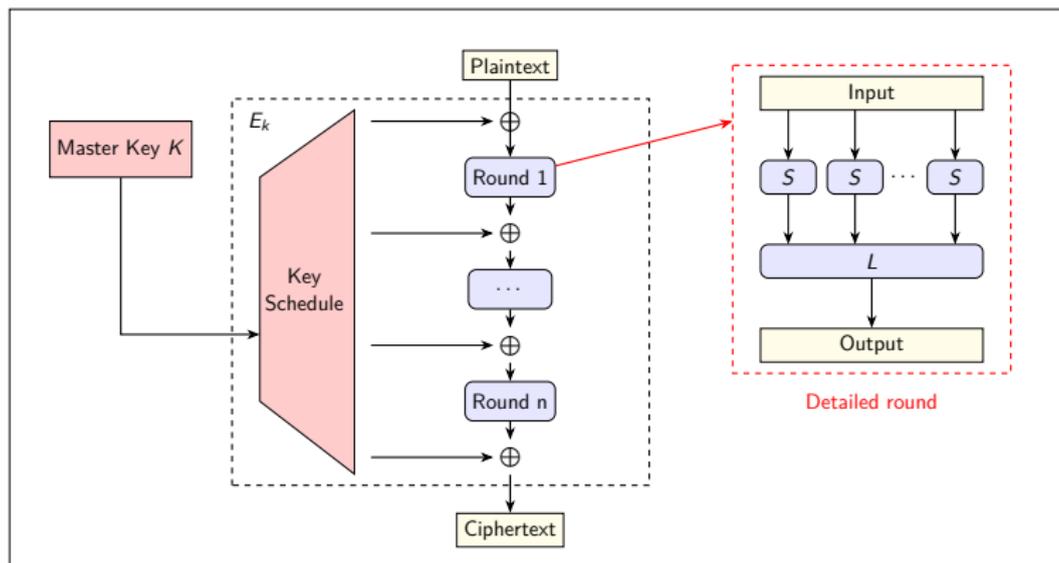


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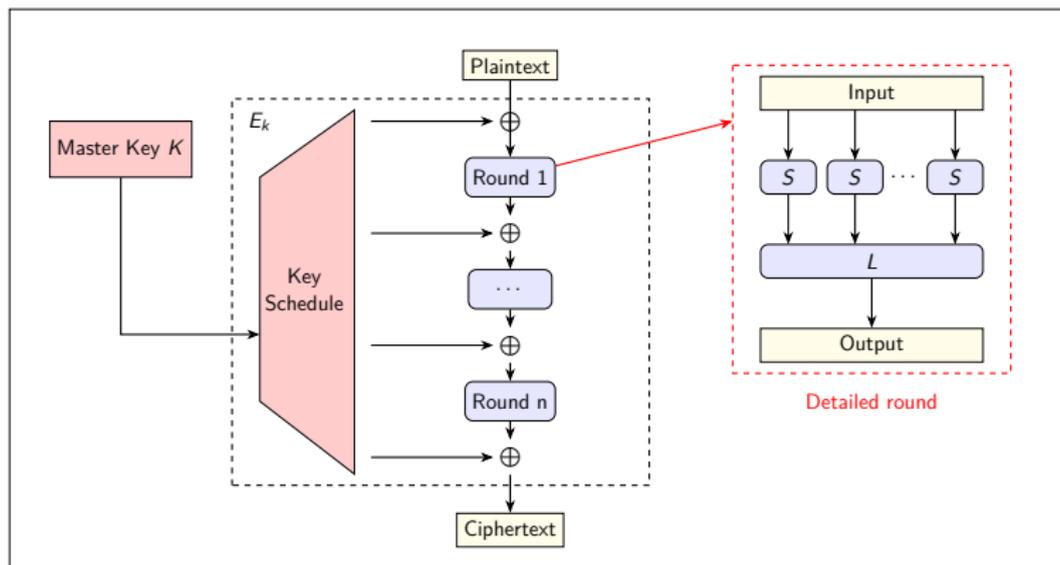
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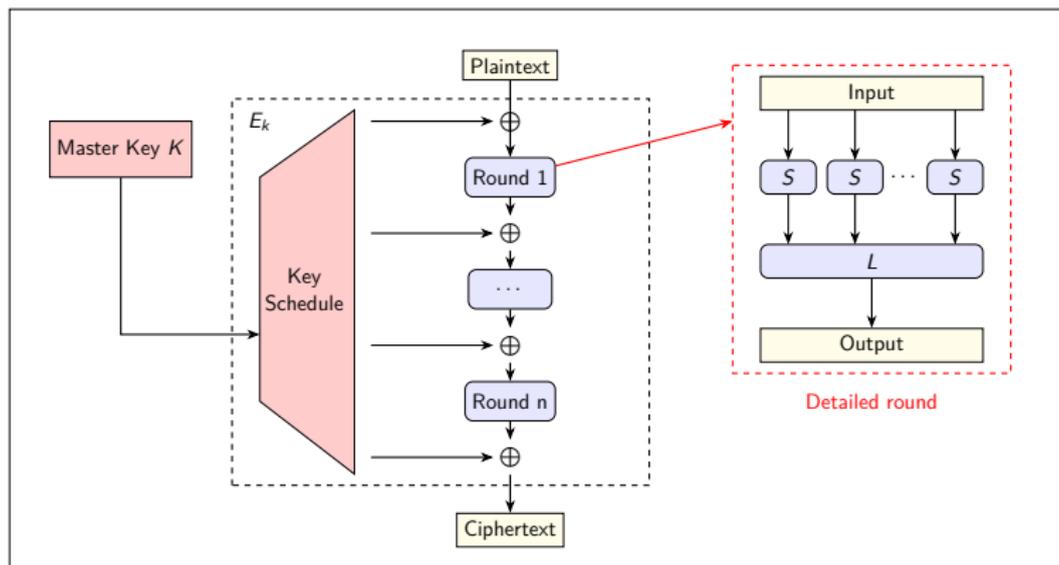
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Cryptographic implementations must be **protected** against physical attacks. A usual countermeasure is **masking**.

