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# DISCRETE LOGARITHM COMPUTATIONS OVER FINITE FIELDS USING REED-SOLOMON CODES

#### D. AUGOT AND F. MORAIN

ABSTRACT. Cheng and Wan have related the decoding of Reed-Solomon codes to the computation of discrete logarithms over finite fields, with the aim of proving the hardness of their decoding. In this work, we experiment with solving the discrete logarithm over  $\mathbb{F}_{q^h}$  using Reed-Solomon decoding. For fixed h and q going to infinity, we introduce an algorithm (RSDL) needing  $\tilde{O}(h!\cdot q^2)$  operations over  $\mathbb{F}_q$ , operating on a  $q\times q$  matrix with (h+2)q nonzero coefficients. We give faster variants including an incremental version and another one that uses auxiliary finite fields that need not be subfields of  $\mathbb{F}_{q^h}$ ; this variant is very practical for moderate values of q and q. We include some numerical results of our first implementations.

#### 1. Introduction

The fastest known algorithms for computing discrete logarithms in a finite field  $\mathbb{F}_{p^n}$  all rely on variants of the number field sieve or the function field sieve. The former is used when n=1 (see [Gor93, Sch93, SWD96, Web96, JL03, Sch05, CS06]) or p is medium ([JLSV06] improving on [JL06]). The latter is used for fixed p and n going to infinity (see [Adl94, AH99, JL02, GHP+04] and [Cop84] for p=2 generalized in [Sem98]). Some related computations are concerned with computing discrete logarithms over tori [GV05a]. All complexities are  $L_{p^n}[c, 1/3]$  where as usual

$$L_x[c,\alpha] = \exp((c + o(1))(\log x)^{\alpha}(\log\log x)^{1-\alpha})$$

as x goes to infinity, c > 0 and  $0 \le \alpha < 1$  being constants.

Traditional index calculus methods over  $\mathbb{F}_{q^h} = \mathbb{F}_q[X]/(Q(X))$  (where Q has degree h) look for relations of the type

(1) 
$$X^{u} \mod Q(X) =: P(X) = \prod_{i=1}^{n} p_{i}(X)^{\alpha_{u,i}},$$

where u varies and the  $p_i$  belong to a factor base  $\mathcal{B}$  containing irreducible polynomials in  $\mathbb{F}_q$ . The polynomial P(X) generically has degree h-1, and we must find a way to factor it over  $\mathcal{B}$  using elementary division or sieving techniques. This collection phase yields a linear system over  $\mathbb{Z}/(q^h-1)\mathbb{Z}$  that has to be solved in order to find  $\log p_i$ . Very often, the system is sparse and suitable methods are known (structured elimination, block Lanczos [Mon95], block Wiedemann [Cop94]).

The second phase (search phase) requires finding a factorization of  $X^u f(X)$ , where we want the discrete logarithm of f(X).

Our aim in this work is to investigate the use of decoding Reed-Solomon codes instead of factorization of polynomials in the core of index calculus methods, following the approach of [CW07, CW04]. Superficially, the code-based algorithm (called

RSDL) replaces relations of the type (1) by

$$X^u \equiv f_A(X) := \prod_{a \in A} (X - a) \bmod Q(X),$$

where A is a subset of a fixed set  $S \subset \mathbb{F}_{q^h}$ . Such a relation exists if and only if  $X^u \mod Q(X)$  can be decoded. In case of successful decoding, the set A (or its complement) is recovered via factorization. If S has cardinality n,  $f_A(X)$  will be of degree n - h, which highlights one of the differences with a classical scheme.

It will turn out that taking  $S = \mathbb{F}_q$ , so that n = q, is often the sensible choice to do and therefore our method is interesting in the case q relatively small. Very much like in Gaudry's setting [Gau09], we will end up with a method of complexity  $\tilde{O}(h! \cdot q^2)$  operations over  $\mathbb{F}_q$ , for fixed h and q tending to infinity. The dependency on h can be dramatically lowered using a variant based on helper fields, auxiliary finite fields that need not be subfields of  $\mathbb{F}_{q^h}$ , making the variant very practical for moderate q and h.

The article starts with a review of the theory and practice of Reed-Solomon codes (Sections 2 and 3). Section 4 comes back to the computation of discrete logarithms. The analysis will be carried out in Section 5. In Section 6, we give an incremental version of our algorithm, which is faster in practice. Section 7 will be concerned with the use of helper fields and their Galois properties.

#### 2. Reed-Solomon codes

2.1. **Definition and properties.** Let  $\mathbb{F}$  be a field, and  $S = \{x_1, x_2, \dots, x_n\} \subset \mathbb{F}^n$  be fixed, with  $x_i \neq x_j$  for  $i \neq j$ . Define the evaluation map:

$$\operatorname{ev}_S: \ \mathbb{F}[X] \to \mathbb{F}^n$$

$$r(X) \mapsto (r(x_1), \dots, r(x_n)).$$

For a given  $1 \leq k \leq n$ , the Reed-Solomon code  $C_k$  over F, with support S and dimension k is

$$\{\operatorname{ev}_S(r(X)) | r(X) \in \mathbb{F}[X], \operatorname{deg} r(X) < k\} \subset \mathbb{F}^n,$$

and the set S is called the *support* of the code, see [Rot06] It is a *linear code* whose elements are called *codewords*. The (Hamming) distance between  $y, z \in \mathbb{F}$  is

$$d(y,z) = |\{i \in [1,n] | y_i \neq z_i\}|,$$

and r(X) is at distance  $\tau$  from  $y = (y_1, \ldots, y_n)$  if  $d(\text{ev}_S(r(X)), y) \leq \tau$ . The minimum distance of a general code is the smallest distance between two different codewords, and the minimum distance of  $C_k$  is known to be equal to d = n - k + 1.

