

Equation Solvers over GF(2)

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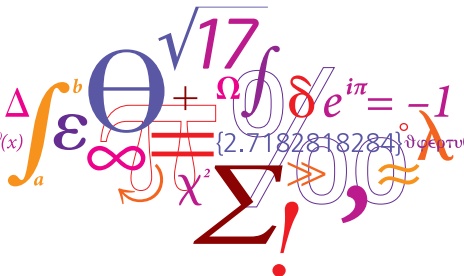
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14th December 2023

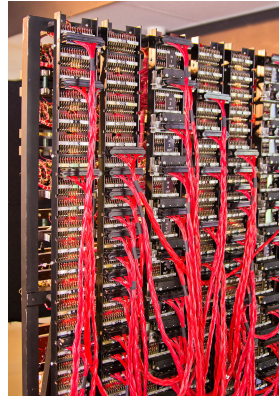


$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$



Sustainable Cryptanalysis

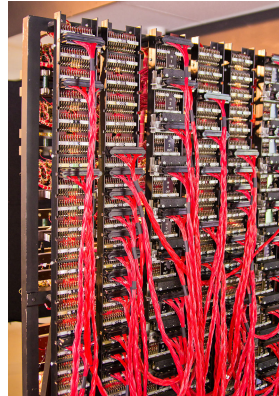
- The idea of making dedicated machines to attack ciphers is not new.



Source: Wikipedia

Sustainable Cryptanalysis

- Why should we care to make circuits ?



Source: Wikipedia

- Most general purpose CPU's have the following structure
 - [A] A processing unit having logic gates and registers
 - [B] A control unit having an instruction register and a program counter
 - [C] Primary memory that stores data and instructions
 - [D] Secondary memory, usually an external mass storage.
- Any computational step of the algorithm
 - First broken down into a sequence of instructions
 - Resides in the control unit.
 - Fetches data from the primary/secondary memory,
 - Processed in the processing unit.

- Dedicated Circuit
 - [A] Collection of logic elements
 - [B] Assembled specifically for the given task
 - [C] Thus executes them with much greater efficiency.
 - [D] Faster and more energy efficient.
- Gaussian Elimination of a 656×656 matrix
 - Dedicated circuit [SMITH, Bogdanov et al. 07]
 - Requires 86 ms.
 - On a Linux 800 MHz PIII PC would take around 40 minutes,
 - We can execute computationally heavy tasks on such circuits.

Introduction: Solving an Equation System



- Given m eqns P_1, P_2, \dots, P_m of n variables over $\text{GF}(2)$ of max degree d .
 - Usually $m = n$, sometimes $m > n$
 - Each equation is a multivariate polynomial over $\text{GF}(2)$
 - The algebraic degree d is usually small.
 - Task: find a common root: $r \in \{0, 1\}^n$ such that $P_i(r) = 0, \forall i$.
- Problem arises in many cryptographic contexts.
 - Block ciphers with low multiplicative complexities like LowMC
 - Given single pt/ct: solving low degree polynomials.
 - Signature schemes like UOV.
 - Cryptanalysis: solving quadratic polynomials over $\text{GF}(2)$.

Linear Systems

If Equations are Linear ($d = 1$)

LSE (m equations, n variables)

- Typical LSE

$$a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1$$

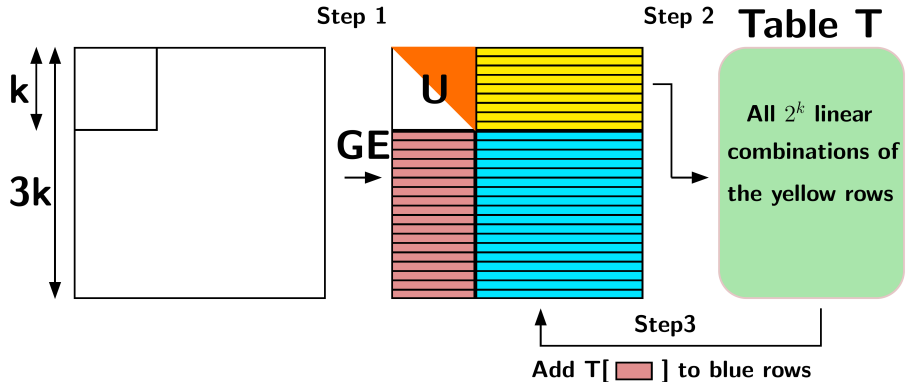
$$a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2$$

$$\vdots$$

$$a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n = b_m$$

- Equivalently $A\vec{x} = \vec{b}$
- Linear equations can be Solved by Gaussian Elimination (GE) efficiently.
- GE takes n^3 operations in the worst case.
- Given a linear system of form $A\vec{x} = \vec{b}$
 - Convert to equivalent system $U \cdot \vec{x} = \vec{b}'$, where U is upper-triangular.
 - Done by applying elementary row operations.

MR4I



Popular in SW

- used in computer algebra packages like SAGE.
- It is interesting to see how far this can be applied in HW

Gaussian Elimination

Gaussian_Elimination(A, n)

Input: Matrix $A \in \{0, 1\}^{n \times n}$: Input matrix

for each column $k = 1 \rightarrow n$ **do**

$s := k$;

while $a_{sk} = 0$ **do**

$s := s + 1$;

end

 Swap row \vec{a}_k with row \vec{a}_s ;

for each row $i = 1 \rightarrow n$ **do**

if $i \neq k$ and $a_{ik} = 1$ **then**

$a_{ij} := a_{ij} \oplus a_{kj}$

end

end

end

SMITH - Architecture

- GE requires 2 operations: row addition/row swap

1	0	0	1	0	1	1	0
0	0	0	1	0	1	0	1
1	1	0	0	1	0	1	0
0	0	1	1	0	1	0	1
1	1	0	0	0	1	0	1
0	0	1	1	0	0	0	1
0	1	0	1	1	0	1	1
1	0	0	0	1	1	0	1

- Pivot is the top-left element of unprocessed rows.

 **Pivot**

SMITH - Architecture

- GE requires 2 operations: row addition/row swap




- Column sweep once pivot is fixed.

Pivot

SMITH - Architecture

- GE requires 2 operations: row addition/row swap

1	0	0	1	0	1	1	0
0	0	0	1	0	1	0	1
0	1	0	1	1	1	0	0
0	0	1	1	0	1	0	1
0	1	0	1	0	0	1	1
0	0	1	1	0	0	0	1
0	1	0	1	1	0	1	1
0	0	0	1	1	0	1	1



 Pivot

- Pivot can not be zero. So swap with next available row which is **unprocessed**.

SMITH - Architecture

- GE requires 2 operations: row addition/row swap

1	0	0	1	0	1	1	0
0	1	0	1	1	1	0	0
0	0	0	1	0	1	0	1
0	0	1	1	0	1	0	1
0	1	0	1	0	0	1	1
0	0	1	1	0	0	0	1
0	1	0	1	1	0	1	1
0	0	0	1	1	0	1	1

↓

↻

- Swap is done. Ready for next row operation.

Pivot

SMITH - Architecture

- GE requires 2 operations: row addition/row swap

$$\begin{array}{cccccccc}
 1 & 0 & 0 & 1 & 0 & 1 & 1 & 0 \\
 0 & 1 & 0 & 1 & 1 & 1 & 0 & 0 \\
 0 & 0 & 0 & 1 & 0 & 1 & 0 & 1 \\
 0 & 0 & 1 & 1 & 0 & 1 & 0 & 1 \\
 0 & 0 & 0 & 0 & 1 & 1 & 1 & 1 \oplus \\
 0 & 0 & 1 & 1 & 0 & 0 & 0 & 1 \\
 0 & 0 & 0 & 0 & 0 & 1 & 1 & 1 \oplus \\
 0 & 0 & 0 & 1 & 1 & 0 & 1 & 1
 \end{array}$$






 **Pivot**

- Second column is cleared.

Circuit Issues

- Assume each element is stored in a single flip-flop.

1	0	0	1	0	1	1	0
0	0	0	1	0	1	0	1
0	1	0	1	1	1	0	0
0	0	1	1	0	1	0	1
0	1	0	1	0	0	1	1
0	0	1	1	0	0	0	1
0	1	0	1	1	0	1	1
0	0	0	1	1	0	1	1



 **Pivot**

- Pivot is ever changing. How to keep track of it ?
- Can not swap with already processed row...
- How to select next row for swap?

SMITH - Architecture

- Initially both circuit and state are same
- Two additional registers Vec1 and Vec2
- To keep track of control flow.

