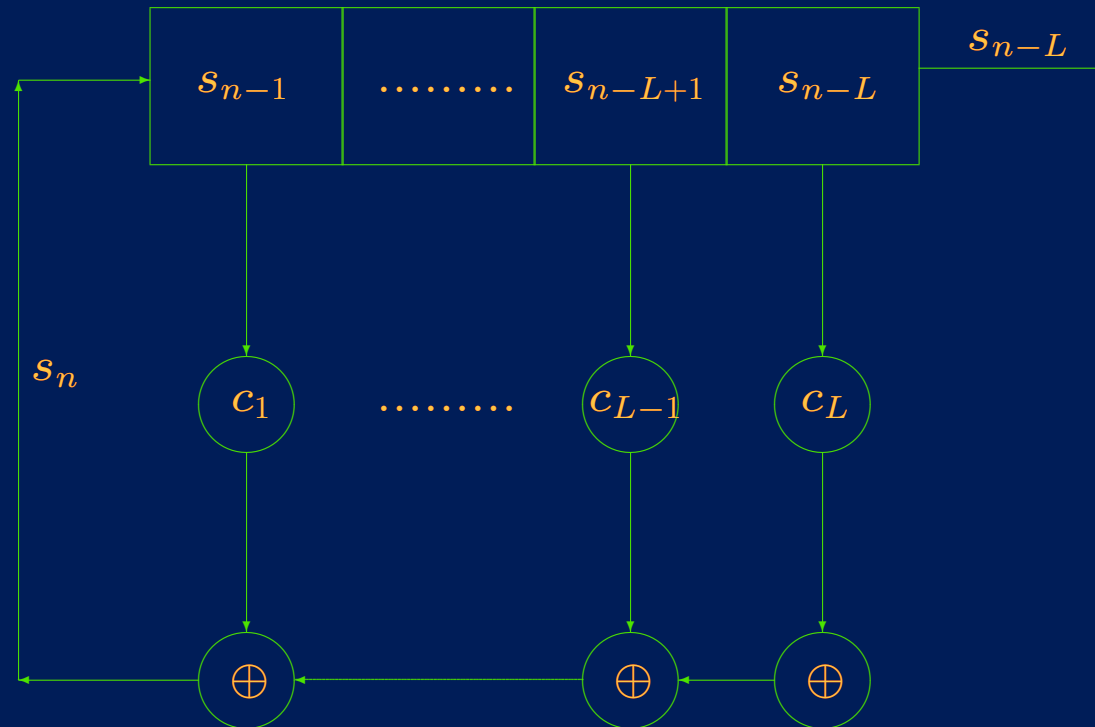


Counting first order correlation-immune functions

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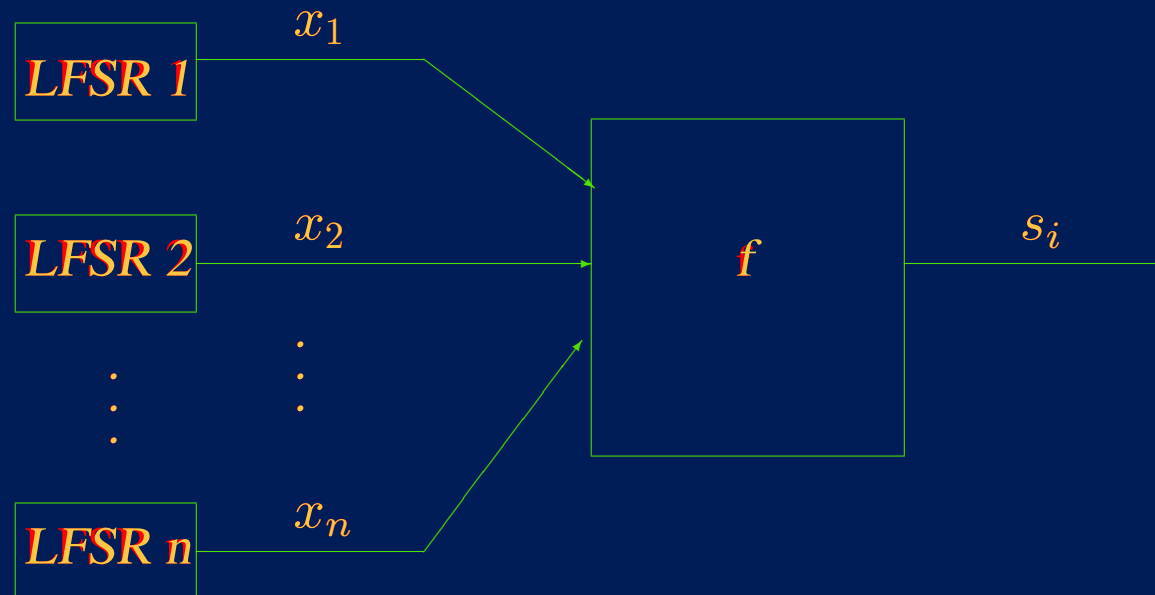
Linear Feedback Shift Registers (LFSR)