2.2. The decoding problem. Given  $C_k$  as above, the decoding problem is: given  $y \in \mathbb{F}^n$ , and  $\tau \leq n$ , find the codewords  $c \in C_k$  within Hamming distance  $\tau$  of y. This problem and its complexity depend  $\tau$ . It is a NP-complete problem [GV05b] for general finite fields, n, k and  $\tau$ .

For Reed-Solomon codes, this amounts to finding, for any  $y \in \mathbb{F}^n$ , the set:

$$F_{\tau}(y) = \{ r(X) \in \mathbb{F}[X] | \deg f(X) < k, \ d(\text{ev}_S(r(X)), y) \le \tau \}.$$

A given algorithm is said to decode up to  $\tau$  if it finds  $F_{\tau}(y)$  for any y. If  $\tau > n - k$  tall solutions can be found by Lagrange interpolation, and there are  $\binom{n}{\tau}q^{k-n-\tau}$  of them. On the other hand, when  $\tau$  is small enough, we have:

**Proposition 1.** (Unique decoding) Let k be fixed and let  $\tau \leq \lfloor \frac{n-k}{2} \rfloor$ . Then, for any  $y \in \mathbb{F}^n$ , one has  $|F_{\tau}(y)| \leq 1$ .

The decoding problem is a list decoding problem when  $\lfloor \frac{n-k}{2} \rfloor < \tau < n-k$ , and an a priori combinatorial problem is to determine how large is the size  $\ell$  of  $F_{\tau}(y)$ , in the worst case over y. Of interest is to find  $\tau = \tau(n,k)$  such that  $\ell = \ell(n,k)$  is small and  $\tau = \lfloor n - \sqrt{(k-1)n} \rfloor$  was achieved, in the breakthrough papers [Sud97, GS99]. In the present paper, we consider only unique decoding, since unique decoding algorithms are simpler and faster.

# 3. A fast algorithm for uniquely decoding Reed-Solomon codes

Among the many algorithms for decoding Reed-Solomon codes, we have focused our attention on a variant of the Euclidean algorithm of [SKHN75]. This version is due to Gao [Gao02].

Let  $y=(y_i)\in\mathbb{F}^n$  to be decoded,  $c=(c_i)\in C_k$  be at distance  $\tau$  from y, if it exists,  $e=y-c=(e_i)$  the error vector, and  $E=\{i|e_i\neq 0\}$ . The locator polynomial of e is  $v(X)=\prod_{i\in E}(X-x_i)$ , and the decoding problem often reduces to finding this polynomial. Given a decoding radius  $\tau$ , the correct behaviour of a decoding radius is to report failure, when the number of errors is larger than  $\tau$ . The following algorithm is correct for Reed-Solomon codes and  $\tau=\lfloor\frac{n-k}{2}\rfloor$  (unique decoding).

3.1. **Gao's algorithm.** For convenience, we reproduce Algorithm 1a in [Gao02]. We let  $(x_i)$  be the support of the code and  $(y_i)$  a received word. Remember that k = n - d + 1. In our case, we will have  $k \simeq n$  and therefore d small. We denote by PartialEEA $(s_0, s_1, D)$  the algorithm that performs the euclidean algorithm on  $(s_0, s_1)$  and stops when a remainder has degree < D. In other words, when this algorithm terminates, we have computed polynomials u and v such that

$$s_0(X)u(X) + s_1(X)v(X) = g(X)$$

where g is the first remainder that has degree < D. We note  $P(X) \div X^k$  for the quotient of P(X) by  $X^k$ .

## Algorithm 1a

INPUT:  $(x_i) \in \mathbb{F}^n$ ,  $(y_i) \in \mathbb{F}^n$ 

OUTPUT: the error locator polynomial in case of successful decoding; failure otherwise.

Step 0. (Compute G) Compute  $G(X) = \prod_{i=1}^{n} (X - x_i)$ .

Step 1. (Interpolation) Compute I(X) such that  $I(x_i) = y_i$  for all i.

Step 2. (Partial gcd) Perform PartialEEA with inputs  $s_0 = G \div X^k$  (of degree d-1),  $s_1 = I \div X^k$  (of degree d-1), D = (d-1)/2, at which time

$$u(X)s_0(X) + v(X)s_1(X) = g(X)$$

with  $\deg(q) < (d-1)/2$ .

Step 3. (Division) divide G(X) by v(X) to get  $G(X) = h_1(X)v(X) + r(X)$ . If  $r \equiv 0$ , return v(X), otherwise return failure.

The original algorithm adds another step for recovering the codeword in case of success, but we do not need it for our purposes. In our case, we will need to factor v(X) to get the error locations.

This algorithm has been analyzed in [CY08], where fast multiplication and gcd algorithms are considered (for the characteristic 2 case). We briefly summarize the results.

Let M(n) be the cost to perform a multiplication of two polynomials of degree n with coefficients in  $\mathbb{F}$ , counted in terms of operations in  $\mathbb{F}$ . Following the algorithms of [GG99], we find that Step 0 costs O(M(n)) and Step 1 costs  $O(M(n)\log n)$ . Step 2 requires computing  $G(X) \div X^k$  and  $I(X) \div X^k$ , which is just coefficient extraction. PartialEEA requires  $O(M(d)\log d)$  operations (note that precise constants are given in [CY08]). Step 3 requires a division of a polynomial of degree n by one of degree n which costs O(M(n)). The cost of computing the roots of v(X) will depend on the base field.