1	0	0	1	0	1	1	0
0	0	0	1	0	1	0	1
1	1	0	0	1	0	1	0
0	0	1	1	0	1	0	1
1	1	0	0	0	1	0	1
0	0	1	1	0	0	0	1
0	1	0	1	1	0	1	1
1	0	0	0	1	1	0	1

 Pivot

1	0	0	1	0	1	1	0
0	0	0	1	0	1	0	1
1	1	0	0	1	0	1	0
0	0	1	1	0	1	0	1
1	1	0	0	0	1	0	1
0	0	1	1	0	0	0	1
0	1	0	1	1	0	1	1
1	0	0	0	1	1	0	1

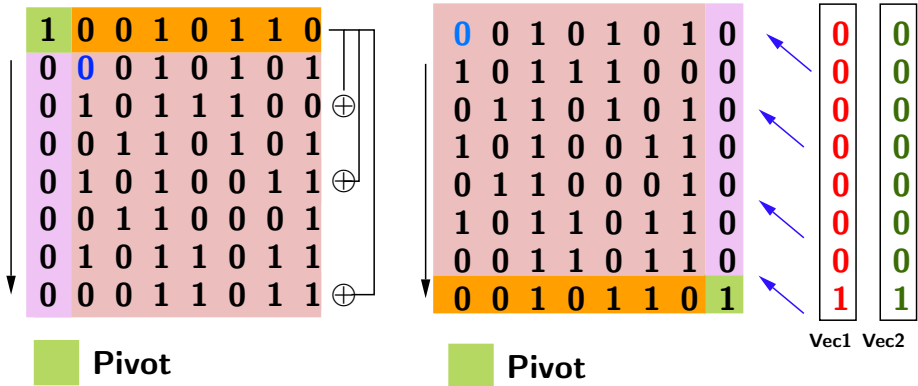
 Pivot

0	1
0	0
0	0
0	0
0	0
0	0
0	0
0	0

Vec1 Vec2

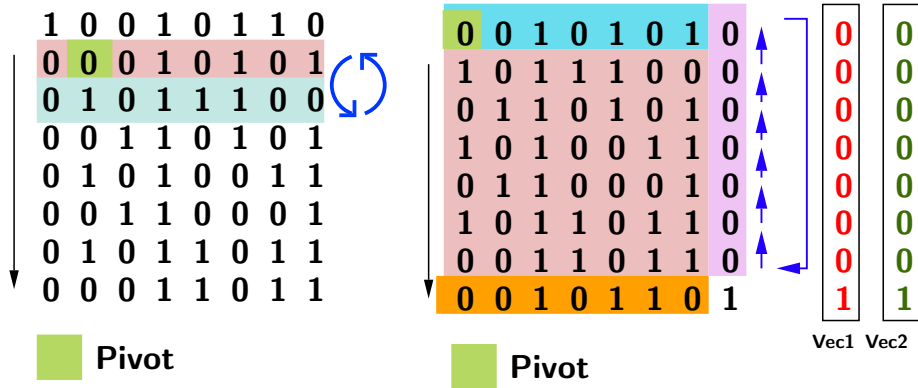
SMITH - Architecture

- Diagonal flow brings a_{11} to bottom-left (next pivot shifts to (1,1)!!!)
- The parts in red, blue and orange move accordingly.
- Orange part moves without xor operation/ purple is all zero



SMITH - Architecture

- Xor rule: New $a_{ij} = a_{i+1, j+1} \oplus a_{1, j+1} \cdot a_{i+1, 1}$
- Vec1=Vec1 $\ll 1 \parallel 1$ (last row is processed).
- Vec2=Vec2 $\lll 1$ (last row is current row).



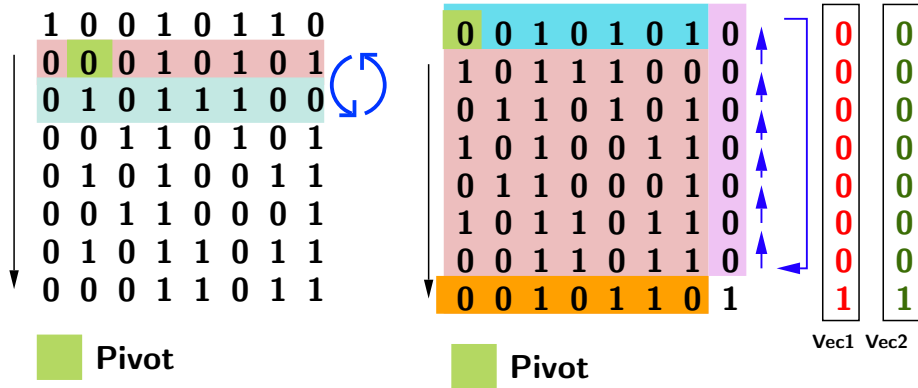
SMITH - Xor rule

Row operations

$$\begin{bmatrix} 1 & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \rightarrow \begin{bmatrix} a_{22} \oplus (a_{12} \cdot a_{21}) & a_{23} \oplus (a_{13} \cdot a_{21}) & \cdots & 0 \\ a_{32} \oplus (a_{12} \cdot a_{31}) & a_{33} \oplus (a_{13} \cdot a_{31}) & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ a_{n2} \oplus (a_{12} \cdot a_{n1}) & a_{n3} \oplus (a_{13} \cdot a_{n1}) & \cdots & 0 \\ a_{12} & a_{13} & \cdots & 1 \end{bmatrix}$$

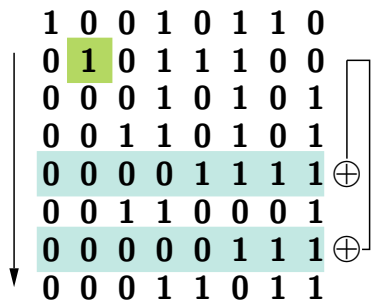
SMITH - Architecture

- Pivot is already at (1,1) and zero \rightarrow Now row swaps are required
- Instead of row swap, we circularly rotate unprocessed rows till $a_{11} \neq 0$.
- Need to do same operation to the \vec{b} in $A\vec{x} = \vec{b}$.

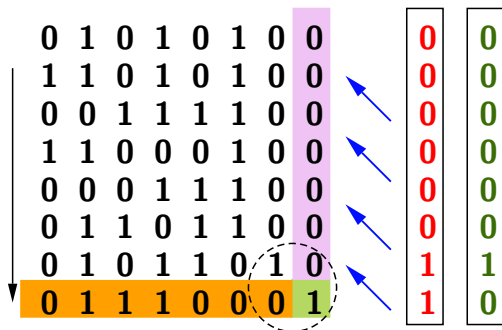


SMITH - Architecture

- After row operation: last 2 cols are cleared.
- Bottom left corner becomes identity.
- Stops when Vec1 is all one.



 Pivot



 Pivot

Vec1 Vec2

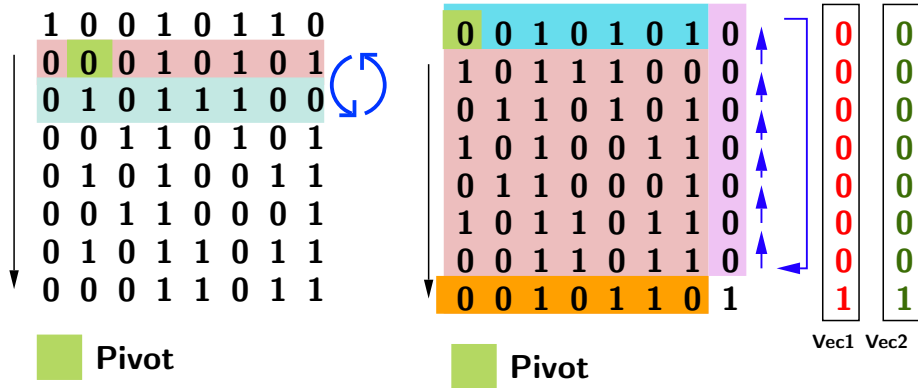
SMITH - Swap rule

Row Swap

$$\begin{bmatrix} \vec{a}_1 \\ \vec{a}_2 \\ \vec{a}_3 \\ \vdots \\ \vec{a}_i \\ \vec{a}_{i+1} \\ \vec{a}_{i+2} \\ \vec{a}_{i+3} \\ \vdots \\ \vec{a}_n \end{bmatrix} \rightarrow \begin{bmatrix} \vec{a}_i \\ \vec{a}_1 \\ \vec{a}_2 \\ \vdots \\ \vec{a}_{i-1} \\ \vec{a}_{i+1} \\ \vec{a}_{i+2} \\ \vec{a}_{i+3} \\ \vdots \\ \vec{a}_n \end{bmatrix}$$

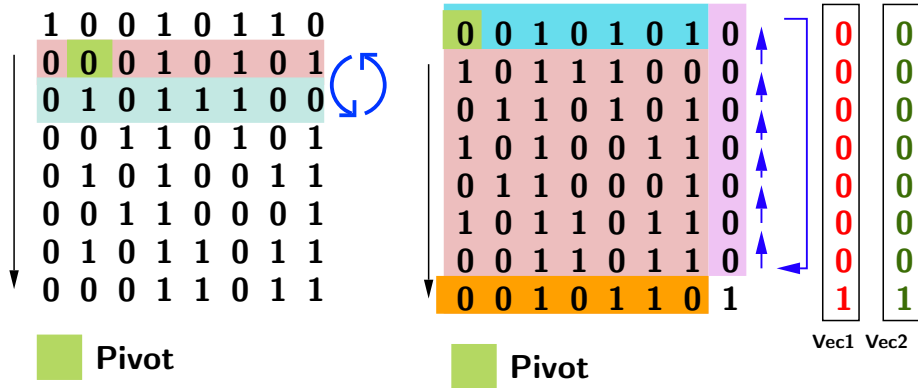
SMITH - RUNTIME

- What is the average Runtime for $n \times n$ matrix???
- At least n row xors are required.
- Additional time depends on number of row swaps.



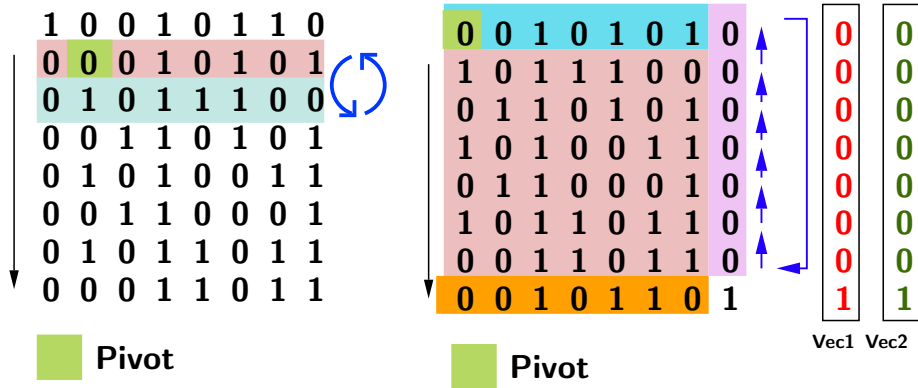
SMITH - RUNTIME

- If Pivot == 1 ($p = \frac{1}{2}$), no swaps are required (zero additional time)
- After one circular rotate if pivot==1, no further swaps are necessary.
- Thus one additional cycle in this case ($p = \frac{1}{4}$).



