
Lecture Notes on Cryptographic Boolean Functions

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Chapter 1

Boolean functions

1.1 Boolean functions and their representations

Definition 1.1 (Boolean function). *A Boolean function of n variables is a function from \mathbb{F}_2^n into \mathbb{F}_2 . Its value vector is the binary vector v_f of length 2^n composed of all $f(x)$ when $x \in \mathbb{F}_2^n$.*

1.1.1 Truth table and Algebraic normal form

A Boolean function is usually defined by its *truth table*, which gives the images of all elements in \mathbb{F}_2^n . For instance, Table 1.1 is the truth table of a Boolean function of 3 variables. The

x_1	0	1	0	1	0	1	0	1
x_2	0	0	1	1	0	0	1	1
x_3	0	0	0	0	1	1	1	1
$f(x_1, x_2, x_3)$	0	1	0	0	0	1	1	1

Table 1.1: Truth table of a Boolean function of 3 variables.

value vector of f is the vector of \mathbb{F}_2^8 corresponding to the last row in the truth table.

Boolean functions are often identified with their value vectors. In particular, the weight and the support of a Boolean function f refer to the weight and the support of its value vector v_f . Most cryptographic applications use *balanced* Boolean functions, i.e., Boolean functions f whose output is uniformly distributed. This equivalently means that the weight of v_f is half of its length.

Besides the truth table, there are several other representations of Boolean functions which may be more appropriate in some contexts. In coding theory and in cryptography, a very natural representation is the so-called *algebraic normal form* (ANF), which corresponds to the expression of a Boolean function as a multivariate polynomial. Since the n inputs of the function take their values in \mathbb{F}_2 , they must be considered modulo $X^2 + X$. Therefore, this polynomial has degree at most 1 in each input variable. It follows that any monomial of this polynomial is the product of some input variables. Each monomial can then be characterized by a subset of $I = \{1, \dots, n\}$, i.e., $\prod_{i \in I} x_i$, or equivalently by the n -bit vector u having I as support. This second notation will be extensively used in the context of Boolean functions.

Notation 1.2. For any $u \in \mathbb{F}_2^n$, x^u denotes the monomial in $\mathbb{F}_2[x_1, \dots, x_n]/(x_1^2 + x_1, \dots, x_n^2 + x_n)$ defined by

$$\prod_{i=1}^n x_i^{u_i}.$$

The following theorem then shows that any Boolean function can be uniquely represented by a multivariate polynomial, and it also gives a simple formula for computing this polynomial from the value vector of the function.

Theorem 1.3 (Algebraic normal form). *Let f be a Boolean function of n variables. Then, there exists a unique multivariate polynomial in $\mathbb{F}_2[x_1, \dots, x_n]/(x_1^2 + x_1, \dots, x_n^2 + x_n)$ such that*

$$f(x_1, \dots, x_n) = \sum_{u \in \mathbb{F}_2^n} a_u x^u, \text{ with } a_u \in \mathbb{F}_2.$$

This multivariate polynomial is called the algebraic normal form (ANF) of f .

Moreover, the coefficients of the ANF and the values of f satisfy:

$$a_u = \sum_{x \preceq u} f(x) \text{ and } f(u) = \sum_{x \preceq u} a_x,$$

where the sums are in \mathbb{F}_2 and $x \preceq y$ if and only if $x_i \leq y_i$ for all $1 \leq i \leq n$.

Proof. We first show by induction on n that the ANF of an n -variable Boolean function can be uniquely computed from its truth table.

- For $n = 1$, it is easy to check that the polynomial $a_1x + a_0$ with $a_1 = f(0) + f(1)$ and $a_0 = f(0)$ is the unique polynomial equal to f .
- Induction step. Given an n -variable function f , we consider the two $(n-1)$ -variable Boolean functions g and h defined by

$$g(x_1, \dots, x_{n-1}) = f(x_1, \dots, x_{n-1}, 0) \text{ and } h(x_1, \dots, x_{n-1}) = f(x_1, \dots, x_{n-1}, 1).$$

Then, we have

$$f(x_1, \dots, x_n) = (1 + x_n)g(x_1, \dots, x_{n-1}) + x_nh(x_1, \dots, x_{n-1})$$

or equivalently,

$$f(x_1, \dots, x_n) = g(x_1, \dots, x_{n-1}) + x_n(g(x_1, \dots, x_{n-1}) + h(x_1, \dots, x_{n-1})).$$

We apply the induction hypothesis and denote by α_u (resp. β_u) for $u \in \mathbb{F}_2^{n-1}$ the coefficients of the ANF of g (resp. of h). We know that

$$\alpha_u = \sum_{x \preceq u} g(x) = \sum_{x \preceq u} f(x, 0) \text{ and } \beta_u = \sum_{x \preceq u} h(x) = \sum_{x \preceq u} f(x, 1).$$

We then deduce that

$$\begin{aligned} f(x_1, \dots, x_n) &= g(x_1, \dots, x_{n-1}) + x_n(g(x_1, \dots, x_{n-1}) + h(x_1, \dots, x_{n-1})) \\ &= \sum_{u \in \mathbb{F}_2^{n-1}} \alpha_u \prod_{i=1}^{n-1} x_i^{u_i} + \sum_{u \in \mathbb{F}_2^{n-1}} (\alpha_u + \beta_u) \left(\prod_{i=1}^{n-1} x_i^{u_i} \right) x_n. \end{aligned}$$

Therefore, the coefficients a_v , $v = (v_1, \dots, v_n) \in \mathbb{F}_2^n$, of the ANF of f are given by

$$a_v = \begin{cases} \alpha_{v_1, \dots, v_{n-1}} & \text{if } v_n = 0 \\ \alpha_{v_1, \dots, v_{n-1}} + \beta_{v_1, \dots, v_{n-1}} & \text{if } v_n = 1 \end{cases}$$

From the expressions of the coefficients α and β , we deduce that

$$a_v = \begin{cases} \sum_{u \preceq (v_1, \dots, v_{n-1})} f(u, 0) & \text{if } v_n = 0 \\ \sum_{u \preceq (v_1, \dots, v_{n-1})} f(u, 0) + \sum_{u \preceq (v_1, \dots, v_{n-1})} f(u, 1) & \text{if } v_n = 1 \end{cases}$$

implying that

$$a_v = \sum_{u \preceq v} f(u) .$$

Conversely, the values of f are uniquely determined by its ANF since the function over $\mathbb{F}_2^{2^n}$ which maps the value vector of f to the vector of coefficients of its ANF is an involution. Indeed, for any $y \in \mathbb{F}_2^n$, we have

$$\begin{aligned} \sum_{u \preceq y} a_u &= \sum_{u \preceq y} \sum_{x \preceq u} f(x) \\ &= \sum_{x \preceq y} f(x) |\{u \in \mathbb{F}_2^n : x \preceq u \preceq y\}| \\ &= \sum_{x \preceq y} 2^{wt(y) - wt(x)} f(x) . \end{aligned}$$

All terms in this sum are then zero modulo 2 unless $x = y$. Thus

$$\sum_{u \preceq y} a_u = f(y) ,$$

which means that the transformation we consider is an involution. \diamond

Example 1.1. Computing the ANF of the function defined in Table 1.1. Using the previous theorem, we compute the coefficients of the ANF of this Boolean function:

$$\begin{aligned} a_{000} &= f(000) = 0 \\ a_{100} &= f(100) + f(000) = 1 \\ a_{010} &= f(010) + f(000) = 0 \\ a_{110} &= f(110) + f(010) + f(100) + f(000) = 1 \\ a_{001} &= f(001) + f(000) = 0 \\ a_{101} &= f(101) + f(001) + f(100) + f(000) = 0 \\ a_{011} &= f(011) + f(001) + f(010) + f(000) = 1 \\ a_{111} &= \sum_{x \in \mathbb{F}_2^3} f(x) = wt(f) \bmod 2 = 0 . \end{aligned}$$

Thus, the ANF of f is

$$x_1 + x_1x_2 + x_2x_3 .$$

The *degree* of f is then the degree of the largest monomial in the ANF of f , i.e.,

$$\deg f = \max_{u \in \mathbb{F}_2^n : a_u \neq 0} wt(u) .$$

For instance the function considered in the previous example has degree 2.

It is worth noticing that there exist several other representations of Boolean functions which may be more convenient than the ANF in some other contexts. For instance, the disjunctive normal form represents the function by some products between variables and negations of variables, which are added by an OR. A disjunctive normal form of the function defined in Table 1.1 is

$$x_1 \overline{x_2} \overline{x_3} \text{ OR } x_1 \overline{x_2} x_3 \text{ OR } \overline{x_1} x_2 x_3 \text{ OR } x_1 x_2 x_3 .$$

Such a representation may be more appropriate than the ANF when we want to determine the smallest circuit which implements the function in a context where only AND, NOT and OR gates are available, see [Weg87] for more details.

1.1.2 Computing the Algebraic Normal Form

The general form of the transformation which associates the coefficients of the ANF to the value vector is

$$\begin{aligned} \mathcal{M}_n : \quad \mathbb{F}_2^{2^n} &\rightarrow \mathbb{F}_2^{2^n} \\ a = (a_u, u \in \mathbb{F}_2^n) &\mapsto (b_u, u \in \mathbb{F}_2^n) \quad \text{with } b_u = \sum_{v \preceq u} a_v . \end{aligned}$$

This function is called *the binary Möbius transform*. Indeed, Möbius inversion is a method for inverting sums over a partially ordered sets. This general inversion formula appears in many contexts in combinatorics. For instance, it leads to the principle of inclusion-exclusion and to the expression of Euler ϕ -function [Rot64, Moe12]. In our context, we have proved in Theorem 1.3 that the transformation \mathcal{M}_n is an involution, so any algorithm for computing the binary Möbius transform can be used both for computing the ANF from the value vector and for computing the value vector from the ANF.

The naive method for computing $\mathcal{M}_n(a)$ consists in evaluating the sum of the coordinates a_v of a over all positions $v \preceq u$ for the 2^n successive elements $u \in \mathbb{F}_2^n$. Since the sum defining b_u has $2^{wt(u)}$ terms, the overall complexity of this algorithm is

$$\sum_{i=0}^n \binom{n}{i} 2^i = 3^n .$$

But there exists a faster algorithm for computing the image of an element by \mathcal{M}_n which has complexity $n2^{n-1}$ only. This algorithm exploits the fact that if we decompose any vector $a = (a_u, u \in \mathbb{F}_2^n)$ into two halves, namely $L(a) = (a_{u,0}, u \in \mathbb{F}_2^{n-1})$ and $R(a) = (a_{u,1}, u \in \mathbb{F}_2^{n-1})$, we get the following recursive formula:

$$L(\mathcal{M}_n(a)) = \mathcal{M}_{n-1}(L(a)) \text{ and } R(\mathcal{M}_n(a)) = \mathcal{M}_{n-1}(L(a)) + \mathcal{M}_{n-1}(R(a))$$

where the addition denotes the addition in $\mathbb{F}_2^{2^{n-1}}$. The corresponding algorithm then starts from $(a_u, u \in \mathbb{F}_2^n)$ where the values of u are written in lexicographic order. The k -th step, for $1 \leq k \leq n$, then computes the images by \mathcal{M}_k of the 2^{n-k} vectors of 2^k consecutive bits composing a . The result at Step k is then obtained from the result at Step $(k-1)$ by

splitting the vector into blocks of 2^k consecutive bits, and for each block, the first half of the block remains unchanged while the second half is replaced by the sum of both halves. This iterative process is described by Algorithm 1. In this algorithm, the vectors $(a_u, u \in \mathbb{F}_2^n)$ are equivalently represented by 2^n -bit arrays $(a[i], 0 \leq i < 2^n)$, where n -bit integers are identified with n -bit vectors.

Algorithm 1 Evaluating the Möbius transform \mathcal{M}_n .

```

Input:  $(a[i], 0 \leq i < 2^n)$ 
Output:  $b = \mathcal{M}_n(a)$ 
for  $i$  from 0 to  $2^n - 1$  do
     $b[i] \leftarrow a[i]$ 
end for
for  $k$  from 1 to  $n$  do
    for  $i$  from 0 to  $2^{n-k} - 1$  do
        // Compute the image of the  $i$ -th  $2^k$ -bit block under  $\mathcal{M}_k$ 
        for  $j$  from 0 to  $2^{k-1} - 1$  do
             $b[2^k i + 2^{k-1} + j] \leftarrow b[2^k i + j] + b[2^k i + 2^{k-1} + j] \bmod 2$ 
        end for
    end for
end for
return  $b$ 

```

Example 1.2. Computing the ANF of a 3-variable Boolean function. Let us denote by $f[0], \dots, f[7]$ the array representing the 8-bit value vector of the 3-variable Boolean function f , namely $f[i] = f(i_0, i_1, i_2)$ where $i = \sum_{j=0}^2 i_j 2^j$. Then, the operations performed during the three successive steps of Algorithm 1 are described in Table 1.2.