- At each clock-cycle computes $\bigoplus_{i=1}^L c_i s_{n-i}$ and outputs s_{n-L} .
- Generates an ultimately periodic sequence with period at most $2^L - 1$.
- The **linear complexity** of such a sequence is the length L of the minimum LFSR producing the same sequence.

Boolean functions and LFSR

- LFSR are cryptographically weak.
 - If L is the linear complexity of a sequence (unknown from the attacker), with $2L$ consecutive bits known, the **Berlekamp-Massey** algorithm *recovers* L , c_i 's and initial values (**key bits**).
 - In practice the attacker only needs to know around 20 consecutive bits.
- **Combining boolean functions** are used to avoid this attack.
- Period at most the LCM of the periods of the sequences generated by the LFSRs.
- Length of the key is $L_1 + L_2 + \dots + L_n$.



Cryptographic criteria for boolean functions

- High algebraic degree.
- Large Hamming distance to all affine functions.
- Balanced functions.
- Correlation-immune and resilient functions to prevent correlation attacks [Siegenthaler 1984; Meier & Staffelbach 1988; Johansson & Jönsson 1999, 2000; Canteaut & Trabbia 2000; and more ...].
- Strict avalanche criterion [Webster & Tavares 1985].
- Propagation criterion [Preneel, Van Leekwijck, Van Linden, Govaerts & Vandevalle 1991].
- [Carlet 2007] has a complete survey on boolean functions for cryptography and error correcting codes.
- Another useful source of information is [Gouguet 2004].

Decomposition of boolean functions

- A boolean function in n variables is a function $f : \{0, 1\}^n \rightarrow \{0, 1\}$.
- We encode a boolean function f_n by a word in $\{0, 1\}^{2^n}$ indexed by $x_n \dots x_1$.
- We may see f_n as the concatenation of 2^{n-j} boolean functions in j variables each, by an operator we call \star .

Example:

