

Optimizing Implementations of Boolean Functions

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Goal:

- ▶ improve the understanding of the circuit complexity of Boolean functions and vectorial Boolean functions;
- ▶ develop new techniques for constructing better circuits for use by academia and industry.

Example circuits: ¹

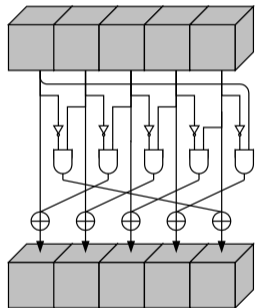
Circuit	Gate count					Depth	
	All	AND	XOR	XNOR	NOT	Total	AND
AES S-Box	113	32	77	4	0	27	6
AES-128(k,m)	28 600	6400	21 356	844	0	326	60
SHA-256(m)	115 882	22 385	89 248	3894	355	5403	1604

¹Project webpage: <https://csrc.nist.gov/Projects/circuit-complexity>

- ▶ Boolean circuits
- ▶ Optimizing linear circuits & Paar's heuristic
- ▶ Extending Paar's heuristic

A *Boolean circuit* with n inputs and m outputs is a **directed acyclic graph** (DAG), where

- ▶ the inputs and the gates are *nodes*,
- ▶ the edges correspond to Boolean-valued *wires*,
- ▶ *fanin/fanout* of a node is the number of wires going in/out the node,
- ▶ the nodes with fanin zero are called *input nodes*,
- ▶ the nodes with fanout zero are called *output nodes*.



Circuit for Keccak s-box
<https://keccak.team/figures.html>

Problem: Given a set of Boolean gates (e.g., AND, NAND, XOR, NOR), construct a circuit that computes a Boolean function that is optimal w.r.t. a target metric.

Target metric depends on the application.

- ▶ *Number of gates:* for *lightweight cryptography applications* running on constrained devices.
- ▶ *Number of nonlinear gates:* for *secure multi-party computation, zero-knowledge proofs and side channel protection.*
- ▶ *AND-depth:* for homomorphic encryption schemes.
- ▶ etc.

- ▶ Linear layers
 - ▶ provides **diffusion**
 - ▶ e.g., bit permutations, multiplication with a binary matrix
 - ▶ implementations by XOR, NOT gates
- ▶ Nonlinear layers
 - ▶ provides **confusion**
 - ▶ e.g., s-boxes
 - ▶ implementations by AND, NAND, XOR, NOT gates

Constructing efficient circuits for these layers are challenging, even for the linear ones.

Linear layers can be represented using a an $m \times n$ binary matrix M , applied to n input variables (x_1, \dots, x_n) to calculate m output variables (y_1, \dots, y_m) .

The linear layer

$$x_0 + x_1 + x_2 = y_0$$

$$x_1 + x_3 + x_4 = y_1$$

$$x_0 + x_2 + x_3 + x_4 = y_2$$

$$x_1 + x_2 + x_3 = y_3$$

$$x_0 + x_1 + x_3 = y_4$$

$$x_1 + x_2 + x_3 + x_4 = y_5$$

Matrix representation

$$\begin{pmatrix} 1 & 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 & 0 \\ 1 & 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 1 \end{pmatrix} \cdot \begin{pmatrix} x_0 \\ x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} y_0 \\ y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \end{pmatrix}$$

For a given binary $m \times n$ matrix M , the goal is to minimize the number of XOR operations.

- ▶ **Direct XOR (d-XOR)**

- ▶ Implements each row individually, corresponds to $(weight(M) - m)$ XORs.

- ▶ **Sequential XOR (s-XOR)**

- ▶ Counts the number of XOR operations of the form $x_i = x_i \oplus x_j$, that updates the value of input x_i
- ▶ Relevant for quantum implementations
- ▶ Known techniques (e.g., Gauss-Jordan elimination)

- ▶ **General XOR (g-XOR)**

- ▶ Corresponds to the number of operations of the form $x_i = x_j \oplus x_k$
- ▶ *The Shortest Linear Program (SLP) problem*: Minimizing the number of XORs (i.e., determining g-XOR) to compute Mx is known to be NP-hard. (Boyar et al., 2013)

Main idea:

- ▶ Determines the frequency for each possible pairs of input variable x_i, x_j ($i \neq j$) that are XORed together in m linear functions
- ▶ Compute the pair with highest frequency and place it to the matrix as a new variable
- ▶ Repeat until all outputs have been computed

Two options in a tie:

- ▶ Choose the first pair in lexicographical order
- ▶ Exhaust all equally frequent options

Example: Paar's Heuristic (1)

Matrix representation

$$\begin{pmatrix} 1 & 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 & 0 \\ 1 & 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 1 \end{pmatrix} \cdot \begin{pmatrix} x_0 \\ x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} y_0 \\ y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \end{pmatrix}$$

Pair	Frequency	Pair	Frequency
(x_0, x_1)	2	(x_1, x_3)	4
(x_0, x_2)	2	(x_1, x_4)	2
(x_0, x_3)	2	(x_2, x_3)	3
(x_0, x_4)	1	(x_2, x_4)	2
(x_1, x_2)	3	(x_3, x_4)	3

The first selected pair is (x_1, x_3) with frequency 4. So, the first step of the implementation is $t_0 = x_1 \oplus x_3$.