Example 1.3. Computing the ANF of a 5-variable Boolean function in C. If the value vector of the function is stored as 32-bit integer x , then the corresponding ANF is computed by the following program.

```

x ^= (x & 0x55555555) << 1;
x ^= (x & 0x33333333) << 2;
x ^= (x & 0x0f0f0f0f) << 4;
x ^= (x & 0x00ff00ff) << 8;
x ^= x << 16;

```

A more general program for any number of variables is given in [Jou09, Page 287] and can be downloaded from http://www.joux.biz/algcrypt/PROGRAMS/Walsh_9-2.html.

1.1.3 Reed-Muller codes

Reed-Muller codes are named after Reed [Ree54] and Muller [Mul54]: Muller described the codes while Reed proposed a majority-logic decoding algorithm for them. Reed-Muller codes can be defined over \mathbb{F}_q but we here focus on the binary case. Binary Reed-Muller codes can be defined very easily in terms of Boolean functions.

Table 1.2: Computing the ANF of a 3-variable Boolean function f with Algorithm 1.

i	0	1	2	3	4	5	6	7
$f[i]$	$f[0]$	$f[1]$	$f[2]$	$f[3]$	$f[4]$	$f[5]$	$f[6]$	$f[7]$
Step 1	$f[0]$	$f[0] + f[1]$	$f[2]$	$f[2] + f[3]$	$f[4]$	$f[4] + f[5]$	$f[6]$	$f[6] + f[7]$
Step 2	$f[0]$	$f[0] + f[1]$	$f[0] + f[2]$	$f[0] + f[1]$ $+ f[2] + f[3]$	$f[4]$	$f[4] + f[5]$	$f[4] + f[6]$	$f[4] + f[5]$ $+ f[6] + f[7]$
Step 3	$f[0]$	$f[0] + f[1]$	$f[0] + f[2]$	$f[0] + f[1]$ $+ f[2] + f[3]$	$f[0] + f[4]$	$f[0] + f[1]$ $f[4] + f[5]$	$f[0] + f[2]$ $+ f[4] + f[6]$	$f[0] + f[1]$ $+ f[2] + f[3]$ $+ f[4] + f[5]$ $+ f[6] + f[7]$

Definition 1.4 (Reed-Muller codes). *Let n be a positive integer and r an integer such $0 \leq r \leq n$. The r -th order binary Reed-Muller code of length 2^n , denoted by $\mathcal{R}(r, n)$, is the set of the value vectors of all Boolean functions of n variables with degree at most r :*

$$\mathcal{R}(r, n) = \{(f(x), x \in \mathbb{F}_2^n), f : \mathbb{F}_2^n \rightarrow \mathbb{F}_2 \text{ with } \deg f \leq r\}.$$

In particular $\mathcal{R}(0, n)$ is composed of the all-zero and the all-one 2^n -bit words. It is also known as the *repetition code* of length 2^n . On the other extreme, $\mathcal{R}(n, n)$ contains all 2^n -bit words.

Reed-Muller codes satisfy the following simple properties.

Proposition 1.5. *Let n be a positive integer and r an integer such $0 \leq r \leq n$.*

1. $\mathcal{R}(r, n)$ is a linear code;
2. The value vectors of all monomials of degree at most r form a basis of $\mathcal{R}(r, n)$;
3. The dimension of $\mathcal{R}(r, n)$ is

$$\dim \mathcal{R}(r, n) = \sum_{i=0}^r \binom{n}{i};$$

4. $\mathcal{R}(r-1, n) \subset \mathcal{R}(r, n)$.

Example 1.4. $\mathcal{R}(1, 3)$ is a linear code of length 8 and dimension $1 + 3 = 4$. A generator matrix of $\mathcal{R}(1, 3)$ consists of the value vectors of all monomials of degree 0 or 1 in x_1, x_2, x_3 :

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

Indeed, the rows of this matrix are equal to the value vectors of the functions $f(x_1, x_2, x_3)$ with respective ANF $1, x_1, x_2$ and x_3 , where the inputs (x_1, x_2, x_3) of the value vectors are ordered lexicographically.

1.2 Weight of a Boolean function

For studying the properties of a Boolean function with a given degree, the following decomposition provides a useful tool since it allows recursive constructions (and proofs by induction). For any Boolean function f of n variables and degree at most r , there exist two functions of $n-1$ variables, namely g and h , with $\deg g \leq r$ and $\deg h \leq r-1$ such that

$$f(x_1, \dots, x_n) = g(x_1, \dots, x_{n-1}) + x_n h(x_1, \dots, x_{n-1}).$$

The value vector v_f of f can then be decomposed into two halves corresponding to the inputs with $x_n = 0$ (resp. with $x_n = 1$). Then, we get

$$v_f = (v_g | v_g + v_h),$$

where v_g and v_h denote the value vectors of g and h and belong to $\mathcal{R}(r, n-1)$ and $\mathcal{R}(r-1, n-1)$ respectively.

This decomposition corresponds to a general construction, named the $(u|u+v)$ construction, described by Plotkin [Plo60], combining two codes of the same length to derive a new code which is twice longer.

1.2.1 Minimum weight of a Boolean function in $\mathcal{R}(r, n)$

The minimum weight of a function in $\mathcal{R}(r, m)$ can be deduced from the previous decomposition.

Theorem 1.6. *Let n be a positive integer and r an integer such $0 \leq r \leq n$. Then, the smallest weight for a nonzero function of n variables and degree at most r is 2^{n-r} . Moreover, this bound is tight by the monomials of degree r .*

Proof. Clearly, any monomial of n variables and degree r has weight 2^{n-r} . Then, we only have to prove that any function of degree at most r has weight $wt(f) \geq 2^{n-r}$, by induction on n .

- For $n = 1$ and $r = 0$, the only nonzero constant function is the all-one function, which has weight 2. Moreover, the value vectors of all nonzero functions correspond to all nonzero vectors in \mathbb{F}_2^2 , implying that the minimum weight for a function in $\mathcal{R}(1, 1)$ is 1.
- Induction step. We now use that any f in $\mathcal{R}(r, n)$ can be decomposed as

$$f(x_1, \dots, x_n) = g(x_1, \dots, x_{n-1}) + x_n h(x_1, \dots, x_{n-1}) ,$$

i.e.,

$$v_f = (v_g | v_g + v_h) ,$$

with $v_g \in \mathcal{R}(r, n-1)$ and $v_h \in \mathcal{R}(r-1, n-1)$. Then,

$$wt(f) = wt(g) + wt(g + h) .$$

If h is the zero function, then $wt(f) = 2wt(g) \geq 2 \times 2^{(n-1)-r} = 2^{n-r}$ by induction hypothesis.

If $h \neq 0$, then, we use that, for any m -bit vectors x and y ,

$$\begin{aligned} wt(x + y) &= wt(x) + wt(y) - 2|\text{Supp } x \cap \text{Supp } y| \\ &\geq wt(x) + wt(y) - 2wt(x) \\ &\geq wt(y) - wt(x) . \end{aligned}$$

Hence,

$$wt(f) = wt(g) + wt(g + h) \geq wt(h) \geq 2^{(n-1)-(r-1)}$$

where the last inequality is deduced from the induction hypothesis.

◇

1.2.2 Weights of affine functions

Proposition 1.7. *All n -variable Boolean functions of degree 1 have weight 2^{n-1} .*

Proof. A function of degree 1 can be written as $f(x) = a \cdot x + \varepsilon$ with $a \in \mathbb{F}_2^n \setminus \{0\}$ and $\varepsilon \in \mathbb{F}_2$. Then,

$$wt(f) = \{x \in \mathbb{F}_2^n : f(x) = 1\} = \{x \in \mathbb{F}_2^n : a \cdot x = 1 + \varepsilon\} = \begin{cases} \langle a \rangle^\perp & \text{if } \varepsilon = 1 \\ \mathbb{F}_2^n \setminus \langle a \rangle^\perp & \text{if } \varepsilon = 0 \end{cases}$$

The support of a function of degree 1 is then a hyperplane or the complement of a hyperplane, and has size 2^{n-1} .

◇

1.2.3 Weights of functions of degree $(n - 1)$

Obviously, half of the Boolean functions of n variables have degree n . These functions are characterized by the following property.

Proposition 1.8. *A Boolean function f of n variables has an odd weight if and only if it has degree n .*

Proof. This comes directly from Theorem 1.3 which shows that the coefficient of degree n in the ANF of an n -variable Boolean function f is equal to the parity of the weight of its value vector. \diamond

This implies that Boolean functions with maximal degree cannot be used in most cryptographic applications since their output distribution is biased.

1.2.4 Weights of functions of degree 2

The list of all possible weights for a Boolean function of degree 2 is also known [SB70]. The proof can be found in [SB70] or in Chapter 15, § 2 of [MS77].

Proposition 1.9. *The weights of the Boolean functions of n variables with degree 2 are of the form*

$$w = 2^{n-1} \text{ or } w = 2^{n-1} \pm 2^{n-1-h} \text{ with } 0 \leq h \leq \left\lfloor \frac{n}{2} \right\rfloor.$$

1.2.5 Weight divisibility of Boolean functions

Another information is that the weight of any n -variable Boolean function of degree at most r is divisible by some power of 2 whose exponent depends on r and n .

Proposition 1.10. *Let n be a positive integer and r an integer such $0 < r \leq n$. Then, the weight of any n -variable Boolean function of degree at most r is divisible by*

$$2^{\lceil \frac{n}{r} \rceil - 1}.$$

This theorem was originally proved by Solomon and McEliece [SM66], but it is usually presented as a consequence of a more general theorem due to McEliece [McE72] on the divisibility of the weight of cyclic codes. A simpler proof can be derived from a classical formula for computing the weight of a Boolean function from its ANF (see e.g. [MM94] or [CHLL97, Page 240]).

1.3 Walsh transform

1.3.1 Definitions

Definition 1.11. *The bias (aka, correlation, or imbalance) of an n -variable Boolean function f is*

$$\mathcal{E}(f) = \sum_{x \in \mathbb{F}_2^n} (-1)^{f(x)} = 2^n - 2wt(f).$$

In other words,

$$\Pr_X[f(X) = 1] = \frac{wt(f)}{2^n} = \frac{1}{2} \left(1 - \frac{\mathcal{E}(f)}{2^n} \right).$$

Most notably, a Boolean function f is balanced if and only if $\mathcal{E}(f) = 0$.

In the following, we denote by φ_a the linear Boolean function $x \mapsto a \cdot x$.

We now reformulate Prop. 1.8 in terms of biases of linear functions.

Lemma 1.12. *Let $a \in \mathbb{F}_2^n$. Then*

$$\mathcal{E}(\varphi_a) = \sum_{x \in \mathbb{F}_2^n} (-1)^{a \cdot x} = \begin{cases} 0 & \text{if } a \neq 0 \\ 2^n & \text{if } a = 0. \end{cases}$$

A useful tool for studying Boolean functions is the Walsh transform.

Definition 1.13 (Walsh transform). *Let f be a Boolean function of n variables. The Walsh transform of f is the function*

$$\begin{aligned} \mathbb{F}_2^n &\rightarrow \mathbb{Z} \\ a &\mapsto \mathcal{E}(f + \varphi_a) = \sum_{x \in \mathbb{F}_2^n} (-1)^{f(x) + a \cdot x}. \end{aligned}$$

The value $\mathcal{E}(f + \varphi_a)$ is called the Walsh coefficient of f at point a and the multiset composed of all Walsh coefficients of f is called the Walsh spectrum of f .

The Walsh spectrum of f corresponds to the biases of all approximations of f by a linear function. This quantity, especially its maximum, plays a major role in linear cryptanalysis.

1.3.2 Computing the Walsh transform.

A useful property of the Fourier transform is that it can be computed by a fast algorithm. While the naive method for computing all values of the Walsh transform of an n -variable Boolean function has complexity 2^{2n} , the complexity of the fast algorithm described in Algorithm 2 is proportional to $n2^n$.

