Lightweight MDS Involution Matrices

Siang Meng Sim¹, Khoongming Khoo², Frédérique Oggier¹, and Thomas Peyrin¹

> ¹ Nanyang Technological University, Singapore ² DSO National Laboratories, Singapore

ssim011@e.ntu.edu.sg, kkhoongm@dso.org.sg, frederique@ntu.edu.sg, thomas.peyrin@ntu.edu.sg

Abstract. In this article, we provide new methods to look for lightweight MDS matrices, and in particular involutory ones. By proving many new properties and equivalence classes for various MDS matrices constructions such as circulant, Hadamard, Cauchy and Hadamard-Cauchy, we exhibit new search algorithms that greatly reduce the search space and make lightweight MDS matrices of rather high dimension possible to find. We also explain why the choice of the irreducible polynomial might have a significant impact on the lightweightness, and in contrary to the classical belief, we show that the Hamming weight has no direct impact. Even though we focused our studies on involutory MDS matrices, we also obtained results for non-involutory MDS matrices. Overall, using Hadamard or Hadamard-Cauchy constructions, we provide the (involutory or non-involutory) MDS matrices with the least possible XOR gates for the classical dimensions 4×4 , 8×8 , 16×16 and 32×32 in GF(2⁴) and GF(2⁸). Compared to the best known matrices, some of our new candidates save up to 50% on the amount of XOR gates required for an hardware implementation. Finally, our work indicates that involutory MDS matrices are really interesting building blocks for designers as they can be implemented with almost the same number of XOR gates as non-involutory MDS matrices, the latter being usually non-lightweight when the inverse matrix is required.

Key words: lightweight cryptography, Hadamard matrix, Cauchy matrix, involution, MDS.

1 Introduction

Most symmetric key primitives, like block ciphers, stream ciphers or hash functions, are usually based on various components that provide confusion and diffusion. Both concepts are very important for the overall security and efficiency of the cryptographic scheme and extensive studies have been conducted to find the best possible building blocks. The goal of diffusion is basically to spread the internal dependencies as much as possible. Several designs use a weak yet fast diffusion layer based on simple XOR, addition and shifting operation, but another trend is to rely on strong linear diffusion matrices, like Maximal Distance Separable (MDS) matrices. A typical example is the AES cipher [17], which uses a 4×4 matrix in GF(2⁸) to provide diffusion among a vector of 4 bytes. These mathematical objects ensure the designers a perfect diffusion (the underlying linear code meets the Singleton bound), but can be quite heavy to implement. Software performances are usually not so much impacted as memory is not really constrained and table-based implementations directly incorporate the field multiplications in the stored values. However, hardware implementations will usually suffer from an important area requirement due to the Galois field multiplications. The impact will also be visible on the efficiency of software bitslice implementations which basically mimic the hardware computations flow.

Good hardware efficiency has became a major design trend in cryptography, due to the increasing importance of ubiquitous computing. Many lightweight algorithms have recently been proposed, notably block ciphers [12, 14, 19, 9] and hash functions [4, 18, 11]. The choice of MDS matrices played an important role in the reduction of the area required to provide a certain amount of security. In PHOTON, the hash function [18] that was proposed a new type of MDS matrix that can be computed in a serial or recursive manner. This construction greatly reduces the temporary memory (and thus the hardware area) usually required for the computation of the matrix. Such matrices were later used in LED [19] block cipher, or PRIMATES [1] authenticated encryption scheme, and were further studied and generalized in subsequent articles [30, 34, 3, 2, 10]. Even though these serial matrices provide a good way to save area, this naturally comes at the expense of an increased number of cycles to apply the matrix. In general, they are not well suited for round-based or low-latency implementations.

Another interesting property for an MDS matrix to save area is to be involutory. Indeed, in most use cases, encryption and decryption implementations are required and the inverse of the MDS matrix will have to be implemented as well (except for constructions like Feistel networks, where the inverse of the internal function is not needed for decryption). For example, the MDS matrix of AES is quite lightweight for encryption, but not really for decryption³. More generally, it is a valuable advantage that one can use *exactly* the same diffusion matrix for encryption and decryption. Some ciphers like ANUBIS [5], KHAZAD [6], ICEBERG [33] or PRINCE [13] even pushed the involution idea a bit further by defining a round function that is almost entirely composed of involution operations, and where the non-involution property of the cipher is mainly provided by the key schedule.

³ The serial matrix construction proposed in [18, 19] allows an efficient inverse computation if the first coefficient is equal to 1. However, we recall that serial matrices are not well suited for round-based or low-latency implementations.

There are several ways to build a MDS matrix [35, 24, 27, 31, 20, 15], a common method being to use a circulant construction, like for the AES block cipher [17] or the WHIRLPOOL hash function [8]. The obvious benefit of a circulant matrix for hardware implementations is that all of its rows are similar (up to a right shift), and one can trivially reuse the multiplication circuit to save implementation costs. However, it has been proven in [23] that circulant matrices of order 4 cannot be simultaneously MDS and involutory. And very recently Gupta *et al.* [21] proved that circulant MDS involutory matrices do not exist. Finding lightweight matrices that are both MDS and involutory is not an easy task and this topic has attracted attention recently. In [31], the authors consider Vandermonde or Hadamard matrices, while in [35, 20, 15] Cauchy matrices were used. Even if these constructions allow to build involutory MDS matrices for big matrix dimensions, it is difficult to find the most lightweight candidates as the search space can become really big.

Our contributions. In this article, we propose a new method to search for lightweight MDS matrices, with an important focus on involutory ones. After having recalled the formula to compute the XOR count, i.e. the amount of XORs required to evaluate one row of the matrix, we show in Section 2 that the choice of the irreducible polynomial is important and can have a significant impact on the efficiency, as remarked in [26]. In particular, we show that the best choice is not necessarily a low Hamming weight polynomial as widely believed, but instead one that has a high standard deviation regarding its XOR count. Then, in Section 3, we recall some constructions to obtain (involutory) MDS matrices: circulant, Hadamard, Cauchy and Cauchy-Hadamard. In particular, we prove new properties for some of these constructions, which will later help us to find good matrices. In Section 4 we prove the existence of equivalent classes for Hadamard matrices and involutory Hadamard-Cauchy matrices and we use these considerations to conceive improved search algorithms of lightweight (involutory) MDS matrices. We describe these new algorithms in Section 5 for lightweight involutory MDS matrices. Our methods can also be relaxed and applied to the search of lightweight non-involutory MDS matrices as explained in Section 6. These algorithms are significant because they are feasible exhaustive search while the search space of the algorithms described in [20, 15] is too big to be exhausted⁴. Our algorithms guarantee that the matrices found are the lightest according to our metric.

Overall, using Hadamard or Hadamard-Cauchy constructions, we provide the smallest known (involutory or noninvolutory) MDS matrices for the classical dimensions 4×4 , 8×8 , 16×16 and 32×32 in GF(2⁴) and GF(2⁸). The designers of one of the CAESAR competition candidates, Joltik [22], have used one of the matrices that we have found to build their primitive. All our results are summarized and commented in Section 7. Surprisingly, it seems that involutory MDS matrices are not much more expensive than non-involutory MDS ones, the former providing the great advantage of a free inverse implementation as well. We recall that in this article we are not considering serial matrices, as their evaluation either requires many clock cycles (for serial implementations) or an important area (for round-based implementations).

Notations and preliminaries. We denote by $GF(2^r)$ the finite field with 2^r elements, $r \ge 1$. This field is isomorphic to polynomials in GF(2)[X] modulo an irreducible polynomial p(X) of degree r, meaning that every field element can be seen as a polynomial $\alpha(X)$ with coefficients in GF(2) and of degree r - 1: $\alpha(X) = \sum_{i=0}^{r-1} b_i X^i$, $b_i \in GF(2)$, $0 \le i \le r-1$. The polynomial $\alpha(X)$ can also naturally be viewed as an r-bit string $(b_{r-1}, b_{r-2}, ..., b_0)$. In the rest of the article, an element α in $GF(2^r)$ will be seen either as the polynomial $\alpha(X)$, or the r-bit string represented in a hexadecimal representation, which will be prefixed with 0x. For example, in $GF(2^8)$, the 8-bit string 00101010 corresponds to the polynomial $X^5 + X^3 + X$, written 0x2a in hexadecimal.

The addition operation on $GF(2^r)$ is simply defined as a bitwise XOR on the coefficients of the polynomial representation of the elements, and does not depend on the choice of the irreducible polynomial p(X). However, for multiplication, one needs to specify the irreducible polynomial p(X) of degree r. We denote this field as $GF(2^r)/p(X)$, where p(X) can be given in hexadecimal representation⁵. The multiplication of two elements is then the modulo p(X)reduction of the product of the polynomial representations of the two elements.

Finally, we denote by M[i, j] the (i, j) entry of the matrix M, we start the counting from 0, that is M[0, 0] is the entry corresponding to the first row and first column.

2 Analyzing XOR count according to different finite fields

In this section, we explain the XOR count that we will use as a measure to evaluate the lightweightness of a given matrix. Then, we will analyze the XOR count distribution depending on the finite field and irreducible polynomial considered. Although it is known that finite fields of the same size are isomorphic to each other and it is believed that

⁴ The huge search space issue can be reduced if one could search intelligently only among lightweight matrix candidates. However, this is not possible with algorithms from [20, 15] since the matrix coefficients are known only at the end of the matrix generation, and thus one cannot limit the search to lightweight candidates only.

⁵ This should not be confused with the explicit construction of finite fields, which is commonly denoted as $\operatorname{GF}(2^r)[X]/(P)$, where (P) is an ideal generated by irreducible polynomial P.

the security of MDS matrices is not impacted by this choice, looking at the XOR count is a new aspect of finite fields that remains unexplored in cryptography.

2.1 The XOR count

It is to note that the XOR count is an easy-to-manipulate and simplified metric, but MDS coefficients have often been chosen to lower XOR count, e.g. by having low Hamming weight. As shown in [26], low XOR count is strongly correlated minimization of hardware area.

Later in this article, we will study the hardware efficiency of some diffusion matrices and we will search among huge sets of candidates. One of the goals will therefore be to minimize the area required to implement these lightweight matrices, and since they will be implemented with XOR gates (the diffusion layer is linear), we need a way to easily evaluate how many XORs will be required to implement them. We explain our method in this subsection.

In general, it is known that low Hamming weight generally requires lesser hardware resource in implementations, and this is the usual choice criteria for picking a matrix. For example, the coefficients of the AES MDS matrix are 1, 2 and 3, in a hope that this will ensure a lightweight implementation. However, it was shown in [26] that while this heuristic is true in general, it is not always the case. Due to some reduction effects, and depending on the irreducible polynomial defining the computation field, some coefficients with not-so-low Hamming weight might be implemented with very few XORs.

Definition 1 The XOR count of an element α in the field $GF(2^r)/p(X)$ is the number of XORs required to implement the multiplication of α with an arbitrary β over $GF(2^r)/p(X)$.

For example, let us explain how we compute the XOR count of $\alpha = 3$ over $GF(2^4)/0\times 13$ and $GF(2^4)/0\times 19$. Let (b_3, b_2, b_1, b_0) be the binary representation of an arbitrary element β in the field. For $GF(2^4)/0\times 13$, we have:

$$(0,0,1,1) \cdot (b_3,b_2,b_1,b_0) = (b_2,b_1,b_0 \oplus b_3,b_3) \oplus (b_3,b_2,b_1,b_0) = (b_2 \oplus b_3,b_1 \oplus b_2,b_0 \oplus b_1 \oplus b_3,b_0 \oplus b_3),$$

which corresponds to 5 XORs⁶. For $GF(2^4)/0x19$, we have:

$$(0,0,1,1) \cdot (b_3,b_2,b_1,b_0) = (b_2 \oplus b_3,b_1,b_0,b_3) \oplus (b_3,b_2,b_1,b_0) = (b_2,b_1 \oplus b_2,b_0 \oplus b_1,b_0 \oplus b_3)$$

which corresponds to 3 XORs. One can observe that XOR count is different depending on the finite field defined by the irreducible polynomial.

In order to calculate the number of XORs required to implement an entire row of a matrix, we can use the following formula given in [26]:

XOR count for one row of
$$M = (\gamma_1, \gamma_2, ..., \gamma_k) + (n-1) \cdot r,$$
 (1)

where γ_i is the XOR count of the *i*-th entry in the row of M, n being the number of nonzero elements in the row and r the dimension of the finite field.

For example, the first row of the AES diffusion matrix being (1, 1, 2, 3) over the field $GF(2^8)/0x11b$, the XOR count for the first row is $(0 + 0 + 3 + 11) + 3 \times 8 = 38$ XORs (the matrix being circulant, all rows are equivalent in terms of XOR count).

2.2 XOR count for different finite fields

We programmed a tool that computes the XOR count for every nonzero element over $GF(2^r)$ for r = 2, ..., 8 and for all possible irreducible polynomials are provided in Appendix D. By analyzing the outputs of this tool, we could make two observations that are important to understand how the choice of the irreducible polynomial affects the XOR count. Before presenting our observations, we state some terminologies and properties related to reciprocal polynomials in finite fields.

Definition 2 A reciprocal polynomial $\frac{1}{p}(X)$ of a polynomial p(X) over $\operatorname{GF}(2^r)$, is a polynomial expressed as $\frac{1}{p}(X) = X^r \cdot p(X^{-1})$. A reciprocal finite field, $\mathbf{K} = \operatorname{GF}(2^r)/\frac{1}{p}(X)$, is a finite field defined by the reciprocal polynomial which defines $\mathbf{F} = \operatorname{GF}(2^r)/p(X)$.

In other words, a reciprocal polynomial is a polynomial with the order of the coefficients reversed. For example, the reciprocal polynomial of p(X) = 0x11b in $GF(2^8)$ is $\frac{1}{p}(X) = 0x\frac{1}{11b} = 0x1b1$. It is also to be noted that the reciprocal polynomial of an irreducible polynomial is also irreducible.

⁶ We acknowledge that one can perform the multiplication with 4 XORs as $b_0 \oplus b_3$ appears twice. But that would require additional cycle and extra memory cost which completely outweighed the small saving on the XOR count.

The total XOR count. Our first new observation is that even if for an individual element of the field the choice of the irreducible polynomial has an impact on the XOR count, the total sum of the XOR count over all elements in the field is independent of this choice. We state this in the following theorem, the proof being provided in Appendix A.

Theorem 1 The total XOR count for a field $GF(2^r)$ is $r \sum_{i=2}^r 2^{i-2}(i-1)$, where $r \ge 2$.

From Theorem 1, it seems that there is no clear implication that one irreducible polynomial is strictly better than another, as the mean XOR count is the same for any irreducible polynomial. However, the irreducible polynomials have different distribution of the XOR count among the field elements, that is quantified by the standard deviation. A high standard deviation implies that the distribution of XOR count is very different from the mean, thus there will be more elements with relatively lower/higher XOR count. In general, the order of the finite field is much larger than the order of the MDS matrix and since only a few elements of the field will be used in the MDS matrices, there is a better chance of finding an MDS matrix with lower XOR count.

Hence, our recommendation is to choose the irreducible polynomial with the highest standard deviation regarding the XOR count distribution. From previous example, in $GF(2^4)$ (XOR count mean equals 4.25 for this field dimension), the irreducible polynomials 0x13 and 0x19 lead to a standard deviation of 2.68, while 0x1f leads to a standard deviation of 1.7075. Therefore, the two first polynomials seem to be a better choice. This observation will allow us to choose the best irreducible polynomial to start with during the searches. We refer to Appendix D for all the standard deviations according to the irreducible polynomial.

We note that the folklore belief was that in order to get lightweight implementations, one should use a low Hamming weight irreducible polynomial. The underlying idea is that with such a polynomial less XORs might be needed when the modular reduction has to be applied during a field multiplication. However, we have shown that this is not necessarily true. Yet, by looking at the data from Appendix D, we remark that the low Hamming weight irreducible polynomials usually have a high standard deviation, which actually validates the folklore belief. We conjecture that this heuristic will be less and less exact when we go to higher and higher order fields.

Matching XOR count. Our second new observation is that the XOR count distribution implied by a polynomial will be the same compared to the distribution of its reciprocal counterpart. We state this observation in the following theorem, the proof being provided in Appendix B.

Theorem 2 There exists an isomorphic mapping from a primitive $\alpha \in \operatorname{GF}(2^r)/p(X)$ to another primitive $\beta \in \operatorname{GF}(2^r)/\frac{1}{p}(X)$ where the XOR count of α^i and β^i is equal for each $i = \{1, 2, ..., 2^r - 1\}$.

In Appendix C, we listed all the primitive mapping from a finite field to its reciprocal finite field for all fields $GF(2^r)$ with r = 2, ..., 8 and for all possible irreducible polynomials. We give an example to illustrate our theorem. For $GF(2^4)$, there are three irreducible polynomials: 0x13, 0x19 and 0x1f and the XOR count for the elements are shown in Appendix D. From the binary representation we see that $0x\frac{1}{13} = 0x19$. Consider an isomorphic mapping ϕ : $GF(2^4)/0x13 \rightarrow GF(2^4)/0x19$ defined as $\phi(2) = 12$, where 2 and 12 are the primitives for the respective finite fields. Table 4 of Appendix C shows that the order of the XOR count is the same.

We remark that for a self-reciprocal irreducible polynomial, for instance 0x1f in $GF(2^4)$, there also exists an automorphism mapping from a primitive to another primitive with the same order of XOR count (see Appendix C).