SMITH - RUNTIME

- If Pivot == 1 after t rotations ($p = \frac{1}{2^{t+1}}$), t additional time.
- $E = 0 \cdot \frac{1}{2} + 1 \cdot \frac{1}{4} + 2 \cdot \frac{1}{8} + \dots + t \cdot \frac{1}{2^{t+1}} + \dots \approx 1$.
- Every row xor needs one extra cycle on average $\rightarrow 2n$ cycles.



Matrix Square and Full rank

- Stops when Vec1 is all one...
- At the end of computation the FF array holds the identity matrix
- Same operations done on \vec{b} .
- If \vec{b}^* is the final state of \vec{b}
→ $\vec{x} = \vec{b}^*$ is the unique solution of $A\vec{x} = \vec{b}$.

Proof

Original Equation $A\vec{x} = \vec{b}$

Unique solution to this is $\vec{x} = A^{-1}\vec{b}$

$A \rightarrow I$ is only possible

→ Iff product of all linear operations on A equals A^{-1}

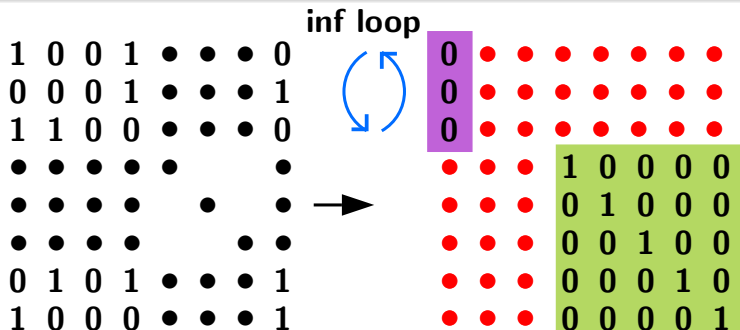
Same operations on b gives $\vec{b} \rightarrow \vec{b}^* = A^{-1}\vec{b}$

QED

SMITH - Use cases

Matrix Square and NOT Full rank

- Never Stops \rightarrow Vec1 is never all one...
- At the end of computation bottom part holds the identity matrix
- Does not yield any meaningful solution.
- Additional counter logic required to stop infinite loop
 \rightarrow Stop if counter $>$ # Remaining rows.



Matrix Non-Square and Overdefined

- That is $m > n$, there are more equations than variables.
- The system is solvable iff

$$A \rightarrow \begin{pmatrix} 0 \\ I_n \end{pmatrix}, \quad \vec{b} \rightarrow \begin{pmatrix} 0 \\ \vec{b}^* \end{pmatrix}$$

- Again Additional counter logic required to stop infinite loop
→ Stop if counter $> m - n$.
- No solution if for any $\vec{t} \neq 0$:

$$A \rightarrow \begin{pmatrix} 0 \\ I_n \end{pmatrix}, \quad \vec{b} \rightarrow \begin{pmatrix} \vec{t} \\ \vec{b}^* \end{pmatrix}$$

SMITH - Other Uses

Matrix Multiplication

- Observe the following

$$D = \begin{pmatrix} I_n & A & 0 \\ 0 & I_n & B \\ 0 & 0 & I_n \end{pmatrix}, \quad D^{-1} = \begin{pmatrix} I_n & A & A \cdot B \\ 0 & I_n & B \\ 0 & 0 & I_n \end{pmatrix}$$

- Matrix multiplication $A \cdot B$ is possible given $3n \times 3n$ space
- Also observe if

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{pmatrix}, \quad B = \begin{pmatrix} \vec{b}_1 \\ \vec{b}_2 \\ \vdots \\ \vec{b}_n \end{pmatrix}$$

- Then we have

$$A \cdot B = \begin{pmatrix} a_{11} \vec{b}_1 \\ a_{21} \vec{b}_1 \\ \vdots \\ a_{n1} \vec{b}_1 \end{pmatrix} \oplus \begin{pmatrix} a_{12} \vec{b}_2 \\ a_{22} \vec{b}_2 \\ \vdots \\ a_{n2} \vec{b}_2 \end{pmatrix} \oplus \cdots \oplus \begin{pmatrix} a_{1n} \vec{b}_n \\ a_{2n} \vec{b}_n \\ \vdots \\ a_{nn} \vec{b}_n \end{pmatrix}$$

SMITH - Other Uses

Matrix Multiplication

$$A \cdot B = \begin{pmatrix} a_{11}\vec{b}_1 \\ a_{21}\vec{b}_1 \\ \vdots \\ a_{n1}\vec{b}_1 \end{pmatrix} \oplus \begin{pmatrix} a_{12}\vec{b}_2 \\ a_{22}\vec{b}_2 \\ \vdots \\ a_{n2}\vec{b}_2 \end{pmatrix} \oplus \cdots \oplus \begin{pmatrix} a_{1n}\vec{b}_n \\ a_{2n}\vec{b}_n \\ \vdots \\ a_{nn}\vec{b}_n \end{pmatrix}$$

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SMITH - Other Uses

Matrix Multiplication

$$A \cdot B = \begin{pmatrix} a_{11}\vec{b}_1 \\ a_{21}\vec{b}_1 \\ \vdots \\ a_{n1}\vec{b}_1 \end{pmatrix} \oplus \begin{pmatrix} a_{12}\vec{b}_2 \\ a_{22}\vec{b}_2 \\ \vdots \\ a_{n2}\vec{b}_2 \end{pmatrix} \oplus \cdots \oplus \begin{pmatrix} a_{1n}\vec{b}_n \\ a_{2n}\vec{b}_n \\ \vdots \\ a_{nn}\vec{b}_n \end{pmatrix}$$

	A			A	AB				
\oplus	0 0 1	1 0 0		0 0 0	1 0 1	$a_{11}\vec{b}_1$	$a_{12}\vec{b}_2$	$a_{13}\vec{b}_3$	
Id_3	0 1 0	0 1 1		0 0 0	0 1 1	$a_{21}\vec{b}_1$	$\oplus a_{22}\vec{b}_2$	$\oplus a_{23}\vec{b}_3$	
\oplus	0 0 1	0 0 0		0 0 0	0 0 1	$a_{31}\vec{b}_1$	$a_{32}\vec{b}_2$	$a_{33}\vec{b}_3$	
Ze_3	1 0 0	1 0 0	B	1 0 0	1 0 0				
	0 1 0	1 1 1		Ze_3	0 1 0	1 1 1	B		
	0 0 1	0 0 1			0 0 1	0 0 1			
Ze_3		Ze_3		Ze_3	Ze_3				

SMITH- Hermite Canonical form



Over/Under-determined systems

- Given $Ax = b$. A and b is first padded with null rows/cols to get a square matrix A^* and extended column vector b^* .
- A^* converted to its Hermite Canonical Form H
- Do row operations $[A^* : I : b^*] \rightarrow [H : G : d]$
- Any general solution to $Ax = b$ is of the form $d + (I + H)z$ for any z ,
→ The columns of $I + H$ form a basis for the null space of A

SMITH- Hermite Canonical form

1	*	0	0	*	0	0	0
0	0	0	0	0	0	0	0
0	0	1	0	*	0	0	0
0	0	0	1	*	0	0	0
0	0	0	0	0	0	0	0
0	0	0	0	0	1	0	0
0	0	0	0	0	0	1	0
0	0	0	0	0	0	0	1

Over/Under-determined systems

- Upper triangular.
- If $h_{ii} = 0$, the row must be null
- If $h_{ii} = 1$, the col must be unit vector

SMITH- Hermite Canonical form

$$\begin{array}{cccccccc}
 1 & * & 0 & 0 & * & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 1 & 0 & * & 0 & 0 & 0 \\
 0 & 0 & 0 & 1 & * & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1
 \end{array}$$

Over/under-determined systems

- The SMITH circuit alone is insufficient.
- Other operations are required
- Depth is no longer constant (OPEN PROBLEM)

Quadratic Systems

Quadratic Systems



When $d = 2$

- Solving degree $d > 2$ equations is NP-complete.
- Hardness of solving quadratics leveraged to construct PKC's.
→ HFE, QUARTZ, UOV, SFLASH etc.
- Quadratic over $GF(2)$ has a maximum of $S = 1 + n + \binom{n}{2}$ non-zero coefficients:
$$P : c + a_1x_1 + a_2x_2 + \cdots + a_nx_n + a_{12}x_1x_2 + a_{13}x_1x_3 + \cdots a_{n-1,n}x_{n-1}x_n$$
- Evaluating one poly over one point needs approx S bit-ops.

Standard Exhaustive search

When $d = 2$, m equations, n variables

- Quadratic over $GF(2)$ has a maximum of $S = 1 + n + \binom{n}{2}$ non-zero coefficients:

$$P : c + a_1x_1 + a_2x_2 + \cdots + a_nx_n + a_{12}x_1x_2 + a_{13}x_1x_3 + \cdots + a_{n-1,n}x_{n-1}x_n$$

- Evaluating one poly over 2^n point needs approx $S \cdot 2^n$ bit-ops.
- Half of them are roots of P
 - Evaluate them 2^{n-1} over the next polynomial.
 - Total of $S \cdot 2^{n-1}$ operations.
- Third equation needs $S \cdot 2^{n-2}$ operations and so on ...

$$Comp = 2^n \cdot \left[S + \frac{S}{2} + \frac{S}{4} + \cdots \right] \approx S \cdot 2^{n+1} = O(n^2 2^n)$$

- Roots are points that are zeros that survive till the end.