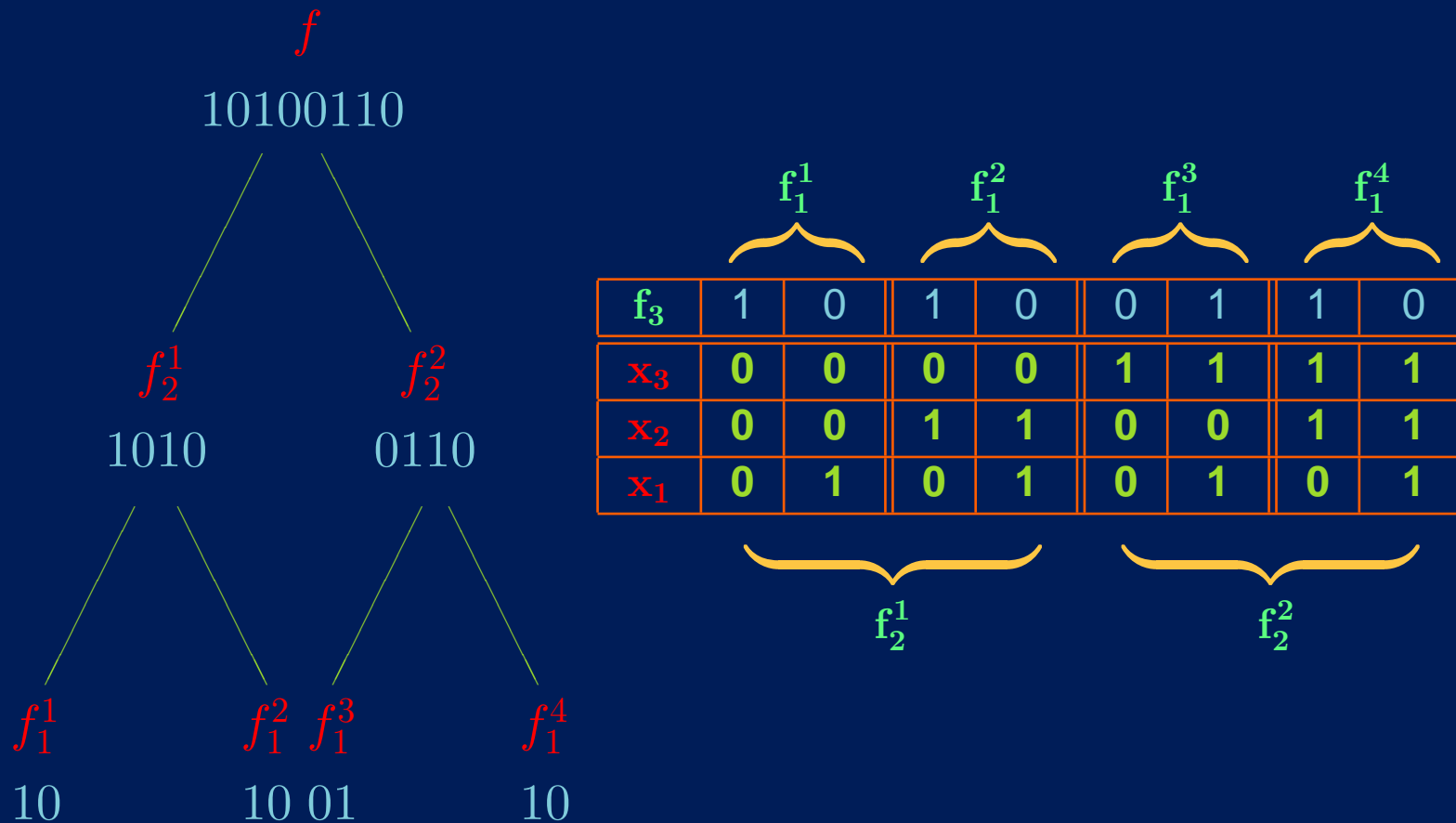
	f_1^1		f_1^2		f_1^3		f_1^4	
f_3	1	0	1	0	0	1	1	0
x_3	0	0	0	0	1	1	1	1
x_2	0	0	1	1	0	0	1	1
x_1	0	1	0	1	0	1	0	1
	f_2^1				f_2^2			

- We have $f_3 = f_2^1 \star f_2^2$ and $f_3 = f_1^1 \star f_1^2 \star f_1^3 \star f_1^4$.
- Functions f_j^k may be seen as f_n **conditioned** to some arguments x_i being some **fixed** a_i .
- Then, we have $f_1^3 = f_3 |_{x_2=0, x_3=1}$, $f_2^1 = f_3 |_{x_3=0}$ and so on.

Tree decomposition of a boolean function

- Decomposition can be described by a complete binary tree of depth $n - 1$ being f_n is the root and the 2^{n-1} functions in 1 variable the leaves.

Example:

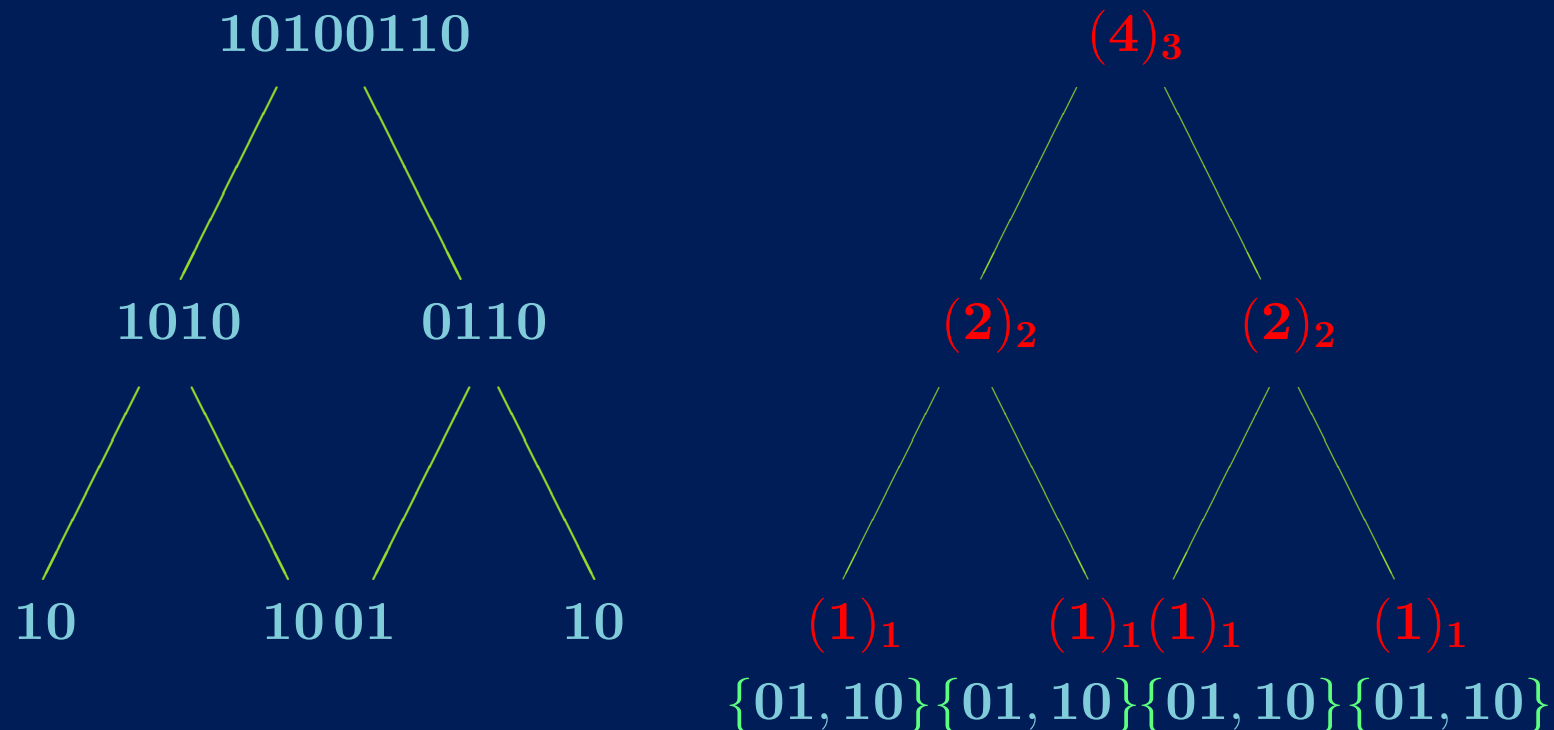


Hamming tree of boolean classes

- The sets $(i)_n = \{f_n \mid w_H(f_n) = i\}$ **partition** $\{0, 1\}^{2^n}$ into **Hamming classes**.
- Similarly to boolean functions we have **Hamming trees** of boolean **classes**.
- Each tree may be **shared** by several functions.

Example:

- The Hamming tree of f_3 is shared by 16 functions.
- These functions are constructed by combining 01 and 10 in all possible ways at the leaves.