Theorem 2 is useful for understanding that we do not need to consider $\operatorname{GF}(2^r)/\frac{1}{p}(X)$ when we are searching for lightweight matrices. As there exists an isomorphic mapping preserving the order of the XOR count, any MDS matrix over $\operatorname{GF}(2^r)/\frac{1}{p}(X)$ can be mapped to an MDS matrix over $\operatorname{GF}(2^r)/p(X)$ while preserving the XOR count. Therefore, it is redundant to search for lightweight MDS matrices over $\operatorname{GF}(2^r)/\frac{1}{p}(X)$ as the lightest MDS matrix can also be found in $\operatorname{GF}(2^r)/p(X)$. This will render our algorithms much more efficient: when using exhaustive search for low XOR count MDS over finite field defined by various irreducible polynomial, one can reduce the search space by almost a factor 2 as the reciprocal polynomials are redundant.

3 Types of MDS matrices and properties

In this section, we first recall a few properties of MDS matrices and we then explain various constructions of (involutory) MDS matrices that were used to generate lightweight candidates. Namely, we will study 4 types of diffusion matrices: circulant, Hadamard, Cauchy, and Hadamard-Cauchy. We recall that we do not consider serially computable matrices in this article, like the ones described in [18, 19, 30, 34, 3, 2], since they are not adapted to round-based implementations.

3.1 Maximum Distance Separable matrices

Maximum Distance Separable matrices are crucial components in cryptographic designs, as they ensure a perfect diffusion layer. Since we will search among many lightweight candidate matrices and only keep the MDS ones, we recall in this subsection a few definitions and properties regarding these mathematical objects. We denote by I_k the $k \times k$ identity matrix.

Definition 3 The branch number of a $k \times k$ matrix M over $GF(2^r)$ is the minimum number of nonzero entries in the input vector v and output vector $v \cdot M = u$ (denoted wt(v) and wt(u) respectively), as we range over all $v \in [GF(2^r)]^k - \{0\}$. I.e. the branching number is equal to $\min_{x\neq 0} \{wt(v) + wt(u)\}$, and when the optimal value k + 1is attained, we say M is an MDS matrix.

Definition 4 A length n, dimension k and distance d binary linear code [n, k, d] is called a MDS code if the Singleton bound k = n - d + 1 is met.

From [16, Section 4], we have the following proposition to relate an MDS matrix to a MDS code.

Proposition 1 A $k \times k$ matrix M is an MDS matrix if and only if the standard form generator matrix $[I_k|M]$ generates a (2k, k, k+1)-MDS code.

There are various ways to verify if a matrix is MDS, in this paper we state two of the commonly used statements that can be used to identify MDS matrix.

Proposition 2 ([29], page 321, Theorem 8 - [28], page 53, Theorem 5.4.5) Given a $k \times k$ matrix M, it is an MDS matrix if and only if

1. every square submatrix (formed from any i rows and any i columns, for any i = 1, 2, ..., k) of M is nonsingular, 2. any k columns of $[I_k|M]$ are linearly independent.

The two following corollaries are directly deduced from the first statement of Proposition 2 when we consider submatrices of order 1 and 2 respectively.

Corollary 1 All entries of an MDS matrix are nonzero.

Corollary 2 Given a $k \times k$ matrix M, if there exists pairwise distinct $i_1, i_2, j_1, j_2 \in \{0, 1, ..., k-1\}$ such that $M[i_1, j_1] = M[i_1, j_2] = M[i_2, j_1] = M[i_2, j_2]$, then M is not an MDS matrix.

3.2 Circulant matrices

A common way to build an MDS matrix is to start from a circulant matrix, reason being that the probability of finding an MDS matrix would then be higher than a normal square matrix [16].

Definition 5 $A \ k \times k$ matrix C is circulant when each row vector is rotated to the right relative to the preceding row vector by one element. The matrix is then fully defined by its first row.

An interesting property of circulant matrices is that since each row differs from the previous row by a right shift, a user can just implement one row of the matrix multiplication in hardware and reuse the multiplication circuit for subsequent rows by just shifting the input. However in this paper, we will show in Section 5.1 and 6.1 that these matrices are not the best choice.

3.3 Hadamard matrices

Definition 6 ([20]) A finite field Hadamard (or simply called Hadamard) matrix H is a $k \times k$ matrix, with $k = 2^s$, that can be represented by two other submatrices H_1 and H_2 which are also Hadamard matrices:

$$H = \begin{pmatrix} H_1 & H_2 \\ H_2 & H_1 \end{pmatrix}.$$

Similarly to [20], in order to represent a Hadamard matrix we use notation $had(h_0, h_1, ..., h_{k-1})$ (with $h_i = H[0, i]$ standing for the entries of the first row of the matrix) where $H[i, j] = h_{i \oplus j}$ and $k = 2^s$. It is clear that a Hadamard matrix is bisymmetric. Indeed, if we define the left and right diagonal reflection transformations as $H_L = T_L(H)$ and $H_R = T_R(H)$ respectively, we have that $H_L[i, j] = H[j, i] = H[i, j]$ and $H_R[i, j] = H[k-1-i, k-1-j] = H[i, j]$ (the binary representation of $k - 1 = 2^s - 1$ is all 1, hence $k - 1 - i = (k - 1) \oplus i$).

Moreover, by doing the multiplication directly, it is known that if $H = had(h_0, h_1, ..., h_{k-1})$ is a Hadamard matrix, then $H \times H = c^2 \cdot I$, with $c^2 = h_0^2 + h_1^2 + h_2^2 + ... + h_{k-1}^2$. In other words, the product of a Hadamard matrix with itself is a multiple of an identity matrix, where the multiple c^2 is the sum of the square of the elements from the first row.

A direct and crucial corollary to this fact is that a Hadamard matrix over $GF(2^r)$ is involution if the sum of the elements of the first row is equal to 1. Now, it is important to note that if one deals with a Hadamard matrix for which the sum of the first row over $GF(2^r)$ is nonzero, we can very simply make it involutory by dividing it with the sum of its first row.

We will use these considerations in Section 5.2 to generate low dimension diffusion matrices (order 4 and 8) with an innovative exhaustive search over all the possible Hadamard matrices. We note that, similarity to a circulant matrix, an Hadamard matrix will have the interesting property that each row is a permutation of the first row, therefore allowing to reuse the multiplication circuit to save implementation costs.

3.4 Cauchy matrices

Definition 7 A square Cauchy matrix, C, is a $k \times k$ matrix constructed with two disjoint sets of elements from $GF(2^r)$, $\{\alpha_0, \alpha_1, ..., \alpha_{k-1}\}$ and $\{\beta_0, \beta_1, ..., \beta_{k-1}\}$ such that $C[i, j] = \frac{1}{\alpha_i + \beta_i}$.

It is known that the determinant of a square Cauchy matrix, C, is given as

$$\det(C) = \frac{\prod_{0 \le i < j \le k-1} (\alpha_j - \alpha_i)(\beta_j - \beta_i)}{\prod_{0 \le i < j \le k-1} (\alpha_i + \alpha_j)}$$

Since $\alpha_i \neq \alpha_j$, $\beta_i \neq \beta_j$ for all $i, j \in \{0, 1, ..., k-1\}$, a Cauchy matrix is nonsingular. Note that for a Cauchy matrix over GF(2^r), the subtraction is equivalent to addition as the finite field has characteristic 2. As the sets are disjoint, we have $\alpha_i \neq \beta_j$, thus all entries are well-defined and nonzero. In addition, any submatrix of a Cauchy matrix is also a Cauchy matrix as it is equivalent to constructing a smaller Cauchy matrix with subsets of the two disjoint sets. Therefore, by the first statement of Proposition 2, a Cauchy matrix is an MDS matrix.

3.5 Hadamard-Cauchy matrices

The innovative exhaustive search over Hadamard matrices from Section 5.2 is sufficient to generate low dimension diffusion matrices (order 4 and 8). However, the computation for verifying the MDS property and the exhaustive search space grows exponentially. It eventually becomes impractical to search for higher dimension Hadamard matrices (order 16 or more). Therefore, we use the Hadamard-Cauchy matrix construction, proposed in [20] as an evolution of the involutory MDS Vandermonde matrices [30], that guarantees the matrix to be an involutory MDS matrix.

In [20], the authors proposed a $2^s \times 2^s$ matrix construction that combines both the characteristics of Hadamard and Cauchy matrices. Because it is a Cauchy matrix, a Hadamard-Cauchy matrix is an MDS matrix. And because it is a Hadamard matrix, it will be involutory when c^2 is equal to 1. Therefore, we can construct a Hadamard-Cauchy matrix and check if the sum of first row is equal to 1 and, if so, we have an MDS and involutory matrix. A detailed discussion on Hadamard-Cauchy matrices is given in Section 5.3.

4 Equivalence classes of Hadamard-based matrices

Our methodology for finding lightweight MDS matrices is to perform an innovative exhaustive search and by eventually picking the matrix with the lowest XOR count. Naturally, the main problem to tackle is the huge search space. By exploiting the properties of Hadamard matrices, we found ways to group them in equivalent classes and significantly reduce the search space. In this section, we introduce the equivalence classes of Hadamard matrices and the equivalence classes of involutory Hadamard-Cauchy matrices. It is important to note that these two equivalence classes are rather different as they are defined by very different relations. We will later use these classes in Sections 5.2, 5.3, 6.2 and 6.3.

4.1 Equivalence classes of Hadamard matrices

It is known that a Hadamard matrix can be defined by its first row, and different permutation of the first row results in a different Hadamard matrix with possibly different branch number. In order to find a lightweight MDS involution matrix, it is necessary to have a set of k elements with relatively low XOR count that sum to 1 (to guarantee involution). Moreover, we need all coefficients in the first row to be different. Indeed, if the first row of an Hadamard matrix has 2 or more of the same element, say H[0, i] = H[0, j], where $i, j \in \{0, 1, ..., k-1\}$, then in another row we have $H[i \oplus j, i] = H[i \oplus j, j]$. These 4 entries are the same and by Corollary 2, H is not MDS.

By permuting the entries we hope to find an MDS involution matrix. However, given k distinct nonzero elements, there are k! ways to permute the first row of the Hadamard matrix, which can quickly become intractable. Therefore, we introduce a relation that relates certain permutations that lead to the same branch number.

Definition 8 Let H and $H^{(\sigma)}$ be two Hadamard matrices with the same set of entries up to some permutation σ . We say that they are related, $H \sim H^{(\sigma)}$, if every pair of input vectors, $(v, v^{(\sigma)})$ with the same permutation σ , to H and $H^{(\sigma)}$ respectively, have the same set of elements in the output vectors.

For example, let us consider the following three Hadamard matrices

$$H = \begin{pmatrix} w \ x \ y \ z \\ x \ w \ z \ y \\ z \ w \ x \\ z \ y \ x \ w \end{pmatrix}, \qquad H^{(\sigma_1)} = \begin{pmatrix} y \ z \ w \ x \\ z \ y \ x \ w \\ x \ y \ z \\ x \ w \ z \ y \end{pmatrix}, \qquad H^{(\sigma_2)} = \begin{pmatrix} w \ x \ z \ y \\ x \ w \ y \ z \\ y \ w \ x \\ y \ z \ x \ w \end{pmatrix},$$

One can see that $H^{(\sigma_1)}$ is defined by the third row of H, i.e. the rows are shifted by two positions and $\sigma_1 = \{2, 3, 0, 1\}$. Let us consider an arbitrary input vector for H, say v = (a, b, c, d). Then, if we apply the permutation to v, we obtain $v^{(\sigma_1)} = (c, d, a, b)$. We can observe that:

$$v \cdot H = (aw + bx + cy + dz, ax + bw + cz + dy, ay + bz + cw + dx, az + by + cx + dw)$$

 $v^{(\sigma_1)} \cdot H^{(\sigma_1)} = (cy + dz + aw + bx, cz + dy + ax + bw, cw + dx + ay + bz, cx + dw + az + by),$

It is now easy to see that $v \cdot H = v^{(\sigma_1)} \cdot H^{(\sigma_1)}$. Hence, we say that $H \sim H^{(\sigma_1)}$. Similarly, with $\sigma_2 = \{0, 1, 3, 2\}$, we have $v^{(\sigma_2)} = (a, b, d, c)$ and:

$$v \cdot H = (aw + bx + cy + dz, ax + bw + cz + dy, ay + bz + cw + dx, az + by + cx + dw),$$

 $v^{(\sigma_2)} \cdot H^{(\sigma_2)} = (aw + bx + dz + cy, ax + bw + dy + cz, az + by + dw + cx, ay + bz + dx + cw),$

and since $v \cdot H$ and $v^{(\sigma_2)} \cdot H^{(\sigma_2)}$ are the same up to the permutation σ_2 , we can say that $H \sim H^{(\sigma_2)}$.

Definition 9 An equivalence class of Hadamard matrices is a set of Hadamard matrices satisfying the equivalence relation \sim .

Proposition 3 Hadamard matrices in the same equivalence class have the same branch number.

Proof. If two Hadamard matrices H_1 and H_2 are equivalent, $H_1 \sim H_2$, then for every pair of input and output vectors for H_1 , there is a corresponding pair of vectors for H_2 with the same sum of nonzero components. Therefore, by taking the minimum over all pairs, we deduce that both matrices have the same branch number.

When searching for an MDS matrix, we can make use of this property to greatly reduce the search space: if one Hadamard matrix in an equivalence class is not MDS, then all other Hadamard matrices in the same equivalence class will not be MDS either. Therefore, it all boils down to analyzing how many and which permutation of the Hadamard matrices belongs to the same equivalence classes. Using the two previous examples σ_1 and σ_2 as building blocks, we generalize them and present two lemmas.

Lemma 1 Given a Hadamard matrix H, any Hadamard matrix $H^{(\alpha)}$ defined by the $(\alpha + 1)$ -th row of H, with $\alpha = 0, 1, 2, ..., k - 1$, is equivalent to H.

Proof. By definition of the Hadamard matrix, we can express the two matrices as $H = had(h_0, h_1, ..., h_{k-1})$ and $H^{(\alpha)} = had(h_\alpha, h_{\alpha\oplus 1}, ..., h_{\alpha\oplus (k-1)})$. Let $v = (v_0, v_1, ..., v_{k-1})$ and $v^{(\alpha)} = (v_\alpha, v_{\alpha\oplus 1}, ..., v_{\alpha\oplus (k-1)})$ be the input vector for H and $H^{(\alpha)}$, u and $u^{(\alpha)}$ be the output vector respectively.

From our example with σ_1 , we see that if the same permutation α is applied to H and to the input vector v, the output vectors are equal, i.e. $u^{(\alpha)} = u$. This is indeed true in general, it is known that the (j + 1)-th component of the output vector is the sum (or XOR as we are working over $GF(2^r)$) of the product of the input vector and (j + 1)-th column of the matrix. We can express the (j + 1)-th component of $u^{(\alpha)}$ as

$$u_j^{(\alpha)} = \bigoplus_{i=0}^{k-1} v_i^{(\alpha)} H^{(\alpha)}[i,j] = \bigoplus_{i=0}^{k-1} v_{\alpha \oplus i} h_{\alpha \oplus i \oplus j},$$

since XOR is commutative, the order of XOR is invariant, therefore $u_i^{(\alpha)} = u_i$.

Next, let us consider the other type of permutation. We can see in the example with σ_2 that up to the permutation applied to the Hadamard matrix, input and output vectors are the same. Let $H^{(\sigma)}$, $v^{(\sigma)}$ and $u^{(\sigma)}$ denote the permuted Hadamard matrix, the permuted input vector and its corresponding permuted output vector. We want the permutation

to satisfy $u_{\sigma(j)} = u_j^{(\sigma)}$, where $j \in \{0, 1, ..., k-1\}$. That is the permutation of the output vector of H is the same as the permuted output vector of $H^{(\sigma)}$. Using the definition of the Hadamard matrix, we can rewrite it as

$$\bigoplus_{i=0}^{k-1} v_i h_{i \oplus \sigma(j)} = \bigoplus_{i=0}^{k-1} v_i^{(\sigma)} H^{(\sigma)}[i,j]$$

Using the definition of the permutation and by the fact that it is one-to-one mapping, we can rearrange the XOR order of the terms on the left-hand side and we obtain

$$\bigoplus_{i=0}^{k-1} v_{\sigma(i)} h_{\sigma(i) \oplus \sigma(j)} = \bigoplus_{i=0}^{k-1} v_{\sigma(i)} h_{\sigma(i \oplus j)}.$$

Therefore, we need the permutation to be linear with respect to XOR: $\sigma(i \oplus j) = \sigma(i) \oplus \sigma(j)$. This proves our next lemma.

Lemma 2 For any linear permutation σ (w.r.t. XOR), the two Hadamard matrices H and $H^{(\sigma)}$ are equivalent.

We note that the permutations in Lemma 1 and 2 are disjoint, except for the identity permutation. This is because for the linear permutation σ , it always maps the identity to itself: $\sigma(0) = 0$. Thus, for any linear permutation, the first entry remains unchanged. On the other hand, when choosing another row of H as the first row, the first entry is always different.

With these two lemmas, we can now partition the family of Hadamard matrices into equivalence classes. For Lemma 1, we can easily see that the number of permutation is equal to the order of the Hadamard matrix. However, for Lemma 2 it is not so trivial. Therefore, we have the following lemma.

Lemma 3 Given a set of 2^s nonzero elements, $S = \{\alpha_0, \alpha_1, ..., \alpha_{2^s-1}\}$, there are $\prod_{i=0}^{s-1}(2^s - 2^i)$ linear permutations w.r.t. XOR operation.