Fast Exhaustive search [BCC+10/BCC+13]



Intuition

- Use Gray codes: $g_i = i \oplus (i \gg 1)$.
- Gray codes of successive integers differ by only 1 bit.

$$g_0 = 000$$

$$g_1 = 001$$

$$g_2 = 011$$

$$g_3 = 010$$

$$g_4 = 110$$

$$g_5 = 111$$

$$g_6 = 101$$

$$g_7 = 100$$

- We traverse the input space of f in a Gray code manner.
- From knowledge of $f(g_i)$: we can evaluate $f(g_{i+1})$ efficiently without having to evaluate the entire function.

Taylor Expansion

The Idea

- $f(g_0) = f(\vec{0})$ is just the constant term of f .
- t is the bit-position where g_j and g_{j+1} differ
→ Let's say we already have the value of $f_i(g_j)$

$$f(g_{j+1}) = f(g_j) \oplus \frac{\delta f}{\delta x_t}(g_j). \quad (1)$$

- Here $\frac{\delta f}{\delta x_t}$ is the 1st order derivative of the function f at the point x_t .
- For example if $f = x_1x_2 \oplus x_3 \oplus x_1x_4x_5$,
→ $\frac{\delta f}{\delta x_1} = x_2 \oplus x_4x_5$ and $\frac{\delta f}{\delta x_2} = x_1$, $\frac{\delta f}{\delta x_3} = 1$ etc.
- Derivative has degree one less than f and is easier to compute.

Taylor Expansion

Example

- If $f = x_1x_2 \oplus x_3 \oplus x_1x_4 \oplus x_5$,
 $\rightarrow \frac{\delta f}{\delta x_1} = x_2 \oplus x_4$ and $\frac{\delta f}{\delta x_2} = x_1$, $\frac{\delta f}{\delta x_3} = 1$, $\frac{\delta f}{\delta x_4} = x_1$ and $\frac{\delta f}{\delta x_5} = 1$.
- If original is quadratic derivatives are linear!!
 $\rightarrow \frac{\delta^2 f}{\delta x_1 x_2} = 1$, $\frac{\delta^2 f}{\delta x_1 x_3} = 0$ etc
 \rightarrow Second derivatives are constant..
- Start with $f(00000) = 0$, next we find $f(g_1) = f(00001)$

$$f(00001) = f(00000) \oplus \frac{\delta f}{\delta x_1}(00000) = 0 \oplus x_2 + \oplus x_4|_{00000} = 0$$

- Then $f(g_2) = f(00011)$

$$f(00011) = f(00001) \oplus \frac{\delta f}{\delta x_2}(00001) = 0 \oplus x_1|_{00001} = 1$$

- Each next step takes evaluation of linear equation

Efficient Exhaustive Search for solutions over F_2



Main Theorem

All the zeroes of a single multivariate polynomial f in n variables of degree d can be found in essentially $d \cdot 2^n$ bit operations (plus a negligible overhead), using n^{d-1} bits of read-write memory, and accessing n^d bits of constants, after an initialization phase of negligible complexity $O(n^{2d})$.

- You need to pre-compute all derivatives.
- Precomputation needs time and energy and space.
- Precomputation required for each new equation system.
- Works best if $d = 2$ or lower.

Systems of arbitrary degree

Construct Truth tables

Truth Tables

$x_0x_1x_2$	P_0	P_1	P_2	\dots	P_m	$\bigvee P_i$
000	0	1	1		0	1
001	1	0	0		1	1
010	0	1	1		1	1
011	1	1	0		0	1
100	0	0	0		0	0
				• •		
110	0	1	0		1	1
111	0	1	1		0	1

Root=100

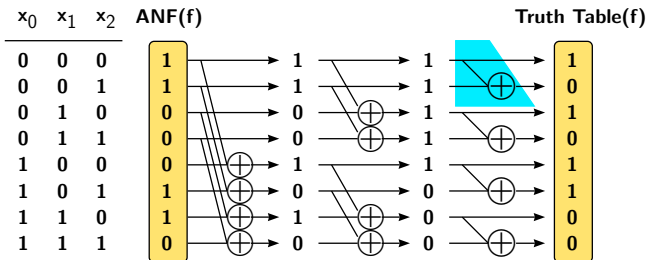
Truth Tables

- Evaluation of a function at all points of its space. How can they help?

Möbius Transform

- Given the algebraic equation of any n -variable Boolean function, how to evaluate it over all the 2^n points of its input domain (i.e. find truth table) ?
- Given truth table of a Boolean function how to deduce its algebraic equation ?
- Answer to both the above is Möbius Transform.
- It is a linear, involutive transform that does both the above.
- Requires $n \cdot 2^{n-1}$ bit-operations.

Möbius Transform



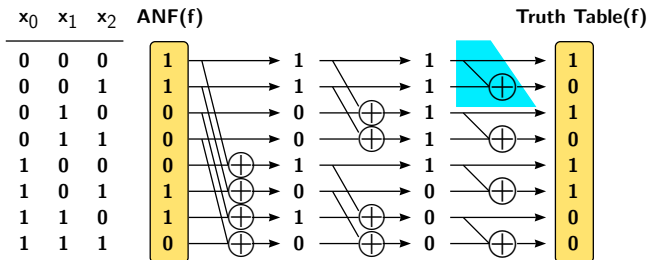
$$f = 1 + x_0x_1 + x_2 + x_0x_2$$

Figure: Möbius transform on $f = 1 \oplus x_0x_1 \oplus x_2 \oplus x_0x_2$. The blue shaded component represents one butterfly unit.

Salient Points

- Note we have lexicographical indexing.
- $t_6 = 1 \Rightarrow 6 = (110)_2 \Rightarrow$ the ANF contains the $x_0x_1 = x_0^1 \cdot x_1^1 \cdot x_2^0$ term.

Möbius Transform



$$f = 1 + x_0x_1 + x_2 + x_0x_2$$

Figure: Möbius transform on $f = 1 \oplus x_0x_1 \oplus x_2 \oplus x_0x_2$. The blue shaded component represents one butterfly unit.

Salient Points

- n stages and 2^{n-1} xors per stage.
- Involutive: the same operations on ANF will give back TT.

- If $\vec{v} = [v_0, v_1, \dots, v_{2^n-1}]$ be the truth-table of f (note $v_i = f(i)$).
- If $\vec{u} = [u_0, u_1, \dots, u_{2^n-1}]$ be the ANF of f .
- Then it is well known that

$$\vec{v} = M_n \cdot \vec{u}$$

- Note $M = m_{ij}$ is such that

$$m_{ij} = 1 \text{ if } j \preceq i \text{ and } 0 \text{ otherwise.}$$

- Eg $100 \preceq 101$, but $011 \not\preceq 100$ since 011 exceeds 100 in the last 2 bit-locations.

The Mathematics

- M_n is well studied in literature: Lower triangular + Involution.
- Since $M_n = M_n^{-1}$, both $\vec{v} = M_n \cdot \vec{u}$ and $\vec{u} = M_n \cdot \vec{v}$ hold.
- Define $M_1 = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}$, then for all $n > 1$, we have $M_n = M_1 \otimes M_{n-1}$, where \otimes is the matrix tensor product.

$$M_3 = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 \\ 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

Exponential circuits: The circuit Expmob1

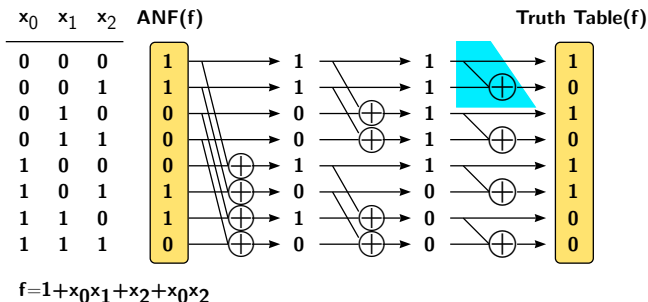


Figure: Möbius transform on $f = 1 \oplus x_0x_1 \oplus x_2 \oplus x_0x_2$. The blue shaded component represents one butterfly unit.

- Huge combinatorial circuit that stacks the stages one by one.
- Calculates in one single clock cycle: $n \cdot 2^{n-1}$ xor gates.

Exponential circuits: The circuit Expmob2

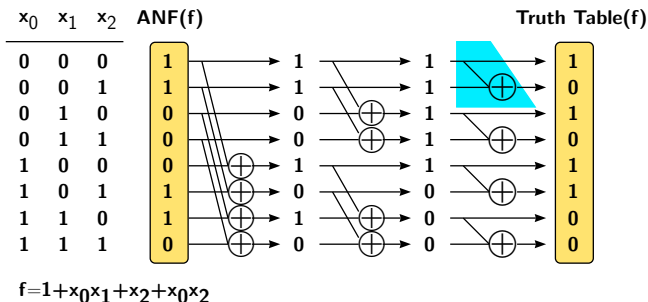


Figure: Möbius transform on $f = 1 \oplus x_0x_1 \oplus x_2 \oplus x_0x_2$. The blue shaded component represents one butterfly unit.

- Round based circuit: One stage in one clock cycle.
- Calculates in one n clock cycles: 2^{n-1} xor gates + Register of 2^n bits.

