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Fast multiplication for skew polynomials

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ABSTRACT

We describe an algorithm for fast multiplication of skew polynomials. It is based on fast modular multiplication of such skew polynomials, for which we give an algorithm relying on evaluation and interpolation on normal bases. Our algorithms improve the best known complexity for these problems, and reach the optimal asymptotic complexity bound for large degree. We also give an adaptation of our algorithm for polynomials of small degree. Finally, we use our methods to improve on the best known complexities for various arithmetics problems.

Introduction

The present paper is dedicated to the description of algorithms for fast arithmetics in skew polynomial rings. Since they were first introduced by Ore, skew polynomials and their variants have been widely studied in several areas of mathematics. In particular, skew polynomials over finite fields have various applications in coding theory [14], cryptography see [2], for p -adic Galois representations [10]. Fast arithmetics for manipulating these objects is useful for such applications, and has been improved over time since the first breakthrough paper on computational skew polynomials over finite fields, due to Giesbrecht [8].

Let K be a field and let L be a finite extension of K , endowed with the endomorphism σ . We assume that σ has order $r \geq 1$ and that $K = L^\sigma$. We consider the ring $L[X, \sigma]$ of skew polynomials with coefficients in L . This is a non commutative ring where the relation $Xa = \sigma(a)X$ holds for all $a \in L$ (for more detail about the definitions, see section 1.1). The main problem addressed in this paper is the fast multiplication of elements of $L[X, \sigma]$. The complexity of algorithms is described in terms of the number of elementary operations in K with respect to the degree d of the skew polynomials to be multiplied, and the degree r of L over K .

State of the art. The naïve method for multiplication of skew polynomials of degree $\leq d$ yields an algorithm that has complexity $O(d^2 r^2)$ operations in K . In [8], this complexity

is improved to $O(dr^2 + d^2 r)$. Let ω denote the exponent of matrix multiplication. The authors of the present paper gave several algorithms for multiplication in [3], with best complexity $\tilde{O}(dr^{\omega-1})$ achieved for $d > r^2$. The most recent results by Puchinger and Wachter-Zeh [12] give a bound of $\tilde{O}(d^{\frac{\omega+1}{2}} r)$ operations in K for multiplication in $L[X, \sigma]$, which improves on the previous results [3] when $d \in \Theta(r)$, which is the most relevant case for applications in coding theory (see [12], §4.2). In the context of differential operators (which share many similarities with skew polynomials), Benoit, Bostan and Van der Hoeven have obtained a complexity of $\tilde{O}(\min\{d, r\}^{\omega-2} dr)$ (see [1], Theorem 1) for multiplication in $L[x]\langle \partial \rangle$. We expect that this complexity should be doable in $L[X, \sigma]$ as well, but we have only achieved it for $d \geq r$.

Contributions of the paper. This paper's main algorithm improves the complexity of the best known algorithms for multiplication in $L[X, \sigma]$ to $\tilde{O}(dr^{\omega-1})$ when $d \geq r$. For $d \in \Theta(r)$, this gives a complexity of $\tilde{O}(r^\omega)$ operations in K . This is quasi-optimal in the sense that matrix multiplication can be reduced to skew polynomial multiplication (this is for example a consequence of Proposition 1.6 below), so that any improvement on the exponent of skew polynomial multiplication would lead to a similar improvement for matrix multiplication. We also design a new algorithm for multiplication of polynomials of small degree $d \ll r$ in $L[X, \sigma]$, whose complexity is $\tilde{O}(d^{\omega-2} r^2)$.

We also show that our method can be used to improve the best known complexities of various related problems, such as multi-point evaluation, minimal subspace polynomial, and interpolation which are studied in [12]. We also improve the complexities for greatest common divisors and least common multiples.

Organization of the paper. The first section of the paper focuses on elementary operations for skew polynomials with normal bases: evaluation and interpolation. More precisely, if $P \in L[X, \sigma]$, then $P(\sigma)$ is an endomorphism of the K -algebra L , and the map $P \mapsto P(\sigma)$ is a morphism of K -algebras. In this section, we describe how we can compute efficiently $P(\sigma)$ using a normal basis and, conversely, how to recover P (the reduction modulo $X^r - 1$ of P from the datum of $P(\sigma)$ (see Proposition 1.6). We also look into more detail how the can solve the same evaluation/interpolation problems with P of small degree n at only the first n elements of a normal basis.

In the second section, we present our algorithm for fast multiplication of skew polynomials. First, we study how the

multiplication can be done efficiently modulo $X^r - 1$ through evaluation/interpolation on a normal basis and matrix multiplication. We generalize this study to multiplication modulo $Z(X^r)$ for any irreducible polynomial $Z \in K[T]$. This allows us to give an algorithm for multiplication of skew polynomials of degree d that works in $O(dr^{\omega-1})$ operations in K (where r^ω denotes the complexity of multiplication of square matrices of size r).

In the third section, we give several applications to fast arithmetics for skew polynomials. We show how we can perform general multi-point evaluation, minimal subspace polynomial, and interpolation, as well as usual operations on skew polynomials such as (extended) Euclidean division, greatest common divisor, least common multiple.

1. FAST EVALUATION AND INTERPOLATION

In this section, we present the notion of skew polynomials, and we study the problems of their evaluation and interpolation using normal bases.

1.1 Definitions and notations

Let K be a field and let L be an étale K -algebra (since K is a field, this just means that L is isomorphic to a product of field extensions of K). Let σ be an automorphism of L . We assume that σ has finite order r and that $K = L^\sigma$. The ring $L[X, \sigma]$ of skew polynomials with coefficients in K is the ring whose underlying group is $L[X]$ and whose multiplication is determined by the relation

$$\forall \alpha \in L, X\alpha = \sigma(\alpha)X.$$

The ring $L[X, \sigma]$ is not commutative unless $r = 1$.