Problem: Masking is very expensive for non-linear operations (AND, OR).

→ The S-box is then the costliest part of the cipher.

Context: choice of metrics

We focused on two main cost metrics for our implementation:

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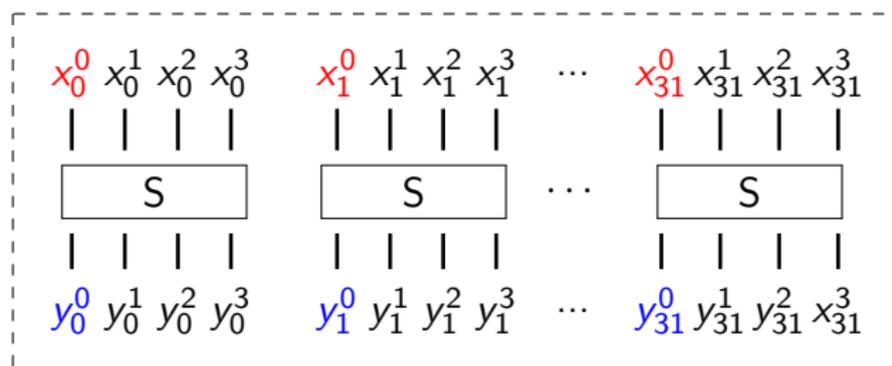
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- **multiplicative depth**, largest number of multiplication operations on any path from input variables to output variables of the circuit
⇒ main factor of the **latency** of the circuit,
- **multiplicative complexity**, total number of multiplications in the implementation.
⇒ main factor of the circuit **size**.

Bitsliced Implementation

Idea : Use only bitwise operations.

Interest : Process the same S-box several times in parallel



$$x_0 = (x_0^0 \ x_1^0 \ x_2^0 \ \dots \ x_{31}^0) \rightarrow \mathbf{S} \rightarrow y_0 = (y_0^0 \ y_1^0 \ y_2^0 \ \dots \ y_{31}^0)$$

S-boxes and Algebraic Normal Form (ANF)

$$S : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^m, \quad S(x_1, \dots, x_n) = (y_1, \dots, y_m),$$

can be described by its **Algebraic Normal Form (ANF)**:

$$y_j = \bigoplus_{I \subseteq \{1, \dots, n\}} a_I \prod_{i \in I} x_i, \quad a_I \in \mathbb{F}_2.$$

Look-Up-Table (LUT):

x	0	1	2	3	4	5	6	7
$S(x)$	0	1	3	4	5	6	7	2

→ ANF is already a bitsliced implementation.

Multivariate representation (ANF):

$$S : \mathbb{F}_2^3 \rightarrow \mathbb{F}_2^3$$

$(x_0, x_1, x_2) \mapsto (y_0, y_1, y_2)$ where

$$y_0 = x_0 \oplus x_1 \oplus x_2 \oplus x_1 x_2$$
$$y_1 = x_1 \oplus x_0 x_1 \oplus x_0 x_2$$
$$y_2 = x_0 x_1 \oplus x_2$$

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Polynomial interpretation:

- y_j can be seen as a polynomial in the ring $\mathbb{F}_2[x_1, \dots, x_n] / \langle x_i^2 + x_i \rangle$.
- Then:

XOR \Leftrightarrow addition in \mathbb{F}_2 , AND \Leftrightarrow multiplication in \mathbb{F}_2 .

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\Rightarrow **Minimising AND gates = minimising polynomial multiplications.**

Formalisation of the problem

Specifications required in the implementation

- Use only bitwise operations (AND, XOR, NOT)
- Minimise multiplicative depth and complexity (AND gates)

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We want to obtain:

$$\begin{aligned}l_0 &= x_0 \oplus x_3 \\q_0 &= x_1 \times l_0 \\l_1 &= x_1 \oplus x_3 \\q_1 &= x_2 \times l_1 \\y_1 &= q_0 \\y_0 &= y_1 \oplus q_1\end{aligned}$$

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2 AND - 3 XOR

Objective: To obtain a tool that automates this process and gives an implementation of a given S-box having the **best multiplicative depth** and **few multiplications**.

Existing tools

Several tools exist to implement small functions. There are 2 major algorithmic approaches:

Depth-First-Search in graphs:

- Defined in 2011 by Ullrich.
- Used in LIGHTER in 2017 and then in PEIGEN in 2019.

SAT-solvers:

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Limitations of these tools: Functions operating on less than 5 bits, sometimes 6 bits for quadratic functions, or to very simple functions. (High computing time)

Our Contributions

Objectives with regards to existing tools:

- Design a *dedicated algorithm* for finding implementations with **optimal multiplicative depth** and **low multiplicative complexity**.
- *Reduce execution time* for more complex functions in order to *extend applicability* to functions on a larger number of bits.

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Achieved results:

- Development of an efficient approach for the *quadratic case* — results published in IEEE TCAS.
- Extension of the method to *higher-degree functions* — accepted to CHES.

Some intuitions

First idea: Try to find a lightweight implementation with pen, paper and coffee.

$$\begin{aligned}y_0 &= x_0x_1 \oplus x_1x_2 \oplus x_1x_3 \oplus x_2x_3 \\y_1 &= x_0x_1 \oplus x_1x_3\end{aligned}$$



$$\begin{aligned}l_0 &= x_0 \oplus x_3 \\q_0 &= x_1 \times l_0 \\l_1 &= x_1 \oplus x_3 \\q_1 &= x_2 \times l_1 \\y_1 &= q_0 \\y_0 &= y_1 \oplus q_1\end{aligned}$$

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Objective: Make this algorithm *feasible* for a computer.

PART 2: DESCRIPTION OF THE ALGORITHM FOR THE QUADRATIC CASE

Notations: Useful polynomial subsets

Degree of a monomial: number of variables that appear in the monomial.