Algorithm 2 Evaluating the Walsh transform of an n -variable Boolean function f .

```

Input: ( $f[a], 0 \leq a < 2^n$ )
Output: ( $e[a] = \mathcal{E}(f + \varphi_a), 0 \leq a < 2^n$ )
for  $i$  from 0 to  $2^n - 1$  do
     $e[i] \leftarrow (-1)^{f[i]}$ 
end for
for  $k$  from 1 to  $n$  do
    for  $i$  from 0 to  $2^{n-k} - 1$  do
        // Compute the image of the  $i$ -th  $2^k$ -bit block
        for  $j$  from 0 to  $2^{k-1} - 1$  do
             $e'[2^k i + j] \leftarrow e[2^k i + j] + e[2^k i + 2^{k-1} + j] \bmod 2$ 
             $e'[2^k i + 2^{k-1} + j] \leftarrow e[2^k i + j] - e[2^k i + 2^{k-1} + j] \bmod 2$ 
        end for
    end for
     $e \leftarrow e'$ 
end for
return  $e$ 

```

Example 1.5. Let us compute the Walsh transform of the 3-variable function

$$f(x_1, x_2, x_3) = x_1 + x_1x_2 + x_2x_3.$$

$f(a)$	0	1	0	0	0	1	1	1
$(-1)^{f(a)}$	1	-1	1	1	1	-1	-1	-1
step 1	0	2	2	0	0	2	-2	0
step 2	2	2	-2	2	-2	2	2	2
$\mathcal{E}(f + \varphi_a)$	0	4	0	4	4	0	-4	0

We deduce that the highest bias (in magnitude) of a linear approximation of this function is 4 and is obtained for instance for $\mathcal{E}(f + x_1)$. This equivalently means

$$\Pr[f(x_1, x_2, x_3) + x_2 = 0] = \frac{1}{2} \left(1 + \frac{4}{2^3} \right) = \frac{3}{4}.$$

1.3.3 Basic properties of the Walsh transform

Since it corresponds to a discrete Fourier transform, The Walsh transform enjoys all mathematical properties of a Fourier transform. For instance, the Walsh transform is (up to a constant factor) an involution.

Proposition 1.14. *Let f be a Boolean function of n variables. For all $b \in \mathbb{F}_2^n$, we have*

$$\sum_{a \in \mathbb{F}_2^n} (-1)^{a \cdot b} \mathcal{E}(f + \varphi_a) = 2^n (-1)^{f(b)}.$$

Proof.

$$\begin{aligned} \sum_{a \in \mathbb{F}_2^n} (-1)^{a \cdot b} \mathcal{E}(f + \varphi_a) &= \sum_{a \in \mathbb{F}_2^n} \sum_{x \in \mathbb{F}_2^n} (-1)^{a \cdot b} (-1)^{f(x) + a \cdot x} \\ &= \sum_{x \in \mathbb{F}_2^n} (-1)^{f(x)} \sum_{a \in \mathbb{F}_2^n} (-1)^{a \cdot (x+b)} \\ &= \sum_{x \in \mathbb{F}_2^n} (-1)^{f(x)} \mathcal{E}(\varphi_{x+b}) = 2^n (-1)^{f(b)} \end{aligned}$$

where the last equality is derived from Lemma 1.12 which states that $\mathcal{E}(\varphi_{x+b}) = 0$ unless $x = b$. \diamond

As any Fourier transform, the Walsh transform satisfies the Parseval relation.

Proposition 1.15 (Parseval relation). *Let f be a Boolean function of n variables. Then,*

$$\sum_{a \in \mathbb{F}_2^n} [\mathcal{E}(f + \varphi_a)]^2 = 2^{2n}.$$

Proof.

$$\begin{aligned}
\sum_{a \in \mathbb{F}_2^n} \mathcal{E}^2(f + \varphi_a) &= \sum_{a \in \mathbb{F}_2^n} \left(\sum_{x \in \mathbb{F}_2^n} (-1)^{f(x) + a \cdot x} \right) \left(\sum_{y \in \mathbb{F}_2^n} (-1)^{f(y) + a \cdot y} \right) \\
&= \sum_{x \in \mathbb{F}_2^n} \sum_{y \in \mathbb{F}_2^n} (-1)^{f(x) + f(y)} \left(\sum_{a \in \mathbb{F}_2^n} (-1)^{a \cdot (x+y)} \right) \\
&= 2^n \sum_{x \in \mathbb{F}_2^n} (-1)^{f(x) + f(x)} = 2^{2n}
\end{aligned}$$

where the last-but-one equality is derived from Lemma 1.12. \diamond

1.4 Linearity of a Boolean function

Since linear cryptanalysis exploits highly biased linear approximations, a natural question for a designer who needs to choose a good nonlinear function is to know the smallest possible value that can be achieved for the bias of the best linear approximation. This value is determined by the so-called *linearity* of the function, in the sense of the following definition.

Definition 1.16 (Linearity of a Boolean function). *Let f be a Boolean function of n variables. The linearity of f is the highest magnitude of its Walsh coefficients, i.e.,*

$$\mathcal{L}(f) = \max_{a \in \mathbb{F}_2^n} |\mathcal{E}(f + \varphi_a)| .$$

1.4.1 Optimal linearity

The following lower bound on the linearity of a Boolean function is directly derived from Parseval relation.

Proposition 1.17. *Let f be a Boolean function of n variables. Then*

$$\mathcal{L}(f) \geq 2^{\frac{n}{2}} .$$

The functions for which equality holds are called bent functions. They exist for even n only, and are not balanced.

Proof. We first observe that a function f satisfying $\mathcal{L}(f) < 2^{\frac{n}{2}}$ cannot exist. Otherwise, from Parseval relation, we would have

$$2^{2n} = \sum_{a \in \mathbb{F}_2^n} \mathcal{E}^2(f + \varphi_a) < 2^n \times 2^n ,$$

a contradiction.

Obviously, this bound on the linearity is tight when n is even only. Let now f be a bent function. Parseval relation implies that, for all $a \in \mathbb{F}_2^n$,

$$\mathcal{E}^2(f + \varphi_a) = 2^n .$$

In other words, the Walsh transform of a bent function has constant magnitude. In particular, for $a = 0$, we get that $\mathcal{E}(f) = \pm 2^{\frac{n}{2}}$ implying that f is not balanced. \diamond

Since their output is biased, bent functions are of little use in cryptography. Most notably, any linear combination of the coordinates of a permutation is balanced, and then satisfies $\mathcal{L}(f) > 2^{\frac{n}{2}}$.

When n is odd, bent functions do not exist, and the lowest possible value for $\mathcal{L}(f)$ is unknown for $n \geq 9$. The quadratic function of $(2t + 1)$ variables

$$f(x_1, \dots, x_{2t+1}) = x_1x_2 + x_3x_4 + \dots + x_{2t-1}x_{2t} + x_{2t+1}$$

satisfies

$$\mathcal{L}(F) = 2^{\frac{n+1}{2}}.$$

We then deduce the following proposition.

Proposition 1.18. *The lowest linearity for a Boolean function of n variables, n odd, satisfies*

$$2^{\frac{n}{2}} < \min_{f \in \text{Bool}_n} \mathcal{L}(f) \leq 2^{\frac{n+1}{2}}$$

where the upper bound is tight for $n \leq 7$ and is not tight for $n \geq 9$.

The fact that the previous upper bound is not tight for $n \geq 9$ was a long-standing open problem solved in 2006 by [KMY07]. More precisely, Table 1.3 gives the lowest possible linearities for an n -variable Boolean function for n odd, $5 \leq n \leq 15$. Note that, by definition, the values of the Walsh transform, including the linearity, are always even.

n	5	7	9	11	13	15
$\min_{f \in \text{Bool}_n} \mathcal{L}(f)$	8	16	24 – 30	46 – 60	92 – 120	182 – 216

Table 1.3: Smallest possible linearity for an n -variable Boolean function, where $a - b$ means that the lowest linearity can be any even integer in this range.

As previously mentioned, most cryptographic applications require the use of balanced functions. The lowest linearity for a balanced function of n variables is also unknown for n even. Only an upper bound on its value is derived from a recursive construction due to Dobbertin [Dob94]. Table 1.4 gives the lowest possible linearities for an n -variable *balanced* Boolean function for $4 \leq n \leq 10$. It is worth noticing that, since the weight of a balanced function is even, its degree is at most $(n - 1)$. Then, all $(f + \varphi_a)$ have degree at most $(n - 1)$ and an even degree, implying that their biases are divisible by 4.

n	4	5	6	7	8	9	10
$\min_{f \in \text{Bool}_n} \mathcal{L}(f)$	8	8	12	16	{20, 24}	{24, 28, 32}	{36, 40}

Table 1.4: Smallest possible linearities for an n -variable balanced function.

1.4.2 Link with Reed-Muller codes.

The previously mentioned problems on the lowest possible linearity for a Boolean function have been extensively investigated in coding theory since they are related to the determination of the covering radius of the first-order Reed-Muller code. Indeed, the Walsh coefficients of a given Boolean function are determined by the weights of the coset of $\mathcal{R}(1, n)$ defined by the value vector of f , as detailed in the following proposition.

Proposition 1.19. *Let f be a Boolean function of n variables. Then, the Hamming weights of all functions in the set $f + \mathcal{R}(1, n)$ are*

$$\{2^{n-1} - \frac{1}{2}\mathcal{E}(f + \varphi_a); 2^{n-1} + \frac{1}{2}\mathcal{E}(f + \varphi_a), a \in \mathbb{F}_2^n\}.$$

In particular, the Hamming distance of f to the set of all affine functions, called the nonlinearity of f , is given by

$$d(f, \mathcal{R}(1, n)) = 2^{n-1} - \frac{1}{2}\mathcal{L}(f).$$

Proof. By definition of $\mathcal{R}(1, n)$, the weights of $f + \mathcal{R}(1, n)$ correspond to the weights of all functions $c = f + \varphi_a + \varepsilon$, $a \in \mathbb{F}_2^n, \varepsilon \in \mathbb{F}_2$. If $\varepsilon = 0$, we get that

$$wt(f + \varphi_a) = 2^{n-1} - \frac{1}{2}\mathcal{E}(f + \varphi_a).$$

If $\varepsilon = 1$, we have

$$wt(f + \varphi_a + 1) = 2^n - wt(f + \varphi_a) = 2^{n-1} + \frac{1}{2}\mathcal{E}(f + \varphi_a).$$

The expression of $d(f, \mathcal{R}(1, n))$ directly follows. \diamond

Finding a function with minimal linearity then boils down to finding a value vector which lies as far as possible from the code $\mathcal{R}(1, n)$. This corresponds to the well-known notion of *covering radius*.

Definition 1.20 (Covering radius). *Let \mathcal{C} be a code of length n . The covering radius of \mathcal{C} is the highest Hamming distance between \mathcal{C} and a word in \mathbb{F}_2^n :*

$$\rho(\mathcal{C}) = \max_{c \in \mathbb{F}_2^n} d(c, \mathcal{C}).$$

Therefore, most of the previously mentioned results on the best linearity for a Boolean functions have been first proved in terms of covering radius of the first-order Reed-Muller codes, e.g. [Myk80, PW83, Hou93, Hou96b, Hou96a].

1.4.3 Affine equivalence for Boolean functions

The number of Boolean functions of n variables is 2^{2^n} , which makes very difficult to find functions with good cryptographic properties, like a high nonlinearity, for practical values of n . A useful tool for avoiding examining all functions is provided by any equivalence relation under which the involved properties are invariant. The most prominent equivalence relation is called *affine equivalence*.

Definition 1.21 (Affine equivalence). *Two n -variable Boolean functions are said to be affine equivalent if there exists an affine permutation A of \mathbb{F}_2^n such that*

$$g(x) = f(A(x)) .$$

Obviously, the degree is invariant under affine equivalence. This is also the case of the unsigned Walsh spectrum (and then of the linearity) as shown in the following proposition.