Examples. The following situations are examples of the general setting that we are considering:

- $L = K^r$, and σ is the shift operator $(x_0, \dots, x_{r-1}) \mapsto (x_1, \dots, x_{r-1}, x_0)$,
- (Extensions of finite fields) $K = \mathbf{F}_q$, $L = \mathbf{F}_{q^r}$ and $\sigma : x \mapsto x^q$ is the Frobenius endomorphism of L ,
- (Cyclotomic extensions) $K = \mathbf{Q}$ and $L = \mathbf{Q}(\zeta_{p^n})$ where ζ_{p^n} is a primitive p^n -th root of unity and p is prime; σ is a generator of the Galois group $\text{Gal}(L/K)$ (which is the cyclic group $(\mathbf{Z}/p^n\mathbf{Z})^\times$).
- (Kummer extensions) K contains a primitive r -th root ζ_r of 1, $L = K(\sqrt[r]{a})$ for some suitable $a \in K$ and σ takes $\sqrt[r]{a}$ to $\zeta_r \sqrt[r]{a}$.

The two last examples are addressed in [13] and have applications to space-time codes.

Remark 1.1. Usually, L is assumed to be a field extension of K . We are considering the more general context of an étale K -algebra because it is stable under base change: if L/K is étale and K' is an extension of K , then $L' = L \otimes_K K'$ is étale over K' (but it is not a field in general, even if L is). This feature is used mostly in Section 2.1.2, and does not make the classical results any more difficult to prove.

Definition 1.2. A *normal basis* of L/K is a basis (b_0, \dots, b_{r-1}) of L over K such that $\sigma(b_{i+1}) = b_i$ (the indices being taken modulo r).

Proposition 1.3 ([5], Satz 1). *Assuming σ has order r and $K = L^\sigma$, L has a normal basis.*

The problem of the construction of normal bases has been widely studied, see for example [7] for the case of finite fields, and [9] for the case of number fields. In both cases of cyclotomic extensions and Kummer extensions, it is easy to exhibit a normal basis: in the cyclotomic case, the basis starting with $b_0 = \zeta_{p^n}$ does the job while in the Kummer case, one can take:

$$b_0 = 1 + \sqrt[r]{a} + \sqrt[r]{a^2} + \dots + \sqrt[r]{a^{r-1}} = \frac{a-1}{\sqrt[r]{a}-1}.$$

From now on, we assume that we have fixed a normal basis (b_0, \dots, b_{r-1}) of L together with a working basis in which the elements of L are represented. Let Ω be the matrix of change of basis from the working basis to the normal basis. We assume that the multiplication in L and the application of σ can be both performed in $\tilde{O}(r)$ operations in K in the working basis.

1.2 Evaluation and interpolation on a normal basis

We introduce a relation between polynomials that allows to evaluate the linear map associated to a skew polynomial at the elements of the normal basis (b_0, \dots, b_{r-1}) .

Lemma 1.4. *The map:*

$$\begin{aligned} L[X, \sigma] &\rightarrow \text{End}_K(L) \\ A = \sum_{i \geq 0} a_i X^i &\mapsto A(\sigma) = \sum_{i \geq 0} a_i \sigma^i \end{aligned}$$

is a homomorphism of K -algebras. It induces an isomorphism of K -algebras:

$$\varepsilon : L[X, \sigma]/(X^r - 1) \simeq \text{End}_K(L).$$

Proof. The first map is a homomorphism because for all $a \in L$, $Xa = \sigma(a)X$ in $L[X, \sigma]$. Since σ has order r , $X^r - 1$ lies in the kernel of this map, so ε is well-defined. Both $L[X, \sigma]/(X^r - 1)$ and $\text{End}_K(L)$ are K -vector spaces of dimension r^2 , hence it suffices to prove injectivity. By Artin's Lemma on independence of characters, $\{id, \sigma, \dots, \sigma^{r-1}\}$ is a linearly independent family over L , so that if $P(\sigma) = 0$ for some $P \in L[X, \sigma]$ of degree $< r$, then $P = 0$. \square

Lemma 1.4 shows that multiplication of skew polynomials modulo $X^r - 1$ is essentially the same as multiplication of $r \times r$ matrices over K , assuming that the isomorphism ε can be computed efficiently (in both ways). We now address this question.

Notation 1.5. Throughout this paper, we will denote $P(x)$ for $P(\sigma)(x) = \varepsilon(P)(x)$ if $P \in L[X, \sigma]$ and $x \in L$.

Let T be a new (commutative) variable and consider the classical polynomial ring $L[T]$. Let $B = \sum_{i=0}^{r-1} b_i T^i \in L[T]$ be the polynomial whose coefficients are the elements of the normal basis.

Proposition 1.6. *Let $A = \sum_{i=0}^{r-1} a_i X^i \in L[X, \sigma]$ and let $\tilde{A}(T) = \sum_{i=0}^{r-1} a_i T^i \in L[T]$. Let $c_j = A(b_j)$ and let $C(T) = \sum_{j=0}^{r-1} c_j T^j$. Then*

$$C(T) = \tilde{A}(T)B(T) \pmod{T^r - 1}.$$

Proof. By linearity, it is enough to check that the relation holds when $A = X^i$ for $0 \leq i \leq r-1$. Let $0 \leq i \leq r-1$. We have $X^i(b_j) = \sigma^i(b_j) = b_{j-i}$, where indices are taken

modulo r .