# 3.2. Improvements.

- 3.2.1. Computing G. We may compute the highest terms of  $G \div X^k$  in time O(M(n)) (with a small constant, since the last step in the product tree will be computing the highest terms).
- 3.2.2. Interpolation. The input to the PartialEEA is

$$s_1(X) = I(X) \div X^k = \sum_{i=1}^n \frac{y_i}{G'(x_i)} (I_i(X) \div X^k) = \sum_{i=1}^n y_i H_i(X).$$

Note that the  $H_i(X)$  are polynomials of degree  $\leq d-2$ . We can compute  $I_i(X) \div X^k$  by appropriately modifying the last step of the algorithm using product trees, so as to compute only the higher order terms of  $I_i(X)$ . This will not modify the complexity, but will decrease the constant.

3.2.3. Reusing data. If the  $x_i$  are fixed (this will be our case), then G(X) can be precomputed (and  $s_0$  deduced from it), as well as  $G'(x_i)$ . The polynomials  $H_i(X)$  can also be precomputed. Instantiating the formula for  $s_1(X)$  will require O(nd) operations, which is interesting when d is much smaller than n.

# 3.3. The special case $S = \mathbb{F}_q$ .

3.3.1. First simplifications. We can write the cost of our modifications of Algorithm 1a as follows

$$T_G + T_{G \div X^k} + T_{I \div X^k} + T_{PEEA} + T_{v|G?},$$

where the notation  $T_X$  should be self-explanatory, the last one accounting for testing whether  $v \mid G$ . Since  $G(X) = X^q - X$ , we have  $T_G = O(1)$  and  $T_{G \div X^k} = O(1)$ .

Since S may be seen as an arithmetic progression, computing I or  $T_{I \div X^k}$  costs O(M(n)) using the techniques of [BS05]. We still have  $T_{PEEA} = O(M(d) \log d)$ .

3.3.2. Discarding v. Step 3 amounts to checking whether v(X) factors into linear factors. The ordinary algorithm requires division of G(X) by v(X) and in case of success, finding the roots of v(X).

When q is very small, we can find the roots of v(X) in  $\mathbb{F}_q$  via successive evaluation of v(a) for  $a \in \mathbb{F}_q$  in O(q) additions. This cost would therefore be neglectible.

For larger q, we can use the Cantor-Zassenhaus or Berlekamp algorithms, starting with the computation of  $X^q \mod v$  at a cost of  $O(M(d)\log q)$ . In that case, we can speed up the factoring process of v(X) when needed (storing  $X^{(q-1)/2}$  for future use when q is odd, etc.). The test  $v \mid G$  will cost  $O(M(d)\log q)$  for all relations, and

in case of success, will be followed by the total cost to find (d-1)/2 roots, that is to say  $O(dM(d) \log q)$  operations (assuming gcd to cost less than exponentiations).

Also, some product tree of the v's could be contemplated.

We can discard some polynomials v(X) by using Swan's theorem [Swa62], via computation of the discriminant of v(X), for a cost of  $O(M(d^2))$  operations.

3.3.3. Final cost. In summary, we find

$$\begin{split} T_G = O(1), \quad T_{G \div X^k} = O(1), \quad T_{I \div X^k} = O(M(q)), \\ T_{EEA} = O(M(d)\log d), \quad T_{X^q \bmod v} = O(M(d)\log q), \quad T_{roots} = O(dM(d)\log q). \end{split}$$

#### 4. Discrete logarithms

4.1. Connection with decoding Reed-Solomon codes. Consider  $\mathbb{F}_{q^h}$  realized as  $\mathbb{F}_q[X]/(Q(X))$ , and let S be any subset of  $\mathbb{F}_{q^h}$ , such that  $Q(a) \neq 0$  for any  $a \in S$ , and n = |S|. Let  $S_{\mu}$  the set of subsets of size  $\mu$  of S. For  $A \in S_{\mu}$ , define

$$f_A(X) = \prod_{a \in A} (X - a).$$

We extend [CW07] in a more general context: the field is not necessarily finite, and Q(X) is not irreducible. Indeed, [CW07] considered only finite fields, and  $S \subset \mathbb{F}_q$ .

**Theorem 2.** Consider F/K a field extension. Let be fixed a monic  $Q(X) \in K[X]$ , with  $\deg Q(X) = h$ , and  $S \subset F$  have size n, such that  $Q(a) \neq 0$  for all  $a \in S$ . Let  $1 \leq \mu \leq n$ . For any  $f(X) \in K[X]$ ,  $\deg f(X) < \mu$ , there exists  $A \in S_{\mu}$ , such that

(2) 
$$\prod_{a \in A} (X - a) \equiv f(X) \bmod Q(X)$$

if and only if the word

$$y = \operatorname{ev}_S\left(-f(X)/Q(X) - X^k\right)$$

is exactly at distance  $n - \mu$  from the Reed-Solomon code  $C_k$  of dimension  $k = \mu - h$  and support S. All the sets A such that (2) holds can be found by decoding y up to the radius  $n - \mu$ .