Proposition 1.22. *Let f and g be two n -variable Boolean functions such that $g(x) = f(Mx + a)$ for some invertible $n \times n$ matrix M and $a \in \mathbb{F}_2^n$. Then, for any $\alpha \in \mathbb{F}_2^n$,*

$$\mathcal{E}(g + \varphi_\alpha) = (-1)^{\alpha \cdot M^{-1}a} \mathcal{E}(f + \varphi_{(M^{-1})^T \alpha}) .$$

Proof.

$$\mathcal{E}(g + \varphi_\alpha) = \sum_{x \in \mathbb{F}_2^n} (-1)^{g(Mx+a) + \alpha \cdot x} = (-1)^{\alpha \cdot M^{-1}a} \sum_{y \in \mathbb{F}_2^n} (-1)^{g(y) + \alpha \cdot (M^{-1}y)} .$$

For any $y \in \mathbb{F}_2^n$, we have

$$\begin{aligned} \alpha \cdot (M^{-1}y) &= \sum_{i=1}^n \alpha_i (M^{-1}y)_i \\ &= \sum_{i=1}^n \sum_{j=1}^n \alpha_i M_{ij}^{-1} y_j \\ &= \sum_{j=1}^n y_j \left(\sum_{i=1}^n \alpha_i M_{ij}^{-1} \right) = y \cdot ((M^{-1})^T \alpha) . \end{aligned}$$

We then deduce

$$\mathcal{E}(g + \varphi_\alpha) = (-1)^{\alpha \cdot M^{-1}a} \sum_{y \in \mathbb{F}_2^n} (-1)^{g(y) + y \cdot ((M^{-1})^T \alpha)} = (-1)^{\alpha \cdot M^{-1}a} \mathcal{E}(g + \varphi_{(M^{-1})^T \alpha}) .$$

◇

Since they mainly focus on the linearity (or on the magnitude of the Walsh coefficients), all existing classifications of Boolean functions are up to affine equivalence. For instance, Berlekamp and Welch [BW72] have classified all functions up to 5 variables.

1.5 Existence of an approximation with fewer variables

Some divide-and-conquer attacks, like (fast) correlation attacks against combination keystream generators, exploit the fact that the involved Boolean function f can be approximated by a function g depending on fewer variables. Typically, in a combination generator composed of n LFSRs, if the combining function f can be approximated by a function g of $k < n$ variables, then the attack involves the initial states of only k out of the n constituent LFSRs.

1.5.1 Correlation-immunity order

A natural counter-measure to avoid these attacks consists then in using as a building-block a *correlation-immune function* in the sense of the following definition.

Definition 1.23 (Correlation-immunity [Sie84]). *A Boolean function f is t -th order correlation-immune if the probability distribution of its output is unaltered when any t input variables are fixed. Balanced t -th order correlation-immune functions are called t -resilient.*

Note that a t -th order correlation-immune function is k -th order correlation-immune for any $k \leq t$. The correlation-immunity order of a function f then implicitly refers to the highest integer t such that f is t -th order correlation-immune.

In a combination generator, the correlation-immunity order t of the combining function determines the minimal number of LFSRs which must be considered together in a correlation attack. Indeed, the keystream produced by the combination generator is then independent from the probability distribution of any set of t constituent LFSRs. The smallest number of LFSRs involved in a correlation attack is therefore $t + 1$. But the correlation-immunity order of a Boolean function cannot be chosen as high as we wish: it is limited by the algebraic degree of the function as shown in the next proposition.

Proposition 1.24. [Sie84] *Let f be a Boolean function of n variables. Then its correlation-immunity order t satisfies*

$$t + \deg f \leq n .$$

Moreover, if f is balanced and $t < n - 1$, then

$$t + \deg f \leq n - 1 .$$

Proof. Let $u \in \mathbb{F}_2^n$ such that $wt(u) \geq n - t$. We compute the coefficients of the ANF of f with Theorem 1.3. For $L = \{1, \dots, n\} \setminus \text{Supp}(u)$, we have

$$a_u = \sum_{x_i=0, i \in L} f(x_1, \dots, x_n) \bmod 2 = 2^{-(n-wt(u))} wt(f) \bmod 2$$

where the last equality comes from the fact that the probability distribution of the output of f is unchanged when the $(n - wt(u)) \leq t$ variables defined by L are set to 0. If f is balanced, i.e., $wt(f) = 2^{n-1}$, we get that, for all u with $wt(u) \geq n - t$,

$$a_u = 2^{wt(u)-1} \bmod 2 = 0$$

since $wt(u) - 1 \geq n - t - 1 > 0$. In other words, all coefficients of degree greater than or equal to $(n - t)$ in the ANF of f are equal to zero, which means that $\deg f < n - t$.

If f is not balanced, we select some word u^* of weight $(n - t)$. Then,

$$a_{u^*} = 2^{-t} wt(f) \bmod 2$$

implying that $wt(f) = 2^t a_{u^*} + \Lambda 2^{t+1}$ for some integer Λ . Now, for any u of weight $(n - t + w)$ with $w \geq 1$, we get

$$a_u = 2^{-t+w} wt(f) = 2^w a_{u^*} + \Lambda 2^{w+1} = 0 \bmod 2 .$$

This means that all coefficients in the ANF of degree greater than or equal to $(n - t + 1)$ vanish, i.e. $\deg f \leq n - t$. \diamond

The correlation-immunity order of a Boolean function can be characterized by its Walsh transform. To this end, we need the following lemma which provides a very useful link between the Walsh transform and the restrictions of a function to some subspace.

Definition 1.25 (Restriction of a Boolean function). *Let f be a Boolean function of n variables, and V be a linear subspace of \mathbb{F}_2^n of dimension k . Let W be the supplementary space of V with respect to \mathbb{F}_2^n i.e., V and W are in direct sum.*

Then, for any $w \in W$, the restriction of f to $w + V$ is the Boolean function of k variables, denoted by f_{w+V} , defined by

$$f_{w+V} : x \in V \times W \mapsto f(x + w)$$

The notion of restriction is of particular interest when V is the linear space spanned by some vectors of the canonical basis, i.e., $V = \langle e_i, i \in I \rangle$ for some $I \subset \{1, \dots, n\}$, where e_i is the vector with support $\{i\}$. In this case, the restriction of f to a coset of V corresponds to f when the input variables $x_i, i \notin I$ are fixed.

Lemma 1.26 (Walsh transform and restriction). *Let f be a Boolean function of n variables. Let V be a linear subspace of \mathbb{F}_2^n . Then, for any a , we have*

$$\sum_{v \in V} (-1)^{a \cdot v} \mathcal{E}(f + \varphi_v) = 2^{\dim V} \mathcal{E}(f_{a+V^\perp}) .$$

Proof.

$$\begin{aligned} \sum_{v \in V} (-1)^{a \cdot v} \mathcal{E}(f + \varphi_v) &= \sum_{v \in V} (-1)^{a \cdot v} \sum_{x \in \mathbb{F}_2^n} (-1)^{f(x) + v \cdot x} \\ &= \sum_{x \in \mathbb{F}_2^n} (-1)^{f(x)} \sum_{v \in V} (-1)^{(a+x) \cdot v} \\ &= 2^{\dim V} \sum_{x \in a+V^\perp} (-1)^{f(x)} \end{aligned}$$

where the last equality comes from the fact that, for any linear space $V \subset \mathbb{F}_2^n$,

$$\sum_{v \in V} (-1)^{a \cdot v} = \#\{v \in V : v \in \langle a \rangle^\perp\} - \#\{v \in V : v \notin \langle a \rangle^\perp\} = \begin{cases} 2^{\dim V} & \text{if } a \in V^\perp \\ 0 & \text{otherwise.} \end{cases}$$

◇

Proposition 1.27 ([XM88]). *Let f be a Boolean function of n variables. Then, f is t -th order correlation-immune if and only if $\mathcal{E}(f + \varphi_\alpha) = 0$ for any $\alpha \in \mathbb{F}_2^n$ with $1 \leq \text{wt}(\alpha) \leq t$.*

Proof. By definition, f is t -th order correlation-immune if and only if, for any $0 \leq k \leq t$ and any set $I \subset \{1, \dots, n\}$ of size k , the restriction of f to $a + \langle e_i, i \notin I \rangle$ has the same probability distribution as f :

$$2^{-(n-k)} \mathcal{E}(f_{a+\langle e_i, i \notin I \rangle}) = 2^{-n} \mathcal{E}(f) .$$

Let $V = \langle e_i, i \notin I \rangle$. We deduce from the previous lemma that this is equivalent to

$$\mathcal{E}(f) = 2^k \mathcal{E}(f_{a+V}) = \sum_{v \in V^\perp} (-1)^{a \cdot v} \mathcal{E}(f + \varphi_v) ,$$

This condition is obviously satisfied if all $\mathcal{E}(f + \varphi_v) = 0$ for $v \in \langle e_i, i \in I \rangle \setminus \{0\}$. Conversely, we get by the inverse Walsh transform that,

$$\begin{aligned} 2^k \sum_{a \in V^\perp} (-1)^{a \cdot b} \mathcal{E}(f_{a+V}) &= \sum_{a \in V^\perp} \sum_{v \in V^\perp} (-1)^{a \cdot (b+v)} \mathcal{E}(f + \varphi_v) \\ &= \sum_{v \in V^\perp} \mathcal{E}(f + \varphi_v) \left(\sum_{a \in V^\perp} (-1)^{a \cdot (b+v)} \right) \\ &= 2^{n-k} \mathcal{E}(f + \varphi_b), \end{aligned}$$

using that, for $V = \langle e_i, i \notin I \rangle$, we have $V \cap V^\perp = \{0\}$. Therefore, if all restrictions satisfy $2^k \mathcal{E}(f_{a+V}) = \mathcal{E}(f)$, then, for any nonzero $b \in \langle e_i, i \in I \rangle$,

$$2^{n-k} \mathcal{E}(f + \varphi_b) = \sum_{a \in V^\perp} (-1)^{a \cdot b} \mathcal{E}(f) = 0.$$

◇

1.5.2 Approximation of a function by a function of $(t + 1)$ variables.

Since the combining function in a combination generator is balanced and nonlinear, its correlation-immunity order is at most $(n - 3)$. It follows that we can always apply a correlation attack by considering at most $(n - 2)$ LFSRs together. We now show how to determine the best approximation of f by a Boolean function g depending on a fixed subset of k variables.

Proposition 1.28. [Can02, Zha00] *Let f be a function of n variables and let V be the subspace spanned by the k elements of the canonical basis of indices in $I = \{i_1, \dots, i_k\}$. The highest possible value of $\mathcal{E}(f + g)$, over all k -variable functions g depending on x_{i_1}, \dots, x_{i_k} , is achieved if and only if*

$$\begin{cases} g(x) = 1 & \text{if } \mathcal{E}(f_{x+V^\perp}) < 0 \\ g(x) = 0 & \text{if } \mathcal{E}(f_{x+V^\perp}) > 0 \end{cases}$$

Moreover,

$$\max_g \mathcal{E}(f + g) = \sum_{x \in V} |\mathcal{E}(f_{x+V^\perp})|.$$

Proof. Using that V and V^\perp are in direct sum, we have that, for any function g defined over V ,

$$\begin{aligned} \mathcal{E}(f + g) &= \sum_{x \in V} \sum_{y \in V^\perp} (-1)^{f(x+y)+g(x)} \\ &= \sum_{x \in V} (-1)^{g(x)} \sum_{y \in V^\perp} (-1)^{f(x+y)} \\ &= \sum_{x \in V} (-1)^{g(x)} \mathcal{E}(f_{x+V^\perp}). \end{aligned}$$

It follows that $\mathcal{E}(f + g)$ is maximal if and only if all terms in the sum are greater than or equal to zero, or equivalently

$$g(x) = \begin{cases} 0 & \text{if } \mathcal{E}(f_{x+V^\perp}) > 0, \\ 1 & \text{if } \mathcal{E}(f_{x+V^\perp}) < 0. \end{cases} \quad (1.1)$$

Note that the value of $g(x)$ can be arbitrarily chosen when $\mathcal{E}(f_{x+V^\perp}) = 0$. The maximal value of $\mathcal{E}(f + g)$ directly follows. ◇

The previous proposition obviously implies that, for a balanced function f , the maximal value of $\mathcal{E}(f + g)$ is 0 if k is less than or equal to the correlation-immunity order of f since all $\mathcal{E}(f_{x+V^\perp}) = 0$ in this case. Another consequence is that, for $k = t+1$ where t is the correlation-immunity order of f , the maximal value of $\mathcal{E}(f + g)$ is achieved by an affine function.