On the other hand, doing the calculations modulo $T^r - 1$, $T^i B(T) = \sum_{j=0}^{r-1} b_{j-i} T^j$. \square

Proposition 1.6, although elementary, shows that the isomorphism ε of Lemma 1.4 can be computed efficiently. Moreover, it also shows how the inverse isomorphism can be computed. More precisely:

Corollary 1.7. *Multiplication in $L[X, \sigma]/(X^r - 1)$ can be performed in $O(r^\omega)$ operations in K .*

Proof. Let $A_1, A_2 \in L[X, \sigma]/(X^r - 1)$. Let $\tilde{A}_1(T), \tilde{A}_2(T) \in L[T]$ be the commutative polynomials with the same coefficients as A_1, A_2 respectively. Let $C_1(T) = \tilde{A}_1(T)B(T) \in L[T]/(T^r - 1)$ and $C_2(T) = \tilde{A}_2(T)B(T) \in L[T]/(T^r - 1)$. Both C_1 and C_2 can be computed in $\tilde{O}(r^2)$ operations in K . Now let M_1 (resp. M_2) be the matrix whose j -th column is the decomposition of the j -th coefficient of C_1 (resp. C_2) in the working basis. By Proposition 1.6, M_1 (resp. M_2) is the matrix of $\varepsilon(A_1)$ (resp. $\varepsilon(A_2)$) where the codomain in endowed with the normal basis and the codomain is endowed with the working basis. Set $M = M_1 \Omega M_2$; this product can be computed within $O(r^\omega)$ operations in K . We know that M is the matrix of $\varepsilon(A_1 A_2)$ where again the codomain in endowed with the normal basis and the codomain is endowed with the working basis. Let

$$C(T) = (b_0 \quad b_1 \quad \cdots \quad b_{r-1}) M \begin{pmatrix} 1 \\ T \\ \vdots \\ T^{r-1} \end{pmatrix},$$

and compute $\tilde{A}(T) = C(T)B(T)^{-1} \pmod{T^r - 1} = \sum_{i=0}^{r-1} a_i T^i$, which can also be computed in $\tilde{O}(r^2)$ operations in K . Then, again by Proposition 1.6, $A_1 A_2 = \sum_{i=1}^{r-1} a_i X^i$. This shows that the global complexity of this computation is $O(r^\omega)$. \square

In Section 2, we will generalize this algorithm and show how it yields a fast multiplication algorithm for skew polynomials (not only in the modular case).

1.3 Evaluation and interpolation at an incomplete normal basis

Evaluation. We shall see later how we can compute the product of two skew polynomials of small degree d by determining how their product acts on $2d$ elements of a normal basis. With this motivation in mind, let us describe how we can compute efficiently the image of the first few elements of a normal basis under the action of the skew polynomial $A \in L[X, \sigma]$. Recall that, using Proposition 1.6 with $\lambda = 1$, and writing $B(T) = \sum_{i=0}^{r-1} b_i T^i$, we know that

$$\tilde{A}(T)B(T) \equiv C(T) \pmod{T^r - 1},$$

where $C(T) = \sum_{i=0}^{r-1} A(b_i)T^i$. Let $n < r$, and let $A \in L[X, \sigma]$ of degree n . We are interested in computing only $A(b_i)$ for $0 \leq i \leq n - 1$.

Lemma 1.8. *Let $A \in L[X, \sigma]$ of degree n and let $c_i = A(b_i)$ for $0 \leq i \leq n - 1$. Let $U(T) = \sum_{i=0}^n b_i T^i$ and $V(T) = \sum_{i=0}^n b_{r-i} T^{r-i}$. Then, for $0 \leq i \leq n$:*

$$c_i = \gamma_i + \tilde{\gamma}_i,$$

where $\tilde{A}(T)U(T) = \sum_{i=0}^{r-1} \gamma_i T^i$ and $\tilde{A}(T)V(T) = \sum_{i=0}^{r-1} \tilde{\gamma}_i T^i$ (the products being taken modulo $T^r - 1$).

Proof. Since $c_i = \sum_{j+j'=i \pmod{r}} a_j b_{j'}$, and $a_j = 0$ for $j > n$, we are left with the formula:

$$c_i = \sum_{j'=0}^i a_{i-j'} b_{j'} + \sum_{j'=r-i}^{r-1} a_{i-j'+r} b_{j'},$$

and both sums correspond precisely to the coefficients of $\tilde{A}U$ and $\tilde{A}V$ respectively. \square

Corollary 1.9. *Let $A \in L[X, \sigma]$ of degree $\leq n$, then the collection of $A(b_0), \dots, A(b_{n-1})$ can be computed in $\tilde{O}(nr)$ operations in K .*

Proof. By Lemma 1.8, the evaluation of A at b_0, \dots, b_{n-1} can be obtained by two multiplications of (classical) polynomials of degree n with coefficients in L , hence with complexity $\tilde{O}(nr)$ operations in K . \square

Interpolation. Still bearing in mind the aim of multiplying two skew polynomials by composing the corresponding linear maps, we are interested in the following question of interpolation: given n values $\alpha_0, \dots, \alpha_{n-1} \in L$, find $A \in L[X, \sigma]$ of degree n such that $A(b_i) = \alpha_i$ for all $0 \leq i \leq n - 1$.

Let us explain first how the solution to this problem can be computed when $\alpha_0 = \dots = \alpha_{n-1} = 0$. In this case, the skew polynomial we are looking for is the so-called *minimal subspace polynomial* corresponding to the span $\langle b_0, \dots, b_{n-1} \rangle$. A generic fast algorithm for solving this problem has been proposed by Puchinger and Wachter-Zeh in [12], Theorem 26; it has complexity $\tilde{O}(n^{\max\{\log_2(3), \frac{n+1}{2}\}} r)$ operations in K . In the special case we are considering, we shall see that this can be improved to $\tilde{O}(nr)$.

Let $B_n(T) = \sum_{i=0}^{r-1} b_{i+n} T^i$, so that $B_n(T) \equiv T^{-n} B(T) \pmod{T^r - 1}$. If A is such that $A(b_i) = 0$ for $0 \leq i \leq n - 1$, then there exists $Q \in L[T]$ of degree $\leq r - n - 1$ such that $\tilde{A}(T)B(T) \equiv T^n Q(T) \pmod{T^r - 1}$. Of course, the converse is also true, and this equation is equivalent to:

$$\tilde{A}(T)B_n(T) \equiv Q(T) \pmod{T^r - 1},$$

with $\deg \tilde{A} \leq n$ and $\deg Q \leq r - n - 1$. The latter equation can be solved thanks to the extended Euclidean algorithm. Indeed, computing the gcd of $T^r - 1$ and $B_n(T)$ and stopping after the first remainder of degree $< r - i$, we get a relation of the form:

$$U_i(T)B_n(T) + V_i(T)(T^r - 1) = Q_i(T),$$

with $\deg U_i \leq i$ and $\deg Q_i \leq r - 1 - i$, which yields a solution to the problem when $i = n$. This computation can be done in $\tilde{O}(nr)$ operations in K thanks to the half-gcd algorithm (see [6], Theorem 11.5).