Exponential circuits: The circuit Expmob2

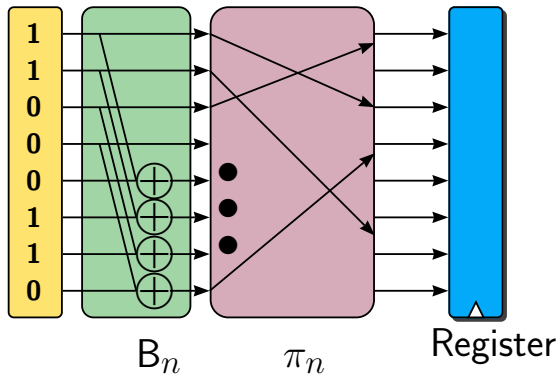
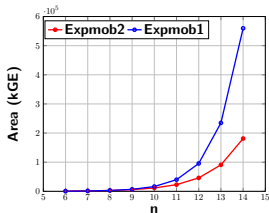


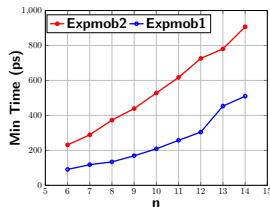
Figure: Round based Circuit.

- $\pi_n(2x) = x$, and $\pi_n(2x + 1) = 2^{n-1} + x$ for all $0 \leq x < 2^{n-1}$
- If P_n is the permutation matrix for π_n , it can be shown $M_n = (P_n \cdot B_n)^n$.

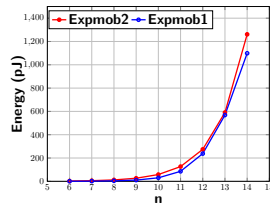
Results (Nangate 15nm Open Cell Library)



(a) Area



(b) Time



(c) Energy

Figure: Synthesis results for **Expmob1** and **Expmob2** circuits

Polynomial number of Coefficients

- ANF of Linear function: $n + 1$ coefficients.
- ANF of Quadratic function: $\binom{n}{2} + n + 1$ coefficients.
- ANF of Degree d function: $\binom{n}{\downarrow d} = \sum_{i=0}^d \binom{n}{i}$ coefficients $\in O(n^d)$.
- Challenge: With a register of size $\binom{n}{\downarrow d}$, can we compute the transform?

Take a look back

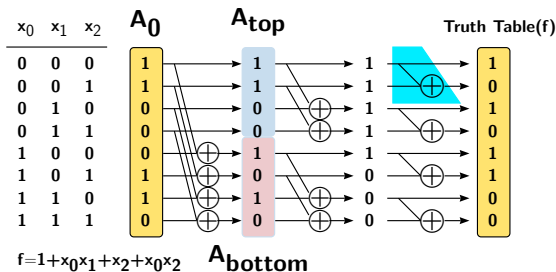


Figure: Round based Circuit.

- First stage $A_0 \rightarrow$ vectors A_{top} and A_{bottom} .
- A_{top} is actually ANF vector for $f(0, x_1, x_2)$ (in $n - 1$ variables!!)
- A_{bottom} is actually ANF vector for $f(1, x_1, x_2)$ (in $n - 1$ variables!!)
- Recursively apply Möbius Transform to these smaller vectors

Algorithm 1: Recursive Möbius Transform

Möbius (A_0, n, d)

Input: A_0 : The compressed ANF vector of a Boolean function f

Input: n : Number of variables, d : Algebraic degree

Output: The Truth table of f

```
/* Final step, i.e. leaf nodes of recursion tree */
if  $n=d$  then
    Use the formula  $B = M_n \cdot A_0$  to output partial truth table  $B$ .
    /* Use either Expmob1/Expmob2 to do this */
end
else
    Declare an array  $T$  of size  $\binom{n-1}{\downarrow d}$  bits.
    /* Compute the 2 operations of the butterfly layer */
1 Store 1st butterfly output i.e.  $A_{\text{top}}$  in  $T$  (requires no xors).
   Call Möbius  $(T, n-1, d)$ 
2 Store 2nd butterfly output i.e.  $A_{\text{bottom}}$  in  $T$  (requires some xors).
   Call Möbius  $(T, n-1, d)$ 
end
```

Recursion tree

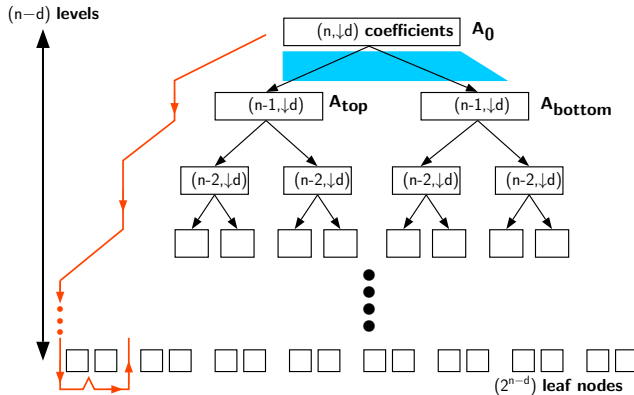


Figure: Recursion tree for the Möbius Transform algorithm. The blue shaded component roughly represents one arm of the butterfly unit.

- The Tree requires Depth first Traversal
- In Software this requires context switches, every time we traverse one level down.
- Mapping to hardware non trivial.

Circuit Sketch Polymob1

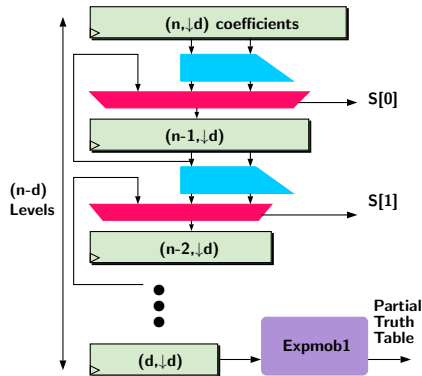


Figure: Hardware architecture **Polymob1** for the Möbius Transform algorithm. The blue shaded part roughly represents one arm of the butterfly unit.

- Primitive attempt to map algorithm to hw: can this work ?
- Each level needs own storage of size $\binom{n-i}{\downarrow d}$
- Task 1: Can we prove the space requirement is $O(n^{d+1})$??

Proof (contd)

- Applying the hockey-stick identity on each column we get

$$S(n, d) < \binom{n+1}{d+1} + \binom{n+1}{d} + \cdots + \binom{n+1}{1}$$

Using mathematical induction it is easy to prove the hypothesis $\mathcal{P}(d) : \sum_{i=0}^d \binom{n}{i} < n^d$, for all $d \geq 2$, $n > d$. The base case for $d = 2$, amounts to $n(n-1)/2 + n + 1 < n^2 \Rightarrow n^2 > n + 2$, which holds for all $n > 2$. Taking $\mathcal{P}(d)$ to be true we have

$$\begin{aligned} \mathcal{P}(d+1) : \sum_{i=0}^{d+1} \binom{n}{i} &< n^d + \binom{n}{d+1} \\ &< n^d + \frac{n^{d+1}}{(d+1)!} = n^d \left(1 + \frac{n}{(d+1)!} \right) < n^{d+1} \end{aligned}$$

Therefore we have $S(n, d) < (n+1)^{d+1}$, from which we can conclude it is $O(n^{d+1})$.

Circuit Sketch Polymob1

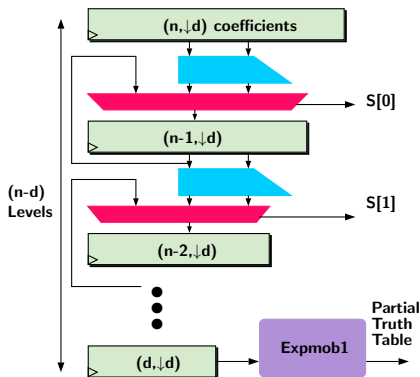


Figure: Hardware architecture **Polymob1** for the Möbius Transform algorithm. The blue shaded part roughly represents one arm of the butterfly unit.

- One reg of size $\binom{n}{\downarrow d}$ for A_0 , but only one reg of size $\binom{n-1}{\downarrow d}$.
- If level 2 stores A_{top} , it must preserve this till its entire left sub-tree is executed.
- Only then overwrite to A_{bottom} .

Circuit Sketch Polymob1

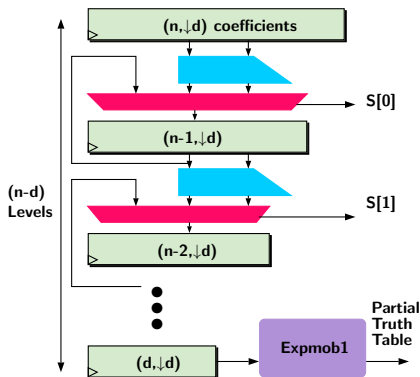


Figure: Hardware architecture **Polymob1** for the Möbius Transform algorithm. The blue shaded part roughly represents one arm of the butterfly unit.

- Multiplexer select signals control the flow.
- 3:1 multiplexer \rightarrow Either preserve state or overwrite with $A_{\text{top/bottom}}$
- However only 2:1 mux is sufficient.

A bit of notation

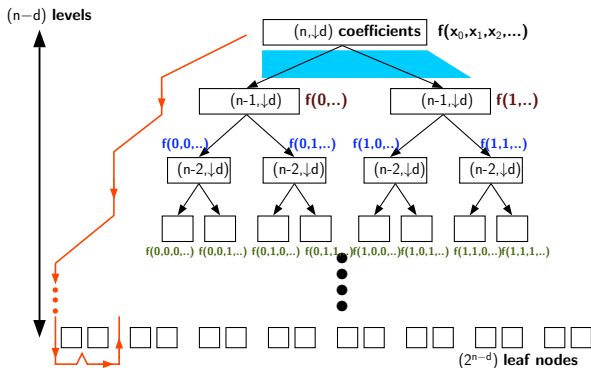


Figure: Hardware architecture **Polymob1** for the Möbius Transform algorithm. The blue shaded part roughly represents one arm of the butterfly unit.