Proof. Let  $f(X) \in K[X]$  be given,  $\deg f(X) < \mu$ , and suppose that there exists  $A \in S_{\mu}$ , such that  $\prod_{a \in A} (x-a) \equiv f(x) \mod Q(x)$ . Then there exists  $t(X) \in F[X]$ ,  $\deg t(X) = \mu - h = k$ , such that  $\prod_{a \in A} (x-a) = f(x) + t(x)Q(x)$ . We remark that t(X) is monic, and we write  $t(X) = X^k + r(X)$ , with  $\deg r(X) < k$ . Then

$$f(X) + (X^k + r(X))Q(X) = \prod_{a \in A} (X - a),$$

which implies that  $r(a) = -f(a)/Q(a) - a^k$  for  $a \in A$ . Since  $|A| = \mu$ , the word  $\operatorname{ev}_S(-f(X)/Q(X) - X^k)$  is at distance  $n - \mu$  from  $\operatorname{ev}_S(r(X)) \in C_k$ .

Conversely, if  $\operatorname{ev}_S\left(-f(X)/Q(X) - X^k\right)$  is at distance exactly  $n - \mu$  from  $C_k$ , there exists  $A \in S_\mu$  and r(X) with  $\operatorname{deg} r(X) < k$ , such that  $r(a) = -f(a)/Q(a) - a^k$  for  $a \in A$ . Then

$$\prod_{a \in A} (X - a) \mid f(X) + (X^k + r(X))Q(X),$$

and the equality of the degrees imply the equality,

$$\prod_{a\in A}(X-a)=f(X)+(X^k+r(X))Q(X)$$

which is a relation of type (2).

Remarks. When  $\mu$  and k are such that  $n - \mu$  is half the minimum distance of  $C_k$ , the mapping

$$A \in S_{\mu} \mapsto \prod_{a \in A} (X - a) \bmod Q(X)$$

is one-to-one, since we have unique decoding. Furthermore, when  $S \subset \mathbb{F}_q$ , the number of relations of type (2) is  $\binom{n}{\mu}$ , and the probability of finding one is thus  $\binom{n}{\mu}/q^h$  when  $f(X) \in \mathbb{F}_q[X]$  is picked at random of degree less than h. When some elements of S lie in some extension of  $\mathbb{F}_q$ , the probability is more intricate because of the action of the Galois group, see Section 7.

4.2. The RSDL algorithm for computing discrete logarithms. The basic idea is to decompose polynomials using decoding of Reed-Solomon codes in the inner loop. For ease of presentation, we suppose that  $F = \mathbb{F}_{q^h}$ . In Section 7, we will present a more general setting.

INPUT: a)  $\mathbb{F}_{q^h} = \mathbb{F}_q[X]/(Q(X))$  where Q(X) is primitive of degree h over  $\mathbb{F}_q$ ;

 $\mathbb{F}_{a^h}^* = \langle \omega \rangle.$ 

b) Two parameters n and  $\mu$ , describing a Reed-Solomon code  $[n, k = \mu - h, d = n - k + 1]$ ; a subset S of  $\mathbb{F}_{q^h}$  of cardinality n.

OUTPUT: the logarithm  $\log_{\omega}(\omega - a)$  for all  $a \in S$ .

Step 1. (Randomize) Compute  $f(X) = X^u \mod Q(X)$  for a random u.

Step 2. (Decode) Find  $A \in S_{\mu}$  such that

$$f_A(X) \equiv f(X) \bmod Q(X)$$

using decoding. If this fails then pick another random u.

Step 3. (Recover support) given the error-locator polynomial v(X), compute  $f_A(X) = G(X)/v(X) = \prod_{a \in A} (X - a)$ ; from which we get the relation

$$u \equiv \sum_{a \in A} \log(\omega - a) \mod (q^h - 1).$$

If we have less than n relations, goto step 1.

Step 4. (Linear algebra) solve the  $n \times n$  linear system over  $\mathbb{Z}/(q^h-1)\mathbb{Z}$ , which yields the logarithms of  $\log(\omega-a)$ .

From  $f_A(X) = G(X)/v(X)$ , we can rewrite a relation as

$$X^u v(X) \equiv G(X) \bmod Q(X)$$
.

The corresponding row of the relation matrix will have as many non-zero coefficients as the degree of v, which will be shown to be small.

The search phase (finding individual logarithms) follows the same scheme.

4.3. Numerical example. Consider  $\mathbb{F}_{13^3} = \mathbb{F}_{13}[X]/(X^3 + 2X + 11)$ . We use  $(n, k, \mu) = (13, 7, 10)$ , which gives d = 7. The support is  $S = \{0, 1, \dots, 12\}$ . The probability of decomposition is  $\approx 0.1302$ . We find for instance that

$$X^{15} \equiv X^2 + 9X + 1 \mod (Q(X), 13).$$

We have to decode the word:

$$y = \text{ev}_S(-X^{15}/Q(X) - X^7) = (7, 1, 1, 0, 1, 3, 6, 8, 9, 12, 4, 11, 10).$$

The PartialEEA procedure yields

$$u(X) = X^2 + 5X + 3$$
,  $v(X) = 5X^3 + 2X^2 + 3$ ,  $q(X) = 7X + 6$ ,

And the polynomial v factors as (X-3)(X-8)(X-12), so that

$$X^{15}(X-3)(X-8)(X-12) \equiv G(X) \bmod (Q(X), 13).$$

Write  $13^3 - 1 = 2^2 \cdot 3^2 \cdot 61$ . Logarithms modulo  $2^2$  and  $3^2$  are easy to compute. The matrix M modulo 61 is given in 1. Its kernel is generated by

$$V = \begin{pmatrix} 1 & 3 & 52 & 24 & 57 & 9 & 41 & 54 & 42 & 27 & 41 & 35 & 5 & 36 \end{pmatrix}^{t}$$
.