In the general case, let $\alpha_0, \dots, \alpha_{n-1} \in L$, and let $C(T) = \sum_{i=0}^{n-1} \alpha_i T^i$. We are looking for $A \in L[X, \sigma]$ with degree $\leq n$ and $Q \in L[T]$ with degree $\leq r - n - 1$ such that $A(T)B(T) \equiv C(T) + T^n Q(T) \pmod{T^r - 1}$. This equation is equivalent to $A(T)B_n(T) \equiv \sum_{i=0}^{n-1} \alpha_i T^{r-n+i} \pmod{T^r - 1}$.

Lemma 1.10. *Let $R_0(T) = T^r - 1$, $R_1(T) = B(T)$ and for $i \geq 2$, let R_i be the remainder of the Euclidean division of R_{i-2} by R_{i-1} . Then for $0 \leq i \leq r$, $\deg R_i = r - i$.*

Proof. Consider the map

$$\begin{aligned} \varphi_i : L[T]_{<i} \times L[T]_{<i-1} &\longrightarrow L[T]_{<r+i-1}/L[T]_{<r-i} \\ (U, V) &\mapsto UR_1 + VR_0 \end{aligned}$$

It is well-defined, linear, and both sides have the same dimension over L . Moreover, the determinant of this map is nonzero if and only if $\deg R_i = r - i$ (see [15], §4.1). Therefore, it is sufficient to prove that φ_i is injective.

Let us consider (U, V) in the kernel of φ_i . By definition, $\deg(UR_1 + VR_0) < r - i$, so that $U(T)B(T) \equiv W(T) \pmod{T^r - 1}$, where $\deg W(T) < r - i$. By Proposition 1.6, the skew polynomial $\sum_{j=0}^{i-1} u_j X^j$ (whose coefficients are the coefficients of U) evaluates to 0 at b_{r-i}, \dots, b_{r-1} . Hence, it is a left multiple of the minimal subspace polynomial M of $\langle b_{r-i}, \dots, b_{r-1} \rangle$. Since $(b_{r-i}, \dots, b_{r-1})$ is linearly independent over K , M has degree i (it is a generator of the kernel of the K -linear map $L[X, \sigma] \rightarrow L^i$ mapping P to $(P(b_{r-i}), \dots, P(b_{r-1}))$). In particular, since $\deg U < i$, $U = 0$, so $V = 0$ and φ_i is injective. Hence $\det \varphi_i \neq 0$ and R_i has the required degree. \square

Theorem 1.11. *Let $n \leq r$ and $\alpha_0, \dots, \alpha_{n-1} \in L$. Then there exists $U, V, H \in L[T]$, with $\deg U \leq n - 1$, $\deg V \leq n$ and $\deg H \leq r - n$ such that*

$$U(T^r - 1) + VB(T) = H(T) + T^{r-n+1}(\alpha_0 + \dots + \alpha_{n-1}T^{n-1}).$$

Moreover, Algorithm `SmallDegreeInterpolation` outputs U and V for a cost of $\tilde{O}(rn)$ operations in K .

Sketch of the proof. The result follows from the correctness of Algorithm 1, but is also a theoretical consequence of Lemma 1.10. Indeed, this lemma shows that there exists a linear combination of $R_0 = T^r - 1, R_1 = B(T), \dots, R_{n-1}$ whose higher degree terms have coefficients c_0, \dots, c_{n-1} , and the bounds on the degrees follow from the fact that for $i \leq n$, $R_i = U_i R_0 + V_i R_1$ with $\deg U_i \leq i - 1$, $\deg V_i \leq i$. Algorithm 1 is an adaptation of the half-gcd algorithm, which computes simultaneously the sequence of the remainders in the extended Euclidean division of R_0 and R_1 , and the combination of R_1 and R_0 that has the given higher degree terms. \square

Thanks to Corollary 1.9, Theorem 1.11 and Algorithm 1, we can solve the problem of evaluation and interpolation at the first n elements of an incomplete normal basis in $\tilde{O}(nr)$ operations in K .

2. FAST MULTIPLICATION

In this section, we study the problem of multiplying efficiently two elements $A_1, A_2 \in L[X, \sigma]$ both of degree $\leq d$. The complexity is the number of operations in K , given as a function of d and $r = \dim_K L$.

2.1 Modular multiplication

2.1.1 Multiplication modulo $X^r - a$

We consider the ring $L[X, \sigma]$. Let $\lambda \in L^\times$, and let $a = N_{L/K}(\lambda)$. We are now going to describe an algorithm for multiplication in $L[X, \sigma]$ modulo $X^r - a$.

Proposition 2.1. *The map*

$$\begin{aligned} L[X, \sigma] &\longrightarrow L[X, \sigma]/(X^r - 1) \\ A(X) = \sum a_i X^i &\mapsto A(\lambda X) = \sum_i \lambda \sigma(\lambda) \cdots \sigma^{i-1}(\lambda) a_i X^i \end{aligned}$$

factors as an isomorphism $L[X, \sigma]/(X^r - a) \simeq L[X, \sigma]/(X^r - 1)$.