- Every level sets one bit in the function argument.

A bit of notation

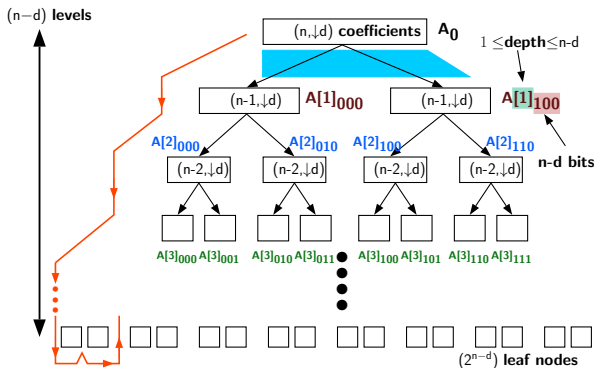


Figure: Hardware architecture **Polymob1** for the Möbius Transform algorithm. The blue shaded part roughly represents one arm of the butterfly unit.

- Let us label each ANF as $A[\text{depth}]_{\text{bits}}$

A bit of notation

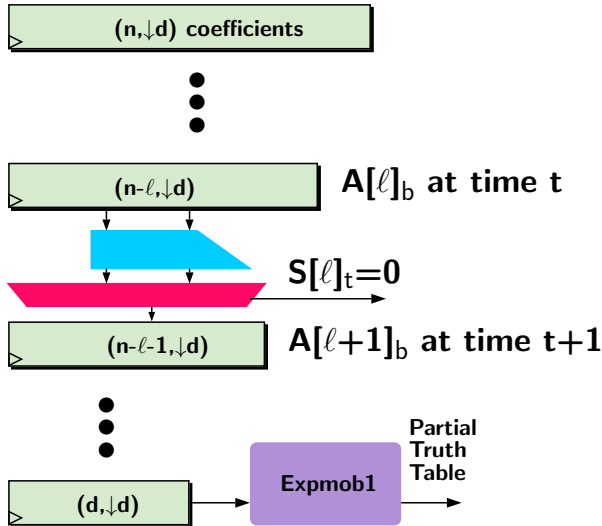


Figure: Hardware architecture **Polymob1** for the Möbius Transform algorithm. The blue shaded part roughly represents one arm of the butterfly unit.

A bit of notation

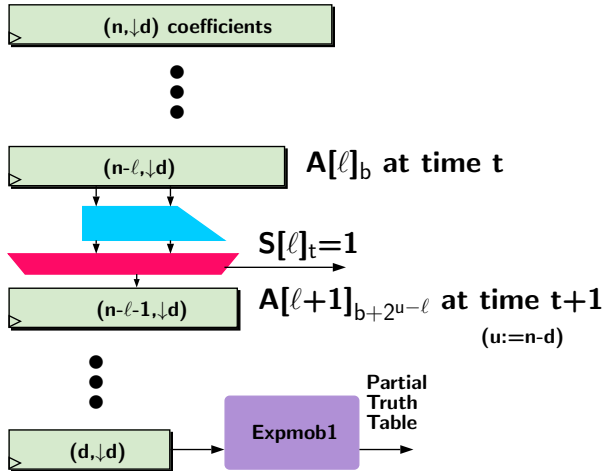
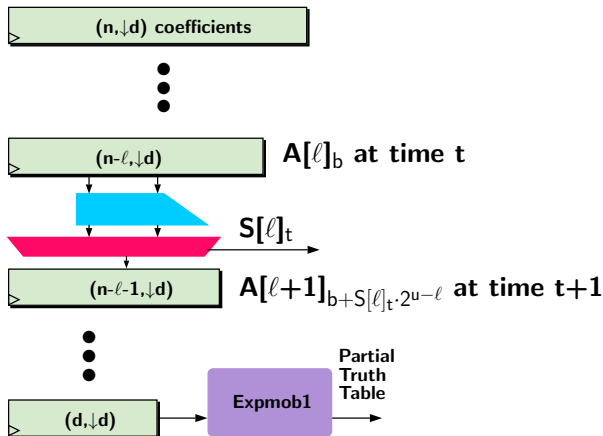
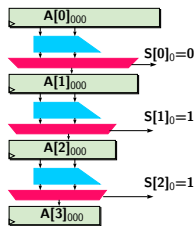


Figure: Hardware architecture **Polymob1** for the Möbius Transform algorithm. The blue shaded part roughly represents one arm of the butterfly unit.

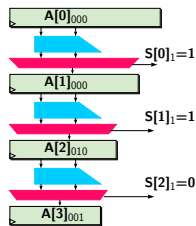
A bit of notation



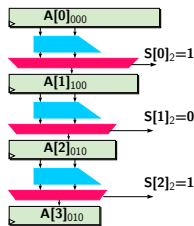
Simulation $n = 5, d = 2$



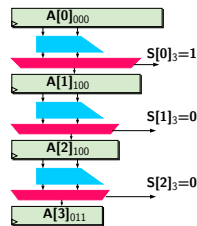
(a) $t=0$



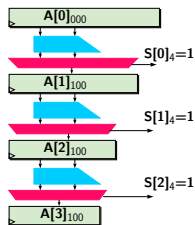
(b) $t=1$



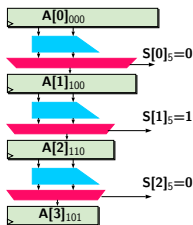
(c) $t=2$



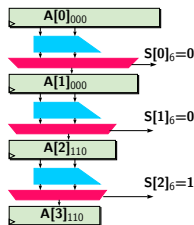
(d) $t=3$



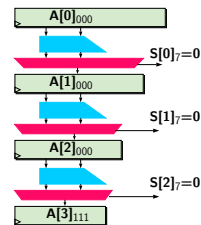
(e) $t=4$



(f) $t=5$



(g) $t=6$



(h) $t=7$

Convert to Set of Equations

t	$\ell = 0$	$\ell = 1$	$\ell = 2$	$\ell = 3$
0	0	0	0	0
1	0	$4 \cdot S[0]_0$	$2 \cdot S[1]_0$	$S[2]_0$
2	0	$4 \cdot S[0]_1$	$4 \cdot S[0]_0 + 2 \cdot S[1]_1$	$2 \cdot S[1]_0 + S[2]_1$
3	0	$4 \cdot S[0]_2$	$4 \cdot S[0]_1 + 2 \cdot S[1]_2$	$4 \cdot S[0]_0 + 2 \cdot S[1]_1 + S[2]_2$
4	0	$4 \cdot S[0]_3$	$4 \cdot S[0]_2 + 2 \cdot S[1]_3$	$4 \cdot S[0]_1 + 2 \cdot S[1]_2 + S[2]_3$
5	0	$4 \cdot S[0]_4$	$4 \cdot S[0]_3 + 2 \cdot S[1]_4$	$4 \cdot S[0]_2 + 2 \cdot S[1]_3 + S[2]_4$
6	0	$4 \cdot S[0]_5$	$4 \cdot S[0]_4 + 2 \cdot S[1]_5$	$4 \cdot S[0]_3 + 2 \cdot S[1]_4 + S[2]_5$
7	0	$4 \cdot S[0]_6$	$4 \cdot S[0]_5 + 2 \cdot S[1]_6$	$4 \cdot S[0]_4 + 2 \cdot S[1]_5 + S[2]_6$

- Left Column needs to be $0, 1, 2, 3, \dots, 7$
- Solve the integer equation system: look for solutions in $\{0, 1\}$

General Case ($u := n - d$)

$$\begin{array}{rcccccccc}
 & & & & & & & 2 \cdot S[u-2]_0 & + S[u-1]_0 & = 1 \\
 & & & & & & & & + S[u-1]_1 & = 2 \\
 & & & & & & & & & \vdots \\
 & & & & & & & & + S[u-1]_i & = i + 1 \\
 & & & & & & & & & \vdots \\
 2^{u-1} \cdot S[0]_0 & + 2^{u-2} \cdot S[1]_1 & + \cdots + & 2^i \cdot S[j]_i & + \cdots & + S[u-1]_{u-1} & = u \\
 2^{u-1} \cdot S[0]_1 & + 2^{u-2} \cdot S[1]_2 & + \cdots + & 2^i \cdot S[j]_{j+1} & + \cdots & + S[u-1]_u & = u + 1 \\
 & & & & & & \vdots \\
 2^{u-1} \cdot S[0]_{2^u - u - 1} & + 2^{u-2} \cdot S[1]_{2^u - u} & + \cdots + & 2^i \cdot S[j]_{-i + 2^u - 2} & + \cdots & + S[u-1]_{2^u - 2} & = 2^u - 1
 \end{array}$$

- Solve the integer equation system: look for solutions in $\{0, 1\}$
- Does Solution exist ? Is solution implementable ?