Degree of a polynomial: maximum degree of its monomials.

- \mathcal{L}_n : set of **linear** polynomials (degree ≤ 1)
- \mathcal{Q}_n : set of **quadratic** polynomials (degree ≤ 2)
- $\mathcal{Q}_{1,n}$: non-linear truncations of $l_1 \times l_2$, $l_i \in \mathcal{L}_n$

Example: $(x_0 \oplus x_1) \times (x_1 \oplus x_2) = x_0x_1 \oplus x_0x_2 \oplus x_1 \oplus x_1x_2$
 $\Rightarrow p = x_0x_1 \oplus x_0x_2 \oplus x_1x_2 \in \mathcal{Q}_{1,n}$.

- $\mathcal{Q}_{1,n}^L = \mathcal{Q}_{1,n} \oplus \mathcal{L}_n$
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- $\mathcal{Q}_{2,n} = \mathcal{Q}_{1,n} \oplus \mathcal{Q}_{1,n}$

Cost interpretation: A polynomial in $\mathcal{Q}_{x,n}$ or $\mathcal{Q}_{x,n}^L$ requires exactly x AND gates.

Reformulation of the problem in the quadratic case

General principle:

- Consider non-linear truncation of the polynomials in the ANF. (Y)
If $y = x_0x_1 \oplus x_3 \oplus x_1x_2$, we only look at $x_0x_1 \oplus x_1x_2$.

Formalisation in matrix form

$$Y = \begin{pmatrix} x_1x_2 \\ x_0x_1 \oplus x_0x_2 \\ x_0x_1 \end{pmatrix}$$

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- Store all possible quadratic factorisation schemes.
→ corresponds to the set $\mathcal{Q}_{1,n}$

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$$\mathcal{Q}_{1,3} = \begin{pmatrix} x_0x_1 \\ x_0x_2 \\ x_1x_2 \\ x_0x_1 \oplus x_0x_2 \\ x_0x_1 \oplus x_1x_2 \\ x_0x_2 \oplus x_1x_2 \\ x_0x_1 \oplus x_0x_2 \oplus x_1x_2 \end{pmatrix} Y = \begin{pmatrix} x_1x_2 \\ x_0x_1 \oplus x_0x_2 \\ x_0x_1 \end{pmatrix}$$

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We search for a binary matrix S such that:

$$S\mathcal{Q}_{1,3} = Y$$

Reformulation of the problem in the quadratic case

To obtain the best implementation, we look for a solution S having **as many zero columns as possible**.

The matrix

$$S_1 = \begin{matrix} & q_0 & q_1 & q_2 & q_3 & q_4 & q_5 & q_6 \\ \begin{pmatrix} 0 & \mathbf{1} & 0 & 0 & 0 & \mathbf{1} & 0 \\ 0 & 0 & 0 & \mathbf{1} & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & \mathbf{1} & \mathbf{1} \end{pmatrix} \end{matrix}$$

is a solution of the equation and induces the implementation :

$$y_0 = \mathbf{q_1} \oplus \mathbf{q_5} = x_0x_2 \oplus x_0x_2 \oplus x_1x_2$$

$$y_1 = \mathbf{q_3} = x_0x_1 \oplus x_0x_2$$

$$y_2 = \mathbf{q_5} \oplus \mathbf{q_6} = x_0x_2 \oplus x_1x_2 \oplus x_0x_1 \oplus x_0x_2 \oplus x_1x_2$$

→ Each non-zero column implies that the operation is used in the implementation and then costs 1 AND.

The algorithm chosen for the non-linear part:

Objective: Express each output bit as a XOR-sum of elements of $\mathcal{Q}_{1,n}$ by constructing iteratively a set **Op_selec** containing polynomials in $\mathcal{Q}_{1,n}$.
⇒ At the end, this set will contain all the polynomials to implement the S-box.

Adding polynomials to Op_selec:

Principle function

Inputs: An output bit y , Op_selec

Outputs: Op_selec \cup

$\{v_0, \dots, v_b \in \mathcal{Q}_{1,n} \mid \exists u_i \in \text{Op_selec}, \bigoplus u_i \oplus \bigoplus v_i = y\}$, with b **minimal**.

Illustration for the non-linear part

Consider: $y_0 = x_0x_2$

$$y_1 = x_0x_1 \oplus x_2x_3$$

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Op_selec

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First step: y_0

- We cannot get y_0 adding 0 polynomials.
 - $y_0 = x_0 \times x_2$ then $y_0 \in \mathcal{Q}_{1,4}$
- We add $q_1 = x_0x_2$ in Op_selec .



Op_selec

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Op_selec

Second step: y_1

- We cannot get y_1 adding 0 polynomials.
- We cannot get y_1 adding 1 polynomial.
- We can get y_1 by adding only 2 polynomials of $\mathcal{Q}_{1,4}$. We get several possibilities that all need to be tested:

Illustration for the non-linear part

Consider: $y_0 = x_0x_2$

$y_1 = x_0x_1 \oplus x_2x_3$

$y_2 = x_0x_3 \oplus x_1x_2 \oplus x_2x_3$

Second step: y_1

• $y_1 = (x_0 \times x_1) \oplus (x_2 \times x_3)$

→ We add $q_2 = x_0x_1$ and $q_3 = x_2x_3$ in Op_selec.

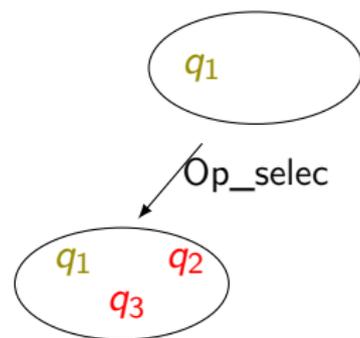


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→ We add $q_2 = x_0x_1$ and $q_3 = x_2x_3$ in Op_selec.