Algorithm 1: `SmallDegreeInterpolation`

Input: $R_0, R_1 \in L[T]$, $a_0, \dots, a_{k-1} \in L$, with $k \leq n_0$

Output: $M \in L[T]^2$ such that $M \begin{pmatrix} R_0 \\ R_1 \end{pmatrix} = \begin{pmatrix} R_{k-1} \\ R_k \end{pmatrix}$

and $N \in L[T]^{1 \times 2}$ such that $N \begin{pmatrix} R_0 \\ R_1 \end{pmatrix} = S_k$

with $\deg S_k - (a_{k-1} + a_{k-2}T + \dots + a_0 T^{k-1}) T^{n_0 - k + 1} \leq n_0 - k$

1 $h := \lfloor k/2 \rfloor$

2 $\tilde{R}_0 = R_0 \text{ quo } T^{2h}, \tilde{R}_1 = R_1 \text{ quo } T^{2h-1}$

3 $M_1, N_1 =$

`SmallDegreeInterpolation`($\tilde{R}_0, \tilde{R}_1, a_0, \dots, a_{h-1}$)

4 $\begin{pmatrix} R_{h-1} \\ R_h \end{pmatrix} := M_1 \begin{pmatrix} R_0 \\ R_1 \end{pmatrix}, S := N_1 \begin{pmatrix} R_0 \\ R_1 \end{pmatrix} =$

$\sum_{i=0}^{n_0-h} s_i T^i + \sum_{i=0}^{h-1} a_i T^{n_0-i}$

5 Make the Euclidean divisions:

6 $R_{h-1} = Q_h R_h + \tilde{R}_{h+1}$

7 $R_{h-1} - a_h T^{n_0-h} = \tilde{Q}_h R_h + \tilde{R}_{h+1}$

8 $M_2, N_2 = \text{SmallDegreeInterpolation}(R_h, R_{h+1}, a_{h+1} - s_{n_0-h}, \dots, a_{2h} - s_{n_0-2h+1})$

9 **return** $M_2 \begin{pmatrix} 0 & 1 \\ 1 & -Q_h \end{pmatrix} M_1, N_1 + (1 \quad -\tilde{Q}_h) M_1 + M_2 N_2$

Proof. This maps X^r to $\lambda \sigma(\lambda) \cdots \sigma^{r-1}(\lambda) X^r = a X^r$, thus mapping $X^r - a$ to $a(X^r - 1)$. \square

Corollary 2.2. *Multiplication in $L[X, \sigma]/(X^r - a)$ can be performed in $O(r^\omega)$ operations in K .*

Proof. By Proposition 2.1 and Proposition 1.6, it is enough to show that for $A \in L[X, \sigma]/(X^r - a)$, $A(\lambda X)$ can be computed in $O(r^\omega)$ operations in K . For this we write $\lambda_i = \lambda \sigma(\lambda) \cdots \sigma^{i-1}(\lambda)$ and remark that the λ_i 's ($0 \leq i < r$) can be all computed within $\tilde{O}(r^2)$ operations in K thanks to the recurrence formula $\lambda_{i+1} = \lambda \cdot \sigma(\lambda_i)$. Now evaluating the formula $A(\lambda X) = \sum_i \lambda_i a_i X^i$ allows us to compute $A(\lambda X)$ in $\tilde{O}(r^2)$ operations in K . \square

We could use the proof of Corollary 2.2 directly to design an algorithm for multiplication modulo $X^r - a$. Such an algorithm would require computing $A_1(\lambda X)$ and $A_2(\lambda X)$ each time we use it to compute $A_1 A_2$. Alternatively, we can slightly modify the basis on which we are evaluating the corresponding maps, which can provide a gain if there are many multiplications to do modulo $X^r - a$.

Let $\lambda \in L^\times$, and let $\sigma_a = \lambda \sigma$. Let $\tilde{b}_{r-1} \in L$, and for $0 \leq i \leq r - 2$, $\tilde{b}_i = \sigma_a^{r-1-i}(\tilde{b}_{r-1})$, such that $\tilde{\mathcal{B}} = (\tilde{b}_0, \dots, \tilde{b}_{r-1})$ is a basis of L over K . By construction, we have for $1 \leq i \leq r - 1$, $\sigma_a(\tilde{b}_i) = \tilde{b}_{i-1}$, and $\sigma_a(\tilde{b}_0) = a \tilde{b}_{r-1}$. For example, if $\mathcal{B} = (b_0, \dots, b_{r-1})$ is a normal basis of L over K , then $\tilde{b}_{r-1} = b_{r-1}$ and $\tilde{b}_i = \lambda \sigma(\lambda) \cdots \sigma^{i-1}(\lambda) b_i$ defines a suitable basis. Now, let $\tilde{B} = \sum_{i=0}^{r-1} \tilde{b}_i T^i \in L[T]$.

Proposition 2.3. *Let $A = \sum_{i=0}^{r-1} a_i X^i \in L[X, \sigma]$ and let $\tilde{c}_j = A(\sigma_a)(\tilde{b}_j)$. Let $\tilde{A}(T) = \sum a_i T^i \in L[T]$. Let $\tilde{C}_a = \sum_{j=0}^{r-1} \tilde{c}_j T^j$. Then*

$$\tilde{C}_a(T) = \tilde{A}(T) \tilde{B}(T) \pmod{T^r - a}.$$

Proof. The proof is similar to that of Proposition 1.6. By linearity, it is enough to check that the relation holds for $A = X^i$ for $0 \leq i \leq r-1$. Let $0 \leq i \leq r-1$. We have :

$$\sigma_a^i(b_j) = \begin{cases} b_{j-i} & \text{if } j \geq i \\ ab_{r+j-i} & \text{if } i > j \end{cases}.$$

On the other hand, doing the calculations modulo $T^r - a$:

$$T^i B(T) = \sum_{j=0}^{r-1} b_j T^{i+j} = \sum_{j=i}^{r-1} b_{j-i} T^i + \sum_{j=0}^{i-1} ab_{r+j-i} T^j.$$

Hence, $T^i B(T) = C_{X^i}(T)$ for all $0 \leq i \leq r-1$, so $C_a(T) = \tilde{A}(T)B(T)$ for all $A \in L[X, \sigma]/(X^r - a)$. \square

Algorithm `ModMult` below makes precise the algorithmical content of Proposition 2.3; it uses a primitive `Matwork` that takes as input a tuple $(x_1, \dots, x_r) \in L^r$ and outputs the $r \times r$ matrix whose j -th column are the coordinates of x_j in the working basis.