Computing the logarithm of  $X^2 + 1$  is done using the relation

$$(X^2 + 1)X \equiv G(X)/((X(X - 2)(X - 8))) \mod Q(X)$$

and therefore

$$\log(X^2 + 1) = 417,$$

using the Chinese remaindering theorem. (Note that this is a toy example, the logarithm of  $X^2 + 1$  could have been computed in different ways, factoring it over the factor base directly for instance.)

FIGURE 1. Matrix modulo 61 for the example.

4.4. **Algorithmic remarks.** The inner loop of the algorithm is the computation of

$$y = \operatorname{ev}\left(-\frac{f(X)}{Q(X)} - X^k\right) \in \mathbb{F}_q^n,$$

followed by the interpolation of y on the support, to get I(X). We can greatly simplify the work by noting that

**Lemma 3.** Let  $\tilde{Q}(X)$  the inverse of -Q(X) modulo G(X). Then

$$I(X) = (f(X)\tilde{Q}(X) \bmod G(X)) - X^k.$$

Since  $\tilde{Q}(X)$  is computed only once, the cost of evaluating I(X) is just O(M(n)). From a practical point of view, this is multiplication by a fixed polynomial modulo a fixed polynomial, a very well known operation that is very common in computer algebra packages (in particular NTL).

Moreover, this result shows that we do not need the explicit points of the support, but rather their minimal polynomial(s). This will be the key to the incremental version of Section 6.

#### 5. Selecting optimal parameters

5.1. **Unique decoding.** Given q and h, we aim to build an optimal  $[n, k, n-k+1]_q$  Reed-Solomon code for finding relations (2). While Theorem 2 was used in [CW07] in a negative way for proving hardness of decoding up to a certain radius, we consider it in a positive way for solving discrete logarithm problem using unique decoding. We will consider list decoding in a subsequent work.

**Proposition 4.** In the context of Theorem 2, to be able to use a unique decoding algorithm of the code  $C_k$ , the parameters should be chosen as follows:  $\tau = h$ ,  $\mu = n - h$ , and k = n - h.

*Proof.* For Reed-Solomon codes, unique decoding holds for  $\tau = \lfloor \frac{n-k}{2} \rfloor$ . From  $k = \mu - h = n - \tau - h$ , it follows that  $\tau = h$ .

It should be noted that  $\mu$  and  $\tau$  play a symmetrical role.

# 5.2. Analyses.

5.2.1. Set up. For any integer s > 0, we assume that any elementary operation over  $\mathbb{F}_{q^s}$  takes  $O(M(\log q^s)) = O(M(s))$  operations over  $\mathbb{F}_q$ . In the same vein, an operation over  $\mathbb{Z}/(q^s-1)\mathbb{Z}$  takes M(s) operations over  $\mathbb{F}_q$ . Given that  $\tau = h$  and d = 2h + 1, we will write our complexities in terms of h (which is the degree of the error-locator polynomial v(X)).

The typical analysis involves the probability  $\varpi$  to get a relation (here getting a decoded word). Since we need n relations, each relation is found after  $1/\varpi$  attempts and c operations, leading to  $O(n\frac{1}{\varpi}c)$ . Using the decoding approach of Section 3, we see that a more precise count is

$$T_G + T_{G \div X^k} + n \frac{1}{\varpi} (T_{I \div X^k} + T_{EEA} + T_{v|G?}) + n T_{roots},$$

where we account for reusing G and  $G \div X^k$  and perform root searching of v only in case of success.

The cost of solving a  $n \times n$  linear system with h non-zero coefficients per row is  $O(h \cdot n^2)$  operations over  $\mathbb{Z}/(q^h - 1)\mathbb{Z}$ , yielding  $O(h \cdot n^2 \cdot M(h))$  operations over coefficients of size  $\log q$ .

We will be fixing h and letting q go to infinity.

5.2.2. The ordinary case. In case S is ordinary, that is  $S \subset \mathbb{F}_{q^h}$ , all polynomial operations are to be understood in  $\mathbb{F}_{q^h}$ . We inject the complexities of Section 3. We have  $T_G = O(M(n))$ . The additional cost will be

$$O\left(\left(n\frac{1}{\varpi}\left(M(n)+M(h)\log h+M(n)\right)+nhM(h)\log q\right)M(h)\right),$$

so that the total cost is

$$O\left(\left(n\frac{1}{\varpi}\left(M(n)+M(h)\log h\right)+nhM(h)\log q+h\cdot n^2\right)M(h)\right).$$

5.2.3. The case  $S \subset \mathbb{F}_q$ . This implies that  $n \leq q$ . Moreover, With  $\mathcal{Q} = q^h$ , we get

$$\varpi = \frac{\binom{n}{\mu}}{Q} = \frac{\binom{n}{n-\tau}}{Q} = \frac{\binom{n}{\tau}}{Q} = \frac{\binom{n}{h}}{Q} \approx \frac{n^h}{h! \cdot Q},$$

since h is fixed.

Using the fact that most of the operations are performed in  $\mathbb{F}_q$ , instead of  $\mathbb{F}_{q^h}$ , we obtain

$$O\left(n\frac{1}{\varpi}\left(M(n) + M(h)\log h\right) + nhM(h)\log q\right) + O(h \cdot n^2M(h)).$$

If  $n > \log q$  and n > h, this simplifies to

$$O\left(h!(q/n)^h nM(n)\right) + O(h \cdot n^2 M(h)),$$

and the first term always dominates. In order to have something not too slow, we are driven to taking n = q, for a cost of

$$O(h! \cdot qM(q)) + O(hM(h) \cdot q^2) = O(h! \cdot qM(q)) = \tilde{O}(q^2).$$

Note that both costs are asymptotically  $\tilde{O}(q^2)$ , but with different constants. We cannot balance these two phases easily, since h and q are given. The only thing we can do is relax the condition  $n \leq q$  using Galois properties (see Section 7).