Theorem 1.29. [CT00] *Let f be a t -resilient function of n variables and let V be the subspace spanned by the $(t+1)$ elements of the canonical basis of indices in $I = \{i_1, \dots, i_{t+1}\}$. The highest possible value of $\mathcal{E}(f + g)$, over all $(t+1)$ -variable functions g depending on $x_{i_1}, \dots, x_{i_{t+1}}$, is achieved by the affine function*

$$g(x_{i_1}, \dots, x_{i_{t+1}}) = \sum_{i \in I} x_i + \varepsilon$$

with $\varepsilon \in \mathbb{F}_2$.

Proof. We use the same notation as in the previous proposition. We have proved that, for any g defined over V

$$\mathcal{E}(f + g) = \sum_{x \in V} (-1)^{g(x)} \mathcal{E}(f_{x+V^\perp}).$$

But, we know from Lemma 1.26 that

$$\mathcal{E}(f_{x+V^\perp}) = 2^{-(t+1)} \sum_{v \in V} (-1)^{x \cdot v} \mathcal{E}(f + \varphi_v).$$

Moreover, since f is t -resilient, all its Walsh coefficients $\mathcal{E}(f + \varphi_v)$ for all v of weight at most t vanish. Then, the only term remaining in the previous sum is when v equals the all-one vector in V , i.e., the vector 1_I having I as support. We deduce

$$\mathcal{E}(f + g) = 2^{-(t+1)} \sum_{x \in V} (-1)^{g(x)} (-1)^{1_I \cdot x} \mathcal{E}(f + \varphi_{1_I}) = 2^{-(t+1)} \mathcal{E}(f + \varphi_{1_I}) \mathcal{E}(g + \varphi_{1_I}).$$

This quantity is then maximized when $g = \varphi_{1_I} + \varepsilon$, and we get

$$\mathcal{E}(f + g) = \pm \mathcal{E}(f + \varphi_{1_I}).$$

◇

This result is of great interest in the context of fast correlation attacks against combination generators. The correlation-immunity order of a Boolean function f can also be characterized by a combinatorial property of the code formed by the support of f , i.e., by $f^{-1}(1)$, namely by the so-called dual distance of this code [Del73, SM95].

Chapter 2

Cryptographic Sboxes

Many primitives like block ciphers use as building blocks some Boolean functions with several output bits. Such functions are called *vectorial Boolean functions*, or *S(ubstitution)-boxes*, by analogy with the former block-cipher standard DES.

2.1 Representations of an Sbox

2.1.1 Components of an Sbox

An Sbox from \mathbb{F}_2^n into \mathbb{F}_2^m (i.e., with n inputs and m outputs) is usually represented by a collection of m Boolean functions of n variables called the *coordinates of f* .

Example 2.1 (A 4-bit Sbox). The following table defines an Sbox from \mathbb{F}_2^4 into \mathbb{F}_2^4 (with hexadecimal notation).

x	0	1	2	3	4	5	6	7	8	9	a	b	c	d	e	f
$S(x)$	f	e	b	c	6	d	7	8	0	3	9	a	4	2	1	5
$S_1(x)$	1	0	1	0	0	1	1	0	0	1	1	0	0	0	1	1
$S_2(x)$	1	1	1	0	1	0	1	0	0	1	0	1	0	1	0	0
$S_3(x)$	1	1	0	1	1	1	1	0	0	0	0	0	1	0	0	1
$S_4(x)$	1	1	1	1	0	1	0	1	0	0	1	1	0	0	0	0

Table 2.1: Example of a 4-bit Sbox.

The Möbius transform enables us to compute the ANF of the four coordinates of this Sbox:

$$\begin{aligned}
 S_1 &= 1 + x_1 + x_3 + x_2x_3 + x_4 + x_2x_4 + x_3x_4 + x_1x_3x_4 + x_2x_3x_4 \\
 S_2 &= 1 + x_1x_2 + x_1x_3 + x_1x_2x_3 + x_4 + x_1x_4 + x_1x_2x_4 + x_1x_3x_4 \\
 S_3 &= 1 + x_2 + x_1x_2 + x_2x_3 + x_4 + x_2x_4 + x_1x_2x_4 + x_3x_4 + x_1x_3x_4 \\
 S_4 &= 1 + x_3 + x_1x_3 + x_4 + x_2x_4 + x_3x_4 + x_1x_3x_4 + x_2x_3x_4
 \end{aligned}$$

Besides the coordinates, all linear combinations of the coordinates are usually involved for determining the cryptographic properties of an Sbox, in the sense of the following definition.

Definition 2.1. Let S be an Sbox from \mathbb{F}_2^n into \mathbb{F}_2^m . The Boolean components of S are the n -variable Boolean functions

$$S_\lambda : x \mapsto \lambda \cdot S(x)$$

for any $\lambda \in \mathbb{F}_2^m$. The component corresponding to $\lambda = 0$ is called the zero (or trivial) component.

In particular, the fact that an Sbox is invertible (i.e., a permutation) can be characterized by its coordinates.

Proposition 2.2. Let S be an Sbox from \mathbb{F}_2^n into \mathbb{F}_2^m . S is a permutation if and only if all its non-trivial components are balanced.

Proof. Let $\lambda \in \mathbb{F}_2^m \setminus \{0\}$. Then, if S is a permutation,

$$\mathcal{E}(S_\lambda) = \sum_{x \in \mathbb{F}_2^n} (-1)^{\lambda \cdot S(x)} = \sum_{y \in \mathbb{F}_2^m} (-1)^{\lambda \cdot y} = 0.$$

Conversely, for any $a \in \mathbb{F}_2^n$, we have

$$\begin{aligned} \sum_{\lambda \in \mathbb{F}_2^m} (-1)^{a \cdot \lambda} \mathcal{E}(S_\lambda) &= \sum_{\lambda \in \mathbb{F}_2^m} (-1)^{a \cdot \lambda} \sum_{x \in \mathbb{F}_2^n} (-1)^{\lambda \cdot S(x)} \\ &= \sum_{x \in \mathbb{F}_2^n} \left(\sum_{\lambda \in \mathbb{F}_2^m} (-1)^{(a + S(x)) \cdot \lambda} \right) \\ &= 2^n \# \{x \in \mathbb{F}_2^n : S(x) = a\}. \end{aligned}$$

It follows that, if all S_λ , $\lambda \neq 0$ are balanced, then for any $a \in \mathbb{F}_2^n$,

$$2^n \# \{x \in \mathbb{F}_2^n : S(x) = a\} = \mathcal{E}(S_0) = 2^n,$$

i.e., S is a permutation. ◇

The degree of an Sbox is defined as the maximal degree of its Boolean components, since this is the most relevant notion from a mathematical point of view. The minimal degree of a non-trivial component may also be of interest in some attacks, but it won't be considered here.

Definition 2.3 (Degree of an Sbox). Let S be an Sbox from \mathbb{F}_2^n into \mathbb{F}_2^m . The degree of S is the highest degree of the ANF of its components.

2.1.2 Univariate representation

When representing an n -bit Sbox, it may be convenient to identify the vector space \mathbb{F}_2^n with the finite field with 2^n elements, \mathbb{F}_{2^n} . Then, for any fixed basis of \mathbb{F}_2^n defining an isomorphism between \mathbb{F}_2^n and \mathbb{F}_{2^n} , the n -bit Sbox S can be uniquely represented as a univariate polynomial in $\mathbb{F}_{2^n}[X]$:

$$S(X) = \sum_{i=0}^{2^n-1} A_i X^i, A_i \in \mathbb{F}_{2^n}.$$

The coefficients of this polynomial are computed from the values of F by the discrete Fourier Transform (DFT) of F (aka Mattson-Solomon transform) (see e.g. [Bla83, MS77, GG04]).

Proposition 2.4 (Discrete Fourier transform of a function). *Let S be a function from \mathbb{F}_{2^n} into \mathbb{F}_{2^n} . Then, there exists a unique univariate polynomial in $\mathbb{F}_{2^n}[X]/(X^{2^n} + X)$ such that*

$$S(X) = \sum_{i=0}^{2^n-1} A_i X^i .$$

Moreover, $A_0 = S(0)$, $A_{2^n-1} = \sum_{x \in \mathbb{F}_{2^n}} S(x)$ and the coefficients A_i , $1 \leq i \leq 2^n - 2$, are given by the discrete Fourier transform of the values of S at all nonzero inputs, namely

$$A_i = \sum_{k=0}^{2^n-2} S(\alpha^k) \alpha^{-ki}, \quad 0 \leq i \leq 2^n - 2$$

where α is a primitive element in \mathbb{F}_{2^n} .

Proof. We first prove that the univariate polynomial

$$A(X) = \sum_{i=0}^{2^n-1} A_i X^i$$

with the previously defined coefficients takes the same values as S . Obviously,

$$A(0) = A_0 = S(0) .$$

Now, for any nonzero element α^j , we have

$$A(\alpha^j) = \sum_{i=0}^{2^n-1} A_i \alpha^{ij} = (A_0 + A_{2^n-1}) + \sum_{i=1}^{2^n-2} A_i \alpha^{ij} .$$

By definition, the constant coefficient of this polynomial is then

$$A_0 + A_{2^n-1} = \sum_{k=0}^{2^n-1} S(\alpha^k) .$$

It then follows that

$$\begin{aligned} A(\alpha^j) &= \sum_{k=0}^{2^n-1} S(\alpha^k) + \sum_{i=1}^{2^n-2} \left(\sum_{k=0}^{2^n-2} S(\alpha^k) \alpha^{-ki} \right) \alpha^{ij} \\ &= \sum_{i=0}^{2^n-2} \left(\sum_{k=0}^{2^n-2} S(\alpha^k) \alpha^{-ki} \right) \alpha^{ij} \\ &= \sum_{k=0}^{2^n-2} S(\alpha^k) \left(\sum_{i=0}^{2^n-2} \alpha^{i(j-k)} \right) = S(\alpha^j) . \end{aligned}$$

Then, the values of A correspond to the values of S . Moreover, this polynomial is unique. Otherwise, there would exist two distinct polynomials of degree less than or equal to $(2^n - 1)$ taking the same value at 2^n distinct points, which is impossible. \diamond

Example 2.2. Let α be a root of the irreducible polynomial $1 + x + x^4$. We identify \mathbb{F}_2^4 with \mathbb{F}_{2^4} using the basis $\{1, \alpha, \alpha^2, \alpha^3\}$. This leads to

\mathbb{F}_{2^4}	0	1	α	α^2	α^3	α^4	α^5	α^6	α^7
	0	1	α	α^2	α^3	$\alpha + 1$	$\alpha^2 + \alpha$	$\alpha^3 + \alpha^2$	$\alpha^3 + \alpha + 1$
\mathbb{F}_2^4	0000	0001	0010	0100	1000	0011	0110	1100	1011
α^8	α^9	α^{10}	α^{11}	α^{12}	α^{13}	α^{14}			
$\alpha^2 + 1$	$\alpha^3 + \alpha$	$\alpha^2 + \alpha + 1$	$\alpha^3 + \alpha^2 + \alpha$	$\alpha^3 + \alpha^2 + \alpha + 1$	$\alpha^3 + \alpha^2 + 1$	$\alpha^3 + 1$			
0101	1010	0111	1110	1111	1101	1001			

Considering the following Sbox S over \mathbb{F}_{2^4} ,

x	0	1	2	3	4	5	6	7	8	9	a	b	c	d	e	f
$S(x)$	f	e	b	c	6	d	7	8	0	3	9	a	4	2	1	5

we obtain the univariate representation:

$$S(X) = \alpha^{12} + \alpha^2 X + \alpha^{13} X^2 + \alpha^6 X^3 + \alpha^{10} X^4 + \alpha X^5 + \alpha^{10} X^6 + \alpha^2 X^7 + \alpha^9 X^8 + \alpha^4 X^9 + \alpha^7 X^{10} + \alpha^7 X^{11} + \alpha^5 X^{12} + X^{13} + \alpha^6 X^{14}$$

It is worth noticing that the univariate representation of an Sbox depends on the basis chosen for identifying \mathbb{F}_2^n and \mathbb{F}_{2^n} .

A particular family of Sboxes is the class of functions whose univariate representation contains a single monomial (up to a change of basis), since they usually have a lower implementation cost. These functions are called *monomial* or *power* functions. The most prominent example in this family is the AES Sbox.