• $y_1 = x_1 \times (x_0 \oplus x_2) \oplus x_2 \times (x_1 \oplus x_3)$

→ We add $q_4 = x_0x_1 \oplus x_1x_2$ and $q_5 = x_1x_2 \oplus x_2x_3$ in Op_selec.

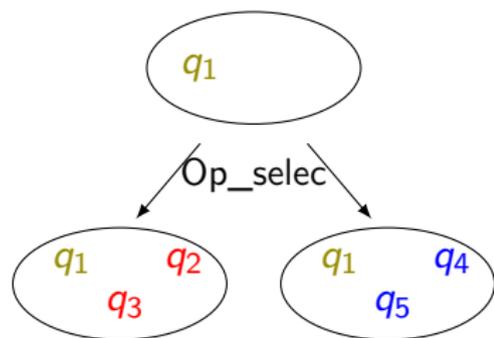


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→ We add $q_2 = x_0x_1$ and $q_3 = x_2x_3$ in Op_selec.

• $y_1 = x_1 \times (x_0 \oplus x_2) \oplus x_2 \times (x_1 \oplus x_3)$

→ We add $q_4 = x_0x_1 \oplus x_1x_2$ and $q_5 = x_1x_2 \oplus x_2x_3$ in Op_selec.

Third step: y_2

• We add $q_6 = x_0x_3$ and $q_7 = x_1x_2$.

Then $y_2 = q_3 \oplus q_6 \oplus q_7$

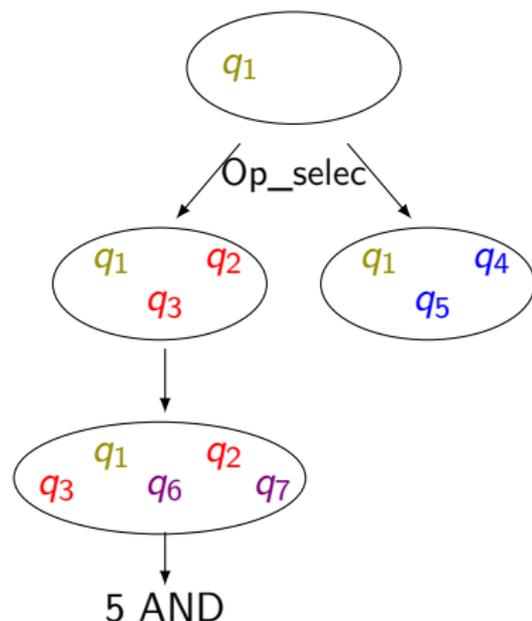


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Consider: $y_0 = x_0x_2$

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→ We add $q_2 = x_0x_1$ and $q_3 = x_2x_3$ in Op_selec.

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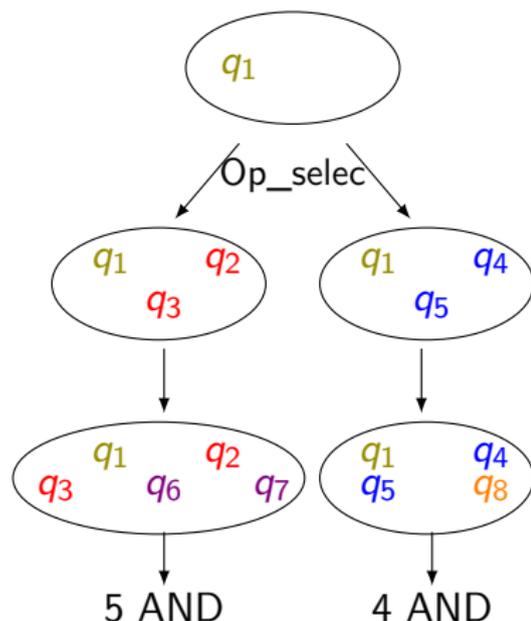
→ We add $q_4 = x_0x_1 \oplus x_1x_2$ and $q_5 = x_1x_2 \oplus x_2x_3$ in Op_selec.

Third step: y_2

• We add $q_6 = x_0x_3$ and $q_7 = x_1x_2$.

Then $y_2 = q_3 \oplus q_6 \oplus q_7$

• We add $q_8 = x_0x_3$. Then $y_2 = q_5 \oplus q_8$



Order of treatment of the output bits

Suppose we use the order $(y_3 - y_0 - y_2 - y_1)$ and obtain the resulting implementation with 7 non-linear operations q_i :

$$y_3 = q_0 \oplus q_4 \oplus q_5$$

$$y_0 = q_5 \oplus q_6$$

$$y_2 = q_2 \oplus q_3$$

$$y_1 = q_0 \oplus q_1$$

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If y_0 is also equal to $q_1 \oplus q_3 \oplus q_4$,

what happens if we change the order ?

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$$y_2 = q_2 \oplus q_3$$

$$y_1 = q_0 \oplus q_1$$

If y_0 is also equal to $q_1 \oplus q_3 \oplus q_4$,

what happens if we change the order ? If we use $(y_1 - y_2 - y_0 - y_3)$, we obtain:

$$y_1 = q_0 \oplus q_1$$

$$y_2 = q_2 \oplus q_3$$

$$y_0 = q_1 \oplus q_3 \oplus q_4$$

$$y_3 = q_0 \oplus q_4 \oplus q_5$$

Results for the quadratic case

S-box	Implementation(AND - XOR)				Timings (in seconds)		
	[BDD ⁺ 20]	[ZH23]	[FWZ ⁺ 24]	Ours	[ZH23]	[FWZ ⁺ 24]	Ours
χ_5	-	5 -	5 - 5	5 - 10	1.2	19.81	0.05
ASCONE	-	5 -	5 -	5 - 15	1.4	26.48	0.1
SYCON	-	-	5 - 17	5 - 15		26.72	0.1
FIDES	7 - 29	7 -	7 - 27	7 - 20	3.6	140.31	0.2
X^3	7 - 29	-	-	7 - 19			< 1
X^5	7 - 26	-	-	7 - 21			< 1

Table: Results for 5-bit S-boxes.