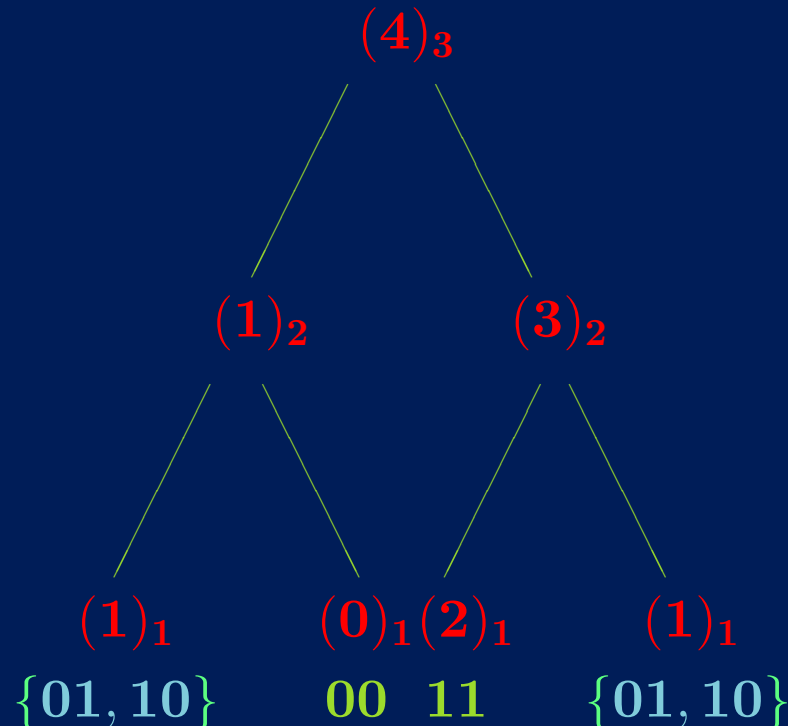


Hamming tree of boolean classes (cont.)

- Not every Hamming tree is shared by the same number of functions.

Example:

- This Hamming tree is shared by only 4 functions, since the class $(0)_1$ only contains the function 00 and $(2)_1$ only contains the function 11.
- These functions are 01001101, 01001110, 10001101 and 10001110.
- Nevertheless these functions are balanced like f_3 (belong to the same Hamming class), although their respective Hamming trees are different.
- **Hamming trees** (and not Hamming classes!) capture the **essential** features for our problem.



Equivalence relation for first-order correlation-immunity

- We define $\delta_i(f_n) = w_H(f_n |_{x_i=0}) - w_H(f_n |_{x_i=1}), 1 \leq i \leq n$.
- Then, f_n is **first-order correlation immune** $\iff \forall i, \delta_i(f_n) = 0$.
- Moreover, f_n is **1-resilient** $\iff \forall i, \delta_i(f_n) = 0, w_H(f_n) = 2^{n-1}$.
- A function f_n belongs to the **class** $\omega = \Omega(f_n) = \langle w_H(f_n), \delta_n(f_n) \dots \delta_1(f_n) \rangle$.

How to find $\Omega(f_3)$

f_3	1	0	1	0	0	1	1	0
x_3	0	0	0	0	1	1	1	1
x_2	0	0	1	1	0	0	1	1
x_1	0	1	0	1	0	1	0	1

$$\Omega(f_3) = \langle 4, 0, , \rangle.$$

How to find $\Omega(f_3)$

f_3	1	0	1	0	0	1	1	0
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x_1	0	1	0	1	0	1	0	1