Example 2.3. The AES Sbox, which operates in parallel on each byte of the input of the round function is defined as a permutation of the finite field with 2^8 elements. The isomorphism between the field \mathbb{F}_{2^8} and the vector space \mathbb{F}_2^8 is given by

$$(x_0, x_1, \dots, x_7) \in \mathbb{F}_2^8 \mapsto \sum_{i=0}^7 x_i X^i$$

where all operations are modulo the irreducible polynomial

$$X^8 + X^4 + X^3 + X + 1.$$

With this identification, the AES Sbox corresponds to the inversion in \mathbb{F}_{2^8} (where the inverse of 0 is 0), i.e.,

$$x \in \mathbb{F}_{2^8} \mapsto x^{2^{54}},$$

followed by an affine function over \mathbb{F}_2^8 :

$$\begin{bmatrix} y_0 \\ y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \\ y_6 \\ y_7 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 1 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 & 0 & 1 & 1 & 1 \\ 1 & 1 & 1 & 0 & 0 & 0 & 1 & 1 \\ 1 & 1 & 1 & 1 & 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} x_0 \\ x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \end{bmatrix} + \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \\ 0 \\ 1 \\ 1 \\ 0 \end{bmatrix}.$$

Some properties of the univariate representation of an Sbox are related to the ANF of its components. For instance, the degree of S can be deduced from the degree of its univariate representation.

Proposition 2.5. *Let S be an n -bit Sbox and let*

$$\sum_{i=0}^{2^n-1} A_i X^i$$

denote its univariate representation in $\mathbb{F}_{2^n}[X]$. Then,

$$\deg(f) = \max\{wt(i) : 0 \leq i < 2^n \text{ and } A_i \neq 0\}$$

Proof. Let $\{\alpha_0, \dots, \alpha_{n-1}\}$ be a basis of \mathbb{F}_{2^n} . We first prove that the degree of S cannot exceed the maximum weight of an exponent i with $A_i \neq 0$. Indeed, the multivariate representation of the monomial X^s over \mathbb{F}_{2^n} is

$$\begin{aligned} X^s &= (\alpha_0 x_0 + \alpha_1 x_1 + \dots + \alpha_{n-1} x_{n-1})^s \\ &= (\alpha_0 x_0 + \alpha_1 x_1 + \dots + \alpha_{n-1} x_{n-1})^{\sum_{i=0}^{n-1} s_i 2^i} \\ &= \prod_{i=0}^{n-1} \left(\alpha_0^{2^i} x_0^{2^i} + \alpha_1^{2^i} x_1^{2^i} + \dots + \alpha_{n-1}^{2^i} x_{n-1}^{2^i} \right)^{s_i}. \end{aligned}$$

Then, we get a product of $wt(s)$ terms of degree 1 implying that the degree of X^s cannot exceed $wt(s)$.

Conversely, we observe that the number of n -variable Boolean functions of degree at most d is $2^{\dim \mathcal{R}(d,n)}$ with

$$\dim \mathcal{R}(d, n) = \sum_{i=0}^d \binom{n}{i},$$

implying that the number of n -bit Sboxes with degree d is

$$2^{n \dim \mathcal{R}(d,n)}.$$

On the other hand, the number of univariate polynomials composed of monomials X^s with $wt(s) \leq d$ is also

$$(2^n)^{\sum_{i=0}^d \binom{n}{i}} = 2^{n \dim \mathcal{R}(d,n)}.$$

It follows that the two sets of functions are the same. \diamond

2.2 Low-degree relations for an Sbox

The basic principle of algebraic attacks goes back to Shannon's work [Sha49, Page 711]: these techniques consist in expressing the whole cipher as a large system of multivariate algebraic equations. The secret key can then be recovered from some plaintext-ciphertext pairs by solving this algebraic system. Therefore, the degree of the underlying system highly influences the complexity of such an attack. But, Courtois and Pieprzyk [CP02] have pointed out that it might be possible to lower the degree of the polynomial system that we need to solve even if the round function has a high degree. Indeed, the attacker may equivalently exploit any Boolean *relation* between the plaintext bits, the ciphertext bits and the key-bits.

Example 2.4. We can check that, even if the Sbox S defined in Table 2.1 has degree 3, there exist some Boolean relations of degree 2 only between its inputs and outputs, for instance it can be checked that

$$x_2x_4 + x_2S_1(x_1, \dots, x_4) + x_2S_2(x_1, \dots, x_4) = 0$$

for all inputs.

An important quantity is then the lowest degree we can reach for a Boolean relation between the inputs and outputs of an Sbox.

Proposition 2.6. *Let S be a function from \mathbb{F}_2^n into \mathbb{F}_2^n . The Boolean relations of degree at most d between the inputs and outputs of S are the elements of the kernel of the binary matrix with $\sum_{i=0}^d \binom{2n}{i}$ rows and 2^n columns whose rows correspond to the value vectors of the n -variable Boolean function*

$$x^u S(x)^v, u, v \in \mathbb{F}_2^n \text{ such that } wt(u) + wt(v) \leq d.$$

Most notably, the number of linearly independent relations of degree at most d is at least

$$\sum_{i=0}^d \binom{2n}{i} - 2^n.$$

Proof. This result comes from the fact that the relations between x and $S(x)$ correspond to the functions

$$\sum_{u,v \in \mathbb{F}_2^n} c_{u,v} x^u [S(x)]^v$$

which vanish for all possible inputs x . Moreover, the degree of such a relation corresponds to the maximal value of $wt(u) + wt(v)$ such that $c_{u,v} = 1$. Any such relation is then defined by a linear combination of the monomials $x^u S(x)^v$ which vanish at all points, i.e., an element in the kernel of the matrix defined by the value vectors of all these monomials. The involved matrix has 2^n columns, and $\sum_{i=0}^d \binom{2n}{i}$ rows (corresponding to all possible pairs (u, v)). Then, the dimension of its kernel exceeds the difference between the number of rows and the number of columns. \diamond

We deduce for instance that any function from \mathbb{F}_2^4 into \mathbb{F}_2^4 has at least

$$\sum_{i=0}^2 \binom{8}{i} - 2^4 = 37 - 16 = 21$$

quadratic relations between its inputs and outputs.

Case of the AES. The AES Sbox can be seen as the composition of the inversion over \mathbb{F}_{2^8} with an affine function. More precisely,

$$S : A \circ \varphi^{-1} \circ (\varphi(x))^{254},$$

where φ is the isomorphism from \mathbb{F}_2^8 into \mathbb{F}_{2^8} defined by the basis $\{1, \alpha, \dots, \alpha^7\}$ and α is a root of $X^8 + X^4 + X^3 + X + 1$. By definition, the inversion Inv over \mathbb{F}_{2^8} satisfies

$$x^2 \text{Inv}(x) = x^2 x^{254} = x.$$

Since $x \mapsto x^2$ is an \mathbb{F}_2 -linear mapping over \mathbb{F}_2^8 , we deduce that this relation corresponds to eight Boolean relations over \mathbb{F}_2 between the inputs and outputs of S . Actually, there exists 39 such quadratic relations for the AES Sbox. This is much higher than expected for a randomly chosen mapping over \mathbb{F}_2^8 , since Proposition 2.6 shows that there is no relation of degree 3 for a random 8-bit Sbox.

These relations of degree 2 can be used for exhibiting a quadratic algebraic system modeling the AES, by introducing intermediate variables that correspond to the outputs of the successive rounds [CP02]. However, solving this system is infeasible due to its size: 8000 quadratic equations of 1600 variables.

Other attacks exploiting a low degree. It is worth noting that there are some other attacks exploiting that the Sbox has a low degree, or more generally that other structural properties of its multivariate representation can be used as integral distinguishers. A first example is the higher-order differential cryptanalysis [Knu95] which exploits the fact that the Boolean function corresponding to a combination of the ciphertext bits has a low degree. A seminal example is the cryptanalysis of the \mathcal{KN} cipher by Jakobsen and Knudsen [JK97]. Higher-order differential attacks on block ciphers usually exploit a low degree of several iterations of the round function, and then require a careful analysis of the growth of the degree when a function is iterated. This analysis depends on the general structure of the cipher (e.g. SPN or Feistel structure). Some results on SPN can be found in [BCD11] for instance. More sophisticated integral distinguishers can be derived from coefficients of the ANF that do not depend on the secret key (or have a low degree in the key bits). They include *cube attacks* [DS09] for instance. Exhibiting such a monomial is a priori much more difficult than finding an upper bound on the degree, but this can be done by a cryptanalytic tool called the *division property*, introduced by Todo [Tod15b]. This tool and its variants [TM16, BC16, HLM⁺20, HSWW20] aim at tracing key-independent values of integrals (or equivalently the presence or absence of monomials) through the rounds of an iterated cipher. It has been proved very helpful for finding integral attacks against several ciphers, including the full Misty-1 [Tod15a].

A different type of attacks is the so-called *interpolation attack* introduced in [JK97]. It exploits the fact that the encryption function has a low *univariate degree*. For any given key, the univariate representation of the encryption (or decryption) function can then be recovered by Lagrange interpolation as soon as enough plaintext-ciphertext pairs are known. The data complexity is then determined by the univariate degree of the function. An interesting point is that this technique provides the attacker with a way to decrypt any ciphertext but does not recover the key.

2.3 Linear properties of Sboxes

2.3.1 Linear cryptanalysis

The principle of linear cryptanalysis has been originally presented by Gilbert, Chassé and Tardy-Corffdir [GC91, TCG91] on the block cipher FEAL, and then applied to DES by Matsui [Mat94, Mat95]. It exploits, as a distinguisher a linear Boolean relation between the input bits, output bits and the key-bits of the cipher, which is biased, i.e., which holds with a probability different from 1/2. In other words, it uses a triple of *masks* $(\alpha, \beta, \gamma) \in \mathbb{F}_2^n \times \mathbb{F}_2^n \times \mathbb{F}_2^\kappa$

such that

$$\Pr_x[\alpha \cdot x + \beta \cdot E_k(x) + \gamma \cdot k = 0] = \frac{1}{2}(1 + \varepsilon) \text{ with } \varepsilon \neq 0.$$

2.3.2 Linearity of an Sbox

In a linear attack, we then search for the most biased linear relation between the input and output bits of the Sbox. For a pair of masks (a, b) , the corresponding bias is determined by the Walsh coefficient of S_b at point a , $\mathcal{E}(S_b + \varphi_a)$. These biases are usually represented as a two-dimensional table $\mathcal{E}(a, b) = \mathcal{E}(S_b + \varphi_a)$ called the *linear-approximation table (LAT)*.

Example 2.5. The LAT of the Sbox defined in Table 2.1 corresponds to the following table.

$a \setminus b$	1	2	3	4	5	6	7	8	9	a	b	c	d	e	f
1	-4	.	4	.	-4	8	-4	4	8	4	.	-4	.	4	.
2	4	-4	.	-4	.	.	4	4	8	.	4	8	-4	-4	.
3	8	4	4	-4	4	4	-4	-4	-4	.	8
4	.	-4	4	4	-4	.	.	-8	.	4	4	4	4	.	8
5	-4	4	.	4	8	.	4	-4	8	.	-4	.	4	-4	.
6	-4	.	4	.	4	8	4	4	-8	4	.	4	.	-4	.
7	.	.	.	8	.	-8	8	.	8	.	.
8	.	-4	4	-8	.	4	4	-8	.	-4	-4	.	.	4	-4
9	-4	-12	.	.	4	-4	.	4	.	.	-4	-4	.	.	4
a	-4	.	-12	-4	.	4	.	-4	.	4	.	.	-4	.	4
b	.	.	.	4	-4	4	-4	.	.	-8	-8	4	-4	-4	4
c	.	.	.	-4	-4	-4	-4	.	.	8	-8	4	4	-4	-4
d	-4	.	4	4	.	-4	.	-4	.	4	.	.	-12	.	-4
e	4	-4	.	.	4	4	-8	-4	.	.	4	-4	.	-8	-4
f	-8	4	4	-8	.	-4	-4	.	.	-4	4	.	.	-4	4

A major parameter is then the highest magnitude of an element in this table, which corresponds to the linearity of the Sbox.

Definition 2.7. Let S be function from \mathbb{F}_2^n into \mathbb{F}_2^m . Then, the linearity of S is the highest linearity of any of its non-trivial components, i.e.

$$\mathcal{L}(S) = \max_{b \in (\mathbb{F}_2^m)^*} \mathcal{L}(S_b) = \max_{a, b \in (\mathbb{F}_2^m)^*} |\mathcal{E}(S_b + \varphi_a)|.$$

The linearity of S is now related to the weights of the following code.