We call RSDL-FQ the corresponding discrete logarithm algorithm with  $S = \mathbb{F}_q$ . One of the advantages of this algorithm is to operate on  $q \times q$  matrices with 2q + hq non-zero coefficients, so that a typical structured Gaussian elimination process will be very efficient.

**Proposition 5.** For fixed h and q tending to infinity, the algorithm RSDL-FQ has running time  $O(h! \cdot qM(q))$  and requires storing O(q) elements of size  $h \log q$ .

As a corollary, we see that the interpolation step dominates. This motivates the following Section, where this cost is decreased.

5.2.4. Looking for a subexponential behavior. It is customary to search for areas in the plane  $(\log q, h)$  yielding a subexponential behavior for the cost function. The analysis of the previous section works also in case  $h \ll n$ . The cost being  $\tilde{O}(h! \cdot q^2)$ , we look for  $0 \le \alpha < 1$  such that

$$2\log q + h\log h \simeq c(\log Q)^{\alpha}(\log\log Q)^{1-\alpha}$$
.

Making the hypothesis that  $h \ll \log q$  implies

$$2\frac{\log \mathcal{Q}}{h} \simeq c(\log \mathcal{Q})^{\alpha} (\log \log \mathcal{Q})^{1-\alpha},$$

or

$$h = \left(\frac{2\log \mathcal{Q}}{c\log\log \mathcal{Q}}\right)^{1-\alpha}.$$

In turn.

$$h \simeq \left(\frac{2h\log q}{c\log\log q}\right)^{1-\alpha}, \text{i.e., } h \simeq \left(\frac{2\log q}{c\log\log q}\right)^{1/\alpha-1}.$$

In order to respect the hypothesis  $h \ll \log q$ , we need  $\alpha \ge 1/2$ , and 1/2 is possible.

### 6. The incremental version of the algorithm

The idea of this variant is to use  $f(X) = X^u$  for increasing values of u, so that we can compute the interpolating polynomial for u + 1 from that of u, noting that I(X) is the real input to Algorithm 1a. We first explain how to do this, and then conclude with the incremental version of our algorithm. We cannot prove that using these polynomials lead to the same theoretical analysis, but it seems to work well in practice. Note that the search phase can benefit from the same idea.

The following result will help us interpolating very rapidly, and is a rewriting of Lemma 3.

**Proposition 6.** For u an integer, put  $f_u(X) = X^u f_0(X) \equiv c_{h-1} X^{h-1} + \cdots + c_{$  $c_0 \mod Q(X)$  and  $I_u$  the interpolation polynomial that satisfies  $I_u(x_i) = y_i$  for all i. Then

$$I_{u+1} \equiv XI(X) + X^{k+1} - X^k + c_{h-1} \mod G(X).$$

For the convenience of the reader, we give a description of the incremental operations performed in the relation collection phase. We claim that we no longer need  $y_i$ , past the initial evaluation.

**procedure** StartDecodingAt $(f_0, (x_i))$ 

- 0. Precompute  $G(X) = \prod_{i=1}^{n} (X x_i)$ ;  $\tilde{Q}(X) \equiv -1/Q(X) \mod G(X)$ ;  $f = f_0$ ; 1. [first interpolation for u = 0:]  $I := \tilde{Q}f \mod G(X) X^k$ ;
- 2. for u := 1 to  $q^h 2$  do c = coefficient of degree h - 1 of f; $\{ \text{ update } I \}$  $I = (XI + X^{k+1} - X^k + c) \mod G;$ { update f to  $X^{u+1} \mod Q(X)$  }

 $f = Xf \mod Q(X);$ if y can be decoded with error-locator polynomial v(X) then compute  $v(X) = \prod_{i=1}^h (X - e_i)$ , set  $A = S - \{e_i\}$ ,

store  $(u, \{e_i\})$  corresponding to the relation

$$X^u \equiv f_A(X) \mod Q(X)$$
 or  $X^u v(X) \equiv G(X) \mod Q(X)$ .

Note that the storage is minimal, we need to store u and h elements of  $\mathbb{F}_q$  for each relation. The corresponding row in the matrix modulo  $P \mid q^h - 1$  will contain one integer modulo P with h values equal to 1.

The analysis of this very heuristic version is similar to that of the original version: we replace some O(M(n)) by O(n) in the updating step for I. We find the same cost. From a practical point of view, we gain a lot, since all operations are now linear in n = q. It is all the more efficient as  $G(X) = X^q - X$  and reduction modulo G costs O(1) operations.

# 7. Galois action

This section is devoted to the case  $S \not\subset \mathbb{F}_q$ , with the idea of increasing the probability of finding relations by using *helper fields*. It turns out that S and the relations must be Galois stable. This is not exactly the same effect as obtained in the NFS/FFS case (see for instance [JL06]), but it results in smaller matrices.

7.1. Galois orbits. We state the property in full generality, for a general field K.

**Theorem 7.** Let F/K be a Galois extension, and  $Q(X) \in K[X]$  have degree h. Let  $\mu > h$  be an integer. Let  $f(X) \in K[X]$ ,  $\deg f(X) < \mu$ , such that there exists a unique  $A \in S_{\mu}$ , such that

$$f(X) \equiv \prod_{a \in A} (X - a) \bmod Q(X).$$

Then A is stable under Gal(F/K).