$$\Omega(f_3) = \langle 4, 0, , \rangle.$$

f_3	1	0	1	0	0	1	1	0
x_3	0	0	0	0	1	1	1	1
x_2	0	0	1	1	0	0	1	1
x_1	0	1	0	1	0	1	0	1

$$\Omega(f_3) = \langle 4, 0, 0, \rangle.$$

How to find $\Omega(f_3)$

f_3	1	0	1	0	0	1	1	0
x_3	0	0	0	0	1	1	1	1
x_2	0	0	1	1	0	0	1	1
x_1	0	1	0	1	0	1	0	1

$$\Omega(f_3) = \langle 4, 0, , \rangle.$$

f_3	1	0	1	0	0	1	1	0
x_3	0	0	0	0	1	1	1	1
x_2	0	0	1	1	0	0	1	1
x_1	0	1	0	1	0	1	0	1

$$\Omega(f_3) = \langle 4, 0, 0, \rangle.$$

f_3	1	0	1	0	0	1	1	0
x_3	0	0	0	0	1	1	1	1
x_2	0	0	1	1	0	0	1	1
x_1	0	1	0	1	0	1	0	1

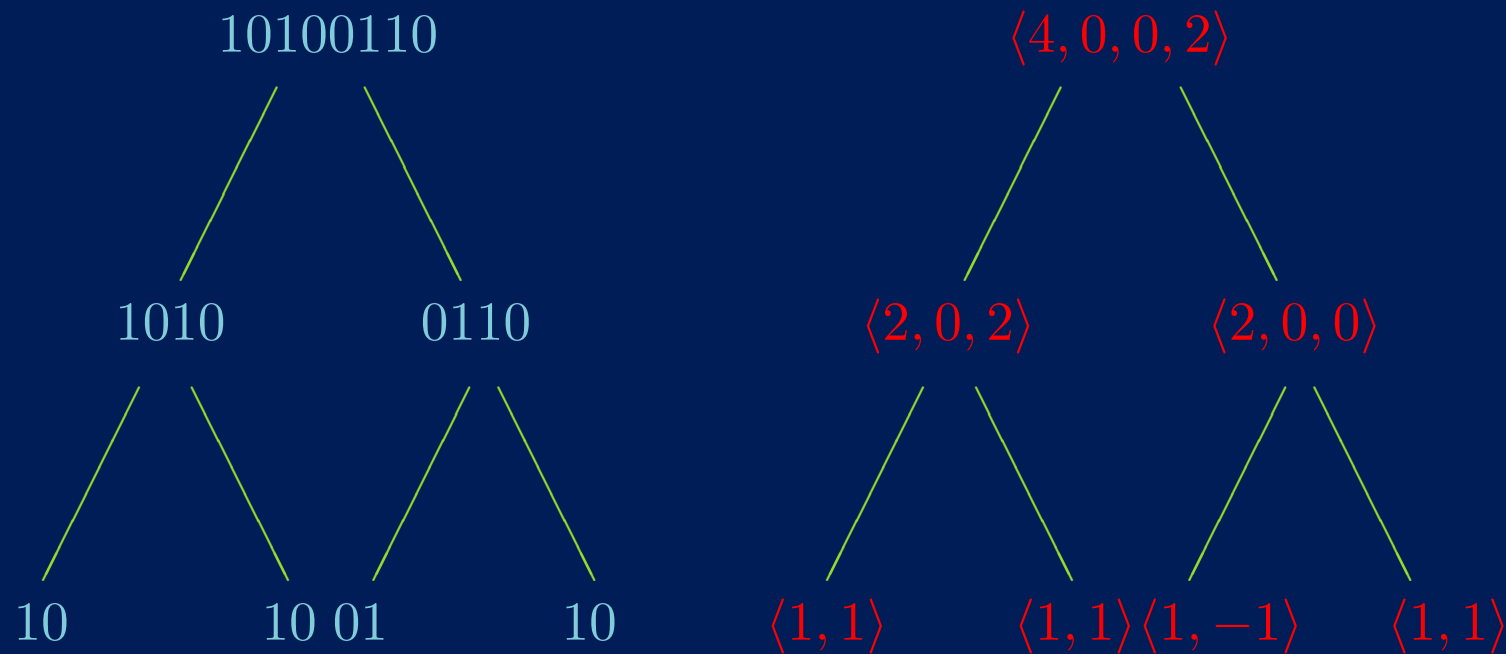
$$\Omega(f_3) = \langle 4, 0, 0, 2 \rangle.$$

First order correlation-immune classes

- Similar to balanced functions we have **first order correlation trees**.

Example:

- This is the first order correlation tree for f_3 .



Recursive construction of correlation classes

Proposition 1 (Recursive construction).

Let

$$\begin{cases} \omega_1 & = \langle p_1, \delta_{n-1}^1, \dots, \delta_1^1 \rangle \in \Omega_{n-1}^{p_1}, \\ \omega_2 & = \langle p_2, \delta_{n-1}^2, \dots, \delta_1^2 \rangle \in \Omega_{n-1}^{p_2}, \\ \omega & = \langle m, \delta_n, \dots, \delta_1 \rangle = \omega_1 \star \omega_2. \end{cases}$$

Then we have

$$\begin{cases} m & = p_1 + p_2 \\ \delta_n & = p_1 - p_2 \\ \delta_i & = \delta_i^1 + \delta_i^2, \quad i \in \{1, \dots, n-1\}. \end{cases}$$

Decomposition of correlation classes.

Theorem 1 (Decomposition of correlation classes).

Let $\omega \in \Omega_n$. Then, with $\omega_1, \omega_2 \in \Omega_{n-1}$ we have

$$\omega^s = \bigcup_{\omega_1 \star \omega_2 = \omega} \omega_1^s \times \omega_2^s.$$

Theorem 2 (Counting correlation functions).

Let $\omega \in \Omega_n$. Then, with $\omega_1, \omega_2 \in \Omega_{n-1}$ we have

$$|\omega^s| = \sum_{\omega_1 \star \omega_2 = \omega} |\omega_1^s| \cdot |\omega_2^s|.$$

Decomposition of correlation-immune classes.