General Case ($u := n - d$)

$$\begin{array}{rcccccccc}
 & & & & & & & 2 \cdot S[u-2]_0 & + S[u-1]_0 & = 1 \\
 & & & & & & & & + S[u-1]_1 & = 2 \\
 & & & & & & & & & \vdots \\
 & & & & & & & & + S[u-1]_i & = i + 1 \\
 & & & & & & & & & \vdots \\
 2^{u-1} \cdot S[0]_0 & + 2^{u-2} \cdot S[1]_1 & + \cdots + & 2^i \cdot S[j]_j & + \cdots & + S[u-1]_{u-1} & = u \\
 2^{u-1} \cdot S[0]_1 & + 2^{u-2} \cdot S[1]_2 & + \cdots + & 2^i \cdot S[j]_{j+1} & + \cdots & + S[u-1]_u & = u + 1 \\
 & & & & & & & & & \vdots \\
 2^{u-1} \cdot S[0]_{2^u - u - 1} & + 2^{u-2} \cdot S[1]_{2^u - u} & + \cdots + & 2^i \cdot S[j]_{-i + 2^u - 2} & + \cdots & + S[u-1]_{2^u - 2} & = 2^u - 1
 \end{array}$$

- Look at the i -th column shaded in green (note $j = u - 1 - i$)
- $S[j]_t$ is the $i + 1$ -th lsb of $(i + 1), (i + 2), \dots$, i.e. the $(i + 1)$ -th lsb of $t + i + 1$.

Circuit is implementable in logarithmic depth

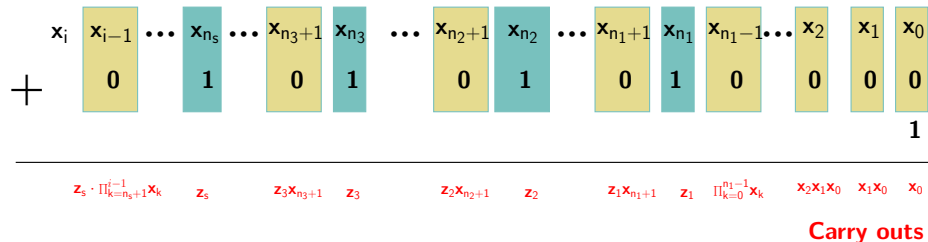


Figure: Visual representation of the addition $t + i + 1$

- Having the whole incrementer circuit is unnecessary.
- We are only interested in $(i + 1)$ -th lsb of $t + i + 1$.
- The expression is $x_i \oplus z_s \prod_{k=n_s+1}^{i-1} x_k$.
- Can be implemented using $2 \log_2 u$ depth.

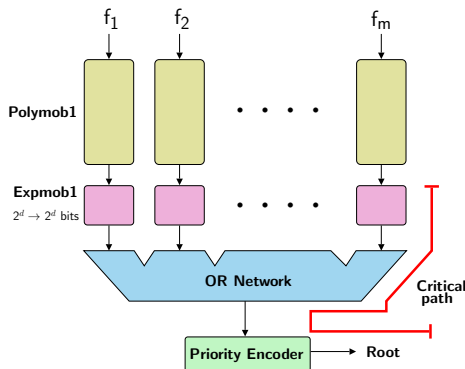


Figure: Hardware Solver **Polysolve1**

- After OR-ing, Priority Encoder gives the location of 1st 0 in the table.
- The solver will extract one root per partial truth table.
- Note large critical path !!

Polysolve2

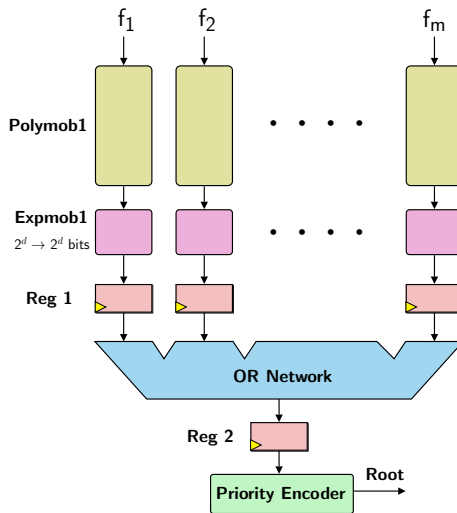
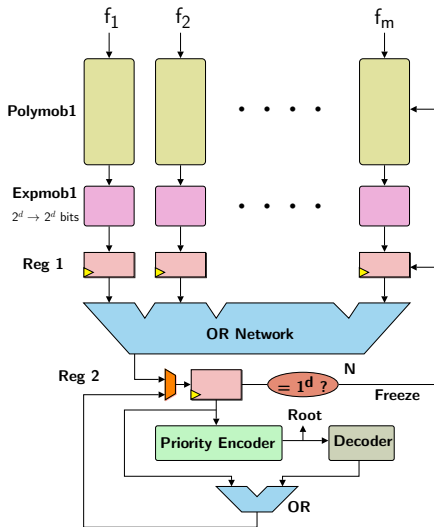


Figure: Hardware Solver **Polysolve2**

- Pipelining reduces the length of critical path.



Example

If $d = 4$, and the **OR** of the truth tables is $T_0 = 1011\ 1111\ 1111\ 0111$

- At $\tau = 0$ Penc outputs 0001
- Decoder op $D_0 = 0100\ 0000\ 0000\ 0000$
- $T_1 = T_0 \vee D_0 = 1111\ 1111\ 1111\ 0111$
- $HW(T_1) = HW(T_0) + 1$, and is written back to **Reg2**.

- At $\tau = 1$ Penc outputs next root 1100
- We have $D_1 = 0000\ 0000\ 0000\ 1000$.
- $T_2 = T_1 \vee D_1 = 1111\ 1111\ 1111\ 1111$ which is now the all one string.

Problem

The critical path of priority encoder+ decoder increases as d increases

- *Task 2: How to reduce it ?*

Problem

The critical path of priority encoder+ decoder increases as d increases

- Task 2: How to reduce it ?

Solution

The Enc+Dec basically flips last 0 from a string

Other solutions exist n OR $n + 1$??

n : 1100 0101 0111

+1

$n+1$: 1100 0101 1000

n : 1100 0101 0111

$n \vee n+1$: 1100 0101 1111

Problem

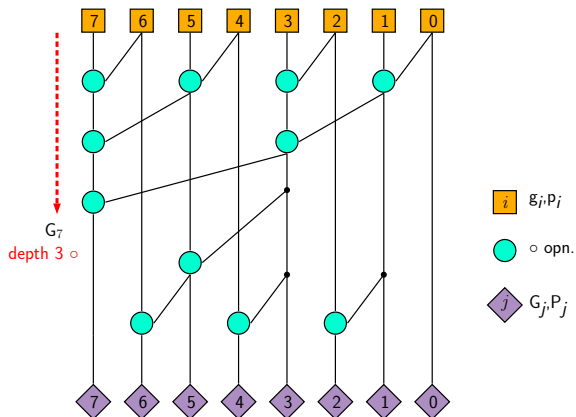
The critical path of priority encoder+ decoder increases as d increases

- *Task 2: How to reduce it ?*

Solution

- In stead of Encoder followed by Decoder, we can do Encoder and n OR $n + 1$ block in parallel.
 - Simultaneously fishes root+ flips zero.

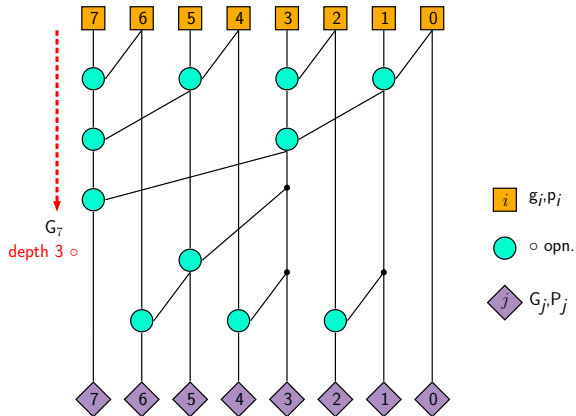
Polysolve3: Brent Kung adder



Generate/propagate

- $g_i = a_i \cdot b_i$ and $p_i = a_i \oplus b_i$
- $G_0, P_0 = g_0, p_0$ and $G_i, P_i = (g_i, p_i) \circ (G_{i-1}, P_{i-1})$
- $(x_1, y_1) \circ (x_2, y_2) = (x_1 \vee (y_1 \cdot x_2), y_1 \cdot y_2)$.

Polysolve3: Brent Kung adder



Generate/propagate

- Has logarithmic depth
- Can be used with carry-select approach
- TO get faster adders for arbitrary d .

Problem

- Plain Circuit takes 2^{n-d} cycles.
- Task 3: How much time does this take if there are R roots ?

Problem

- Plain Circuit takes 2^{n-d} cycles.
- Task 3: How much time does this take if there are R roots ?

Solution

- If partial truth table has $r_i = 0$ roots: no additional cycle.
- If partial truth table has $r_i = 1$ roots: one additional cycle.
- If partial truth table has $r_i = 2$ roots: two additional cycle.
 - Therefore $1 + r_i$ cycles per partial TT

$$\sum_{i=1}^{2^{n-d}} 1 + r_i = 2^{n-d} + \sum r_i = 2^{n-d} + R$$

Wasteful Computation

- Suppose we have 50 equations in 50 variables.
 - The common solution of 1st 10 equations is 100.
 - Evaluating Möbius Transform for the remaining equations ⇒ Evaluating 40 equations at 2^{50} points each.
 - Evaluating 40 equations at 10 points is sufficient !!!!
- We found energy efficient solution for this.
- The idea is to filter any common root of first 10 eqns using Dot-product circuit.