Algorithm 2: `ModMult`

Input: $A_1, A_2 \in L[X, \sigma]$, $\lambda \in L^\times$

Output: $A = A_1 A_2 \pmod{X^r - a}$ where $a = N_{L/K}(\lambda)$

- 1 $a = N_{L/K}(\lambda)$
 - 2 $\{b_0, \dots, b_{r-1}\} = \text{NormalBasis}(L/K)$
 - 3 $\tilde{b}_{r-1} = b_{r-1}$
 - 4 **for** $r-1 \geq i \geq 1$ **do**
 - 5 $\tilde{b}_{i-1} = a\sigma(\tilde{b}_i)$
 - 6 $P = \text{Mat}_{\text{work}}(\tilde{b}_0, \dots, \tilde{b}_{r-1})$
 - 7 $B = \sum_{i=0}^{r-1} \tilde{b}_i T^i$
 - 8 **for** $1 \leq i \leq 2$ **do**
 - 9 $C_i = A_i B \pmod{T^r - a}$, write $C_i = \sum_{i=0}^{r-1} c_{i,j} T^j$
 - 10 $N_i = \text{Mat}_{\text{work}}(c_{i,0}, \dots, c_{i,r-1})$
 - 11 $N = N_1 P N_2$
 - 12 $C = (\beta_0 \ \dots \ \beta_{r-1}) N \begin{pmatrix} 1 \\ T \\ \vdots \\ T^{r-1} \end{pmatrix}$
 - 13 $A = C B^{-1} \pmod{T^r - a}$
 - 14 **return** $A(X)$
-

Proposition 2.4. *Algorithm `ModMult` computes the product $A_1 A_2$ in $L[X, \sigma]/(X^r - a)$ in $O(r^\omega)$ operations in K .*

Proof. Multiplication of polynomials in L modulo $T^r - a$ requires $\tilde{O}(r^2)$ operations in K . Multiplication of matrices of size r in K requires $O(r^\omega)$ operations in K . Hence the global complexity is $O(r^\omega)$ operations in K . \square

2.1.2 Multiplication modulo $Z(X^r)$

Let K'/K be a finite extension. Define $L' = K' \otimes L$; it is an étale K' -algebra endowed with the endomorphism $\sigma' = \text{id} \otimes \sigma$ that extends σ and has order r .

Remark 2.5. The algebra L' is not necessarily a field (for instance, when $K' = L$, it splits as a product L^r). It is the reason why we needed to place this paper in the more general setting of étale algebras.

Let $\lambda \in (L')^\times$. Set $a = N_{L'/K'}(\lambda) = \lambda \sigma(\lambda) \cdots \sigma^{r-1}(\lambda) \in K'$. We assume that $K' = K(a)$. Let $Z \in K[T]$ be the minimal polynomial of a . We want to generalize the results of §2.1.1 to multiplication modulo $Z(X^r)$ (in §2.1.1, we have $K' = K$, $L' = L$ and $\lambda \in L^\times$). Note that if (b_0, \dots, b_{r-1}) is a normal basis of L/K , then $(1 \otimes b_0, \dots, 1 \otimes b_{r-1})$ is a normal basis of L'/K' .

Lemma 2.6. *The canonical morphism $1 \otimes \text{id} : L[X, \sigma] \rightarrow L'[X, \sigma]$ induces an isomorphism*

$$L[X, \sigma]/Z(X^r) \simeq L'[X, \sigma']/(X^r - a).$$

Proof. First note that $(X^r - a)$ is a two-sided ideal of $L'[X, \sigma]$, and that the canonical morphism $L[X, \sigma] \rightarrow L'[X, \sigma]$ induces a morphism $L[X, \sigma] \rightarrow L'[X, \sigma]/(X^r - a)$ which maps X^r to a , hence the latter surjective. Moreover, by K -linearity, $Z(X^r)$ lies in the kernel of this map. We then get a surjective morphism of K -algebras $L[X, \sigma]/Z(X^r) \rightarrow L'[X, \sigma]/(X^r - a)$. Since both sides have dimension $r^2 \deg Z$ over K , this morphism is an isomorphism. \square

We are now back exactly in the situation of Section 2.1.1, where K has been replaced by K' and L by L' : all the computations can be carried out the same way, and passing back through the isomorphism of Lemma 2.6, we can perform fast multiplication modulo $Z(X^r)$. The algorithm is as follows:

Algorithm 3: `ModMultZ`

Input: $A_1, A_2 \in L[X, \sigma]$, K'/K a finite extension, $\lambda \in L' = K' \otimes L$ nonzero, $a = N_{L'/K'}(\lambda) \in K'$ such that $K' = K(a)$, $Z \in K[T]$ the minimal polynomial of a over K .

Output: $A = A_1 A_2 \pmod{Z(X^r)}$ where Z is the minimal polynomial of $a = N_{L'/K'}(\lambda)$ over K .

- 1 Write $A_1 = \sum_{i=0}^{r-1} \alpha_i(X^r) X^i$, $A_2 = \sum_{i=0}^{r-1} \beta_i(X^r) X^i$
 - 2 Let $\tilde{A}_1 = \sum_{i=0}^{r-1} \alpha_i(a) X^i$, $\tilde{A}_2 = \sum_{i=0}^{r-1} \beta_i(a) X^i$
 - 3 Compute $\tilde{A} = \tilde{A}_1 \tilde{A}_2$ using `ModMult` in $L'[X, \sigma]/(X^r - a)$ endowed with the normal basis $(1 \otimes b_i)$
 - 4 Write $A = \sum_{i=0}^r \gamma_i(a) X^i$
 - 5 **return** $A = \sum_{i=0}^r \gamma_i(X^r) X^i$
-

Proposition 2.7. *Algorithm 3 computes the product $A_1 A_2$ in $L[X, \sigma]/(Z(X^r))$ with $O(r^\omega \deg Z)$ operations in K .*

2.2 Reconstruction with CRT

Let $A_1, A_2 \in L[X, \sigma]$ be two skew polynomials. We recall that our aim is to design a fast algorithm for computing the product $P = A_1 A_2$. We set $d = \deg P$.