Proof. For simplicity, we see how the indices of the elements are permuted. As mentioned, we need to map identity to itself, $\sigma(0) = 0$. After index 0 is fixed, index 1 can be mapped to any of the remaining $2^s - 1$ indices. Similarly for index 2, there are $2^s - 2$ choices. But for index 3, because of the linear relation, its image is defined by the mapping of index 1 and 2: $\sigma(3) = \sigma(1) \oplus \sigma(2)$.

Following this pattern, we can choose the permutation for index 4 among the $2^s - 4$ index, while 5, 6 and 7 are defined by $\sigma(1)$, $\sigma(2)$ and $\sigma(4)$. Therefore, the total number of possible permutations is

$$(2^{s}-1)(2^{s}-2)(2^{s}-4)...(2^{s}-2^{s-1}) = \prod_{i=0}^{s-1} (2^{s}-2^{i}).$$

Theorem 3 Given a set of 2^s nonzero elements, $S = \{\alpha_0, \alpha_1, ..., \alpha_{2^s-1}\}$, there are $\frac{(2^s-1)!}{\prod_{i=0}^{s-1}(2^s-2^i)}$ equivalence classes of Hadamard matrices of order 2^s defined by the set of elements S.

Proof. To prove this theorem, we use the *double counting* proof technique that is commonly used in combinatorics. We count the total number of permutations of Hadamard matrices for a given set of elements.

Counting 1: there is a total of $(2^s)!$ permutations for the given set of elements.

Counting 2: for each of the equivalence classes of Hadamard matrix, by Lemma 2 and 3, there are $\prod_{i=0}^{s-1} (2^s - 2^i)$ linear permutations. For each of these permutations, by Lemma 1, there are 2^s permutations by defining a new Hadamard from one of the row. Therefore the total number of permutations is

{# of equivalence classes}
$$\left(\prod_{i=0}^{s-1} (2^s - 2^i)\right) (2^s)$$
.

Equating these two expressions together, we get

$$\{\# \text{ of equivalence classes}\} = \frac{(2^s - 1)!}{\prod_{i=0}^{s-1} (2^s - 2^i)}.$$

For convenience, we call the permutations in Lemma 1 and 2 the \mathcal{H} -permutations. The \mathcal{H} -permutations can be described as a sequence of the following types of permutations on the index of the entries:

- 1. choose $\alpha \in \{0, 1, ..., 2^s 1\}$, define $\sigma(i) = i \oplus \alpha, \forall i = 0, 1, ..., 2^s 1$, and
- 2. fix $\sigma(0) = 0$, in ascending order of the index *i*, choose the permutation if *i* is power of 2, otherwise it is defined by the linear permutation (w.r.t. XOR): $\sigma(i \oplus j) = \sigma(i) \oplus \sigma(j)$.

We remark that given a set of 4 nonzero elements, from Theorem 3 we see that there is only 1 equivalence class of Hadamard matrices. This implies that there is no need to permute the entries of the 4×4 Hadamard matrix in hope to find MDS matrix if one of the permutation is not MDS.

With the knowledge of equivalence classes of Hadamard matrices, what we need is an algorithm to pick one representative from each equivalence class and check if it is MDS. The idea is to exhaust all non- \mathcal{H} -permutations through selecting the entries in ascending index order. Since the entries in the first column of Hadamard matrix are distinct (otherwise the matrix is not MDS), it is sufficient for us to check the matrices with the first entry (index 0) being the smallest element. This is because for any other matrices with the first entry set as some other element, it is in the same equivalence class as some matrix $H^{(\alpha)}$ where the first entry of $(\alpha + 1)$ -th row is the smallest element. For indexes that are powers of 2, select the smallest element from the remaining set. While for the other entries, one can pick any element from the remaining set.

For 8×8 Hadamard matrices for example, the first three entries, α_0 , α_1 and α_2 are fixed to be the three smallest elements in ascending order. Next, by Lemma 2, α_3 should be defined by α_1 and α_2 in order to preserve the linear property, thus to "destroy" the linear property and obtain matrices from different equivalence classes, pick an element from the remaining set in ascending order as the fourth entry α_3 . After which, α_4 is selected to be the smallest element among the remaining 4 elements and permute the remaining 3 elements to be α_5 , α_6 and α_7 respectively. For each of these arrangement of entries, we check if it is MDS using the algorithm discussed in Section 5.2. We terminate the algorithm prematurely once an MDS matrix is found, else we conclude that the given set of elements does not generate an MDS matrix.

It is clear that arranging the entries in this manner will not obtain two Hadamard matrices from the same equivalence class. But one may wonder if it actually does exhaust all the equivalence classes. The answer is yes: Theorem 3 shows that there is a total of 30 equivalence classes for 8×8 Hadamard matrices. On the other hand, from the algorithm described above, we have 5 choices for α_3 and we permute the remaining 3 elements for α_5 , α_6 and α_7 . Thus, there are 30 Hadamard matrices that we have to check.

4.2 Equivalence classes of involutory Hadamard-Cauchy matrices

Despite having a new technique to reduce the search space, the computation cost for checking the MDS property is still too huge when the order of the Hadamard matrix is larger than 8. Therefore, we use the Hadamard-Cauchy construction for order 16 and 32. Thanks to the Cauchy property, we are ensured that the matrix will be MDS. Hence, the only problem that remains is the huge search space of possible Hadamard-Cauchy matrices. To prevent confusion with Hadamard matrices, we denote Hadamard-Cauchy matrices with K.

First, we restate in Algorithm 1 the technique from [20] to build involutory MDS matrices, with some modifications on the notations for the variables. Although it is not explicitly stated, we can infer from Lemma 6,7 and Theorem 4 from [20] that all Hadamard-Cauchy matrices can be expressed as an output of Algorithm 1.

Algorithm 1 Construction of $2^s \times 2^s$ MDS matrix or involutory MDS matrix over $GF(2^r)/p(X)$.

INPUT: an irreducible polynomial p(X) of $GF(2^r)$, integers s, r satisfying s < r and r > 1, a boolean $B_{involutory}$. **OUTPUT:** $2^s \times 2^s$ Hadamard-Cauchy matrix K, where K is involutory if $B_{involutory}$ is set **True**.

procedure CONSTRUCTH-C $(r, p(X), s, B_{involutory})$

select s linearly independent elements $x_1, x_2, x_{2^2}, ..., x_{2^{s-1}}$ from $GF(2^r)$ and construct S, the set of 2^s elements x_i , where $x_i = \bigoplus_{t=0}^{s-1} b_t x_{2^t}$ for all $i \in [0, 2^s - 1]$ (with $(b_{s-1}, b_{s-2}, ..., b_1, b_0)$ being the binary representation of i) select $z \in GF(2^r) \setminus S$ and construct the set of 2^s elements y_i , where $y_i = z + x_i$ for all $i \in [0, 2^s - 1]$ initialize an empty array ary_{-s} of size 2^s if $(B_{involutory} = False)$ then $ary_{-s}[i] = \frac{1}{y_i}$ for all $i \in [0, 2^s - 1]$ else $ary_{-s}[i] = \frac{1}{c \cdot y_i}$ for all $i \in [0, 2^s - 1]$, where $c = \bigoplus_{t=0}^{s-1} \frac{1}{z+x_t}$ end if construct the $2^s \times 2^s$ matrix K, where $K[i, j] = ary_{-s}[i \oplus j]$ return K end procedure

Similarly to Hadamard matrices, we denote a Hadamard-Cauchy matrix by its first row of elements as $hc(h_0, h_1, ..., h_{2^s-1})$, with $h_i = K[0, i]$. To summarize the construction of a Hadamard-Cauchy matrix of order 2^s mentioned in Algorithm 1, we pick an ordered set of s + 1 linearly independent elements, we call it the basis. We use the first s elements to span an ordered set S of 2^s elements, and add the last element z to all the elements in S. Next, we take the inverse of each of the elements in this new set and we get the first row of the Hadamard-Cauchy matrix. Lastly, we generate the matrix based on the first row in the same manner as an Hadamard matrix. For example, for an 8×8 Hadamard-Cauchy matrix over $GF(2^4)/0x13$, say we choose $x_1 = 1, x_2 = 2, x_4 = 4$, we generate the set $S = \{0, 1, 2, 3, 4, 5, 6, 7\}$, choosing z = 8 and taking the inverses in the new set, we get a Hadamard-Cauchy matrix K = hc(15, 2, 12, 5, 10, 4, 3, 8). To make it involution, we multiply each element by the inverse of the sum of the elements. However for this instance the sum is 1, hence K is already an involutory MDS matrix.

One of the main differences between the Hadamard and Hadamard-Cauchy matrices is the choice of entries. While we can choose all the entries for a Hadamard matrix to be lightweight and permute them in search for an MDS candidate, the construction of Hadamard-Cauchy matrix makes it nontrivial to control its entries efficiently. Although in [20] the authors proposed a backward re-construction algorithm that finds a Hadamard-Cauchy matrix with some pre-decided lightweight entries, the number of entries that can be decided beforehand is very limited. For example, for a Hadamard-Cauchy matrix of order 16, the algorithm can only choose 5 lightweight entries, the weight of the other 11 entries is not controlled. The most direct way to find a lightweight Hadamard-Cauchy matrix is to apply Algorithm 1 repeatedly for all possible basis. We introduce now new equivalence classes that will help us to exhaust all possible Hadamard-Cauchy matrices with much lesser memory space and number of iterations.

Definition 10 Let K_1 and K_2 be two Hadamard-Cauchy matrices, we say they are related, $K_1 \sim_{HC} K_2$, if one can

be transformed to the other by either one or both operations on the first row of entries:

 $1. \ multiply \ by \ a \ nonzero \ scalar, \ and$

2. \mathcal{H} -permutation of the entries.

The crucial property of the construction is the independence of the elements in the basis, which is not affected by multiplying a nonzero scalar. Hence, we can convert any Hadamard-Cauchy matrix to an involutory Hadamard-Cauchy matrix by multiplying it with the inverse of the sum of the first row and vice versa. However, permutating the positions of the entries is the tricky part. Indeed, for the Hadamard-Cauchy matrices of order 8 or higher, some permutations destroy the Cauchy property, causing it to be non-MDS. Using our previous 8×8 example, suppose we swap the first two entries, K' = hc(2, 15, 12, 5, 10, 4, 3, 8), it can be verified that it is not MDS. To understand why, we work backwards to find the basis corresponding to K'. Taking the inverse of the entries, we have $\{9, 8, 10, 11, 12, 13, 14, 15\}$. However, there is no basis that satisfies the 8 linear equations for the entries. Thus it is an invalid construction of Hadamard-Cauchy matrix. Therefore, we consider applying the \mathcal{H} -permutation on Hadamard-Cauchy matrix. Since it is also a Hadamard matrix, the \mathcal{H} -permutation preserves its branch number, thus it is still MDS. So we are left to show that a Hadamard-Cauchy matrix that undergoes \mathcal{H} -permutation is still a Hadamard-Cauchy matrix.

Lemma 4 Given a $2^s \times 2^s$ involutory Hadamard-Cauchy matrix K, there are $2^s \cdot \prod_{i=0}^{s-1} (2^s - 2^i)$ involutory Hadamard-Cauchy matrices that are related to K by the \mathcal{H} -permutations of the entries of the first row.

Proof. We first show that a \mathcal{H} -permutation of the first row of a Hadamard-Cauchy matrix is equivalent to choosing a different set of basis. Let $K = hc(\frac{1}{z}, \frac{1}{z \oplus x_1}, \frac{1}{z \oplus x_2}, \frac{1}{z \oplus x_3}, ..., \frac{1}{z \oplus x_{2^{s-1}}})$ be an involutory Hadamard-Cauchy matrix. Under the type 1 of \mathcal{H} -permutation, for some $\alpha \in \{1, ..., 2^s - 1\}$, we have $K' = hc(\frac{1}{z \oplus x_\alpha}, \frac{1}{z \oplus x_1 \oplus x_\alpha}, \frac{1}{z \oplus x_2 \oplus x_\alpha}, \frac{1}{z \oplus x_3 \oplus x_\alpha}, ..., \frac{1}{z \oplus x_{2^{s-1}}})$. From this, we can see that $z' = z \oplus x_\alpha$ while the first s elements $\{x_{2^j}\}, \forall j = 0, 1, ..., s - 1$, remain unchanged. Since z' is not a linear combination of the s element z remain unchanged. Therefore, it is a linear permutation (w.r.t. XOR) on the set S and the new s elements $\{x'_{2^j}\}, \forall j = 0, 1, ..., s - 1$ are still linearly independent. Again, we have our s + 1 linearly independent elements. Finally, as mentioned before in Lemma 1 and 3, there are $2^s \cdot \prod_{i=0}^{s-1} (2^s - 2^i)$ ways of \mathcal{H} -permutations.

With that, we can define our equivalence classes of involutory Hadamard-Cauchy matrices.

Definition 11 An equivalence class of involutory Hadamard-Cauchy matrices is a set of Hadamard-Cauchy matrices satisfying the equivalence relation \sim_{HC} .

In order to count the number of equivalence classes of involutory Hadamard-Cauchy matrices, we use the same technique for proving Theorem 3. To do that, we need to know the total number of Hadamard-Cauchy matrices that can be constructed from the Algorithm 1 for a given finite field.

Lemma 5 Given two natural numbers s and r, based on Algorithm 1, there are $\prod_{i=0}^{s} (2^r - 2^i)$ many $2^s \times 2^s$ Hadamard-Cauchy matrices over $GF(2^r)$.

Proof. As we can see from Algorithm 1, we need to choose s + 1 many linearly independent ordered elements from $GF(2^r)$ to construct a Hadamard-Cauchy matrix. For the (t + 1)-th element, where $0 \le t \le s$, it cannot be a linear combination of the t previously chosen elements, hence there are $2^r - 2^t$ many choices. Therefore, there are $(2^r - 1)(2^r - 2)(2^r - 4)...(2^r - 2^s)$ ways to choose s + 1 linearly independent ordered elements.

Theorem 4 Given two positive integers s and r, there are $\prod_{i=0}^{s-1} \frac{2^{r-1}-2^i}{2^s-2^i}$ equivalence classes of involutory Hadamard-Cauchy matrices of order 2^s over $GF(2^r)$.

Proof. Again, we use the double counting to prove this theorem. We count the total number of distinct Hadamard-Cauchy matrices that can be generated from Algorithm 1.

Counting 1: by Lemma 5, the total number of distinct Hadamard-Cauchy matrices generated from the algorithm is $\prod_{i=0}^{s} (2^r - 2^i)$.

Counting 2: for each of the equivalence classes of involutory Hadamard-Cauchy matrices, by Lemma 4, there are $2^s \cdot \prod_{i=0}^{s-1} (2^s - 2^i)$ involutory Hadamard-Cauchy matrices that are related. Moreover, for each of the involutory Hadamard-Cauchy matrices, we can multiply by a nonzero scalar to obtain another related Hadamard-Cauchy matrix, thus there are another factor $2^r - 1$ of distinct Hadamard-Cauchy matrices. Therefore, the total number of distinct Hadamard-Cauchy matrices is

$$\{\# \text{ of equivalence classes}\}\left(2^s \cdot \prod_{i=0}^{s-1} (2^s - 2^i)\right) (2^r - 1).$$

Equating these two expressions together, we get

$$\{\# \text{ of equivalence classes}\} = \prod_{i=0}^{s-1} \frac{2^{r-1} - 2^i}{2^s - 2^i}.$$

In [15], the authors introduced the notion of compact Cauchy matrices which are defined as Cauchy matrices with exactly 2^s distinct elements. These matrices seem to include Cauchy matrices beyond the class of Hadamard-Cauchy matrices. However, it turns out that the equivalence classes of involutory Hadamard-Cauchy matrices can be extended to compact Cauchy matrices.

Corollary 3 Any compact Cauchy matrices can be generated from some equivalence class of involutory Hadamard-Cauchy matrices.

Proof. We count the number of distinct compact Cauchy matrices that can be generated from one equivalence class. Taking the first row of an equivalence class of involutory Hadamard-Cauchy matrices, we can multiply it by a nonzero scalar. The inverse of these entries corresponds to a set of 2^s nonzero elements. Each of these elements can be defined to be z and we have a set S and $z \in GF(2^r) \setminus S$. Note that S is closed under XOR operation and in the context of [15], we can regard S as a subgroup of $GF(2^r)$ defined under XOR operation. Finally, by fixing the first element of S and z + S to be 0 and z repectively, we have $(2^s - 1)!$ permutation for each set S and z + S. Each arrangement generates a distinct compact Cauchy matrix. Therefore, considering all equivalence classes, we can obtain

$$\left(\prod_{i=0}^{s-1} \frac{2^{r-1} - 2^i}{2^s - 2^i}\right) (2^r - 1)(2^s) \left((2^s - 1)!\right)^2$$

distinct compact Cauchy matrices, which coherent to Theorem 3 of [15].

Note that since the permutation of the elements in S and z + S only results in rearrangement of the entries of the compact Cauchy matrix, the XOR count is invariant from Hadamard-Cauchy matrix with the same set of entries.

5 Searching for MDS and involutory matrices

In this section, using the previous properties and equivalence classes given in Sections 3 and 4 for several matrix constructions, we will derive algorithms to search for lightweight involutory MDS matrices. First, we point out that the circulant construction can not lead to such matrices, then we focus on the case of matrices of small dimension using the Hadamard construction. For bigger dimension, we add the Cauchy property to the Hadamard one in order to guarantee that the matrix will be MDS. We recall that, similarity to a circulant matrix, an Hadamard matrix will have the interesting property that each row is a permutation of the first row, therefore allowing to reuse the multiplication circuit to save implementation costs.