Example: Paar's Heuristic (2)

Updated matrix:

$$\begin{pmatrix} 1 & 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 & 0 \\ 1 & 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 & 1 \end{pmatrix}$$

Updated frequency table

Pair	Frequency	Pair	Frequency
(x_0, x_1)	1	(x_1, t_0)	0
(x_0, x_2)	2	(x_2, x_3)	1
(x_0, x_3)	1	(x_2, x_4)	2
(x_0, x_4)	1	(x_2, t_0)	2
(x_0, t_0)	1	(x_3, x_4)	1
(x_1, x_2)	1	(x_3, t_0)	0
(x_1, x_3)	0	(x_4, t_0)	2
(x_1, x_4)	0	-	-

Implementation:

$$t_0 = x_1 \oplus x_3$$

$$t_1 = x_0 \oplus x_2$$

$$t_2 = x_4 \oplus t_0$$

$$t_3 = x_1 \oplus t_1$$

$$t_4 = x_3 \oplus x_4$$

$$t_5 = t_1 \oplus t_4$$

$$t_6 = x_2 \oplus t_0$$

$$t_7 = x_0 \oplus t_0$$

$$t_8 = x_2 \oplus t_2$$

The output $(y_0, y_1, y_2, y_3, y_4, y_5)$ is obtained as $(t_3, t_2, t_5, t_6, t_7, t_8)$.

Cancellations in circuits happen when the inputs to an XOR gate are of the form $(x_1 \oplus x_3, x_2 \oplus x_3)$. The XOR gate computes $x_1 \oplus x_2$, and cancels x_3 .

Paar's heuristic is **cancellation-free**, which leads to generating sub-optimal circuits (Boyar et al., 2019).

New heuristics with cancellation property, such as Maximov & Ekdahl, 2019, Banik et al. 2019, Xiang et al, 2020.

Observation

- ▶ Due to cancellation-free property, a modification of Paar's algorithm can be applied to nonlinear Boolean functions.

Represent n -variable Boolean function with m monomials using a $m \times n$ binary matrix.

Example. $f = x_1 + x_2.x_3 + x_0x_1x_3x_4$. Matrix representation is $\begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 1 & 1 \end{pmatrix}$

With this representation, it is possible to apply Paar's heuristic. Note that XOR operations now corresponds to AND.

Not promising approach, as it implements each monomial independently.

Main idea: Decompose Boolean function into homogeneous Boolean functions, and exploit affine equivalence relations to find low-weight matrix representations.

1. Decompose f into d homogeneous Boolean functions,

$$f = a + f_1 \oplus f_2 \oplus \dots \oplus f_d,$$

where f_i is the sum of monomials of f with degree i , and a is the constant term.

Example. $f = x_1 + x_2.x_3 + x_2.x_4 + x_2x_5 + x_1.x_2.x_4 + x_1.x_2.x_5 + 1$ The decomposition is

$$a = 1$$

$$f_1 = x_1$$

$$f_2 = x_2.x_3 + x_2.x_4 + x_2x_5$$

$$f_3 = x_1.x_2.x_4 + x_1.x_2.x_5$$

2. Apply affine transformations to the highest-degree homogeneous function, (i.e., f_d) to reduce the # of monomials. If f'_d includes monomials with degree smaller than d , those monomials are added to the corresponding f_i depending on their degree.

$$a = 1$$

$$f_1 = x_1 \rightarrow f'_1 = x_1$$

$$f_2 = x_2.x_3 + x_2.x_4 + x_2.x_5 \rightarrow f'_2 = x_2.x_3$$

$$f_3 = x_1.x_2.x_4 + x_1.x_2.x_5 \rightarrow f'_3 = x_1.x_2.x_3$$

3. Apply modified Paar's heuristic to find an implementation for the degree d terms of f'_d . (Note that in modified Paar's heuristic each iteration corresponds to modulo 2 multiplication, instead of modulo 2 addition.) Apply the inverse affine transformation to the circuit to construct an implementation for the degree d monomials of f .
4. Repeat the procedure to find an implementation for f'_{d-1} where f'_{d-1} is the XOR of f_d and the new degree $d - 1$ monomials generated during Step 2.
5. Repeat until implementations for all homogeneous function is generated and combine the sub-circuits.

- ▶ Most time consuming phase is finding the *right* affine equivalence class. If a class representative with low degree is available, decomposing functions into homogeneous functions, and reducing the number of monomials of same degree can be done much more efficiently.

Example. Let $n = 6$. There are 150357 affine equivalence classes.

- ▶ Degree=6, # classes = 74596 \rightarrow # monomial = 1
 - ▶ Degree=5, # classes = 73262 \rightarrow # monomial = 1
 - ▶ Degree=4, # classes = 2465 \rightarrow # monomial ≤ 3
 - ▶ Degree=3, # classes = 30 \rightarrow # monomial ≤ 5
 - ▶ Degree=2, # classes = 3 \rightarrow # monomial ≤ 3
-
- ▶ We observe that the technique achieves optimal implementations (in terms of nonlinear gates) for some of the classes for small $n \leq 6$, where it is possible to compare with the optimal values.

- ▶ Proposed a modification to Paar's algorithm to apply to nonlinear Boolean functions (possible due to the cancellation-free property)
- ▶ Technique is currently more efficient when a low-weight representative from the equivalence class of the target function is available ($n \leq 6$).
- ▶ For larger n , our goal is to achieve a generic bound for Boolean function complexity (in term of AND gates), which is better than generic bounds.

Thanks! Questions?

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- ▶ GitHub: <https://github.com/usnistgov/Circuits/>
- ▶ NIST Circuit Complexity Project Webpage:
<https://csrc.nist.gov/Projects/Circuit-Complexity>



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