*Proof.* We have  $\prod_{a \in A} (X - a) = f(X) + t(X)Q(X)$ , for some  $t(X) \in F[X]$ . Then, for any  $\sigma \in \operatorname{Gal}(F/K)$ , we find:

$$\sigma\left(\prod_{\alpha\in A}(X-\alpha)\right) = f(X) + \sigma(t(X))Q(X),$$

where the action of  $\sigma$  is naturally extended to polynomials. Writing  $\sigma(t(X)) = u(X)$  for some  $u(X) \in F[X]$ , and since  $\sigma(f(X)) = f(X)$ , we get

$$\prod_{a \in A} (X - \sigma(a)) = f(X) + u(X)Q(X),$$

i.e.

$$\prod_{a \in A} (X - \sigma(a)) \equiv f(X) \bmod Q(X).$$

From the hypothesis of the unicity of A, we have  $\sigma(A) = A$ .

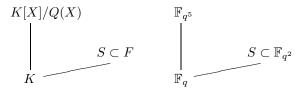
To use the decoding correspondence, we fix a set  $S \subset F$  such that relations of type (2) are sought for sets  $A \subset S$ . Then, we can enforce the uniqueness condition by fixing the parameters n = |S|, and  $\mu$  to have "unique decoding", i.e.  $\mu = n - h$ . From the previous Theorem, S must be a union of orbits under  $\operatorname{Gal}(F/K)$ . We collect these orbits by their size, i.e.

$$S = \bigcup_{i=1}^{e} S_i$$

where  $S_i$  is the union of the orbits of size i contained in S, and e is the maximal orbit size. Defining  $n_i = |S_i|$ , then  $n = \sum_{i=1}^e i n_i$ , and  $(n_1, \ldots, n_e)$  is a partition of n with restricted summands. Given e and n, we call the set of such partition set  $P_n^e$  for short, and its size is asymptotically [FS09]

$$|P_n^e| \sim \frac{1}{e!(e-1)!} n^{e-1}.$$

Before going further, let us mention that F/K does not need to be a subfield of K[X]/Q(X), and the following diagrams are perfectly valid for Theorem 7 to hold and for all the considerations in this Section.



**Proposition 8.** Let  $S = \bigcup_{i=1}^{e} S_i$ , with  $n_i = |S_i|$ ,  $n = \sum_{i=1}^{e} i n_i$ , and suppose that unique decoding holds for the parameters n and  $\mu$ . Then the number of relations (2) is

$$N_e(\mu) = \sum_{(\mu_1, \dots, \mu_e) \in P_u^e} \prod_{i=1}^e \binom{n_i}{\mu_i}.$$

*Proof.* Consider a partition  $(\mu_1, \dots \mu_e)$  of  $\mu$ ,  $\mu = \mu_1 + 2\mu_2 + \dots + e\mu_e$ , and for each i, pick  $\mu_i$  orbits of size i in S, and consider their union  $O_i$ . Then  $\prod_{i=1}^e \prod_{a \in O_i} (X - a)$  is a decomposition of type (2) of size  $\mu$ , which is Galois stable. Conversely, given a relation  $\prod_{a \in A} (X - a) \mod Q(X)$ , with  $|A| = \mu$ , Theorem 7 indicates that A is Galois stable. For each i, letting  $O_i$  be the set of elements of A with orbit size equal to i, and  $\mu_i = |O_i|$ , we can write

$$A = O_1 \cup \cdots \cup O_e$$

with  $\mu = \mu_1 + 2\mu_2 + \dots + e\mu_e$ , i.e. a partition of  $\mu$ . The enumeration formula follows, by considering that there are  $\binom{n_i}{\mu_i}$  ways of choosing  $\mu_i$  orbits between  $n_i$ .

Then, given  $\mathbb{F}_{q^h}$ , in the above situation, the probability of finding a relation is

$$\varpi = \frac{N_e(h)}{q^h} = \frac{1}{q^h} \left( \sum_{(h_1, \dots, h_e) \in P_h^e} \prod_{i=1}^e \binom{n_i}{h_i} \right),$$

from the symmetry of  $\mu$  and  $\tau = n - \mu$ , and using  $\tau = h$ .

7.1.1. Example:  $n = q^e$ . We choose  $S = \mathbb{F}_{q^e}$ ,  $S_i$  being the set of all elements in S whose orbits under Galois have size i. Then  $n_i = \frac{1}{i} \sum_{j|i} \mu(j) q^{\frac{i}{j}} \sim q^i/i$ , if  $i \mid e$ , and

h	3	5	7	11	13	31	67
1/h!	0.167	0.00833	0.000198	$2.51 \ 10^{-8}$	$1.61 \ 10^{-10}$	$1.22 \ 10^{-34}$	$2.74 \ 10^{-95}$
$c_2(h)$	0.667	0.217	0.0460	0.000895	$9.13 \ 10^{-5}$	$4.46 \ 10^{-16}$	$2.36 \ 10^{-45}$
$c_3(h)$		0.175	0.0697	0.00356	0.000783	$1.13 \ 10^{-11}$	$1.32 \ 10^{-31}$
$c_4(h)$		0.467	0.213	0.0333	0.0113	$3.24 \ 10^{-8}$	$1.03 \ 10^{-22}$
$c_6(h)$			0.407	0.117	0.0605	$1.48 \ 10^{-5}$	$4.11 \ 10^{-15}$
$c_8(h)$				0.117	0.0696	$9.79 \ 10^{-5}$	$5.71 \ 10^{-12}$
$c_9(h)$				0.0591	0.0424	$9.06 \ 10^{-5}$	$1.76 \ 10^{-11}$
$c_{12}(h)$					0.227	0.00384	$6.67 \ 10^{-8}$