5.1 Circulant MDS involution matrix does not exist

The reason why we do not consider circulant matrices as potential candidates for MDS involution matrices is simple: it simply does not exist. In [23], the authors proved that circulant matrices of order 4 cannot be simultaneously MDS and involutory. And recently [21] proved that generic circulant MDS involutory matrices do not exist.

5.2 Small dimension lightweight MDS involution matrices

The computation complexity for checking if a matrix is MDS and the huge search space are two main complexity contributions to our exhaustive search algorithm for lightweight Hadamard MDS matrices. The latter is greatly reduced thanks to our equivalence classes and we now need an efficient algorithm for checking the MDS property. In this section, using properties of Hadamard matrix, we design a simple algorithm that can verify the MDS property faster than for usual matrices. First, let us prove some results using Proposition 2. Note that Lemma 6 and Corollary 4 are not restricted to Hadamard matrices. Also, Corollary 4 is the contra-positive of Lemma 6.

Lemma 6 Given a $k \times k$ matrix M, there exists a $l \times l$ singular submatrix if and only if there exists a vector, $v \neq 0$, with at most l nonzero components such that vM = u and the sum of nonzero components in v and u is at most k.

Proof. Suppose there exists a $l \times l$ singular submatrix, by the first statement of Proposition 2, M is not MDS and thus from the second statement, there exists k columns of $[I_k|M]$ that are linearly dependent, in particular, k - l columns from I_k and l columns from M. Let L be the square matrix comprising these k linearly dependent columns. From linear algebra, there exists nonzero vector, v, such that the output is a zero vector, vL = 0. For the columns from I_k , there is exactly one 1 and 0 for the other entries, this implies that the components of v corresponding to these columns are zero, else the output will be nonzero. Therefore, there are at most l nonzero components in v. Now, let us consider vM = u, for the l columns of M that are also in L, the corresponding output components are zero, as vL = 0. Thus, there are at most k - l nonzero components in u. Hence, the sum of nonzero components in v and u is at most k. The converse is similar, we consider $v[I_k|M] = [v|u]$, since there are at most k nonzero components in [v|u], we pick k - l columns of I_k and k columns of M corresponding to the zero components in [v|u] to form a singular square matrix L. The determinant of L is equal to some $l \times l$ submatrix of M, which is also singular.

Corollary 4 Given a $k \times k$ matrix M, the sum of nonzero components of the input and output vector is at least k + 1 for any input vector v with l nonzero components if and only if all $l \times l$ submatrices of M are nonsingular.

One direct way for checking the MDS property is to compute the determinant of all the submatrices of M and terminates the algorithm prematurely by returning **False** when a singular submatrix is found. If no such submatrix has been found among all the possible submatrices, the algorithm can return **True**. Using the fact that the product of a Hadamard matrix with itself is a multiple of an identity matrix, we can cut down the number of submatrices to be checked with the following proposition.

Proposition 4 Given a $k \times k$ Hadamard matrix H with the sum c of first row being nonzero ($c \neq 0$), if all submatrices of order $l \leq \frac{k}{2}$ are nonsingular, then H is MDS.

Proof. Suppose not, there exists submatrix of order $l \ge \frac{k}{2} + 1$ that is singular. By Lemma 6, there exists a vector, $v \ne 0$, with at most l nonzero components such that vH = u and u has at most k - l nonzero components. Right multiply H to the equation and we get $c^2v = uH$, where $c \ne 0$, hence the number of nonzero component of c^2v is the same as v. However, since u has $k - l \le \frac{k}{2}$ nonzero components, by Corollary 4, the sum of nonzero components is at least k + 1. This contradicts that v has at most l nonzero components.

We can further reduce the computation complexity using the fact that Hadamard matrices are bisymmetric. Given a Hadamard matrix, we have four regions dissected by the left and right diagonal, namely top, left, right and bottom quadrant. For convention, we let the diagonal entries to be in both quadrants. See Figure 1 for illustration, where the top four entries "a" belong to both top and left quadrants, while the bottom four "a" belong to both bottom and right quadrant.

Proposition 5 Given a $k \times k$ Hadamard matrix H, if all submatrices L with leading entry L[0,0] in the top quadrant are nonsingular, then H is MDS.

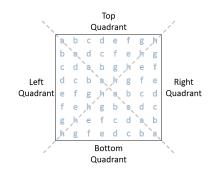


Fig. 1: The four quadrants of Hadamard matrix.

Proof. It is known that the determinant of a matrix remains unchanged when it undergoes left or right diagonal reflection. Thus, it is sufficient to show that for any submatrix, it corresponds to some submatrix with the leading entry in top quadrant. This can be shown by looking at the reflection through the left and/or right diagonal. Consider the submatrices case by case:

• case A: the leading entry is in left quadrant. Through the left diagonal reflection, we can see that it is same as a submatrix with leading entry in top quadrant. Refer to Figure 2a, the red submatrix is reflected at the blue matrix with leading entry in top quadrant.

- case B1: the leading entry is not in left quadrant and ending entry is in right quadrant. Through the right diagonal reflection, the ending entry L[l-1, l-1] of red submatrix is reflected to the leading entry in the top quadrant of the blue submatrix, see Figure 2b. Since the determinant does not change, the red submatrix will be nonsingular if the blue matrix is nonsingular.
- case B2: the leading entry is not in left quadrant and ending entry is in bottom quadrant. From Figure 2c, we see that through left diagonal reflection, the ending entry is now in the right quadrant, which is the case B1.

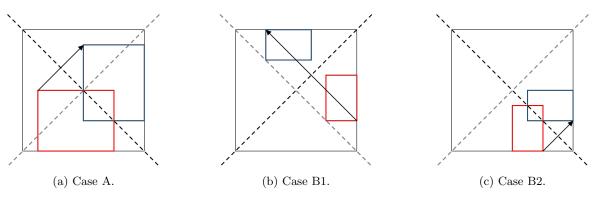


Fig. 2: Submatrices Reflections

Thanks to Propositions 4 and 5, our algorithm for checking the MDS property of Hadamard matrices is now much faster than a naive method. Namely, given a $2^s \times 2^s$ Hadamard matrix, the algorithm to test its MDS property can be as follows. First, we check that all entries are nonzero and for $l = 2, ..., 2^{s-1}$ we check that the determinant of $l \times l$ submatrices with leading entry in top quadrant is nonzero. If one submatrix fails, we output **False**. Once all the submatrices are checked, we can output **True**.

Using this algorithm as the core procedure for checking the MDS property, we can find the lightest-weight MDS involution Hadamard matrix by choosing a set of elements that sum to 1 with the smallest XOR count, permute the entries as mentioned in Section 4.1 and use this core procedure to check if it is MDS. If all equivalence classes of Hadamard matrices are not MDS, we swap some element in the set with another element with a slightly higher XOR count and repeat the process until we find the first MDS involution Hadamard matrix with the lowest possible XOR count. Eventually, we found the lightest-weight MDS involution Hadamard matrix of order 4 and 8 over $GF(2^4)$ and $GF(2^8)$, which can be found in Table 1 in Section 7. We emphasize that our results close the discussions on MDS involution Hadamard matrix of order 4 and 8, since our technique allows to take into account of all possible matrices.

5.3 Large dimension lightweight MDS involution matrices

The algorithm computation complexity grows exponentially with the matrix dimension, it is difficult to go to matrices of higher order. For that reason, we reduce the search space from Hadamard to Hadamard-Cauchy matrices, which guarantee the MDS property. Nevertheless, it is not feasible to generate and store all possible Hadamard-Cauchy matrices. For 16 × 16 Hadamard-Cauchy matrices over $GF(2^8)$, by Lemma 5 we know there are almost a trillion distinct candidates. This is where the idea of equivalence classes comes in handy again. By Theorem 4, instead of storing over 9.7×10^{11} matrices, all we need is to find the 11811 equivalence classes. Even if memory space is not an issue, using Algorithm 1 to exhaustively search for all Hadamard-Cauchy matrices requires about $2^{39.9}$ iterations. In this subsection, we propose a deterministic and randomized algorithm that only takes on average of $2^{16.9}$ iterations to find all the equivalence classes, which is equivalent to finding all possible Hadamard-Cauchy matrices.

First, we present two statements that are useful in designing the algorithm.

Lemma 7 Based on Algorithm 1, given a basis of s + 1 ordered elements $\{x_1, x_2, x_{2^2}, ..., x_{2^{s-1}}, z\}$, any permutation of the first s elements $\{\sigma(x_1), \sigma(x_2), \sigma(x_{2^2}), ..., \sigma(x_{2^{s-1}}), z\}$ will form a Hadamard-Cauchy matrix that belongs to the same equivalence class.

Proof. Since we are taking the span of the ordered set $\{x_1, x_2, x_{2^2}, ..., x_{2^{s-1}}\}$ and adding z to the span, it is obvious that permuting $\{x_{2^i}\}$ will only permute the order of the entries of K.

Proposition 6 Given two positive integers s and r, where s < r, doing exhaustive search through $1 \le x_1 < x_2 < x_{2^2} < ... < x_{2^{s-1}} \le 2^r$ and $1 \le z \le 2^r$ is sufficient to find all possible equivalence classes of involutory Hadamard-Cauchy matrices.

Algorithm 2 Finding all $2^s \times 2^s$ equivalence classes of involutory Hadamard-Cauchy matrices over $GF(2^r)/p(X)$.

INPUT: an irreducible polynomial p(X) of $GF(2^r)$, integers s, r satisfying s < r and r > 1. **OUTPUT:** a list of equivalence classes of involutory Hadamard-Cauchy matrix.

procedure GENECOFINVH-C $(r, p(X), s)$
compute the total number of equivalence classes, $EC = \prod_{i=0}^{s-1} \frac{2^{r-1}-2^i}{2^s-2^i}$
initialize an empty set of arrays <i>list_EC</i>
while $(size of(list_EC) \neq EC)$ do
select s linearly independent elements $x_1, x_2, x_{2^2},, x_{2^{s-1}}$ from $\operatorname{GF}(2^r)/p(X)$ in ascending order
select element z as linearly independent of $x_1, x_2, x_{2^2},, x_{2^{s-1}}$
$temp_mat = \text{CONSTRUCTH-C}^*(r, p(X), s, \text{True}, x_1, x_2, x_{2^2},, x_{2^{s-1}}, z)$
if $temp_mat$ is not a permutation of any matrix in $list_EC$ then
store $temp_mat$ into $list_EC$
end if
end while
$return \ list_EC$
end procedure

Proof. Any linearly independent ordered set of elements $\{x_1, x_2, x_{2^2}, ..., x_{2^{s-1}}\}$ that are not in ascending order is simply some permutation of a set in ascending order. By Lemma 7, it forms a Hadamard-Cauchy matrix of the same equivalence class.

We describe our search method in Algorithm 2 and one can see that it uses most of Algorithm 1 as core procedure. We denote CONSTRUCTH-C^{*} the procedure ConstructH-C from Algorithm 2 where the values $x_1, x_2, x_{2^2}, ..., x_{2^{s-1}}$ and z are given as inputs instead of chosen in the procedure. We first choose s + 1 linearly independent elements and apply Algorithm 1 to generate an involutory Hadamard-Cauchy matrix. We initialise an array temp_mat and a list $list_EC$ to empty. Then, temp_mat is the matrix considered at the current iteration, it will be checked against $list_EC$ which is the list of equivalence classes of involutory Hadamard-Cauchy matrices that have been found. If temp_mat is not a permutation of any matrix in $list_EC$, then a new equivalence class is found and temp_mat will be stored in $list_EC$. When all the equivalence classes are found, we terminates the algorithm, which will dramatically cut down the number of iterations required.

From a representative of an equivalence class, one can obtain all the involutory Hadamard-Cauchy matrices of the same equivalence class through \mathcal{H} -permutations. Note that the \mathcal{H} -permutation is also applicable to non-involutory Hadamard-Cauchy matrices.

We remark that for 2×2 and 4×4 Hadamard-Cauchy matrices, any permutation of the equivalence class is still an involutory Hadamard-Cauchy matrix.

Notice that Algorithm 2 is a deterministic search for the equivalence classes. To further reduce the iterations needed, we propose to choose the s + 1 elements randomly. Using this randomized search, it takes about $2^{16.9}$ iterations before finding all the equivalence classes. Once all the equivalence classes of involutory Hadamard-Cauchy matrices are found, we can check which matrix has the lightest-weight.

Using the randomized search algorithm, we found the lightest-weight involution Hadamard-Cauchy matrix of order 16 and 32 over $GF(2^8)$, which can be found in Table 1.

6 Searching for MDS matrices

The disadvantage of using non-involution matrices is that its inverse may have a high computation cost. But if the inverse is not required, non-involution matrices might be lighter than involutory matrices. In this paper, we look at encryption only and do not consider the reuse of component for encryption/decryption (which can be studied in future work). Note that the inverse of the matrix would not be required for popular constructions such as a Feistel network, or a CTR encryption mode.

6.1 Circulant matrices

As the discussion on lightweight MDS circulant matrix is well-explored in [26], we focus on Hadamard-based matrix and extend the exhaustive search for from involutory to non-involutory lightest-weight MDS matrix.

6.2 Small dimension lightweight MDS matrices

The results in Section 4.1 and 5.2 can also be applied on non-involution Hadamard matrices. Thus the method of finding a lightweight MDS involution matrix is basically the same. We pick a set of low XOR count nonzero elements

that does not sum to 0, else it would be non-MDS, and apply the permutation method which is discussed at the end of Section 5.2 to check through all equivalence classes of Hadamard matrices.

6.3 Large dimension lightweight MDS matrices

After finding all the equivalence classes of involutory Hadamard-Cauchy matrices using the Algorithm 2, we can conveniently use this collection of equivalence classes to find lightest-weight non-involutory Hadamard-Cauchy matrix. That is to multiply by a nonzero scalar to each equivalence classes to generate all possible Hadamard-Cauchy matrices up to permutation. In this way, it is more efficient than exhaustive search on all possible Hadamard-Cauchy matrices as we eliminated all the permutations of the Hadamard-Cauchy matrices that have the same XOR count.

7 Results

We first emphasize that although in [20, 15] the authors proposed methods to construct lightweight matrices, the choice of the entries are limited as mentioned in Section 4.2. This is due to the nature of the Cauchy matrices where the inverse of the elements are used during the construction, which makes it non-trivial to search for lightweight Cauchy matrices⁷. However, using the concept of equivalence classes, we can exhaust all the matrices and pick the lightest-weight matrix.

We applied the algorithms of Section 5 to construct lightweight MDS involutions over $GF(2^8)$. We list them in Table 1 and we can see that they are much lighter than known MDS involutions like the KHAZAD and ANUBIS, previous Hadamard-Cauchy matrices [6, 20] and compact Cauchy matrices [15]. In Table 2, we list the $GF(2^8)$ MDS matrices we found using the methods of Section 6 and show that they are lighter than known MDS matrices like the AES, WHIRLPOOL and WHIRLWIND matrices [17, 8, 7]. We also compare with the 14 lightweight candidate matrices C_0 to C_{13} for the WHIRLPOOL hash functions suggested during the NESSIE workshop [32, Section 6]. Both Tables 1 and 2 are comparing matrices that are explicitly provided in the papers. Recently, Gupta *et al.* [21] constructed some circulant matrices that is lightweight for both itself and its inverse. However we do not compare them in our table because their approach minimizes the number of XORs, look-up tables and temporary variables, which might be optimal for software but not for hardware implementations based purely on XOR count.

By Theorem 2 in Section 2, we only need to apply the algorithms from Section 5 and Section 6 for half the representations of $GF(2^8)$ when searching for optimal lightweight matrices. And as predicted by the discussion after Theorem 1, the lightweight matrices we found in Tables 1, 2 and 3 do come from $GF(2^8)$ and $GF(2^4)$ representations with higher standard deviations.

We provide in the first column of the result Tables 1, 2 and 3 the type of the matrices. They can be circulant, Hadamard or Cauchy-Hadamard. The subfield-Hadamard construction is based on the method of [26, Section 7.2] which we explain here. Consider the MDS involution M = had(0x1, 0x4, 0x9, 0xd) over $GF(2^4)/0x13$ in the first row of Table 1. Using the method of [26, Section 7.2], we can extend it to a MDS involution over $GF(2^8)$ by using two parallel copies of Q. The matrix is formed by writing each input byte x_j as a concatenation of two nibbles $x_j = (x_j^L || x_j^R)$. Then the MDS multiplication is computed on each half $(y_1^L, y_2^L, y_3^L, y_4^L) = M \cdot (x_1^L, x_2^L, x_3^L, x_4^L)$ and $(y_1^R, y_2^R, y_3^R, y_4^R) = M \cdot (x_1^R, x_2^R, x_3^R, x_4^R)$ over $GF(2^4)$. The result is concatenated to form four output bytes (y_1, y_2, y_3, y_4) where $y_j = (y_j^L || y_j^R)$.

We could have concatenated different submatrices and this is done in the WHIRLWIND hash function [7], where the authors concatenated four MDS submatrices over GF(2⁴) to form $(M_0|M_1|M_1|M_0)$, an MDS matrix over GF(2¹⁶). The submatrices are non-involutory Hadamard matrices $M_0 = had(0x5, 0x4, 0x3, 0x6, 0x2, 0xd, 0x8, 0x3)$ and $M_1 = (0x5, 0xe, 0x4, 0x7, 0x1, 0x3, 0xf, 0x8)$ defined over GF(2⁴)/0x13. For fair comparison with our GF(2⁸) matrices in Table 2, we consider the corresponding WHIRLWIND-like matrix $(M_0|M_1)$ over GF(2⁸) which takes half the resource of the original WHIRLWIND matrix and is also MDS.