Results for the quadratic case

Size (on bits)	Implementation (AND - XOR)	
	[BDD ⁺ 20]	Ours
5	7 - 29	7 - 19
6	9 - 43	8 - 37
7	15 - 79	11 - 45
8	-	14 - 67
9	-	19 - 94

Table: Results for the power function X^3

Our tool is able to handle any quadratic functions up to 9 bits.

PART 3: DESCRIPTION OF THE ALGORITHM FOR HIGHER-DEGREE FUNCTIONS

What about higher degrees ?

An example for degree 3:

$$\mathcal{C}_{2,n} \text{ v1} = \{l_1 \times l_2 \times l_3\}, l_i \in \mathcal{L}_n$$

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4	35	141		
5	155	1241		
6	651	10 417		

Table: Cardinalities of polynomial sets for degree 3

What about higher degrees ?

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→ A more relevant approach would be to use the different formulas given above to decompose a given function.

From the quadratic case to higher-degree functions

$$\mathcal{C}_{2,n} \text{ v1} = \{l_1 \times l_2 \times l_3\}, \quad l_i \in \mathcal{L}_n$$

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Previous approach:

Start from the **inputs** and rely on *multiplications tables*.

New approach: Start from the **outputs** and use *division tables*.

High-level description of the algorithm

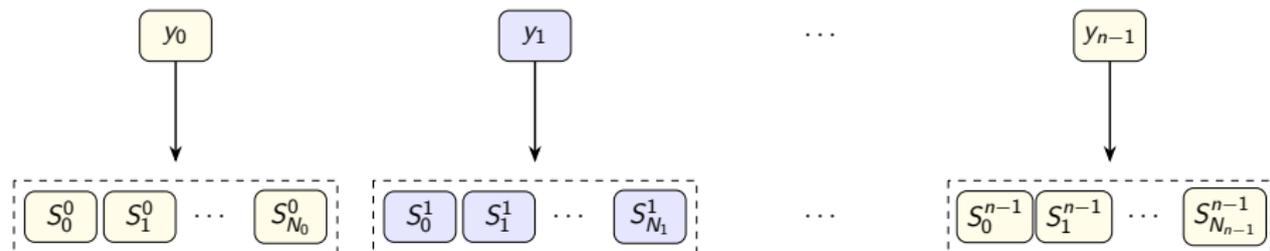
2 different phases

High-level description of the algorithm

2 different phases

Phase 1: AND depth $\geq 2 \rightarrow 1$ Reduce high-degree functions into compositions of quadratic functions based on recognising factorisation patterns (*using formulas*).

Objective: Find efficient implementations (*decomposition sets*) expressed in $\{\mathcal{L}_n \cup \mathcal{Q}_{1,n}\}$ of each output bit y_i .

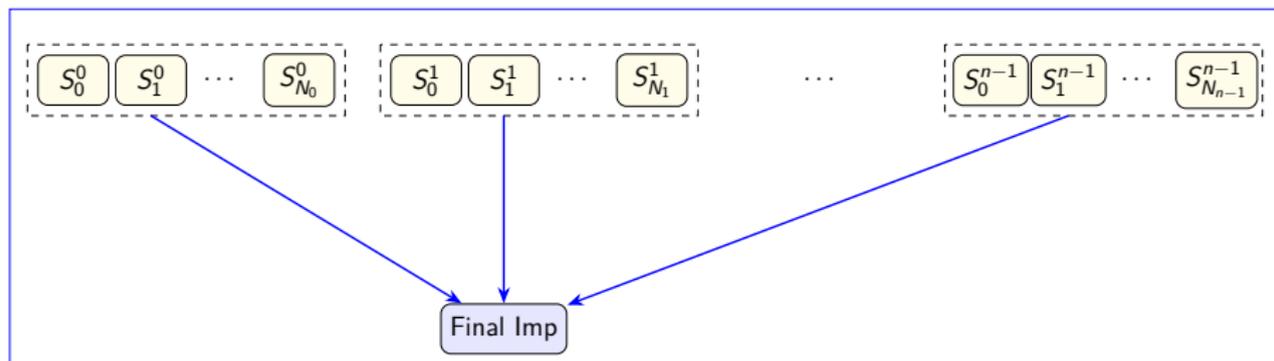


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High-level description of the algorithm

Phase 2: AND depth $1 \rightarrow 0$ Computational phase based on algebraic calculations, different from the previous quadratic algorithm.

Objective: Find a combination of these parallel implementations to minimise cost.



First phase: Construction of the decomposition sets

Decomposition: A decomposition of a polynomial is a formal expression representing the polynomial using multiplications, additions and polynomials of degree at most 2.

Cost of a decomposition = Number of multiplications in the expression.

Example:

$y = (q_1 \times l_1) \oplus l_2$ is a decomposition of a polynomial of degree 3 and its cost is 1.

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Total cost of an implementation:

$$\underbrace{\text{Cost of the decompositions}}_{\# \text{ AND in formal expression}} + \underbrace{\text{Cost of the quadratic polynomials used}}_{\text{from Phase 2}}$$

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- We explore decomposition starting from the lowest cost.
- If valid solutions are found for a given cost, we stop exploring the next, more expensive ones.
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These strategies ensure that only **locally minimal** decompositions are kept.

First phase: Instantiation of the decomposition sets using divisions between multivariate polynomials

Objective: Given $p_1, p_2 \in \mathbb{F}_2[x_1, \dots, x_n]$ with $\deg(p_2) < \deg(p_1)$, division means finding a couple $q, r \in \mathbb{F}_2[x_1, \dots, x_n]$ such that:

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To define this division, we need to:

- define how to divide two monomials,
- set a monomial order to process the monomials.

→ we will successively divide the monomials of the divisor by the monomials of the dividend.

The **difficult part** here is to define a fitting operator “ $<$ ” over $\mathbb{F}_2[x_1, \dots, x_n]$.