Multiplication in large degree. We first assume that the polynomial $P = A_1 A_2$ has degree larger than r . In this case, the idea is to evaluate the P modulo various $Z_i(X^r)$ using Algorithm `ModMultZ` and then to reconstruct the result using a non commutative version of the Chinese Remainder Theorem. The precise result we need is given by the following Proposition.

Proposition 2.8. *Let $Z_1, \dots, Z_m \in K[T]$ be pairwise coprime polynomials, and let $Z = Z_1 \cdots Z_m$. Then the natural map:*

$$L[X, \sigma]/Z(X^r) \rightarrow L[X, \sigma]/Z_1(X^r) \times \cdots \times L[X, \sigma]/Z_m(X^r)$$

The case of number fields. We assume that K and L are number fields. It is then known that the image of the norm map $N_{L/K} : L^* \rightarrow K^*$ has index r in K^* . More precisely, class field theory teaches us that $K^*/N_{L/K}(L^*)$ is canonically isomorphic to the Galois group of the abelian extension L/K , *i.e.* to $\mathbf{Z}/r\mathbf{Z}$. In particular, the image of $N_{L/K}$ is infinite meaning that if we take a finite set of random elements $\lambda \in L$, it is likely that the norm of the λ 's will be pairwise distinct. We can then reapply the strategy used in the case of finite field without having to work with an auxiliary extension K' . We end up this way with a probabilistic Las Vegas algorithm whose complexity is $\tilde{O}(dr^{\omega-1})$ operations in K and whose probability of success is high.

Multiplication in small degree. The idea for fast multiplication in small degree is that if a skew polynomial has degree $d \ll r$, it is determined by its values on $d+1$ linearly independent elements of L . Hence, starting with two skew polynomials A_1, A_2 whose degrees add up to d , we should be able to compute their product by composing of two K -linear maps over vector spaces of dimension $d+1$. However, we know some efficient algorithm for evaluating $A(\sigma)$ only on a subspace of L which is spanned by the first vectors of a normal basis. For this reason, in order to compute $A_1 A_2(b_0), \dots, A_1 A_2(b_{d-1})$, we shall need to know the whole of the linear map $A_1(\sigma)$ (because $A_2(b_0), \dots, A_2(b_{d-1})$ are in general nothing to do with a truncated normal basis).

Algorithm 5: SmallDegreeMultiplication

Input: $A_1, A_2 \in L[X, \sigma]$, $\deg A_1 + \deg A_2 < r$

Output: $P = A_1 A_2$

- 1 Set $d = \deg A_1 + \deg A_2$
 - 2 Compute $A_2(b_0), \dots, A_2(b_d)$ // use Corollary 1.9
 - 3 Compute the matrix of $P(\sigma)$ // use Proposition 1.6
 - 4 Compute $c_0 = A_1 A_2(b_0), \dots, c_d = A_1 A_2(b_d)$
// matrix multiplication of sizes $r \times r$ by $r \times (d+1)$
 - 5 Compute $P \in L[X, \sigma]$ s.t. $P(b_i) = c_i$ and $\deg P \leq d$.
// use Algorithm SmallDegreeInterpolation
 - 6 **return** P
-

The complexity of the above algorithm is given by the next Theorem whose proof is straightforward after what we have already done (the bottleneck comes from the matrix multiplication step).

Theorem 2.12. *Let A_1, A_2 such that $\deg A_1 + \deg A_2 \leq d < r$. Then Algorithm 5 computes the product $A_1 A_2$ with $O(d^{\omega-2} r^2)$ operations in K .*

Conclusion. As a conclusion, several algorithms with different complexities are available for the multiplication of skew polynomials. Precisely, we have designed in this paper one algorithm of complexity $\tilde{O}(dr^{\omega-1})$ when $d \geq r$ and another algorithm of complexity $\tilde{O}(d^{\omega-2} r^2)$ when $d \leq r$. Apart from that, Wachter-Zeh's algorithm [12] performs the same computation with complexity $\tilde{O}(d^{(\omega+1)/2} r)$ without any assumption on d . The corresponding complexity curves are represented on Figure 1. Putting all together, we find that the product in $L[X, \sigma]$ can be performed within

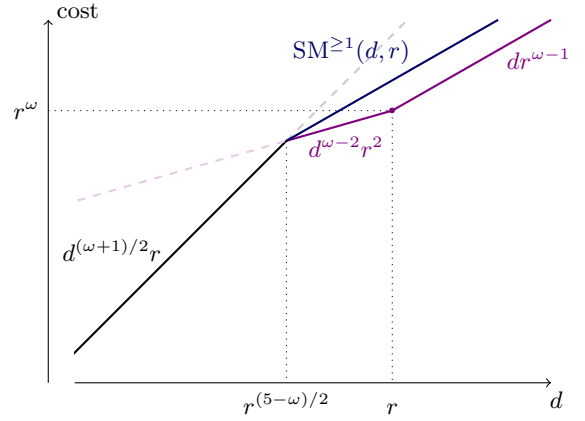


Figure 1: Complexity profiles (log-log scale)

$\tilde{O}(\text{SM}(d, r))$ operations in K where:

$$\begin{aligned} \text{SM}(d, r) &= d^{(\omega+1)/2} r && \text{for } d \leq r^{(5-\omega)/2} \\ &= d^{\omega-2} r^2 && \text{for } r^{(5-\omega)/2} \leq d \leq r \\ &= dr^{\omega-1} && \text{for } d \geq r. \end{aligned}$$

As already discussed in the introduction, we expect to lower the complexity to $\tilde{O}(d^{\omega-1} r)$ in the range $d \leq r$ and, until now, we have not succeeded in doing so.

3. OTHER OPERATIONS AND APPLICATIONS

Classically, fast multiplication algorithms can be used to speed up many other computations. This general philosophy works for skew polynomials as well and was concretized in [3], §3.2. Below, we analyze briefly the impact of the algorithms designed above in this paper.