The second column of the result tables gives the finite field over which the matrix is defined, while the third column displays the first row of the matrix where the entries are bytes written in hexadecimal notation. The fourth column gives the XOR count to implement the first row of the $n \times n$ matrix. Because all subsequent rows are just permutations of the first row, the XOR count to implement the matrix is just n times this number. For example, to compute the XOR count for implementing had(0x1, 0x4, 0x9, 0xd) over $GF(2^4)/0x13$, we consider the expression for the first row of matrix multiplication $0x1 \cdot x_1 \oplus 0x4 \cdot x_2 \oplus 0x9 \cdot x_3 \oplus 0xd \cdot x_4$. From Table 8 of Appendix D, the XOR count of multiplication by 0x1, 0x4, 0x9 and 0xd are 0, 2, 1 and 3, which gives us a cost of $(0 + 2 + 1 + 3) + 3 \times 4 = 18$ XORs to implement one row of the matrix (the summand 3×4 account for the three XORs summing the four nibbles). For the subfield construction over $GF(2^8)$, we need two copies of the matrix giving a cost of $18 \times 2 = 36$ XORs to implement one row.

⁷ Using direct construction, there is no clear implication for the choice of the elements α_i and β_j that will generate lightweight entries c_{ij} . On the other hand, every lightweight entry chosen beforehand will greatly restrict the choices for the remaining entries if one wants to maintain two disjoint sets of elements $\{\alpha_i\}$ and $\{\beta_j\}$.

We also applied the algorithms from Section 5 and Section 6 to find lightweight MDS involution and non-involution matrices of order 4 and 8 over $GF(2^4)$, these matrices are listed in Table 3. By Corollary 2 and the structure of Hadamard matrix, the first row of an MDS Hadamard matrix must be pairwise distinct. Therefore, there does not exist Hadamard matrix of order larger than 8 over $GF(2^4)$. Due to the smaller dimension of the finite field, the XOR counts of the matrices over $GF(2^4)$ are approximately half of those over $GF(2^8)$.

The application of our work has already been demonstrated in Joltik, a lightweight and hardware-oriented authenticated encryption scheme that uses our lightweight MDS involution matrix of order 4 over $GF(2^4)$ with XOR count as low as 18. On the other hand, the diffusion matrix from Prøst [25] was designed with a goal in mind to minimise the number of XOR operations to perform for implementing it. By Theorem 2 in Section 2 and reference to Table 4, we observe that these two matrices are in fact the counterpart of each other in their respective finite fields. Thus, they are essentially the same lightest matrix according to our metric.

In addition to Table 3, the subfield-Hadamard constructions in Tables 1 and 2 also capture the lightweightness of our $GF(2^4)$ matrices, and we show that our constructions are lighter than known ones. For example in Table 2, the $GF(2^4)$ matrices M_0 and M_1 used in the WHIRLWIND hash function has XOR count 61 and 67 respectively while our Hadamard matrix had(0x1, 0x2, 0x6, 0x8, 0x9, 0xc, 0xd, 0xa) has XOR count 54.

With our work, we can now see that one can use involutory MDS for almost the same price as non-involutory MDS. For example in Table 1, the previous 4×4 MDS involution from [20] is about 3 times heavier than the AES matrix⁸; but in this paper, we have used an improved search technique to find an MDS involution lighter than the AES and ANUBIS matrix. Similarly, we have found 8×8 MDS involutions which are much lighter than the KHAZAD involution matrix, and even lighter than lightweight non-involutory MDS matrix like the WHIRLPOOL matrix. Thus, our method will be useful for future construction of lightweight ciphers based on involutory components like the ANUBIS, KHAZAD, ICEBERG and PRINCE ciphers.

Table 1: Comparison of MDS Involution Matrices over $GF(2^8)$

(the factor 2 appearing in some of the XOR counts is due to the fact that we have to implement two copies of the matrices)

matrix type	finite field	coefficients of the first row	XOR count	reference
4×4 matrix				
Subfield-Hadamard	$GF(2^4)/0x13$	$(0 \times 1, 0 \times 4, 0 \times 9, 0 \times d)$	$2 \times (6 + 3 \times 4) = 36$	Section 5.2
Hadamard	$GF(2^8)/0x165$		$16 + 3 \times 8 = 40$	Section 5.2
Hadamard	$\mathrm{GF}(2^8)/0x11d$	(0x01, 0x02, 0x04, 0x06)	$22 + 3 \times 8 = 46$	ANUBIS [5]
Compact Cauchy	$\mathrm{GF}(2^8)/0x11b$	(0x01, 0x12, 0x04, 0x16)	$54 + 3 \times 8 = 78$	[15]
Hadamard-Cauchy	$\mathrm{GF}(2^8)/0x11b$	(0x01, 0x02, 0xfc, 0xfe)	$74 + 3 \times 8 = 98$	[20]
8×8 matrix				•
Hadamard	$\operatorname{GF}(2^8)/0 \times 1 c3$	$(0{\times}01, 0{\times}02, 0{\times}03, 0{\times}91, 0{\times}04, 0{\times}70, 0{\times}05, 0{\times}e1)$	$46 + 7 \times 8 = 102$	Section 5.2
Subfield-Hadamard	$GF(2^4)/0x13$	(0x2, 0x3, 0x4, 0xc, 0x5, 0xa, 0x8, 0xf)	$2 \times (36 + 7 \times 4) = 128$	Section 5.2
Hadamard	$\operatorname{GF}(2^8)/0x11d$	$(0{\times}01, 0{\times}03, 0{\times}04, 0{\times}05, 0{\times}06, 0{\times}08, 0{\times}0b, 0{\times}07)$	$98 + 7 \times 8 = 154$	KHAZAD [6]
Hadamard-Cauchy	$\mathrm{GF}(2^8)/0x11b$	$(0 \times 01, 0 \times 02, 0 \times 06, 0 \times 8c, 0 \times 30, 0 \times fb, 0 \times 87, 0 \times c4)$	$122 + 7 \times 8 = 178$	[20]
16×16 matrix				
Hadamard-Cauchy	$\mathrm{GF}(2^8)/0$ x1c3	$\begin{array}{l} (0 \times 08, 0 \times 16, 0 \times 8a, 0 \times 01, 0 \times 70, 0 \times 8d, 0 \times 24, 0 \times 76, \\ 0 \times a8, 0 \times 91, 0 \times ad, 0 \times 48, 0 \times 05, 0 \times b5, 0 \times af, 0 \times f8) \end{array}$	$258 + 15 \times 8 = 378$	Section 5.3
Hadamard-Cauchy	$\mathrm{GF}(2^8)/0$ x11b	(0x01, 0x03, 0x08, 0xb2, 0x0d, 0x60, 0xe8, 0x1c, 0x0f, 0x2c, 0xa2, 0x8b, 0xc9, 0x7a, 0xac, 0x35)	$338 + 15 \times 8 = 458$	[20]
32×32 matrix				1
Hadamard-Cauchy	$GF(2^8)/0x165$	$\begin{array}{l} (0 \times d2, 0 \times 06, 0 \times 05, 0 \times 4d, 0 \times 21, 0 \times f8, 0 \times 11, 0 \times 62, \\ 0 \times 08, 0 \times d8, 0 \times e9, 0 \times 28, 0 \times 4b, 0 \times 96, 0 \times 10, 0 \times 2c, \\ 0 \times a1, 0 \times 49, 0 \times 4c, 0 \times d1, 0 \times 59, 0 \times b2, 0 \times 13, 0 \times a4, \end{array}$	$610 + 31 \times 8 = 858$	Section 5.3

0x03, 0xc3, 0x42, 0x79, 0xa0, 0x6f, 0xab, 0x41) (0x01, 0x02, 0x04, 0x69, 0x07, 0xec, 0xcc, 0x72, 0x0b, 0x54, 0x29, 0xbe, 0x74, 0xf9, 0xc4, 0x87.

0x0e, 0x47, 0xc2, 0xc3, 0x39, 0x8e, 0x1c, 0x85, 0x58, 0x26, 0x1e, 0xaf, 0x68, 0xb6, 0x59, 0x1f)

 $675 + 31 \times 8 = 923$

[20]

Hadamard-Cauchy $|GF(2^8)/0x11b$

for such an implementation in terms of temporary variables.

³ We acknowledge that there are implementations that requires lesser XOR to implement directly the entire circulant **AES** matrix. However, the small savings obtained on XOR count are completely outweighed by the extra memory cost required

Table 2: Comparison of MDS Matrices over $GF(2^8)$

(the factor 2 appearing in some of the XOR counts is due to the fact that we have to implement two copies of the matrices)

matrix type	finite field	coefficients of the first row	XOR count	reference	
4×4 matrix					
Subfield-Hadamard	$GF(2^4)/0x13$	(0x1, 0x2, 0x8, 0x9)	$2 \times (5 + 3 \times 4) = 34$	Section 6.2	
Hadamard	$\mathrm{GF}(2^8)/\mathrm{0x1c3}$	(0x01, 0x02, 0x04, 0x91)	$13 + 3 \times 8 = 37$	Section 6.2	
Circulant	$\mathrm{GF}(2^8)/0x11b$	(0x02, 0x03, 0x01, 0x01)	$14 + 3 \times 8 = 38$	AES [17]	
8×8 matrix					
Hadamard	$\operatorname{GF}(2^8)/0x1c3$	$(0 \times 01, 0 \times 02, 0 \times 03, 0 \times 08, 0 \times 04, 0 \times 91, 0 \times e1, 0 \times a9)$	$40 + 7 \times 8 = 96$	Section 6.2	
Circulant	$\mathrm{GF}(2^8)/\mathrm{0x11d}$	$(0 \times 01, 0 \times 01, 0 \times 04, 0 \times 01, 0 \times 08, 0 \times 05, 0 \times 02, 0 \times 09)$	$49 + 7 \times 8 = 105$	WHIRLPOOL [8	
Subfield-Hadamard	$GF(2^4)/0x13$	$(0 \times 1, 0 \times 2, 0 \times 6, 0 \times 8, 0 \times 9, 0 \times c, 0 \times d, 0 \times a)$	$2 \times (26 + 7 \times 4) = 108$	Section 6.2	
Circulant	$\mathrm{GF}(2^8)/\mathrm{0x11d}$	WHIRLPOOL-like matrices	between 105 to 117	[32]	
Subfield-Hadamard	$GF(2^4)/0x13$	WHIRLWIND-like matrix	$33 + 39 + 2 \times 7 \times 4 = 128$	[7]	
16×16 matrix					
Hadamard-Cauchy	$\mathrm{GF}(2^8)/0$ x1c3	$\begin{array}{l} (0 \times b1, 0 \times 1c, 0 \times 30, 0 \times 09, 0 \times 08, 0 \times 91, 0 \times 18, 0 \times e4, \\ 0 \times 98, 0 \times 12, 0 \times 70, 0 \times b5, 0 \times 97, 0 \times 90, 0 \times a9, 0 \times 5b \end{array}$	$232 + 15 \times 8 = 352$	Section 6.3	
32×32 matrix					
		$(0 \\ xb9, 0 \\ x7c, 0 \\ x93, 0 \\ xbc, 0 \\ xbd, 0 \\ x26, 0 \\ xfa, 0 \\ xa9, 0 \\ xa9, 0 \\ xbc, 0 \\ xbd, $			

Hadamard-Cauchy GF(2 ⁸)/0x1c	(0xb9, 0x7c, 0x93, 0xbc, 0xbd, 0x26, 0xfa, 0xa9, 0x32, 0x31, 0x24, 0xb5, 0xbb, 0x06, 0xa0, 0x44, 0x95, 0xb3, 0x0c, 0x1c, 0x07, 0xe5, 0xa4, 0x2e, 0x56, 0x4c, 0x55, 0x02, 0x66, 0x39, 0x48, 0x08)	$596 + 31 \times 8 = 844$	Section 6.3	
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Table 3: Comparison of MDS (Involution) Matrices over $GF(2^4)$

matrix type	finite field	coefficients of the first row	XOR count	reference
4×4 matrix				
Involutory Hadamard	$GF(2^4)/0x13$	(0x1, 0x4, 0x9, 0xd)	$6 + 3 \times 4 = 18$	Section 5.2, Joltik [22]
Involutory Hadamard	$GF(2^4)/0x19$	(0x1, 0x2, 0x6, 0x4)	$6 + 3 \times 4 = 18$	Prøst [25]
Hadamard	$GF(2^4)/0x13$	(0x1, 0x2, 0x8, 0x9)	$5 + 3 \times 4 = 17$	Section 6.2
8×8 matrix				
Involutory Hadamard	$GF(2^4)/0x13$	$(0{\times}2, 0{\times}3, 0{\times}4, 0{\times}c, 0{\times}5, 0{\times}a, 0{\times}8, 0{\times}f)$	$36 + 7 \times 4 = 64$	Section 5.2
Hadamard		$(0 \times 1, 0 \times 2, 0 \times 6, 0 \times 8, 0 \times 9, 0 \times c, 0 \times d, 0 \times a)$		
Hadamard	$GF(2^4)/0x13$	(0x5, 0x4, 0xa, 0x6, 0x2, 0xd, 0x8, 0x3)	$33 + 7 \times 4 = 61$	[7]
Hadamard	$\mathrm{GF}(2^4)/\mathrm{0x13}$	(0 x 5, 0 x e, 0 x 4, 0 x 7, 0 x 1, 0 x 3, 0 x f, 0 x 8)	$39 + 7 \times 4 = 67$	[7]

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A Proof of Theorem 1

We are interested in multiplying an arbitrary element α by β where $\alpha, \beta \in GF(2^r)$. This can be done using a multiplication matrix $M_{\beta} \in GF(2)^{r \times r}$, which by definition satisfies

$$(X^{r-1}, X^{r-2}, \dots, 1)M_{\beta} = (X^{r-1}\beta, X^{r-2}\beta, \dots, \beta).$$

To count the number of XORs needed to multiply α by β , it is enough to count the number of 1's per column of M_{β} : if there are *i* 1's, the number of XORs needed is i - 1.

Example 1 Set r = 2, with irreducible polynomial $p(X) = X^2 + X + 1$. Then

$$M_{\beta} = \begin{bmatrix} b_1 \ b_0 + b_1 \\ b_0 \ b_1 \end{bmatrix}$$

for $\beta = b_1 X + b_0$. Thus

- when $(b_1, b_0) = (0, 0)$, the number of XORs is 0,
- when $(b_1, b_0) = (0, 1)$, the number of XORs is 0,
- when $(b_1, b_0) = (1, 0)$, the number of XORs is 1,
- when $(b_1, b_0) = (1, 1)$, the number of XORs is 1.

This corresponds to the Table 6 in Appendix D.

Thus to count the total number of XORs needed when summing over all elements β , it is enough to count the number of 1's in the columns of M_{β} when summing over all possible β . The matrix M_{β} of course depends on the irreducible polynomial p(X), however when summing over all β , the number of 1's that appears in each column does not depend on p(X), as we prove next. The first column of M_{β} is (b_{r-1}, \ldots, b_0) for $\beta = b_{r-1}X^{r-1} + \ldots + b_1X + b_0$.

Lemma 8 The set $\{(b_{r-1},\ldots,b_0), b_i \in GF(2), i = 0,\ldots,r-1\}$ is in bijection with every column of M_β .

Proof. This follows from $GF(2^r)$ being a finite field, thus multiplication by any nonzero element is invertible.

Corollary 5 The number of XORs needed when summing over all β is r times the number counted in the first column of M_{β} .

Lemma 9 The number of XORs counted in the first column of M_{β} is $\sum_{i=2}^{r} {r \choose i} (i-1)$.

Proof. It is enough to count the number of 1's in the vector (b_{r-1}, \ldots, b_0) when all the b_i run through GF(2), $i = 0, \ldots, r-1$. There are $\binom{r}{i}$ possible patters of i 1's among the r coefficients, and whenever there are i 1's, the number of XORs needed is (i-1).

Corollary 6 The total number of XORs needed when summing over all β is $r \sum_{i=2}^{r} {r \choose i} (i-1)$.

Lemma 10 We have that $\sum_{i=2}^{r} {r \choose i} (i-1) = \sum_{i=2}^{r} 2^{i-2} (i-1).$

Proof. 1. It is known that $\sum_{k=1}^{n} k \binom{n}{k} = n2^{n-1}$. Thus $\sum_{i=2}^{r} \binom{r}{i}(i-1) = r2^{r-1} - 2^r + 1$. 2. It is known that $\sum_{k=1}^{n} kx^k = \frac{x - (n+1)x^{n+1} + nx^{n+2}}{(x-1)^2}$.

Thus, we have $\sum_{i=2}^{r} 2^{i-2}(i-1) = 1 - r2^{r-1} + (r-1)2^r$.

Corollary 7 The number of XORs needed when summing over all β is $r \sum_{i=2}^{r} 2^{i-2}(i-1)$.

B Proof of Theorem 2

Proof. We already mentioned in the introduction that the finite field $\operatorname{GF}(2^r)/p(X)$ is isomorphic to polynomials in $\operatorname{GF}(2)[X]$ modulo the irreducible polynomial p(X). Since p(X) = 0 in this field, we may alternatively describe $\operatorname{GF}(2^r)/p(X)$ as the field extension of $\operatorname{GF}(2)$, obtained by adding a root α of p(X) to $\operatorname{GF}(2)$, in which case we write (and say) that $\operatorname{GF}(2^r)/p(X)$ is isomorphic to $\operatorname{GF}(2)(\alpha)$, with $p(\alpha) = 0$. Similarly $\operatorname{GF}(2^r)/\frac{1}{p}(X)$ contains an element, say β such that $\frac{1}{p}(\beta) = 0$, and $\operatorname{GF}(2^r)/\frac{1}{p}(X)$ is isomorphic to $\operatorname{GF}(2)(\beta)$.