Proposition 2.8 ([CCZ98a]). Let S be an Sbox from \mathbb{F}_2^n into \mathbb{F}_2^n . Let us consider the code Γ_S having for generator matrix the following $(2n + 1) \times 2^n$ -matrix

$$\begin{pmatrix} S(0) & S(1) & S(2) & \dots & S(2^n - 1) \\ 0 & 1 & 2 & \dots & 2^n - 1 \\ 1 & 1 & 1 & \dots & 1 \end{pmatrix}$$

where the integers in the first two rows of the matrix correspond to n -bit vectors, i.e., each element in these two rows is an n -bit column vector. The Hamming weights of this code are

$$\{2^{n-1} - \frac{1}{2}\mathcal{E}(S_b + \varphi_a); \ 2^{n-1} + \frac{1}{2}\mathcal{E}(S_b + \varphi_a), a, b \in \mathbb{F}_2^n\}$$

The proof is similar to the proof of Proposition 1.19. Note that the case a and b equal to zero should be included in order to take into account the words of $\mathcal{R}(1, n)$. Based on some relationships satisfied by the weight distribution of a code with such parameters, we can deduce a lower bound on the linearity of an Sbox.

Proposition 2.9. [CV95, CCZ98a] *Let S be a function from \mathbb{F}_2^n into \mathbb{F}_2^n . Then*

$$\mathcal{L}(S) \geq 2^{\frac{n+1}{2}}.$$

The Sboxes for which equality holds are called almost bent (AB) functions. They exist for n odd only.

When n is even, almost bent functions do not exist, and the lowest possible linearity for an n -bit Sbox is not known. But, the best known value is $\mathcal{L}(S) = 2^{\frac{n}{2}+1}$, and this value is tight for a very few families of Sboxes, including the inversion over \mathbb{F}_{2^n} which is used in the AES.

2.4 Differential properties of Sboxes

2.4.1 Differential cryptanalysis

Differential cryptanalysis has been introduced by Biham and Shamir [BS91]. It exploits as a distinguishing property the existence of a *differential*, i.e., of a pair of differences (a, b) in \mathbb{F}_2^n such that

$$\Pr_X[E_k(X + a) + E_k(X) = b]$$

is high. For a randomly chosen permutation, the derivative $X \mapsto E_k(X + a) + E_k(X)$ is expected to take every possible nonzero value with uniform probability. It follows that a differential can be used as a distinguisher if the corresponding probability p is significantly different from 2^{-n} .

2.4.2 Differential uniformity

Then, a relevant parameter for evaluating the resistance of a cipher against differential attacks is the maximal probability that a given nonzero input difference leads to a given output difference for the Sbox. This quantity is determined by the *differential uniformity* of S .

Definition 2.10 (Differential uniformity [Nyb93]). *Let S be a function from \mathbb{F}_2^n into \mathbb{F}_2^m . For any $a \in \mathbb{F}_2^n$ and $b \in \mathbb{F}_2^m$, we define*

$$\delta(a, b) = \#\{x \in \mathbb{F}_2^n : S(x + a) + S(x) = b\}.$$

Then the multi-set $\{\delta(a, b), a \in \mathbb{F}_2^n \setminus \{0\}, b \in \mathbb{F}_2^m\}$ is the differential spectrum of S , and its maximum

$$\delta_S = \max_{a \neq 0, b} \delta(a, b)$$

is the differential uniformity of S .

The values $(\delta(a, b))_{a \in \mathbb{F}_2^n, b \in \mathbb{F}_2^m}$ are usually represented as a two-dimensional array called the *difference distribution table (DDT)* of S .

Example 2.6. The DDT of the Sbox in Table 2.1 is as follows.

$a \setminus b$	1	2	3	4	5	6	7	8	9	a	b	c	d	e	f
1	2	0	4	2	0	2	2	0	0	0	2	0	0	0	2
2	2	2	0	2	4	0	2	0	4	0	0	0	0	0	0
3	2	0	4	0	2	0	0	0	0	6	0	0	0	2	0
4	2	0	2	4	0	0	0	2	2	0	0	2	0	0	2
5	0	4	2	0	0	0	2	2	0	0	4	2	0	0	0
6	4	0	0	0	0	4	0	4	0	0	0	0	4	0	0
7	0	2	0	0	2	2	2	0	2	2	2	0	0	2	0
8	0	4	0	0	0	4	0	0	0	0	0	0	4	0	4
9	2	2	0	2	2	0	0	0	4	0	0	2	0	2	0
a	0	0	2	2	0	2	2	2	0	2	2	0	0	0	2
b	0	0	2	0	4	0	2	2	0	0	0	6	0	0	0
c	0	2	0	0	0	2	0	0	2	2	2	2	0	4	0
d	2	0	0	0	2	0	0	0	0	2	0	0	8	2	0
e	0	0	0	0	0	0	4	0	0	0	4	0	0	4	4
f	0	0	0	4	0	0	0	4	2	2	0	2	0	0	2

The lowest differential uniformity that can be achieved by an Sbox is determined in the following proposition.

Proposition 2.11 ([NK93]). *Let S be a function from \mathbb{F}_2^n into \mathbb{F}_2^n . Then, its differential uniformity satisfies*

$$\delta(S) \geq 2.$$

The Sboxes for which equality holds are called almost perfect nonlinear (APN) functions.

Proof. The proof comes from the fact that

$$\delta(a, b) = \#\{x \in \mathbb{F}_2^n, S(x+a) + S(x) = b\}$$

is always even since, if x is a solution of this equation, $(x+a)$ is a solution too. \diamond

The APN terminology comes from the fact that we call *perfect nonlinear* the functions from \mathbb{F}_2^n into \mathbb{F}_2^m such that $\delta(S) = 2^{n-m}$. It can be proved that the perfect nonlinear functions correspond to the functions with linearity $\mathcal{L}(S) = 2^{\frac{n}{2}}$, i.e., such that all their components are bent. Such functions exist only when $n \geq 2m$ [Nyb91]. Therefore, when $m = n$, the optimal differential uniformity is obtained for the almost perfect nonlinear functions.

2.5 Link between differential uniformity and linearity

The differential spectrum and the Walsh spectrum of an Sbox are strongly related. This link appears under several forms. We first focus on the coding approach.

We have seen in Proposition 2.8 that the Walsh spectrum of an Sbox is related to the weight distribution of the code Γ_S with generator matrix

$$G = \begin{pmatrix} S(0) & S(1) & S(2) & \dots & S(2^n - 1) \\ 0 & 1 & 2 & \dots & 2^n - 1 \\ 1 & 1 & 1 & \dots & 1 \end{pmatrix} \quad (2.1)$$

Now, we show that the APN property is also related to this code.

Proposition 2.12 ([CCZ98a]). *Let S be an Sbox from \mathbb{F}_2^n into \mathbb{F}_2^n . Let us consider the code Γ_S defined by the generator matrix in Eq. (2.1). Then, the minimum distance of Γ_S^\perp is equal to 6 if S is APN and equal to 4 otherwise.*

Proof. It can be easily checked that the minimum distance of Γ_S^\perp is even (since Γ_S contains the all-one codeword), and at least 4. Moreover, it cannot be greater than 6 [DZ84, BT93]. Now, a codeword of weight 4 of Γ_S^\perp corresponds to four columns of the generator matrix G which sum to zero (since the words of Γ_S^\perp are the column vectors c such that $Gc = 0$). Therefore, Γ_S^\perp has minimum distance 4 if and only if there exist four distinct elements x_1, x_2, x_3, x_4 such that

$$x_1 + x_2 + x_3 + x_4 = 0 \text{ and } S(x_1) + S(x_2) + S(x_3) + S(x_4) = 0 .$$

Replacing x_2 by $x_1 + a$, we get that it is equivalent to the existence of x_1, x_3 and a such that

$$S(x_1) + S(x_1 + a) = S(x_3) + S(x_3 + a)$$

since $x_4 = x_3 + (x_1 + x_2) = x_3 + a$. All four x_i are distinct if and only if they correspond to four distinct solutions of the equation

$$S(x + a) + S(x) = b$$

with $b = S(x_1) + S(x_1 + a)$. This equivalently means that S is not APN. \diamond

The APN and AB properties then correspond to a particular value of the minimum distance of the code Γ_S^\perp and Γ_S . Using the relations between the weight distributions of the two codes, it can be deduced that these two properties are related as follows.

Proposition 2.13 ([CV95, CCZ98a]). *Any almost bent Sbox is also APN.*

This result comes from the fact that, if the minimum distance of Γ_S is maximal (i.e., equal to $2^{n-1} - 2^{\frac{n-1}{2}}$), then the minimum distance of Γ_S^\perp is also maximal (i.e., equal to 6). It is worth noticing that the converse does not hold in general, except for Sboxes of degree 2 [CCZ98a].

Another approach used in [CV95] and revisited in [BN13] provides a relation between the DDT and the LAT. Indeed, the squared Walsh coefficients of an Sbox correspond to the Walsh transform of its DDT.

Theorem 2.14. [CV95, BN13] *Let S be a function from \mathbb{F}_2^n into \mathbb{F}_2^n . Let $\delta(a, b)$ denote the entries in the difference table of S . Then, for any λ and μ in \mathbb{F}_2^n , we have*

$$\mathcal{E}^2(S_\mu + \varphi_\lambda) = \sum_{a \in \mathbb{F}_2^n} \sum_{b \in \mathbb{F}_2^n} (-1)^{a \cdot \lambda + b \cdot \mu} \delta(a, b) .$$

Conversely, for any a and b in \mathbb{F}_2^n ,

$$\delta(a, b) = 2^{-2n} \sum_{\lambda \in \mathbb{F}_2^n} \sum_{\mu \in \mathbb{F}_2^n} (-1)^{a \cdot \lambda + b \cdot \mu} \mathcal{E}^2(S_\mu + \varphi_\lambda) .$$

Proof. We prove the second formula only, since the other one is directly derived by inverting the Walsh transform. Let a and b in \mathbb{F}_2^n . Then

$$\begin{aligned}
\mathcal{A} &= 2^{-2n} \sum_{\lambda \in \mathbb{F}_2^n} \sum_{\mu \in \mathbb{F}_2^n} (-1)^{a \cdot \lambda + b \cdot \mu} \mathcal{E}^2(S_\mu + \varphi_\lambda) \\
&= 2^{-2n} \sum_{\lambda \in \mathbb{F}_2^n} \sum_{\mu \in \mathbb{F}_2^n} (-1)^{a \cdot \lambda + b \cdot \mu} \left(\sum_{x \in \mathbb{F}_2^n} (-1)^{\mu \cdot S(x) + \lambda \cdot x} \right) \left(\sum_{y \in \mathbb{F}_2^n} (-1)^{\mu \cdot S(y) + \lambda \cdot y} \right) \\
&= 2^{-2n} \sum_{x \in \mathbb{F}_2^n} \sum_{y \in \mathbb{F}_2^n} \left(\sum_{\lambda \in \mathbb{F}_2^n} (-1)^{\lambda \cdot (x+y+a)} \right) \left(\sum_{\mu \in \mathbb{F}_2^n} (-1)^{\mu \cdot (S(x)+S(y)+b)} \right) \\
&= \# \{ (x, y) \in \mathbb{F}_2^n \times \mathbb{F}_2^n : x + y = a \text{ and } S(x) + S(y) = b \} \\
&= \delta(a, b) .
\end{aligned}$$

◇

It is worth noticing that, while a function can be entirely recovered from the knowledge of its Walsh transform (i.e., of its LAT), this is not the case if only the DDT is known. Indeed, the DDT equivalently provides the magnitude of the Walsh coefficients, but does not give any information on the signs.

2.6 Sboxes with good cryptographic properties

We now give some examples of Sboxes with low linearity and differential uniformity. For small values of n , such Sboxes are classified. As previously explained for Boolean functions, this classification takes advantage of several equivalence relations under which the main cryptographic properties are invariant.