First phase: Monomial Orders

Lexicographic order: Example for $x_0 > x_1 > x_2 > x_3 > 1$:

$$x_0x_1x_2x_3 > x_0x_1x_2 > x_0x_1x_3 > x_0x_2x_3 > x_1x_2x_3 > x_0x_1 > \cdots > x_3 > 1$$

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Consider $p = x_0x_1x_2 \oplus x_0x_2 \oplus x_0x_3 \oplus x_1x_2x_3$ and $d = x_0 \oplus x_1x_2$.

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$\oplus \quad x_0x_1x_2 \oplus x_0$	$x_0 \oplus x_3$
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$$\begin{array}{r|l} x_0x_1x_2 \oplus x_1x_2x_3 \oplus x_0x_2 \oplus x_0x_3 & x_1x_2 \oplus x_0 \\ \hline & x_0 \oplus x_3 \\ \oplus \quad x_0x_1x_2 \oplus x_0 & \\ x_1x_2x_3 \oplus x_0x_2 \oplus x_0x_3 \oplus x_0 & \\ \oplus \quad x_1x_2x_3 \oplus x_0x_3 & \\ & x_0x_2 \oplus x_0 \end{array}$$

Motivation for adaptation:

- In classical lexicographic order, monomials are sorted mainly by degree.
- For effective division, we prefer monomials containing a specific variable (e.g. x_0) to appear first.

First phase: Monomial Orders

Adapted order: We modify the variable hierarchy so that one variable leads the order: $x_i > 1 > \text{remaining variables (in index order)}$.

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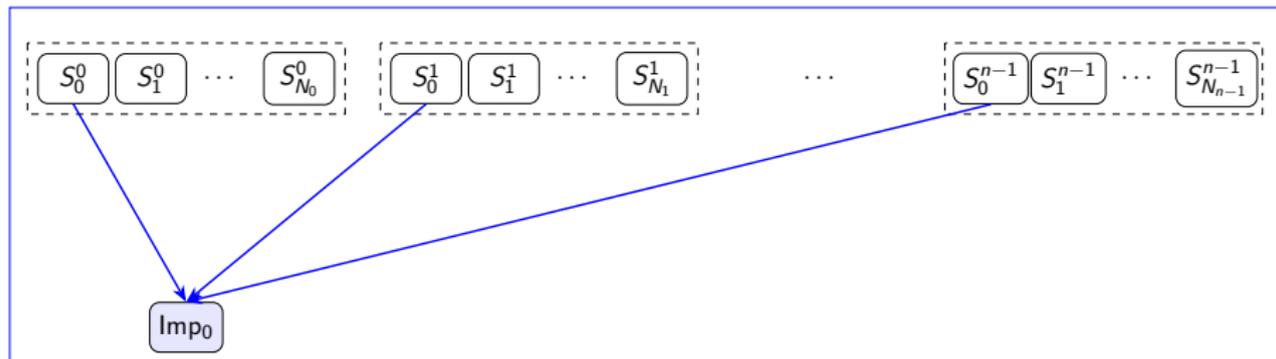
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consider now the order $x_0 > 1 > x_1 > x_2 > x_3$

$x_0x_1x_2 \oplus x_0x_2 \oplus x_0x_3 \oplus x_1x_2x_3$	$x_0 \oplus x_1x_2$
$\oplus x_0x_1x_2 \oplus x_1x_2$	$x_1x_2 \oplus x_2 \oplus x_3$
$\oplus x_0x_2 \oplus x_1x_2$	
$\oplus \underline{x_0x_3 \oplus x_1x_2x_3}$	
0	

Second phase: Aggregation of the decomposition sets

Objective: Find a combination of decomposition sets to obtain a good final implementation.

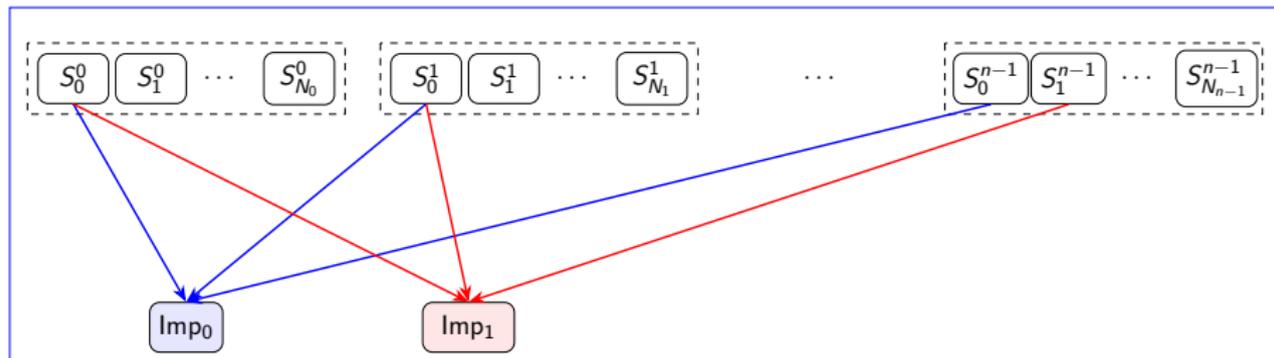


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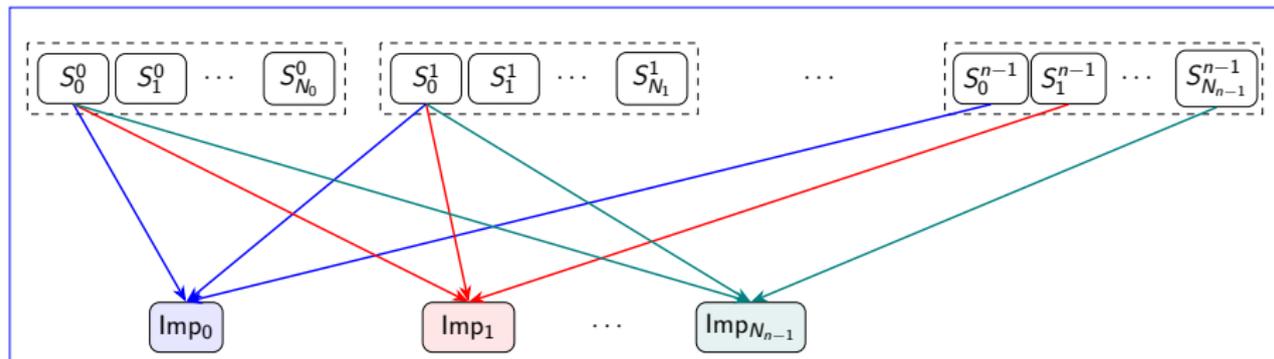