- Let $\omega_1 \in \Omega_{n-1}^{p_1}$, $\omega_2 \in \Omega_{n-1}^{p_2}$ and $m = p_1 + p_2$. From our recursive construction we have the equivalence $\omega_1 \star \omega_2 \in \text{Cor}_n^m \iff \omega_2 = \overline{\omega_1}$.

Theorem 3 (Decomposition of correlation-immune classes).

$$\text{Cor}_n^m = \bigcup_{\omega_1 \in \Omega_{n-1}^m} \omega_1^s \times \overline{\omega_1}^s, \text{ for } 0 \leq m \leq 2^{n-1}.$$

Theorem 4 (Counting correlation-immune functions).

$$|\text{Cor}_n^m| = \sum_{\omega_1 \in \Omega_{n-1}^m} |\omega_1^s|^2,$$

$$|\text{Cor}_n| = \sum_{\omega_1 \in \Omega_{n-1}} |\omega_1^s|^2.$$

Counting 1-resilient boolean functions.

- We denote by \mathcal{B}_n the set of **balanced first-order correlation classes** with n variables.
- We have $|\mathcal{B}_n| = \binom{2^n}{2^{n-1}}$.

Corollary 4 (Counting 1-resilient boolean functions).

Since $Res1_n = Cor_n^{2^{n-2}}$, we have

$$|Res1_n| = \sum_{\omega_1 \in \mathcal{B}_{n-1}} |\omega_1|^2.$$

- Then, to compute $Res1_n$ we only need to know the cardinality of all balanced first-order correlation classes with $n - 1$ variables.
- We find an efficient algorithm by working with correlation classes and not with correlation functions.

n	5	6	7
$Res1_n$	807980	95259103924394	23478015754788854439497622689296
Time	0.028 s	0.526 s	1 h 02 min 27.332 s

Normalized classes

- Let $m \leq 2^{n-1}$ and $\omega = \langle \mathbf{m}, \delta_n, \dots, \delta_1 \rangle \in \Omega_n^m$. There exists a **permutation** $\sigma : \{1, \dots, n\} \rightarrow \{1, \dots, n\}$ which satisfies

$$\begin{cases} \alpha_i = |\delta_{\sigma(i)}| \\ \alpha_n \leq \alpha_{n-1} \leq \dots \leq \alpha_1. \end{cases}$$

- The class $\mathbf{N}(\omega) = \theta = \langle \mathbf{m}, \alpha_n, \dots, \alpha_1 \rangle$ will be called the **normalised class** of ω .
- Every Boolean function in ω may be **transformed** in a **unique** Boolean function in θ .

Example:

- $N(\langle 7, 1, 5, -3, -3 \rangle) = \langle 7, 1, 3, 3, 5 \rangle$.

New characterization of 1-resilient functions.

- Each set ω^s corresponding to a class ω with $\mathbf{N}(\omega) = \theta$ has the **same cardinality as** θ^s . Then,

Theorem 5 (Number of 1-resilient functions).

$$|Res1_n| = \sum_{\theta \in \Theta_n^{2^n-2}} n(\theta) |\theta^s|^2.$$

- **Normalized classes** help to still **speed up** our counting algorithm and find the number of 1-resilient functions for $n = 7$ in **50 seconds!**.
- By only computing a **fraction** of all normalized classes, we obtain the **lower bound** $4 \cdot 10^{67}$ for the number of 1-resilient functions with $n = 8$ variables.

Upper bounds on the number of first-order correlation classes

- Let $\omega = \langle m, \delta_n, \dots, \delta_1 \rangle \in \Omega_n^m$. Then we define $\delta(\omega) = \sum_{i=1}^n |\delta_i|$.
- We may see $\delta(\omega)$ as a measure of *how far from first-order correlation* is ω .
- Define $\delta(\mathbf{n}, \mathbf{m}) = \sup_{\omega \in \Omega_n^m} \delta(\omega)$, $\mu(\mathbf{n}, \mathbf{m}) = \delta(\mathbf{n}, \mathbf{m})/2$ if m is even and $\mu(\mathbf{n}, \mathbf{m}) = (\delta(\mathbf{n}, \mathbf{m}) - \mathbf{n})/2$ otherwise.