2.6.1 Equivalence relations

Definition 2.15. *Two n -bit Sboxes S_1 and S_2 are said to be affine equivalent if there exist two affine permutations of \mathbb{F}_2^n , A_1 and A_2 , such that*

$$S_2 = A_2 \circ S_1 \circ A_1 .$$

Proposition 2.16. *Let S_1 and S_2 be two EA-equivalent Sboxes, with*

$$S_2 = A_2 \circ S_1 \circ A_1 + A_3 .$$

Then, their difference tables and their linear-approximation tables satisfy

$$\begin{aligned}
\delta_{S_2}(a, b) &= \delta_{S_1}(L_1(a), L_2^{-1}(b)) \\
\mathcal{E}_{S_2}(a, b) &= \mathcal{E}_{S_1}((L_1^{-1})^T(a), L_2^T(b))
\end{aligned}$$

where L_1 and L_2 are the linear parts of A_1 and A_2 .

Proof. Let $A_i(x) = L_i(x) + c_i$. We first consider the differential spectrum. An element $x \in \mathbb{F}_2^n$ satisfies $S_2(x + a) + S_2(x) = b$ if and only if

$$L_2 \left[S_1(L_1(x) + L_1(a) + c_1) + S_1(L_1(x) + c_1) \right] = b ,$$

or equivalently

$$S_1(y + L_1(a)) + S_1(y) = L_2^{-1}(b) .$$

For the linear-approximation table, we use that

$$\begin{aligned} b \cdot S_2(x) + a \cdot x &= b \cdot c_2 + b \cdot L_2(S_1(y)) + a \cdot A_1^{-1}(y) \\ &= b \cdot c_2 + L_2^T(b) \cdot S_1(y) + a \cdot L_1^{-1}(c_1) + (L_1^{-1})^T(a) \cdot y . \end{aligned}$$

◇

A more general equivalence notion can be defined as follows.

Definition 2.17. *Two n -bit Sboxes S_1 and S_2 are said to be extended-affine equivalent (EA-equivalent) if there exist two affine permutations of \mathbb{F}_2^n , A_1 and A_2 and an affine function A_3 , such that*

$$S_2 = A_2 \circ S_1 \circ A_1 + A_3 .$$

The most general form of function equivalence that is known to preserve the differential uniformity and the linearity is *CCZ-equivalence*, of which EA-equivalence is a particular case. It was named after Carlet, Charpin and Zinoviev who introduced it in [CCZ98b].

Definition 2.18 (CCZ-equivalence). *Two n -bit Sboxes S_1 and S_2 are said to be CCZ-equivalent if there exists an affine permutation of \mathbb{F}_2^{2n} , A , such that*

$$\{ (x, S_1(x)), x \in \mathbb{F}_2^n \} = A \left(\{ (x, S_2(x)), x \in \mathbb{F}_2^n \} \right) .$$

Obviously, EA-equivalence is a particular case of CCZ-equivalence. In general, given an Sbox S_1 and an affine permutation A of \mathbb{F}_2^{2n} , there a priori does not exist a function S_2 such that

$$A \left(\{ (x, S_1(x)), \forall x \in \mathbb{F}_2^n \} \right) = \{ (x, S_2(x)), \forall x \in \mathbb{F}_2^n \} .$$

Indeed, it is necessary for S_2 to be well-defined that the left-hand side of the output of $x \mapsto A(x, S_2(x))$ is a permutation. As a consequence, only a few permutations A yield valid Sboxes S_2 .

The differential spectrum and the Walsh spectrum are both invariant under CZZ equivalence.

Proposition 2.19. *Let S_1 and S_2 be two n -bit Sboxes such that $\{ (x, S_2(x)), x \in \mathbb{F}_2^n \} = A \left(\{ (x, S_1(x)), x \in \mathbb{F}_2^n \} \right)$ for some affine permutation A of \mathbb{F}_2^{2n} . Let L denote the linear part of A , i.e., $A(x) = L(x) + c$ for some constant $c \in \mathbb{F}_2^{2n}$. Then the DDT and LAT of S_2 are given by*

$$\begin{aligned} \delta_{S_2}(a, b) &= \delta_{S_1}(L^{-1}(a, b)) , \\ \mathcal{E}_{S_2}(\alpha, \beta) &= (-1)^{c \cdot (\alpha, \beta)} \mathcal{E}_{S_1}(L^T(\alpha, \beta)) . \end{aligned}$$

Proof. By definition, $\delta_{S_2}(a, b)$ equals the number of pairs (x, x') of elements in \mathbb{F}_2^n such that

$$x + x' = a \text{ and } S_2(x) + S_2(x') = b .$$

Let $F_1 : x \mapsto A_1(x, F(x))$ and $F_2 : x \mapsto A_2(x, F(x))$ where $A = (A_1, A_2)$. Then, $\delta_{S_2}(a, b)$ equivalently equals the number of $y = F_1^{-1}(x)$ and $y' = F_1^{-1}(x')$ such that

$$F_1(y) + F_1(y') = a \text{ and } F_2(y) + F_2(y') = b .$$

This system corresponds to

$$L_1(y + y', S_1(y) + S_1(y')) = a \text{ and } L_2(y + y', S_1(y) + S_1(y')) = b$$

where L_1 and L_2 are the linear parts of A_1 and A_2 . Equivalently

$$(y + y', S_1(y) + S_1(y')) = L^{-1}(a, b) .$$

The number of pairs satisfying this equation is then $\delta_{S_1}(L^{-1}(a, b))$.

Let us now look at the LAT. The value of $\mathcal{E}_{S_2}(\alpha, \beta)$ is determined by the number of $x \in \mathbb{F}_2^n$ such that

$$\beta \cdot S_2(x) + \alpha \cdot x = 0 .$$

By setting $y = F_1^{-1}(x)$, the left-hand term of this equation can be replaced by

$$\begin{aligned} \beta \cdot F_2(y) + \alpha \cdot F_1(y) &= \beta \cdot A_2(y, S_1(y)) + \alpha \cdot A_1(y, S_1(y)) \\ &= \beta \cdot L_2(y, S_1(y)) + \alpha \cdot L_1(y, S_1(y)) + c \cdot (\alpha, \beta) \\ &= (\alpha, \beta) \cdot L(y, S_1(y)) + c \cdot (\alpha, \beta) \\ &= L^T(\alpha, \beta) \cdot (y, S_1(y)) + c \cdot (\alpha, \beta) \end{aligned}$$

by using that, for any linear function L and any pair (x, y) , $x \cdot L(y) = L^T(x) \cdot y$ where L^T is the linear function corresponding to the transpose of the matrix defining L . \diamond

2.6.2 Odd number of variables

When n is odd, almost bent Sboxes offer the best cryptographic properties since they have both the lowest possible linearity and differential uniformity. The list of all known *almost bent power permutations* of n variables, n odd, is given in Table 2.2.

A few AB permutations have been found which are not equivalent to a power permutation. For instance, when n is odd, divisible by 3 and not by 9.

$$S(x) = x^{2^i+1} + ux^{2^{j\frac{n}{3}}+2^{(3-j)\frac{n}{3}+i}} \text{ with } \gcd(i, n) = 1 \text{ and } j = i\frac{n}{3} \bmod 3$$

is an AB permutation [BCL08]. Other examples are provided in [BL22, LK24] for instance.

2.6.3 4-bit permutations

When n is even, almost bent functions do not exist and finding optimal functions is then much more difficult. For $n = 4$, 4-bit permutations have been classified up to affine equivalence

	exponent s	
Quadratic exponents	$2^i + 1$ with $\gcd(i, n) = 1$, $1 \leq i \leq t$	[Gol68, Nyb93]
Kasami exponents	$2^{2i} - 2^i + 1$ with $\gcd(i, n) = 1$ $2 \leq i \leq t$	[Kas71]
Welch exponent	$2^t + 3$	[Dob99b, CCD00]
Niho exponent	$2^t + 2^{\frac{t}{2}} - 1$ if t is even $2^t + 2^{\frac{3t+1}{2}} - 1$ if t is odd	[Dob99a, HX01]

Table 2.2: Known almost bent power permutations $S : x \mapsto x^s$ over \mathbb{F}_{2^n} with $n = 2t + 1$ (up to inversion).

in [LP07] and in [De 07]. The smallest differential uniformity and linearity which can be achieved for a 4-bit permutation S are

$$\delta(S) \geq 4 \text{ and } \mathcal{L}(S) \geq 8.$$

Up to affine equivalence, there are 16 classes of optimal permutations [LP07]. All of them have degree 3 but half of them have at least one component of degree 2 which may be an unsuitable property. Then, there are eight equivalence classes having all their non-trivial components of degree 3. A representative in each class is given in Table 2.3.

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
G_0	0	1	2	13	4	7	15	6	8	11	12	9	3	14	10	5
G_1	0	1	2	13	4	7	15	6	8	11	14	3	5	9	10	12
G_2	0	1	2	13	4	7	15	6	8	11	14	3	10	12	5	9
G_3	0	1	2	13	4	7	15	6	8	12	5	3	10	14	11	9
G_4	0	1	2	13	4	7	15	6	8	12	9	11	10	14	5	3
G_5	0	1	2	13	4	7	15	6	8	12	11	9	10	14	3	5
G_6	0	1	2	13	4	7	15	6	8	12	11	9	10	14	5	3
G_7	0	1	2	13	4	7	15	6	8	12	14	11	10	9	3	5
G_8	0	1	2	13	4	7	15	6	8	14	9	5	10	11	3	12
G_9	0	1	2	13	4	7	15	6	8	14	11	3	5	9	10	12
G_{10}	0	1	2	13	4	7	15	6	8	14	11	5	10	9	3	12
G_{11}	0	1	2	13	4	7	15	6	8	14	11	10	5	9	12	3
G_{12}	0	1	2	13	4	7	15	6	8	14	11	10	9	3	12	5
G_{13}	0	1	2	13	4	7	15	6	8	14	12	9	5	11	10	3
G_{14}	0	1	2	13	4	7	15	6	8	14	12	11	3	9	5	10
G_{15}	0	1	2	13	4	7	15	6	8	14	12	11	9	3	10	5

Table 2.3: The 8 classes of optimal 4-bit permutations [LP07].

2.6.4 6-bit permutations

It has been conjectured for a long time that APN permutations of an even number of variables did not exist. This has been disproved by Dillon in 2009 who has exhibited an APN permutation of 6 variables [BDMW10]. But this is (up to composition by affine transformations) the only known example of APN permutation of an even number of variables. This permutation is defined by

$$S = \{0, 54, 48, 13, 15, 18, 53, 35, 25, 63, 45, 52, 3, 20, 41, 33, 59, 36, 2, 34, 10, 8, 57, 37, 60, 19, 42, 14, 50, 26, 58, 24, 39, 27, 21, 17, 16, 29, 1, 62, 47, 40, 51, 56, 7, 43, 44, 38, 31, 11, 4, 28, 61, 46, 5, 49, 9, 6, 23, 32, 30, 12, 55, 22\} .$$

It satisfies

$$\delta(S) = 2 , \deg S = 4 \text{ and } \mathcal{L}(S) = 16 .$$

The corresponding univariate polynomial over \mathbb{F}_{2^6} contains 53 nonzero monomials (out of the 56 possible monomials of degree at most 4), see e.g. [BN15, Page 131].

This APN permutation has been derived from a 6-bit non-bijective APN function of degree 2, called the *Kim mapping*, by CCZ-equivalence [BDMW10]. It is worth noting that 6-bit APN functions of degree 2 and 3 are classified [BL08, LSS12] and the Kim mapping is the only function in this list having a permutation in its CCZ-equivalence class.

2.6.5 Sboxes of an even number of variables

For a permutation depending on an even number of variables n , $n \geq 8$, the best known differential uniformity is $\delta(S) = 4$. Then, we usually use permutations S with

$$\delta(S) = 4 \text{ and } \mathcal{L}(S) = 2^{\frac{n+2}{2}} ,$$

even if we do not know whether these two values are optimal or not. Table 2.4 provides the list of all known power permutations over \mathbb{F}_{2^n} , n even, which achieve these values.

$2^i + 1, \gcd(i, n) = 2$	$n \equiv 2 \pmod{4}$	[Gol68]
$2^{2i} - 2^i + 1, \gcd(i, n) = 2$	$n \equiv 2 \pmod{4}$	[Kas71]
$2^n - 2$		[LW90]

Table 2.4: Known power permutations $S : x \mapsto x^s$ over \mathbb{F}_{2^n} with $n = 2t$ (up to inversion) with the best known linearity and differential uniformity.

It is worth noticing that, when n is divisible by 4, the inversion is the only known infinite family of permutations of n variables (up to equivalence) which reaches these two values.

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