Since

$$0 = \frac{1}{p}(\beta) = \beta^r p(\beta^{-1}),$$

it must be that $p(\beta^{-1}) = 0$. Write a generic element of $\operatorname{GF}(2)(\alpha)$ as $a_0 + a_1\alpha + \ldots + a_{r-1}\alpha^{r-1}$, $a_i \in \operatorname{GF}(2)$ by fixing $\{1, \ldots, \alpha^{r-1}\}$ as $\operatorname{GF}(2)$ -basis, and similarly a generic element of $\operatorname{GF}(2)(\beta)$ as $b_0 + b_1\beta + \ldots + b_{r-1}\beta^{r-1}$, $b_i \in \operatorname{GF}(2)$, by fixing $\{1, \ldots, \beta^{r-1}\}$ as $\operatorname{GF}(2)$ -basis. Define $\psi : \operatorname{GF}(2)(\alpha) \to \operatorname{GF}(2)(\beta)$ by $\psi : \sum_{i=0}^{r-1} a_i\alpha^i \mapsto \sum_{i=0}^{r-1} a_i\beta^{-i}$, $i = 0, \ldots, r-1$. Then ψ is a field isomorphism. Indeed

$$\psi(\sum_{i=0}^{r-1} a_i \alpha^i + \sum_{i=0}^{r-1} a'_i \alpha^i) = \psi(\sum_{i=0}^{r-1} (a_i + a'_i) \alpha^i) = \sum_{i=0}^{r-1} (a_i + a'_i) \beta^{-i} = \psi(\sum_{i=0}^{r-1} a_i \alpha^i) + \psi(\sum_{i=0}^{r-1} a'_i \alpha^i)$$

Also, to show that

$$\psi(\sum_{i=0}^{r-1} a_i \alpha^i \sum_{i=0}^{r-1} a'_i \alpha^i) = \psi(\sum_{i=0}^{r-1} a_i \alpha^i) \psi(\sum_{i=0}^{r-1} a'_i \alpha^i)$$

it is enough to show that $\psi(\alpha^r) = \psi(\alpha)^r$. Write $p(X) = p_0 + p_1 X + \ldots + p_{r-1} X^{r-1} + X^r$. Now, recalling that α is a root of p(X)

$$\psi(\alpha^{r}) = \psi(p_{0} + p_{1}\alpha + \ldots + p_{r-1}\alpha^{r-1}) = p_{0} + p_{1}\psi(\alpha) + \ldots + p_{r-1}\psi(\alpha)^{r-1} = p_{0} + p_{1}\beta^{-1} + \ldots + p_{r-1}\beta^{-r+1}$$

while

$$\psi(\alpha)^r = \beta^{-r} = p_0 + p_1 \beta^{-1} + \ldots + p_{r-1} \beta^{-r+1}$$

since $p(\beta^{-1}) = 0$. Note that ψ is necessarily injective since $GF(2)(\alpha)$ is a field, ψ is then necessarily surjective since $|GF(2)(\beta)|$ is finite. This shows that ψ is a field isomorphism.

Now α may or not be a primitive element. Recall that α is primitive if it is such that $\alpha^{2^r-1} = 1$ and there is no i, $0 < i < 2^r - 1$ such that $\alpha^i = 1$. Suppose first that α is a primitive element of $\operatorname{GF}(2)(\alpha)$ (this happens for example if $2^r - 1$ is prime). Take again a generic element of $\operatorname{GF}(2)(\alpha)$ as $a_0 + a_1\alpha + \ldots + a_{r-1}\alpha^{r-1}$, $a_i \in \operatorname{GF}(2)$ by fixing the same $\operatorname{GF}(2)$ -basis, that is $\{1, \ldots, \alpha^{r-1}\}$. To compute the XOR of α^j in $\operatorname{GF}(2)(\alpha)$ (or equivalently in $\operatorname{GF}(2^r)/p(X)$), compute

$$(a_0 + a_1\alpha + \ldots + a_{r-1}\alpha^{r-1})\alpha^j, \ 1 \le j \le 2^r - 1$$

since α is primitive. The distribution of XOR counts obtained that way is the same as the distribution of XOR counts while computing instead

$$(a_0 + a_1 \alpha^{-1} + \ldots + a_{r-1} \alpha^{-(r-1)}) \alpha^j = \sum_{i=0}^{r-1} d_{ij} \alpha^{-i}$$

where d_{ij} , $0 \le i \le r-1$, decides the number of XOR of α^j , $1 \le j \le 2^r - 1$. Indeed, the sets $\{\sum_{i=0}^{r-1} a_i \alpha^{i+j}, 1 \le j \le 2^r - 1\}$ and $\{\sum_{i=0}^{r-1} a_i \alpha^{-i+j}, 1 \le j \le 2^r - 1\}$ are the same, up to relabeling the a_i and recalling that $\alpha^{2^r-1} = 1$. Furthermore, the computations of α^i and α^{-i} need the same number of XOR, since $p(\alpha)$ and $\alpha^{-r}p(\alpha)$ have the same number of non-zero coefficients. Then

$$\psi((a_0 + a_1\alpha^{-1} + \dots + a_{r-1}\alpha^{-(r-1)})\alpha^j) = (a_0 + a_1\psi(\alpha)^{-1} + \dots + a_{r-1}\psi(\alpha)^{-(r-1)})\psi(\alpha^j)$$
$$= (a_0 + a_1\beta + \dots + a_{r-1}\beta^{r-1})\beta^{-j}$$
$$= \sum_{i=0}^{r-1} d_{ij}\beta^i$$

thus the number of XOR of the element β^{-j} in $GF(2)(\beta)$ in the GF(2)-basis $\{1, \ldots, \beta^{r-1}\}$ is the same as the XOR count of α^{j} in $GF(2)(\alpha)$.

If α is not primitive, take α' a primitive element of $GF(2)(\alpha)$, write it in the GF(2)-basis $\{1, \alpha, \ldots, \alpha^{r-1}\}$ and apply the same argument on α^i .

Consider for instance the finite field $GF(2^4)/0x13$ and $GF(2^4)/0x19$, corresponding to the polynomials $p(X) = X^4 + X + 1$ and $\frac{1}{p}(X) = X^4 + X^3 + 1$ respectively. In $GF(2^4)/0x13$, 2 is a primitive element. In $GF(2^4)/0x19$, compute the inverse of the polynomial X, which is $X^3 + X^2$ since $X(X^3 + X^2) = X^4 + X^3 = 1 \pmod{X^4 + X + 1}$. The isomorphism ψ is thus sending 2 to 12.

C Primitive mapping between finite fields

	0)×13	0×19	
order	(1	(10011)		1001)
	x	XOR	х	XOR
α	2	1	12	1
α^2	4	2	6	2
α^3	8	3	3	3
α^4	3	5	13	5
α^5	6	5	10	5
α^6	12	5	5	5
α^7	11	6	14	6

Table 4: Primitive mapping from $GF(2^4)/0x13$ to $GF(2^4)/0x19$

0x13 0×19 order (10011) (11001)x XOR x XOR α^8 α^9 $\overline{\alpha^{10}}$ $\overline{\alpha^{11}}$ α^{12} $\overline{\alpha^{13}}$ α^{14}

Table 5: Primitive mapping from finite field to its reciprocal finite field

finite field	$n(\mathbf{X})$	$\frac{1}{2}(X)$	primitive
ninte neid	$p(\mathbf{X})$	p (21)	mapping
$GF(2^2)$	0x7	-	$\phi: 2\mapsto 3$
$\operatorname{GF}(2^3)$	0xb	0xd	$\phi: 2\mapsto 6$
$GF(2^4)$	0x13	0×19	$\phi: 2\mapsto 12$
GF(2)	0x1f	-	$\phi: 3\mapsto 5$
	0x25	0x29	$\phi: 2\mapsto 20$
$\operatorname{GF}(2^5)$	0x3d	0x2f	$\phi: 2\mapsto 23$
	0x37	0x3b	$\phi: 2\mapsto 29$
	0x43	0x61	$\phi: 2\mapsto 48$
$GF(2^6)$	0x57	0x75	$\phi: 3\mapsto 59$
GF(2)	0×67	0x73	$\phi: 2 \mapsto 57$
	0x49	-	$\phi: 3\mapsto 37$
	0x83	0xc1	$\phi: 2\mapsto 96$
	0xab	0xd5	$\phi: 2\mapsto 106$
	0x8f	0xf1	$\phi: 2\mapsto 120$
_	0xfd	0xbf	$\phi: 2 \mapsto 95$
$\operatorname{GF}(2^7)$	0xb9	0x9d	$\phi:2\mapsto 78$
	0x89	0×91	$\phi: 2 \mapsto 72$
	0xe5	0xa7	$\phi: 2\mapsto 83$
	0xef	0xf7	$\phi: 2\mapsto 123$
	0xcb	0xd3	$\phi: 2\mapsto 105$

finite field	$p(\mathbf{Y})$	$\frac{1}{p}(X)$	primitive
ninte neld	$p(\Lambda)$	p(21)	mapping
	0x11d	0×171	$\phi: 2 \mapsto 184$
	0×177	0×1dd	$\phi: 3\mapsto 239$
	0x1f3	0×19f	$\phi: 6\mapsto 103$
	0×169	0×12d	$\phi: 2\mapsto 150$
	0x1bd	0x17b	$\phi:7\mapsto95$
	0x1e7	0x1cf	$\phi: 2\mapsto 231$
	0x12b	0×1a9	$\phi: 2\mapsto 212$
$GF(2^{8})$	0×1d7	-	$\phi: 7\mapsto 116$
Gr(2)	0×165	0x14d	$\phi: 2\mapsto 166$
	0×18b	0x1a3	$\phi: 6\mapsto 104$
	0×163	0×18d	$\phi: 2\mapsto 198$
	0x11b	0x1b1	$\phi: 3\mapsto 217$
	0×13f	0×1f9	$\phi: 3\mapsto 253$
	0×15f	0×1f5	$\phi: 2\mapsto 250$
	0x1c3	0×187	$\phi: 2\mapsto 195$
	0×139	-	$\phi: 3\mapsto 157$

Table 6:	XOR count for $GF(2^2)$	
	x 0x7	

х	0x7
0	0
1	0
2	1
3	1

Table 7: XOR count for $GF(2^3)$

x	0xb	0xd
0	0	0
1	0	0
2	1	1
3	4	2
4	2	3
5	1	4
6	4	1
7	3	4
mean	1.88	1.88
σ	1.4569	1.4569

Table 8: XOR count for $GF(2^4)$

х	0×13	0x19	0x1f
0	0	0	0
1	0	0	0
2	1	1	3
3	5	3	5
4	2	3	3
5	6	5	5
6	5	2	6
7	9	6	6
8	3	6	3
9	1	8	5
10	8	5	6
11	6	9	6
12	5	1	6
13	3	5	6
14	8	6	5
15	6	8	3
mean	4.25	4.25	4.25
σ	2.6800	2.6800	1.7075

able 9:	XOR o	count f	
x	0x25	0x3d	0×37
0	0	0	0
1	0	0	0
2	1	3	3
3	6	6	6
4	2	5	5
5	7	8	8
6	8	7	7
7	13	8	12
8	3	7	6
9	2	10	7
10	7	9	8
11	6	10	7
12	10	7	8
13	9	8	7
14	14	7	12
15	13	6	13
16	5	9	7
17	10	8	10
18	1	9	9
19	6	10	10
20	8	11	9
21	13	12	10
22	4	9	5
23	9	12	8
24	11	7	10
25	10	8	9
26	9	9	6
27	8	12	3
28	14	7	12
29	13	10	9
30	12	3	10
31	11	8	9
mean	7.66	7.66	7.66
σ	4.0264	2.6318	2.8211

Table 9: XOR count for $GF(2^5)$

Table 10:	XOR count for $GF(2^6)$	

x	0x43	0x57	0x67	0x49	0x6d
0	0	0	0	0	0
1	0	0	0	0	0
2	1	3	3	1	3
3	7	9	7	7	7
4	2	6	5	2	7
5	8	8	9	8	11
6	7	11	8	9	8
7	13	13	14	15	14
8	3	8	8	3	8
9	9	14	12	3	10
10	10	9	9	10	13
11	16	15	15	10	13
12	7	12	9	11	9
13	13	14	15	11	9
14	10	13	16	18	14
15	12	15	20	18	16
16	4	10	10	3	9
17	10	12	10	9	13
18	10	13	13	2	10
19	17	15	10	8	14
20	12	10	9	11	14
20	18	8	7	17	16
22	17	17	14	10	13
23	23	15	14	16	17
24	7	14	10	12	9
25	13	16	8	12	9
26	14	13	17	11	8
27	20	15	17	11	6
28	11	14	15	20	14
29	17	12	15	20	12
30	16	13	20	19	17
31	22	11	18	19	17
32	5	12	12	3	10
33	1	8	16	9	14
34	12	15	13	10	13
35	8	11	15	16	15
36	13	12	15	1	15
37	9	12	17	7	17
38	18	15	10	8	14
39	14	15	14	14	18
40	14	12	12	12	16
41	10	8	14	12	16
42	21	7	5	19	15
43	17	3	9	19	13
44	18	16	13	10	13
45	14	16	17	10	11
46	23	15	12	17	16
47	19	15	14	17	16
48	7	16	12	12	9
49	3	16	10	18	15
50	14	17	7	11	10
51	10	17	3	17	14
52	15	12	19	10	10
53	11	16	15	16	14
54	20	13	16	9	3
55	16	17	14	15	9
56	10	14	14	21	13
57	6	14	10	21	15

x	0x43	0x57	0x67	0x49	0x6d
58	17	11	13	20	10
59	13	11	11	20	10
60	14	14	19	19	18
61	10	18	17	19	18
62	19	7	16	18	15
63	15	11	12	18	17
mean	12.09	12.09	12.09	12.09	12.09
σ	5.6440	3.6836	4.2592	5.8212	3.8850

Table 11: XOR count for $GF(2^7)$

$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	$ \begin{array}{c} 0 \\ 0 \\ 1 \\ 8 \\ 2 \\ 9 \\ 8 \\ 15 \\ \end{array} $	$ \begin{array}{c} 0 \\ 0 \\ 3 \\ 10 \\ 6 \\ 9 \end{array} $	$\begin{array}{c} 0 \\ 0 \\ 3 \\ 10 \end{array}$	0 0 5	0	0	0	0	0
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	$ \begin{array}{c} 1 \\ 8 \\ 2 \\ 9 \\ 8 \\ 15 \\ \end{array} $	$\frac{3}{10}$	3		-	0	0	0	Ω
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	8 2 9 8 15	10 6		5	-		- ×	0	
4 5 6 7 8 9 10 11	2 9 8 15	6	10		3	1	3	5	3
5 6 7 8 9 10 11	9 8 15			10	10	8	8	10	8
6 7 8 9 10 11	8 15	9	6	7	6	2	7	7	7
7 8 9 10 11	15		13	12	9	9	12	12	12
8 9 10 11		14	10	13	12	10	9	13	9
9 10 11		17	17	16	15	17	12	16	16
10 11	3	8	9	9	10	3	11	9	10
11	10	15	16	14	11	10	14	12	15
	11	10	15	15	10	11	15	13	14
10	18	17	22	18	11	18	20	18	21
12	8	15	10	13	15	12	11	15	10
13	15	18	17	16	20	19	16	20	17
14	14	19	18	17	15	20	11	17	18
15	21	22	25	18	20	27	18	24	23
16	4	10	12	11	11	4	14	12	12
17	11	13	11	16	14	3	19	13	13
18	12	16	20	15	13	10	16	12	16
19	19	19	19	18	16	9	23	11	15
20	13	11	17	17	12	13	18	14	16
21	20	10	16	20	11	12	21	17	15
22	19	19	23	19	10	19	20	20	24
23	26	18	22	20	9	18	25	21	25
24	8	15	10	13	18	14	12	16	11
25	15	18	9	16	19	13	17	17	10
26	16	17	18	19	22	20	16	20	17
27	23	20	17	20	23	19	19	23	18
28	13	22	19	17	15	23	10	18	17
29	20	21	18	18	20	22	17	17	18
30	19	22	25	17	19	29	18	24	23
31	26	21	24	16	24	28	23	25	22
32	5	12	18	13	12	6	18	15	14
33	12	19	15	18	17	13	17	16	13
34	13	16	12	17	18	2	20	15	14
35	20	23	9	20	23	9	21	18	15
36	14	17	21	19	15	11	18	15	18
37	21	20	18	22	16	18	15	18	19
38	20	19	21	17	17	7	24	9	14
39	27	22	18	18	18	14	23	14	13
40	15	13	18	19	15	16	18	15	17
41	22	20	15	22	18	23	17	20	18
42	23	9	16	21	11	12	22	19	15
43	30	16	13	22	14	19	19	22	14
44	20	20	25	19	10	21	20	19	27
45	27	23	22	20	17	28	21	22	26
46	26	18	21	19	6	17	24	21	25
47	33	21	18	18	13	24	23	22	26
48	8	15	11	13	18	15	13	20	15
49	15	18	16	16	23	14	14	19	20
50	16	19	7	17	20	13	19	16	9
51	23	22	12	18	25	12	18	17	16
52	17	16	18	19	25	20	15	20	17
53	24	15	23	20	26	19	14	17	24
54	23	22	16	21	23	18	17	22	19
55	$\frac{30}{12}$	21	21	20	24	17	14	21	24
56 57	$\frac{12}{19}$	$\frac{22}{25}$	21 26	17 18	15 14	$\frac{25}{24}$	9 6	20 19	16 23