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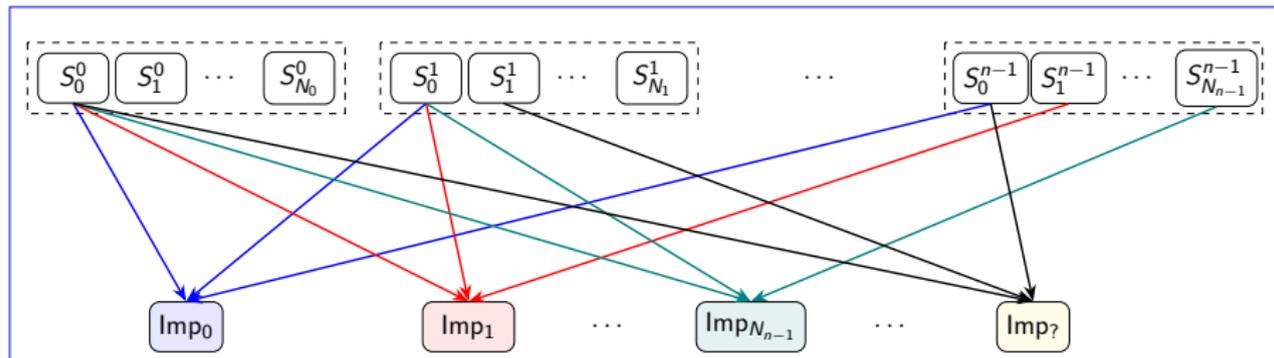


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→ Find a criterion to assess whether a given family of polynomials can be efficiently implemented together or not.

Second phase: Rank computation

$$S^0 = \{x_0x_1 \oplus x_0x_3, x_2x_3\} \quad S^1 = \{x_0x_2 \oplus x_2x_3, x_0x_3\} \quad S^2 = \{x_0x_1, x_2x_3\}$$

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$$\text{Imp} = \bigcup_{i=0}^2 S^i = \{x_0x_1 \oplus x_0x_3, x_2x_3, x_0x_2 \oplus x_2x_3, x_0x_3, x_0x_1, x_2x_3\}$$

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$$M_{\text{imp}} = \begin{pmatrix} & x_0x_1 & x_0x_2 & x_0x_3 & x_1x_2 & x_1x_3 & x_2x_3 \\ \begin{matrix} 1 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \end{matrix} & & \begin{matrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{matrix} & \begin{matrix} 1 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{matrix} & \begin{matrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{matrix} & \begin{matrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{matrix} & \begin{matrix} 0 \\ 1 \\ 1 \\ 0 \\ 0 \\ 1 \end{matrix} \end{pmatrix}$$

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Rank of M = Required number of AND gates to implement all the quadratic polynomials in the set Imp.

Summary of the algorithm

Total cost of an implementation:

$$\underbrace{\text{Cost of the decomposition}}_{\# \text{ AND in formal expression}} + \underbrace{\text{Cost of the quadratic polynomials used}}_{\text{from rank computation}}$$

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- Up to **degree 5** on **6-bit** functions.
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Remarks:

- The algorithm performs well for decompositions from **depth 2** \rightarrow **depth 1** (e.g., degree 4 or 3 to degree 2).
- Increasing the depth makes decompositions more complex and requires more divisions.
- We achieved **depth 3** with degree 5 on 6-bit functions using specific precomputations for the **depth 3** \rightarrow **depth 2** step.

Precomputation Strategy for Degree-5 Polynomials

Phase 1 is time-consuming but uses little memory.

Idea: Use precomputation to optimise the time-memory trade-off

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- Use truncated division tables: $y = q \times c_q \oplus c_r$, with degree constraints $\deg(c_q), \deg(c_r) \leq 3$.
- Only higher-degree terms (degrees 4-5) matter to perform the division lower-degree terms can be absorbed into remainder.
- This reduces the number of polynomials to $2^{\binom{6}{5} + \binom{6}{4}} = 2^{6+15} = 2^{21}$, making precomputation feasible.

Precomputation Strategy for Degree-5 Polynomials

- Decompose output y into high- and low-degree parts:

$$y = y_{\geq 4} \oplus y_{\leq 3}$$

- Using the precomputed table:

$$y_{\geq 4} = (c_1 \times q_1) \oplus c_2 \quad \Rightarrow \quad y = (c_1 \times q_1) \oplus c_3, \quad c_3 = y_{\leq 3} \oplus c_2$$

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What about degree-5 polynomials over 7 bits ?

The number of polynomials containing only monomials of degree 4 and 5 would be $2^{\binom{7}{5} + \binom{7}{4}} = 2^{21+35} = 2^{56}$, making this approach unfeasible for higher sizes.