Lemma 3 (Upper bound for number of classes)

Let $0 \leq m \leq 2^{n-1}$ and U_n^m defined by

$$U_n^m = \begin{cases} \sum_{j=0}^n \binom{\mu(n,m)}{j} \binom{\mu(n,m)+n-j}{\mu(n,m)} & m \text{ even} \\ \binom{\mu(n,m) + n}{n} 2^n & \text{otherwise.} \end{cases}$$

- U_n^m is an **upper bound** for the **number of classes** in Ω_n^m .

New lower bound on the number of 1-resilient functions

Theorem 7 $|Res1_n| \geq \frac{\binom{2^{n-1}}{2^{n-2}}^2}{U_{n-1}^{2^{n-2}}} \geq \frac{\binom{2^{n-1}}{m}^2}{\binom{\mu(n-1, 2^{n-2}) + n - 1}{n-1} 2^{n-1}}.$

Theorem 8 $|Res1_n| \geq \frac{2^{2^n} (n\pi)^{n/2}}{2^{n^2 - \frac{3}{2}n - 1} e^{n-1/2}}.$

- [Maitra & Sarkar 1999]:

$$|Res1_n| \geq 2^{2^{n-2}} + \binom{2^{n-1}}{2^{n-2}} + \binom{2^{n-2}}{2^{n-3}} * \left(\binom{2^{n-2}}{2^{n-3}} - 2 \right) + \binom{2^{n-3}}{2^{n-4}} - 2^{2^{n-3}}.$$

n	[Maitra & Sarkar 1999]	Our lower bound
5	17876	503430
6	7.667 10^8	7.523 10^{12}
7	2.193 10^{18}	1.312 10^{29}
8	2.730 10^{37}	1.134 10^{64}
9	6.342 10^{75}	8.884 10^{136}
10	5.058 10^{152}	2.128 10^{286}

- Dramatic improvements are due to our **general construction**. The previous bounds have been found by building and counting **restricted classes**.

New lower bound on the number of k -resilient functions

- Given a 1-resilient function in $n - k + 1$ variables, we have a construction that leads to a k -resilient function in n variables.
- As a consequence we have the following new lower bound for the number of k -resilient functions:

Theorem 9 Let $k \geq 2$, and $n > k$. The set of k -resilient functions with n variables satisfies $|\mathbf{Res}_{n-k+1}^1| \leq |\mathbf{Res}_n^k|$.

n	Maiorana-McFarland [Camion, Carlet, Charpin & Sendrier 1991]	Our lower bound
\mathcal{Res}_{10}^1	$3.0 \cdot 10^{79}$	$5.1 \cdot 10^{285}$
\mathcal{Res}_{10}^2	$4.3 \cdot 10^{40}$	$3.4 \cdot 10^{136}$
\mathcal{Res}_{10}^3	$1.2 \cdot 10^{21}$	$2.6 \cdot 10^{63}$
\mathcal{Res}_{10}^4	$1.4 \cdot 10^{11}$	$2.3 \cdot 10^{31}$
\mathcal{Res}_{10}^5	$1.1 \cdot 10^6$	$9.5 \cdot 10^{13}$

Summary of results

- We present a complete characterization of 1-correlation immune functions and give efficient algorithms to generate and count them.
- The number of 1-resilient functions in **7** variables is **23478015754788854439497622689296**.
- We drastically improve known bounds specially lower bounds.

n	8	9	10	11	12	13
<i>Maitra</i>	10^{37}	10^{75}	10^{152}	10^{306}	10^{614}	10^{1231}
<i>New Lower Bound</i>	10^{64}	10^{136}	10^{286}	10^{589}	10^{1199}	10^{2426}
<i>New Upper Bound</i>	10^{68}	10^{144}	10^{297}	10^{603}	10^{1218}	10^{2449}
<i>Schneider</i>	10^{71}	10^{147}	10^{299}	10^{606}	10^{1221}	10^{2452}

- We conjecture that the *probability* of a boolean function being *1-resilient* is

$$\sim \frac{2^{-\frac{n^2}{2}}}{\sqrt{2\pi \frac{n+1}{2}}}$$

- Use of the generating function derived from our constructions. Work in progress with **P. Flajolet**, **S. Mesnager**.