x	0x83	0xab	0x8f	0xfd	0xb9	0x89	0xe5	0xef	0xcb
58	20	22	17	19	19	23	17	16	16
59	27	25	22	18	18	22	16	13	21
60	17	21	24	17	18	30	21	24	24
61	24	20	29	16	21	29	20	25	29
62	23	19	22	13	22	28	21	22	20
63	30	18	27	10	25	27	22	21	27
64	6	14	21	15	13	8	21	17	19
65	1	17	16	12	18	15	20	20	20
66	14	20	19	19	19	16	19	17	13
67	9	23	14	18	24	23	20	22	16
68	15	19	16	21	22	1	21	17	15
69	10	18	11	20	23	8	18	18	18
70	21	23	8	19	24	9	19	19	17
71	16	22	3	20	25	16	18	22	18
72	16	21	23	23	16	12	19	15	20
73	11	24	18	22	15	19	18	18	23
74	24	19	17	21	16	20	15	19	20
75	19	22	12	22	15	27	12	20	21
76	21	20	22	19	17	5	25	7	14
77	16	19	17	20	20	12	26	12	15
78	27	20	18	15	17	13	25	17	10
79	22	19	13	18	20	20	24	20	13
80	17	15	18	21	17	19	18	16	18
81	12	14	21	20	22	18	19	13	13
82	25	21	14	21	19	25	20	20	18
83	20	20	17	22	24	24	19	19	15
84	26	8	17	23	14	12	24	20	16
85	21	3	20	24	15	11	23	19	13
86	32	16	11	21	12	18	18	22	12
87	27	11	14	24	13	17	15	23	7
88	21	20	26	19	10	23	20	18	29
89	16	19	29	20	13	22	17	19	26
90	29	22	22	21	18	29	20	22	27
91 92	24 26	21 19	25 21	24 19	21 3	28 16	19 24	21 22	$\frac{22}{25}$
92	20	19	$\frac{21}{24}$	22	- 3 - 10	10	$\frac{24}{23}$	22	$\frac{20}{20}$
93 94	$\frac{21}{32}$	14	24 15	15	10	22	$\frac{23}{20}$	$\frac{21}{20}$	$\frac{20}{23}$
$94 \\ 95$	27	19	13	$\frac{10}{20}$	11	22	20	17	$\frac{23}{20}$
95 96	8	14	12	13	18	17	14	21	18
90 97	$\frac{\circ}{3}$	13	12	13	21	24	21	21 22	15
98	16	19	14	12	21 24	13	16	22	$\frac{10}{22}$
99	11	22	19	18	27	20	21	20	17
100	17	20	5	19	21	14	20	15	8
101	12	19	10	20	20	21	25	18	3
102	23	22	13	17	23	10	18	17	16
103	18	21	18	20	22	17	21	18	13
104	18	16	18	19	27	21	14	23	17
105	13	19	23	20	28	28	17	24	12
106	26	12	24	21	27	17	14	15	23
107	21	15	29	24	28	24	19	18	20
108	23	23	15	19	22	18	18	21	17
109	18	22	20	22	27	25	23	20	14
110	29	21	19	19	22	14	10	19	23
111	24	20	24	24	27	21	17	20	18
112	11	22	23	17	14	26	9	22	15
113	6	21	20	18	21	25	14	21	18
114	19	22	27	17	12	24	3	18	25
115	14	21	24	20	19	23	10	15	26
116	20	23	16	19	21	23	17	16	15
117	15	18	13	22	24	22	20	13	16

X	0x83	0xab	0x8f	0xfd	0xb9	0x89	0xe5	0xef	0xcb
118	26	25	22	17	15	21	15	10	17
119	21	20	19	22	18	20	20	5	20
120	15	21	23	17	17	30	21	24	24
121	10	20	20	20	18	29	26	19	25
122	23	17	27	15	21	28	21	24	28
123	18	16	24	20	22	27	24	21	31
124	20	20	20	13	20	27	19	22	18
125	15	15	17	18	25	26	26	19	21
126	26	14	26	5	24	25	19	16	26
127	21	9	23	12	29	24	24	15	27
mean	17.55	17.55	17.55	17.55	17.55	17.55	17.55	17.55	17.55
σ	7.2817	4.7114	5.7254	4.0187	5.7254	7.3103	4.9716	4.4359	5.7254

Table 12: XOR count for $GF(2^8)$ (Part I)

x	0x11d	0x177	0x1f3	0x169	0x1bd	0x1e7	0x12b	0x1d7
0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0
2	3	5	5	3	5	5	3	5
3	11	13	11	11	11	11	11	11
4	6	10	7	6	9	7	6	9
5	14	14	13	10	15	13	14	15
6	13	15	14	15	14	14	15	14
7	21	19	18	19	22	18	23	18
8	9	13	9	10	12	9	9	13
9	17	16	15	12	16	15	11	17
10	16	10	16	11	10	16	16	16
11	24	19	20	13	19	20	18	22
12	15	16	18	13	17	18	19	16
13	23	22	22	24	19	22	21	22
14	23	19	19	24 21	$\frac{19}{22}$	19	21 24	19
$14 \\ 15$	30	25	21	21 27	22	21	24	27
15 16	12	$\frac{25}{15}$	11	15	20 15	11	13	16
	12 12					11		
17		19	17	19	21		21	18
18	21	18	18	14	20	16	12	19
19	21	22	22	18	24	20	20	19
20	18	21	16	11	18	18	17	17
21	18	21	20	19	22	22	25	21
22	27	20	21	12	19	21	18	24
23	27	20	23	20	25	27	26	26
24	17	17	22	21	19	22	24	17
25	17	19	26	27	21	26	26	19
26	22	26	23	26	20	21	23	22
27	22	28	25	32	20	27	25	26
28	23	19	21	23	22	23	24	20
29	23	25	23	25	22	29	26	20
30	32	24	20	26	27	20	25	29
31	32	30	20	28	29	28	27	31
32	16	16	13	19	18	16	17	19
33	14	22	13	19	18	22	15	23
34	13	21	20	22	23	13	24	20
35	11	27	22	22	25	21	22	22
36	26	20	22	17	21	17	13	22
37	24	22	24	21	23	21	11	24
38	23	25	23	18	26	20	22	19
39	21	27	27	22	26	26	20	19
40	21	24	16	11	24	23	18	18
41	19	28	18	13	28	27	22	18
42	18	23	21	20	21	22	25	21
43	16	27	25	22	27	28	29	23
44	27	20	23	11	19	22	18	27
45	25	28	27	9	25	24	22	29
46	28	19	22	22	24	27	27	26
47	26	27	28	20	28	31	31	30
48	18	19	26	24	21	23	26	19
49	24	25	28	28	19	27	24	25
50	17	18	27	27	22	28	29	18
51	23	24	31	31	22	30	27	22
52	24	27	25	28	22	20	24	22
53	30	29	29	28	22	26	22	30
54	19	30	24	33	19	27	25	29
55	25	32	30	33	17	31	23	35
56	23	19	23	24	23	24	25	22
57	29	19	27	22	25	30	29	28
51		10			20	00	_ 20	20

x	0x11d	0×177	0x1f3	0×169	0x1bd	0x1e7	0x12b	0×1d7
58	22	28	22	25	20	31	24	19
59	28	28	28	23	24	35	28	27
60	33	23	20	26	28	19	27	31
61	39	27	26	28	32	27	31	35
62	32	28	17	25	29	28	24	28
63	38	32	25	27	31	34	28	34
64	19	17	16	24	20	20	$\frac{20}{22}$	22
65	15	15	22	$\frac{24}{30}$	20	18	22	18
		26	13	21	19	25	17	25
66	16							
67	12	24	21	27	17	25	19	23
68	17	23	23	24	25	13	26	23
69	13	25	27	26	23	9	28	21
70	10	28	22	23	28	24	23	22
71	6	30	28	25	28	22	25	22
72	30	23	23	20	22	21	15	25
73	26	27	27	24	26	21	23	25
74	27	22	26	23	25	20	10	26
75	23	26	32	27	27	22	18	24
76	24	25	24	18	27	20	23	20
77	20	25	26	26	29	18	31	18
78	21	28	29	19	26	25	20	17
79	17	28	33	27	30	25	28	13
80	25	26	21	11	25	25	19	24
81	29	24	25	13	27	25	21	26
82	20	29	16	16	32	30	26	17
83	24	27	22	18	32	28	28	21
84	19	24	22	21	20	22	25	21
85	23	26	24	27	20	24	20	21
86	14	20	23	24	20	29	30	21 22
87	14	29	25	30	29	29	32	22
88	28	29	26	11	19	$\frac{23}{24}$	18	24 27
	32					$\frac{24}{22}$		21
89		20	28	19 6	25		26	
90	23	29	27	6	26	23	21	30
91	27	29	31	14	30	19	29	30
92	26	18	21	23	26	27	28	26
93	30	14	21	27	30	27	36	30
94	25	27	28	20	25	28	29	29
95	29	23	30	24	31	26	37	31
96	19	21	27	27	24	24	25	21
97	25	21	29	25	24	24	29	21
98	26	26	30	28	19	29	24	28
99	32	26	30	26	21	27	28	30
100	17	17	28	27	23	31	31	20
101	23	21	32	29	25	29	35	22
102	24	26	33	30	22	30	28	19
103	30	30	35	32	22	26	32	23
104	26	29	24	29	24	19	26	22
105	32	31	28	33	20	21	24	26
106	29	28	29	28	23	26	21	31
107	35	30	31	32	21	26	19	33
108	16	29	23	35	19	28	28	27
109	22	27	29	35	17	28	26	29
110	23	32	30	32	14	29	21	36
111	29	30	34	32	10	27	19	36
112	23	22	26	24	25	25	26	23
$112 \\ 113$	20	18	$\frac{20}{30}$	24	25	23	$\frac{20}{30}$	23
114	32	17	$\frac{30}{27}$	20	24	$\frac{23}{30}$	29	21 28
$114 \\ 115$	$\frac{32}{30}$	13	21	23	$\frac{24}{28}$	$\frac{30}{30}$	$\frac{29}{33}$	28
$115 \\ 116$	21	26	29 21	$\frac{25}{26}$	28 18	$\frac{50}{32}$	33 22	28 18
			$\frac{21}{27}$		$\frac{18}{22}$			
117	19	26	21	24	22	32	26	14

x	0x11d	0×177	0x1f3	0×169	0x1bd	0x1e7	0x12b	0x1d7
118	26	29	28	21	25	35	27	27
119	24	29	32	19	27	37	31	25
120	34	22	21	26	29	18	29	32
121	32	24	27	26	27	14	27	30
122	39	27	24	29	28	25	32	35
123	37	29	28	29	28	23	30	31
124	32	26	14	22	30	27	21	25
125	30	24	22	26	30	25	19	25
126	37	31	23	27	29	32	26	32
127	35	29	29	31	27	32	24	30
128	21	23	20	29	22	23	26	26
129	29	25	16	29	28	25	20	26
130	16	14	23	32	25	20	27	19
131	24	16	21	32	29	24	21	17
132	17	27	13	23	21	28	20	27
133	25	25	7	27	25	28	14	29
133 134	12	26	22	28	16	20	19	$\frac{23}{24}$
$134 \\ 135$	20	20	18	32	22	29	13	24
135	20	24	27	25	22	16	27	$\frac{24}{27}$
$130 \\ 137$	$\frac{20}{28}$	23	27	$\frac{23}{27}$	$\frac{20}{30}$	20	27	27
137	$\frac{20}{15}$	21 26	$\frac{23}{28}$	26	$\frac{30}{23}$	20	21	27
$130 \\ 139$	$\frac{15}{23}$	20	28 28	$\frac{20}{28}$	23	13	$\frac{20}{28}$	22
139	12	22	20	$\frac{20}{25}$	23 31	27	$\frac{20}{25}$	$\frac{24}{22}$
140	$\frac{12}{20}$	29	18	23	31	21	$\frac{25}{25}$	22
$141 \\ 142$	$\frac{20}{3}$	$\frac{29}{30}$	29	23	26	$\frac{29}{20}$	$\frac{23}{24}$	20
$142 \\ 143$	- - 3 - 11	30	$\frac{29}{27}$	24	$\frac{20}{28}$	20	$\frac{24}{24}$	$\frac{21}{21}$
143	35	26	27	22	23	24	17	$\frac{21}{28}$
$144 \\ 145$	35	20	23	22	$\frac{23}{27}$	22	11	$\frac{20}{34}$
145	$\frac{35}{28}$	24 27	23 26	20	26	20	26	$\frac{34}{27}$
140	$\frac{28}{28}$	27	20	29	20	$\frac{23}{25}$	20	31
147	$\frac{28}{27}$	23	20	29	$\frac{20}{28}$	23 19	<u>20</u> 9	$\frac{31}{25}$
$140 \\ 149$	$\frac{27}{27}$	16	20 24	20	$\frac{20}{30}$	25	3	$\frac{23}{29}$
$149 \\ 150$	21	27	31	20	27	25	$\frac{3}{20}$	$\frac{29}{24}$
	$\frac{24}{24}$	21	29	$\frac{27}{27}$	31			$\frac{24}{26}$
$\frac{151}{152}$			29 26	18		30	14	
	26	28			29	21	24	21
153	26	$\frac{24}{23}$	$\frac{26}{25}$	16	29	23	$\frac{24}{29}$	23
154	19			27	28	16		18
155	19	19	27	25	26	16	29	22
$\frac{156}{157}$	$\begin{array}{c} 22 \\ 22 \end{array}$	$\frac{28}{28}$	$\frac{31}{29}$	16 18	$\frac{26}{24}$	$\frac{24}{28}$	$\begin{array}{c} 20 \\ 20 \end{array}$	$\frac{20}{24}$
$\frac{157}{158}$	$\frac{22}{15}$	$\frac{28}{27}$	$\frac{29}{32}$	$\frac{18}{27}$	$\frac{24}{29}$	$\frac{28}{25}$	$\frac{20}{27}$	$\frac{24}{9}$
			$\frac{32}{32}$	27	29 29	$\frac{25}{27}$	27	
$\frac{159}{160}$	$\frac{15}{29}$	$\frac{27}{29}$	$\frac{32}{23}$	$\frac{29}{12}$	$\frac{29}{26}$	$\frac{27}{27}$	19	$\frac{15}{25}$
160	$\frac{29}{27}$		$\frac{23}{23}$		$\frac{20}{24}$	31	$\frac{19}{23}$	
$\frac{161}{162}$		29	$\frac{23}{28}$	20	24 29	$\frac{31}{28}$	$\frac{23}{20}$	$\frac{29}{30}$
	$\frac{30}{28}$	24		11	29 29		$\frac{20}{24}$	$\frac{30}{32}$
163	$\begin{array}{c} 28 \\ 21 \end{array}$	$\frac{24}{31}$	$\frac{26}{14}$	19 18	29 31	30 30	$\frac{24}{27}$	$\frac{32}{16}$
164								
165	19	27	16	22	31	32	31	22
166	26	26	25	19	34	29	30	21
167	24	22	25	23	32	29	34	25
168	20	25	24	22	22	22	28	24
169	18	23	22	24	24	28	26	28
170	25	26	23	27	17	25	25	19
171	23	24	19	29	21	29	23	25
172	12	27	21	26	27	27	32	21
173	10	29	21	32	31	31	30	23
174	17	28	26	29	30	28	31	24
175	15	30	24	35	32	30	29	28
176	29	20	26	11	19	26	18	27
177	35	20	24	15	15	28	22	29

x	0x11d	0×177	0x1f3	0x169	0x1bd	0x1e7	0x12b	0x1d7
178	32	21	29	22	26	23	27	30
179	38	21	25	26	24	27	31	30
180	21	30	27	3	26	25	20	30
181	27	26	27	11	24	29	24	30
182	24	27	32	12	29	16	27	29
183	30	23	30	20	25	22	31	27
184	24	20	21	25	27	27	29	26
185	30	14	17	31	27	27	27	24
186	31	11	18	26	30	26	38	31
187	37	5	12	32	32	28	36	31
188	24	26	28	19	22	28	27	27
189	30	24	26	21	24	30	25	27
190	27	21	27	22	29	21	34	28
191	33	19	23	24	29	25	32	30
192	20	26	28	29	28	25	24	22
193	16	26	26	27	30	25	24	22
194	25	21	31	26	25	26	29	21
195	20	21	27	24	25	24	29	23
196	28	26	29	29	19	28	24	31
197	24	30	25	31	19	26	24	29
198	33	25	30	24	20	20	31	30
199	29	29	$\frac{50}{24}$	24	20	23	31	30
200	17	16	29	20	24	34	33	21
200	13	22	29	31	30	32	27	21
201	26	21	34	30	25	29	34	22
202	22	27	32	34	29	25	28	20
203	25	28	32	29	23	31	29	16
201	20	30	30	29	27	27	23	18
206	30	29	35	34	20	24	32	25
207	26	31	31	34	26	18	26	25
208	28	31	23	30	23	18	27	22
209	32	31	23	32	27	12	27	16
210	31	32	28	31	20	23	24	27
210	35	32	26	33	22	19	24	23
212	28	27	30	28	24	25	21	31
213	32	31	28	26	26	21	21	27
214	35	28	29	31	21	24	16	32
215	39	32	25	29	25	22	16	30
216	13	29	22	36	19	29	30	25
217	17	31	24	36	27	25	24	23
218	20	24	29	35	16	28	27	28
219	24	26	29	35	22	26	21	24
220	21	33	31	32	16	30	20	36
221	25	31	31	36	22	28	14	32
222	28	28	32	29	5	23	15	35
223	32	26	30	33	13	23	9	29
224	24	24	29	24	26	25	27	25
225	30	26	27	30	28	23	25	21
226	19	19	30	29	27	22	32	20
227	25	21	30	35	31	22	30	18
228	32	18	28	20	23	30	29	28
229	38	24	28	22	27	26	27	22
230	31	9	27	23	28	29	32	27
231	37	15	29	25	30	27	30	23
232	21	24	20	28	16	32	20	18
233	27	28	16	32	14	28	20	14
234	16	25	27	23	21	31	25	11
234 235	22	29	25	23	21 21	29	29	5
$\frac{235}{236}$	25	$\frac{29}{30}$	$\frac{25}{25}$	18	21 25	$\frac{29}{35}$	$\frac{29}{26}$	27
230 237	31	30	23	26	$\frac{25}{25}$	$\frac{35}{29}$	$\frac{20}{30}$	25
201	01	00	_ 20		20		00	40

x	0x11d	0×177	0x1f3	0×169	0x1bd	0x1e7	0x12b	0×1d7
238	20	27	30	15	26	36	29	24
239	26	27	30	23	24	32	33	20
240	34	21	22	27	29	18	30	33
241	32	19	18	29	33	14	28	31
242	31	26	25	24	26	11	27	26
243	29	24	23	26	32	5	25	26
244	38	27	27	29	24	23	30	36
245	36	29	25	35	30	21	28	36
246	35	24	24	28	29	22	29	29
247	33	26	24	34	33	18	27	31
248	31	25	11	19	31	27	21	22
249	29	29	5	27	31	25	25	24
250	28	20	20	22	28	22	14	25
251	26	24	16	30	30	18	18	25
252	39	31	22	27	26	30	25	31
253	37	31	18	31	28	30	29	31
254	32	26	25	28	23	31	20	26
255	30	26	23	32	23	29	24	24
mean	24.03	24.03	24.03	24.03	24.03	24.03	24.03	24.03
σ	7.3958	5.3004	5.7544	6.7574	5.2528	5.8838	6.1752	5.7979