Results for 4-bit functions

S-box	Security properties				Implementation properties	
	deg	δ	\mathcal{L}	Bij	Our D/AND/XOR	Previous D/AND/XOR
Present [BKL ⁺ 07]	3	4	8	yes	2/4/18	2/4/14 [FWZ ⁺ 24]
Rectangle [ZBL ⁺ 15]	3	4	8	yes	2/4/18	2/4/11 [FWZ ⁺ 24]
Gift [BPP ⁺ 17]	3	6	8	yes	2/5/18	2/5/16 [FWZ ⁺ 24]
Prince [BEK ⁺ 20]	3	4	8	yes	2/6/23	2/6/24 [BDD ⁺ 20]
Prost [KLL ⁺ 14]	3	4	8	yes	2/4/8	2/4/15 [FWZ ⁺ 24]
Piccolo [SIH ⁺ 11]	3	4	8	yes	2/4/17	2/4/4 [FWZ ⁺ 24]
Elephant [DM19]	3	4	8	yes	2/4/17	2/4/14 [FWZ ⁺ 24]
sc2000-4 [SYY ⁺ 02]	3	4	8	yes	2/6/19	4/5/23 [FWZ ⁺ 24]
Twine [SMMK11]	3	4	8	yes	2/6/22	2/6/21 [FWZ ⁺ 24]
Minalpher [STA ⁺ 14]	3	4	8	yes	2/6/25	2/6/21 [FWZ ⁺ 24]
X^{-1} in $GF(2^4)$	3	4	8	yes	2/6/12	3/5/11 [BP10] 2/7/10 [BP12]

Results for 5-bit functions

S-box	Security properties				Implementation properties	
	deg	δ	\mathcal{L}	Bij	Our D/AND/XOR	Previous D/AND/XOR
$(X^3)^{-1}$	3	2	8	yes	2/10/37	2/12/64 [BDD ⁺ 20]
$(X^5)^{-1}$	3	2	8	yes	2/10/37	2/10/65 [BDD ⁺ 20]
Fides Inv [BBK ⁺ 13]	3	2	8	yes	2/10/40	2/10/62 [BDD ⁺ 20]
Keccak Inv [BDPA09]	3	8	16	yes	2/9/27	2/9/52 [BDD ⁺ 20]
sc2000-5 [SYY ⁺ 02]	3	2	8	yes	2/10/44	7/10/- [FWZ ⁺ 24]
X^{-1}	4	2	12	yes	2/12/51	2/25/79 (ANF)
DB_1 [DB19]	4	4	12	yes	2/9/36	2/23/56 (ANF)
DB_17 [DB19]	4	2	12	yes	2/12/50	2/25/66 (ANF)

Results for 6-bit functions

S-box	Security properties				Implementation properties	
	deg	δ	\mathcal{L}	Bij	Our D/AND/XOR	Previous D/AND/XOR
Q_2256 Inv [DB19]	3	4	16	yes	2/13 /55	2/31/64 [BDD ⁺ 20]
Q_2257 Inv [DB19]	3	4	16	yes	2/13 /46	2/29/94 (ANF)
Q_2258 Inv [DB19]	3	4	16	yes	2/11 /42	2/22/58 [BDD ⁺ 20]
Q_2259 Inv [DB19]	3	4	16	yes	2/13 /51	2/34/115 (ANF)
Q_2260 Inv [DB19]	3	4	16	yes	2/12 /53	2/33/111 (ANF)
Q_2261 Inv [DB19]	3	4	16	yes	2/11 /47	2/29/87 (ANF)
Q_2262 Inv [DB19]	3	4	16	yes	2/13 /57	2/32/102 (ANF)
Q_2263 Inv [DB19]	3	4	16	yes	2/16 /71	2/26/- [BDD ⁺ 20]
Dillon [BDMW10]	4	2	16	yes	2/17 /75	3/12/- [PUB16]
X^{15}	4	8	16	no	2/18 /86	2/50/177 (ANF)
X^{31} (Affine eq. of X^{-1})	5	4	16	yes	3/29/109	3/21 /63 X^{-1} TF
sc2000-6 [SYY ⁺ 02]	5	4	16	yes	3/28 /122	7/19/- [FWZ ⁺ 24]
Speedy [LMMR21]	5	8	28	yes	3/20 /51	5/12/- [FWZ ⁺ 24]

Perspectives and conclusion

Our work: S-box implementations via pattern recognition and polynomial factorisation

Achievements

- Depth 1 (degree 2 up to 9 bits)
- Depth 2 (degree 3 and 4 up to 7 bits)
- Depth 3 (degree 5 up to 6 bits)

→ *New available S-boxes for designers.*

Next steps

- Optimise algorithm and implementation
- Extend to higher depths
- Apply beyond binary case

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Thank you for your attention !

Monomial Division over $\mathbb{F}_2[x_1, \dots, x_n]$

Representation: A multivariate monomial in n variables is encoded by an integer on n bits.

Example: $6_{10} = 0110_2 \Rightarrow m = x_2x_1$

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Division rule:

- Let m_1, m_2 be represented by integers a_1, a_2 with $\text{deg}(m_1) > \text{deg}(m_2)$.
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Example:

$$m_1 = x_1x_2x_3$$

$$a_1 = 1110$$

$$m_2 = x_1x_3$$

$$a_2 = 1010$$

- $a_1 | a_2 = 1110 = a_1 \Rightarrow m_1$ divisible by m_2
- $a_1 \oplus a_2 = 0100 \Rightarrow x_1x_2x_3 = x_1x_3 \times x_2$

Main limitations:

- **Multiplicative depth:**

- ▶ The first phase explores a tree of decompositions whose size grows exponentially with the depth.
- ▶ A larger depth also produces more decompositions, increasing the search space and the number of rank computations in the second phase.

- **S-box size:**

- ▶ For $n \leq 6$, degree-3 or degree-4 polynomials often have cost-1 decompositions.
- ▶ When $n \geq 7$, such low-cost decompositions become rare, forcing the use of higher-cost ones.
- ▶ Each additional cost level implies testing larger divisor sets (\mathcal{L}_n , then $\mathcal{Q}_{1,n}^L$, then $\mathcal{Q}_{2,n}^L$), whose size grows rapidly with n .

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Observation: At $n = 6$, about half of the quadratic polynomials can serve as divisors; at $n = 7$, this ratio drops below 0.15, and continues decreasing, making higher-degree divisions impractical in practice.