In order to state our complexity results more elegantly, we introduce the function $\text{SM}^{\geq 1}$ defined by:

$$\text{SM}^{\geq 1}(d, r) = \sup_{d' \leq d} \left(\text{SM}(d', r) \cdot \frac{d}{d'} \right).$$

A direct computation shows that:

$$\begin{aligned} \text{SM}^{\geq 1}(d, r) &= d^{(\omega+1)/2} r && \text{for } d \leq r^{(5-\omega)/2} \\ &= dr^{4/(5-\omega)} && \text{for } d \geq r^{(5-\omega)/2}. \end{aligned}$$

The function $\text{SM}^{\geq 1}$ (viewed as a function of the variable d) is the smallest function above SM whose “log-log slope” is always at least 1 (see Figure 1). The notation comes from this interpretation.

With $\omega = 2.37$, we have $\text{SM}^{\geq 1}(d, r) \approx d^{1.69} r$ for $d \leq r^{0.76}$ and $\text{SM}^{\geq 1}(d, r) \approx dr^{1.52}$ for larger d .

Euclidean division. An algorithm that performs (right) Euclidean divisions in $L[X, \sigma]$ and takes advantage of fast multiplication algorithm is depicted in [3], §3.2.1 (Algorithm **REuclideanDivision**). Proposition 3.2.3 of *loc. cit.* extends readily to the settings of this paper and shows that the aforementioned algorithm has a complexity cost of $\tilde{O}(\text{SM}^{\geq 1}(d, r))$ operations in K .

GCD and LCM computation. The classical half-gcd algorithm that we already mentioned above (see §1.3 and [6], §11) works in the same way to compute left and right GCD's

of skew polynomials. The precision corresponding algorithm is written in [3], §3.2.2 (Algorithm `FastExtendedRGCD`).

Proposition 3.1. *The algorithm `FastExtendedRGCD` of [3], §3.2.2 (using fast multiplication algorithms described above in this paper as primitives) runs in $\tilde{O}(\text{SM}^{\geq 1}(d, r))$ operations in K .*

Proof. A careful look at the algorithm `FastExtendedRGCD` shows that its complexity in operations in K is bounded by $T(d, r)$ where $T(d, r)$ satisfies the recurrence relation:

$$T(d, r) \leq 2T\left(\frac{d}{2}, r\right) + \tilde{O}(\text{SM}(\frac{d}{2}, r)).$$

By induction, it follows that for $m \geq 0$,

$$\begin{aligned} T(d, r) &\leq 2^m T\left(\frac{d}{2^m}, r\right) + \tilde{O}\left(\sum_{j=1}^m 2^j \text{SM}\left(\frac{d}{2^j}, r\right)\right) \\ &\leq 2^m T\left(\frac{d}{2^m}, r\right) + \tilde{O}\left(m \cdot \text{SM}^{\geq 1}(d, r)\right). \end{aligned}$$

Taking $m = \lceil \log_2 d \rceil$, we get $T(d, r) = \tilde{O}(\text{SM}^{\geq 1}(d, r))$ as expected. \square

Remark 3.2. A similar complexity is available for the computation of LCM's.

Minimal subspace polynomial. Let (x_1, \dots, x_d) be a family of elements of L which is free over K . We are interested in computing the unique monic polynomial $P \in L[X, \sigma]$ of degree d such that $P(x_i) = 0$ for all $i \in \{1, \dots, d\}$.

Lemma 3.3. *For $x \in L$, $x \neq 0$, the value $\frac{P(x)}{x}$ is the remainder in the right Euclidean division of P by $X - \frac{\sigma(x_i)}{x_i}$.*

Proof. It is a direct computation. \square

Lemma 3.3 shows that the polynomial P we are looking for is nothing but the left-lcm of the polynomials $X - \frac{\sigma(x_i)}{x_i}$. As a consequence, P can be computed for a cost of $\tilde{O}(\text{SM}^{\geq 1}(d, r))$ operations in K using fast algorithms for LCM computation together with a “tree division strategy” [6], §10.1.

General multievaluation. We consider again a free family (x_1, \dots, x_d) of elements of K . The general multievaluation problem consists in evaluating a given polynomial $P \in K[X, \sigma]$ of degree d at the x_i 's. Thanks to Lemma 3.3, the value $P(x_i)$ agrees with x_i times the remainder of the right division of P by $X - \frac{\sigma(x_i)}{x_i}$. We are then reduced to compute the reduction of a given polynomials modulo some given moduli. This can be done efficiently using the strategy of [6], §10.1 for a cost of $\tilde{O}(\text{SM}^{\geq 1}(d, r))$ operations in K . If d are r have the same order of magnitude, one can preferably compute the matrix of $P(\sigma)$ using the formula of Proposition 1.6 and derive from it the values of the $P(x_i)$'s thanks to a single matrix multiplication. The cost of the resulting algorithm is $O(r^\omega)$.

Remark 3.4. If the x_i 's are the first vectors of a normal basis of L over K , one can use directly the algorithm of §1.3 which has a better complexity.

General interpolation. We keep the family (x_1, \dots, x_d) and consider in addition some values $y_1, \dots, y_d \in L$. We address the question of computing a polynomial P of degree at most $d-1$ such that $P(x_i) = y_i$ for all i . Thanks to

Lemma 3.3, the above problem reduces to solve the following Chinese Remainder system:

$$P(x_i) \equiv x_i y_i \pmod{X - \frac{\sigma(x_i)}{x_i}}$$

which again can be done for a cost of $\tilde{O}(\text{SM}^{\geq 1}(d, r))$ operations in K .

Remark 3.5. If the x_i 's are the first vectors of a normal basis of L over K , one can use directly the Algorithm `Small-DegreeInterpolation` which has a better complexity.

Gabulin codes. The solution sketched above to the general multievaluation problem allows us to encode messages in the framework of (generalized) Gabidulin codes [13] in complexity $O(n^\omega)$ where n is the length of the code. (Better complexities are possible when the dimension of the code is much smaller than its length.) In the similar fashion, efficient decoding is also possible using the key equation together with the half-GCD algorithm. The resulting algorithms run in $\tilde{O}(\text{SM}^{\geq 1}(n, k))$ operations in K where n and k denotes the length and the dimension of the Gabidulin code respectively.

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