Table 13:	XOR	count	for	GF((2^8)) ((Part II))
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x	0×165	0x18b	0x163	0x11b	0x13f	0×15f	0x1c3	0×139
0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0
2	3	3	3	3	5	5	3	3
3	11	9	11	11	13	13	9	11
4	6	7	6	6	10	10	5	6
5	10	13	10	14	18	14	11	14
6	15	10	13	13	13	15	10	13
7	19	18	17	21	21	19	14	21
8	10	10	12	9	15	12	7	9
9	10	16	14	17	17	20	11	11
10	11	15	11	18	20	17	12	18
11	13	23	13	26	20	25	18	20
12	18	11	14	15	13	18	14	15
13	24	19	20	23	15	22	20	17
14	21	20	17	20	24	19	13	24
15	27	26	23	28	26	23	21	26
16	13	14	17	12	19	14	12	13
10	17	20	21	12	19	14	12	13
18	14	17	18	12	20	21	11	13
19	14	25	22	19	20	21	19	12
$\frac{19}{20}$	13	17	15	22	20	$\frac{20}{20}$	19 13	12
20	21	25	23	22	23	20	17	19
$\frac{21}{22}$	12	23	12	22	23	20	18	22
22	20	34	20	29	22	27	24	22
$\frac{23}{24}$	20	12	15	17	16	20	17	16
$\frac{24}{25}$	21 27	20	21	17	22	20	23	22
$\frac{20}{26}$	24	20	21	26	13	23	23	17
$\frac{20}{27}$	30	21 27	26	26	19	$\frac{23}{27}$	26	23
28	23	21	17	19	24	21	12	26
20	25	21 27	19	19	30	22	20	32
$\frac{20}{30}$	28	26	26	28	27	21	23	27
31	30	34	28	28	33	21	29	33
32	17	17	19	16	22	16	16	18
33	23	17	25	14	20	14	22	16
34	18	22	20	13	20	21	21	15
35	24	20	30	10	19	19	25	13
36	15	18	23	22	24	22	11	14
37	17	16	25	20	22	24	19	12
38	18	27	$\frac{23}{24}$	19	22	24	22	11
39	20	27	24	17	19	29	22	9
40	15	19	19	25	25	$\frac{23}{24}$	17	23
40	19	17	23	23	29	24	23	23
42	26	26	20	20	23	21	16	18
43	30	26	24	24	24	19	24	22
44	11	30	11	31	23	26	18	23
45	19	30	19	29	20	28	22	20
46	20	37	20	30	20	20	23	21
47	28	35	28	28	20	29	29	26
48	20	13	16	20	19	20	20	17
49	26	11	18	26	25	24	20	23
50	20	22	23	17	20	25	25	20
51	29	22	25	23	28	20	31	28
52	23	22	20	26	11	24	27	17
53	30	22	26	32	17	30	33	23
54	29	27	23	27	20	27	26	23
55	35	25	29	33	26	33	34	22
56	24	19	18	21	20	24	11	$\frac{28}{28}$
57	32	19	26	21 27	24	$\frac{24}{26}$	19	28
อก	32	19	20	41	24	_ 20	19	20

x	0×165	0x18b	0×163	0x11b	0x13f	0x15f	0x1c3	0×139
58	25	26	17	16	31	23	22	35
59	33	24	25	22	31	25	28	35
60	30	26	26	27	28	22	24	28
61	34	24	30	33	28	28	30	28
62	29	33	29	26	33	17	29	31
63	33	33	33	32	33	23	33	31
64	21	20	21	19	26	19	20	22
65	19	18	27	27	28	23	24	24
66	26	19	26	16	21	14	23	19
67	20	19	32	24	23	18	29	21
68	19	25	27	17	20	25	25	20
69	21	25	29	25	24	25	31	20
70	21 26	20	32	10	19	20	26	13
71	28	18	34	18	21	20	34	15
72	17	18	27	22	$\frac{21}{25}$	20	11	17
73	17	18	31	30	$\frac{23}{21}$	$\frac{23}{27}$	19	25
74	16	15	28	$\frac{30}{21}$	$\frac{21}{24}$	26	20	12
74 75	16	13	32	21 29	$\frac{24}{20}$	$\frac{20}{30}$	20	20
75 76	21	29	$\frac{52}{25}$	$\frac{29}{20}$	$\frac{20}{25}$	$\frac{30}{29}$	$\frac{20}{22}$	20
			$\frac{25}{33}$					11 19
77	17	27		28	21	29	$\frac{28}{29}$	19 6
78	18	26	26	15	18	28		
79	14	26	34	23	14	28	33	14
80	16	22	22	27	27	23	18	27
81	18	22	24	27	25	23	24	33
82	19	17	25	24	30	24	25	28
83	21	15	27	24	28	24	29	34
84	28	27	24	25	25	25	15	17
85	26	25	30	25	23	21	23	23
86	33	26	27	22	30	18	24	22
87	31	26	33	22	28	14	30	28
88	10	32	10	34	24	25	19	24
89	6	30	18	34	28	25	25	24
90	19	29	17	29	27	30	20	27
91	15	29	25	29	31	30	28	27
92	20	39	20	32	18	27	22	22
93	20	39	24	32	22	23	26	22
94	27	36	27	27	23	28	25	25
95	27	34	31	27	27	24	31	25
96	28	17	18	23	22	25	24	18
97	24	23	26	21	22	23	30	22
98	27	10	17	30	27	24	25	25
99	23	18	25	28	27	22	33	29
100	30	24	26	17	22	27	27	22
101	30	32	30	15	22	29	31	26
102	27	25	25	24	29	26	30	29
103	27	31	29	22	29	28	36	33
104	24	23	20	26	9	25	29	17
105	26	31	22	24	15	23	35	15
106	29	22	27	31	18	32	36	22
107	31	28	29	29	24	30	40	20
108	28	28	20	28	21	27	26	21
109	26	34	26	26	27	29	34	19
110	35	23	27	33	24	30	35	30
111	33	31	33	31	30	32	41	28
111	25	17	19	23	$\frac{50}{25}$	25	10	20
$112 \\ 113$	$\frac{25}{25}$	25	$\frac{19}{23}$	29	$\frac{23}{29}$	$\frac{23}{27}$	10	29
113	$\frac{23}{34}$	18	$\frac{23}{28}$	29	$\frac{29}{22}$	26	10	$\frac{29}{28}$
114	34	24	32	32	22	$\frac{20}{28}$	25	$\frac{20}{28}$
115	$\frac{34}{25}$	24	32 15	13	31	23	23	$\frac{20}{37}$
						$\frac{23}{29}$	$\frac{21}{27}$	
117	21	34	23	19	35	- 29	21	37

x	0×165	0x18b	0×163	0x11b	0x13f	0x15f	0×1c3	0×139
118	32	21	24	20	30	24	28	36
119	28	29	32	26	34	30	32	36
120	31	27	27	26	30	23	25	26
121	29	33	33	32	28	25	29	32
122	34	22	28	31	27	28	28	27
123	32	30	34	37	25	30	34	33
124	25	32	31	24	32	13	30	30
125	27	40	33	30	30	19	36	36
126	30	31	32	33	31	22	31	27
127	32	37	34	39	29	28	39	33
128	24	26	23	22	30	22	25	25
129	26	28	23	16	26	22	21	31
130	21	19	28	27	29	25	26	26
131	23	23	28	21	25	25	20	32
132	26	21	27	18	22	14	24	21
133	32	25	23	12	18	18	22	27
134	25	22	32	27	27	21	31	22
135	31	24	28	21	23	25	27	28
136	20	28	29	21	23	28	30	22
137	28	32	23	15	25	28	26	22
138	23	27	30	28	26	27	33	25
139	31	29	24	22	28	27	31	25
140	28	21	33	9	19	20	27	14
141	32	23	31	3	21	24	21	14
142	29	16	34	20	20	19	36	13
143	33	20	32	14	22	23	32	13
144	19	18	30	22	29	24	11	20
145	25	22	26	24	29	20	5	22
146	18	19	33	31	20	29	20	29
147	24	21	29	33	20	25	16	31
148	15	17	30	22	23	28	22	12
149	17	19	30	24	23	28	18	14
150	16	10	33	31	20	29	25	21
151	18	14	33	33	20	29	23	23
152	25	32	26	21	24	32	22	11
153	29	34	24	23	18	28	20	19
154	18	27	33	28	23	29	29	18
155	22	31	31	30	17	25	25	26
156	19	25	26	13	22	28	31	3
157	27	29	20	15	16	28	27	11
158	10	24	33	20	11	25	32	14
159	18	26	27	22	5	25	26	22
160	17	25	26	28	28	22	19	31
161	21	19	24	32	30	20	17	31
162	20	24	25	27	25	25	26	34
163	24	20	23	31	27	23	22	34
164	19	18	28	24	32	26	28	29
165	27	14	22	28	34	20	24	29
166	20	13	27	23	27	25	29	36
167	28	7	21	27	29	19	23	36
168	29	27	24	27	29	24	14	16
169	35	23	20	31	25	22	8	22
170	26	24	31	24	22	23	23	21
171	32	18	27	28	18	21	19	27
172	33	26	26	23	29	20	25	22
173	35	20	26	27	25	14	21	28
174	32	23	33	20	28	11	28	27
175	34	19	33	20	20	5	26	33
176	10	35	9	36	25	24	20	24
177	18	31	3	32	23	26	17	28
	1.0	01						

x	0×165	0x18b	0×163	0x11b	0x13f	0x15f	0x1c3	0×139
178	3	30	18	35	30	25	24	23
179	11	24	12	31	28	27	22	27
180	22	28	15	32	27	32	18	30
181	26	22	13	28	25	30	12	34
182	13	27	24	27	30	29	27	25
183	17	23	22	23	28	27	23	29
184	20	41	21	31	16	28	22	23
185	22	35	21	27	20	30	18	21
186	19	38	22	32	21	21	23	20
187	21	34	22	28	25	23	17	18
188	26	36	27	27	22	28	21	25
189	32	32	23	23	26	26	19	23
190	27	33	28	24	25	21	28	22
191	33	27	24	20	29	19	24	20
192	31	20	20	27	26	27	27	19
193	31	16	18	21	24	27	23	19
194	26	25	27	22	23	24	32	22
195	26	19	25	16	20	24	30	22
196	20	9	16	31	21	27	26	27
197	25	3	10	25	26	31	20	27
198	20	18	27	$\frac{20}{30}$	20	20	33	30
199	18	14	21	24	25	20	29	30
$\frac{133}{200}$	33	26	21	24	23	29	23	$\frac{30}{22}$
200	31	20	20	14	20	29	20	22
201	30	33	31	13	20	30	31	20
202	28	29	27	7	20	30	25	33
203 204	25	25	28	24	29	$\frac{50}{25}$	29	30
204	25	20	28	18	33	29	23	36
200 206	24	32	20	21	28	25	34	31
200 207	24	26	27	15	32	$\frac{20}{30}$	30	37
201	20	20	23	27	7	$\frac{30}{27}$	31	18
208	24	18	17	29	13	23	29	26
203 210	20	33	20	$\frac{23}{22}$	16	$\frac{23}{22}$	36	13
210	25	29	14	22	22	18	32	21
211 212	28	23	27	31	19	31	38	21 22
212	28	17	25	33	25	31	34	30
$\frac{213}{214}$	31	26	$\frac{23}{28}$	26	23	30	41	17
$\frac{214}{215}$	31	20	26	$\frac{20}{28}$	22	30	35	25
$\frac{210}{216}$	28	20	17	$\frac{20}{28}$	20	$\frac{30}{27}$	26	$\frac{23}{23}$
$\frac{210}{217}$	$\frac{20}{30}$	20	17	$\frac{20}{30}$	22	23	20	$\frac{23}{25}$
217	$\frac{30}{23}$	33	26	$\frac{30}{25}$	22	$\frac{23}{30}$	33	$\frac{23}{16}$
$\frac{218}{219}$	$\frac{25}{25}$	33 27	20	$\frac{25}{27}$	27	$\frac{50}{26}$	35 29	10
$\frac{219}{220}$	$\frac{23}{34}$	21	20	32	21	$\frac{20}{27}$	35	31
220	$34 \\ 32$	15	23	34	22	27	31	33
$\frac{221}{222}$	32 31	28	$\frac{21}{30}$	$\frac{54}{29}$	22	$\frac{27}{30}$	40	$\frac{33}{28}$
222	29	$\frac{28}{24}$	$\frac{50}{26}$	$\frac{29}{31}$	29	$\frac{30}{30}$	$\frac{40}{38}$	$\frac{28}{30}$
$\frac{223}{224}$	$\frac{29}{26}$	24 15	20	25	29	$\frac{50}{25}$	<u> </u>	$\frac{50}{31}$
$\frac{224}{225}$	$\frac{20}{28}$	15	21 21	$\frac{25}{29}$	$\frac{20}{30}$	$\frac{25}{27}$	9 3	$\frac{31}{37}$
$\begin{array}{c} 225 \\ 226 \end{array}$	$\frac{28}{23}$	26	$\frac{21}{22}$	$\frac{29}{30}$	$\frac{30}{29}$	$\frac{27}{26}$	3 16	28
$\frac{220}{227}$	$\frac{23}{25}$	$\frac{20}{28}$	$\frac{22}{22}$	$\frac{30}{34}$	29 33	$\frac{20}{28}$	10	$\frac{28}{34}$
227	$\frac{25}{36}$	28 18	22	$\frac{54}{25}$	20	$\frac{28}{29}$	12	$\frac{54}{25}$
228 229	$\frac{30}{34}$	$\frac{18}{20}$	27	$\frac{25}{29}$	$\frac{20}{24}$	$\frac{29}{27}$	18	$\frac{25}{31}$
$\frac{229}{230}$	$\frac{34}{35}$	20	$\frac{23}{32}$	$\frac{29}{30}$	$\frac{24}{25}$	$\frac{27}{26}$	$\frac{14}{23}$	$\frac{31}{26}$
	35 33			$\frac{30}{34}$				$\frac{26}{32}$
231 232	$\frac{33}{24}$	$\frac{25}{29}$	$\frac{28}{13}$		29 31	$\begin{array}{c} 24 \\ 23 \end{array}$	21 20	$\frac{32}{38}$
				10				
233	20	31	7	14	29	25	18	38
234	19	34	22	17	34	28	25	37
235	15	38	16	21	32	30	21	37
236	32	18	23	18	29	23	27	36
237	32	22	21	22	27	21	23	36

x	0x165	0x18b	0x163	0x11b	0x13f	0x15f	0x1c3	0×139
238	25	27	28	25	34	28	30	35
239	25	29	26	29	32	26	24	35
240	33	29	28	25	29	21	25	24
241	31	31	24	21	25	27	21	22
242	28	32	31	34	28	24	28	33
243	26	36	27	30	24	30	22	31
244	33	20	26	29	29	29	26	26
245	35	24	26	25	25	31	24	24
246	30	31	33	34	22	28	31	31
247	32	33	33	30	18	30	27	29
248	21	31	30	22	32	9	30	29
249	21	35	28	18	34	15	26	33
250	26	40	33	29	27	20	35	36
251	26	42	31	25	29	26	33	40
252	27	28	32	34	28	21	29	23
253	23	30	26	30	30	23	23	27
254	30	33	31	37	25	24	36	30
255	26	37	25	33	27	26	32	34
mean	24.03	24.03	24.03	24.03	24.03	24.03	24.03	24.03
σ	6.8679	7.3618	6.4144	6.7574	5.5773	5.0581	7.4634	7.5303