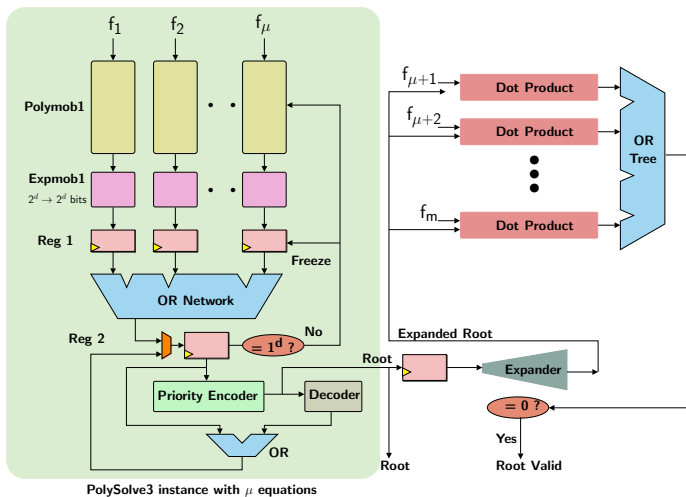
Wasteful Computation

- First run Möbius Transform on a small number of μ equations.
 - Find the common solution set Λ of 1st μ equations.
 - The remaining equations has to be checked only on above set
- So how do we this?
- Each $r \in \Lambda$ has to be evaluated on $m - \mu$ equations.

Circuit components

- Root expander: $RE(n, d): \{0, 1\}^n \rightarrow \{0, 1\}^{\binom{n}{\downarrow d}}$.
 - Eg. $RE(4, 3)$ over the vector $(x_0, x_1, x_2, x_3) = (1, 0, 1, 1)$
 - const value = 1, $x_0x_1 = 0$, $x_0x_2 = 1$, $x_0x_3 = 1$, $x_1x_2 = 0$, $x_1x_3 = 0$, $x_2x_3 = 1$, $x_0x_1x_2 = 0$, $x_0x_1x_3 = 0$, $x_0x_2x_3 = 1$, $x_1x_2x_3 = 0$.
 - The expanded root $\mathbf{r} = 1111\ 0001\ 110$
 - Total hardware overhead is $\binom{n}{\downarrow d} - n$ **AND** gates.
- Dot-Product: Eg $f = 1 \oplus x_0 \oplus x_2 \oplus x_0x_1 \oplus x_2x_3$.
 - Vector Description $\mathbf{v} = 1011\ 0001\ 001$.
 - The dot-product $\mathbf{r} \cdot \mathbf{v} = 0$, equals $f(\mathbf{r})$.
 - $\binom{n}{\downarrow d}$ **AND** gates and $\binom{n}{\downarrow d} - 1$ **XOR**

Circuit



Problem

- *Plain Circuit takes $2^{n-d} + R$ cycles.*
- *Task 3: How much time does this take if there are μ instances ?*

Time taken

Problem

- Plain Circuit takes $2^{n-d} + R$ cycles.
- Task 3: How much time does this take if there are μ instances ?

Lemma

Let f_1, f_2, \dots, f_μ be iid balanced Boolean functions of n variables each. Then the expected cardinality of the solution space of the system of equations $f_1 = f_2 = \dots = f_\mu = 0$ is $2^{n-\mu}$.

- So $R + 2^{n-d} = 2^{n-\mu} + 2^{n-d}$ cycles on average.

Energy

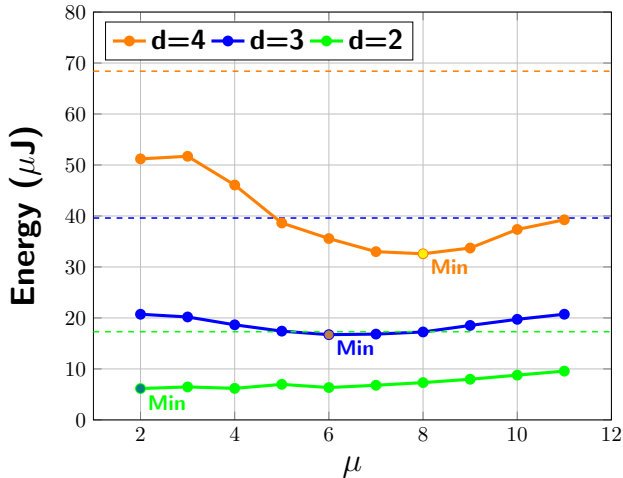


Figure: Energy consumption for varying μ for $n = m = 20$. The colored dashed lines show the energy consumed in the **Polysolve3** circuit for the corresponding equation systems.

Depth Bound trees

Algorithm 2: Recursive Möbius Transform

Möbius (A_0, n, d)

Input: A_0 : The compressed ANF vector of a Boolean function f

Input: n : Number of variables, d : Algebraic degree

Output: The Truth table of f

```
/* Final step, i.e. leaf nodes of recursion tree */
if  $n=d$  then
    | Use the formula  $B = M_n \cdot A_0$  to output partial truth table  $B$ .
    | /* Use either Expmob1/Expmob2 to do this */
end
else
    | Declare an array  $T$  of size  $\binom{n-1}{\downarrow d}$  bits.
    | /* Compute the 2 operations of the butterfly layer */
1 | Store 1st butterfly output i.e.  $A_{\text{top}}$  in  $T$  (requires no xors).
    | Call Möbius ( $T, n-1, d$ )
2 | Store 2nd butterfly output i.e.  $A_{\text{bottom}}$  in  $T$  (requires some xors).
    | Call Möbius ( $T, n-1, d$ )
end
```

Algorithm 3: Recursive Möbius Transform

Möbius (A_0, n, d)

Input: A_0 : The compressed ANF vector of a Boolean function f

Input: n : Number of variables, d : Algebraic degree

Output: The Truth table of f

```
/* Final step, i.e. leaf nodes of recursion tree */
if  $n=h>d$  then
    | Use the formula  $B = M_n \cdot A_0$  to output partial truth table  $B$ .
    | /* Use either Expmob1/Expmob2 to do this */
end
else
    | Declare an array  $T$  of size  $\binom{n-1}{\downarrow d}$  bits.
    | /* Compute the 2 operations of the butterfly layer */
1 | Store 1st butterfly output i.e.  $A_{\text{top}}$  in  $T$  (requires no xors).
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    | Call Möbius ( $T, n-1, d$ )
end
```

Depth boundedness

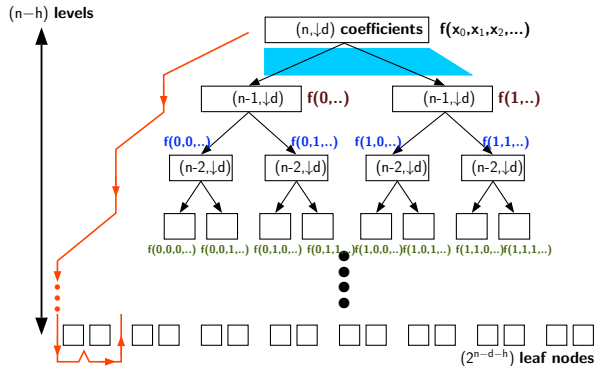


Figure: Hardware architecture **Polymob1** for the Möbius Transform algorithm. The blue shaded part roughly represents one arm of the butterfly unit.

- Number of levels shrink to $n - h$ from $n - d$.
- Faster computation: $2^{n-\mu} + 2^{n-d} \rightarrow 2^{n-\mu} + 2^{n-h}$.

Depth boundedness

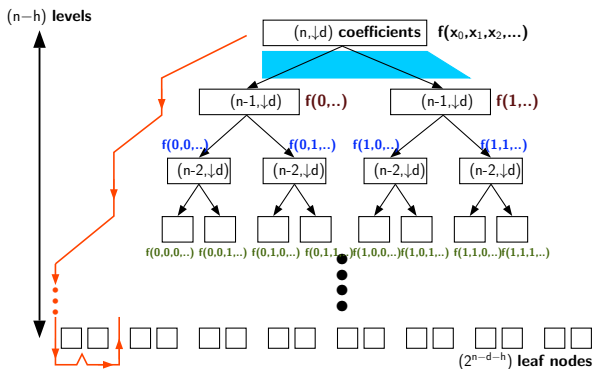


Figure: Hardware architecture **Polymob1** for the Möbius Transform algorithm. The blue shaded part roughly represents one arm of the butterfly unit.

- Downside **Expmob1**, encoder needed over $h > d$ bits.
- Increases the critical path.

Energy reduction

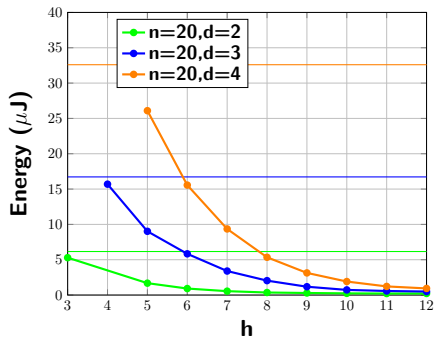


Figure: Energy decrease with increasing h for new solvers for $n = 20$, $h = \mu$. The colored horizontal lines indicate the best possible energy consumption for the full depth circuit for the same equation system.

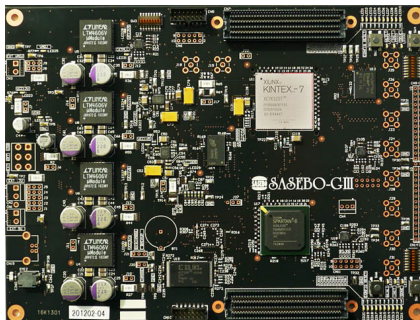


Figure: SAKURA-X

Proof of Concept

- SAKURA-X mainly built for side-channel experiments, limited computational power.
- We could solve quadratic equations of upto 50 variables in 8 hours.
- TODO → Implement on an FPGA cluster and solve upto 100 variables.

Conclusion



- Given m equations in n variables over $GF(2)$.
- Asymptotically, all the solutions can be found using a circuit of area $\propto m \cdot n^{d+2}$.
- This is not energy-efficient however: Möbius Transform does a lot of redundant computations.
- Circuit for energy efficiency also proposed.

THANK YOU