FIGURE 2. The constants 1/h!,  $c_e(h)$ , for e = 2, 3, 4, 6, 8, 9, 12, and h = 3, 5, 7, 11, 13, 31, 67.

zero otherwise. For h constant and growing q, we get a probability of

$$\varpi = \frac{1}{q^h} \left( \sum_{(h_1, \dots, h_e) \in P_h^e} \prod_{i=1}^e \binom{n_i}{h_i} \right) \\
\sim \frac{1}{q^h} \left( \sum_{(h_1, \dots, h_e) \in P_h^e} \prod_{i=1}^e \frac{n_i^{h_i}}{h_i!} \right) \\
\sim \frac{1}{q^h} \left( \sum_{(h_1, \dots, h_e) \in P_h^e} \prod_{i=1}^e \frac{q^{ih_i}}{i^{h_i}h_i!} \right) \\
= \frac{1}{q^h} \left( \sum_{(h_1, \dots, h_e) \in P_h^e} q^{h_1 + 2h_2 + \dots + eh_e} \prod_{i=1}^e \frac{1}{i^{h_i}h_i!} \right) \\
= \sum_{(h_1, \dots, h_e) \in P_h^e} \prod_{i=1}^e \frac{1}{i^{h_i}h_i!} = c_e(h)$$

which does not depend on q. This is much higher than 1/h!, see Table 2.

7.2. **Practice.** Since  $S = \mathbb{F}_{q^e}$ , we have  $G(X) = X^{q^e} - X$ . Decoding over S amounts to testing divisibility of G(X) by an error-location polynomial v(X) whose roots are conjugate under the Frobenius, since S and the corresponding A are. This means that v(X) is a product of minimal polynomials of elements of S. In other words, we can see this as decomposing over the basis containing these minimal polynomials. As a consequence, the matrix of relations will be smaller, its number of columns being  $\sum_i n_i \simeq q^e/e$  instead of  $q^e$ .

It is not difficult to adapt the incremental version of our algorithm to that case. Assuming all operations take place over  $\mathbb{F}_q$ , we thus have a complexity for the relation step which is dominated by

$$C = O\left(n\frac{1}{\varpi}\left(M(n) + M(h)\log q\right)\right).$$

In the case where we take  $n = q^e$ , this yields

$$C = \tilde{O}\left(\frac{q^{2e}}{c_e(h)}\right).$$

Optimizing the value of n is still on-going work.

7.3. Numerical example. Consider  $\mathbb{F}_{7^5} = \mathbb{F}_7[X]/(X^5 + X + 4)$  and a helper field  $\mathbb{F}_{7^2}$ . The decomposition base contains 7 polynomials of degree 1 and 21 of degree 2, and its cardinality is 28. By Table 2, the probability of success is approximately 0.217. We find for instance

$$X^{20}(X+3)(X+4)(X+5)(X^2+X+4) \equiv G(X) := X^{49} - X \mod Q(X).$$

# 8. Numerical examples

8.1. **RSDL-FQ.** We programmed RSDL-FQ in NTL 5.5.2 and made it run on an Intel Xeon CPU E5520 at 2.27GHz. We took p = 65537 and ran the program on several prime values of h (timings are in seconds rounded to the nearest integer):

h	update	EEA	$X^q \bmod v$	roots	$\log_2 P$	linear algebra
3	67	4	4	3	27	213
5	1297	135	104	6	28	3398
7	53007	8086	5745	8	97	124095

Defining polynomials are:

$$W^3 + 6W - 3$$
,  $W^5 + W + 3$ ,  $W^7 + W + 3$ .

For the last column, we indicate the size of the largest prime factor P of  $p^h-1$  and the time needed to perform Gaussian inversion on the system modulo P (using Magma V2.17-1 on the same machine).

8.2. **RSDL-HF.** We programmed the collection phase RSDL-HF in NTL 5.5.2 and made it run on an Intel Xeon CPU E5520 at 2.27GHz, collecting the v(X) unfactored.

We took p=3 and ran the program on h=29, with a helper field of degree e=8 (timings are in seconds rounded to the nearest integer), and the defining polynomial is  $Q:=W^{29}+2W^4+1$ . Another example is p=101, h=11 and e=2. We also include an example over  $\mathbb{F}_2$ , and extension degree h=31, with e=8.

p	h	e	update	EEA	$X^{q^e} \bmod v$	linear algebra
2	31	8	9	271	347	0
3	29	8	2255	12456	8036	2
101	11	2	440	816	589	100

# 9. Concluding remarks

Improvements can certainly be made to the present scheme to tackle more realistic discrete logarithm computations. It seems valuable to have an approach not using smooth polynomials nor using too much algebraic factorizations in discrete logarithm computations. This sheds some light on the relationship between coding theory and classical problems in algorithmic number theory.

Our investigations on the use of Reed-Solomon decoding for discrete logarithm computations have just begun. For the time being, the proposed approach seems to have a worse complexity than its competitor FFS. Many paths are still to follow. In

our setting, the use of so-called large primes is not clear. In our case, we can force them by trying to decode  $P(X)X^u \mod Q(X)$  for fixed P and hoping for several relations, but this does not seem to decrease the cost of the algorithm.

Some other topics of research include the use of list decoding algorithms, variants of Reed-Solomon or more general codes. We could also dream of getting the best of the two worlds, for instance factoring our  $f_A(X)$ 's to get more relations. All this is the subject of on-going work.

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