A Bivariate Preprocessing Paradigm for Buchberger-Möller Algorithm

Xiaoying Wang, Shugong Zhang, Tian Dong

School of Mathematics, Key Lab. of Symbolic Computation and Knowledge Engineering (Ministry of Education), Jilin University, Changchun 130012, PR China

Abstract

For the last almost three decades, since the famous Buchberger-Möller (BM) algorithm emerged, there has been wide interest in vanishing ideals of points and associated interpolation polynomials. Our paradigm is based on the theory of bivariate polynomial interpolation on cartesian point sets that gives us related degree reducing interpolation monomial and Newton bases directly. Since the bases are involved in the computation process as well as contained in the final output of BM algorithm, our paradigm obviously simplifies the computation and accelerates the BM process. The experiments show that the paradigm is best suited for the computation over finite prime fields that have many applications.

Key words: Buchberger-Möller algorithm, Bivariate Lagrange interpolation, Degree reducing interpolation space, Cartesian set

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1. Introduction

For an arbitrary field $\mathbb{F}$, we let $\mathbb{F}_q$ a finite prime field of size $q$ and $\Pi^d := \mathbb{F}[x_1, \ldots, x_d]$ the $d$-variate polynomial ring over $\mathbb{F}$. Given a preassigned set of distinct affine points $\Xi \subset \mathbb{F}^d$, it is well-known that the set of all polynomials...
in \(\Pi^d\) vanishing at \(\Xi\) constitutes a radical zero-dimensional ideal, denoted by \(I(\Xi)\), which is called the \textit{vanishing ideal} of \(\Xi\).

Recent years, there has been considerable interest in vanishing ideals of points in many branches of mathematics such as algebraic geometry\([1]\), multivariate interpolation\([2,3]\), coding theory\([4,5]\), statistics\([6]\), and even computational molecular biology\([7,8]\). As is well known, the most significant milestone of the computation of vanishing ideals is the algorithm presented in \([9]\) by Hans Michael Möller and Bruno Buchberger known as Buchberger-Möller algorithm (BM algorithm for short). For any point set \(\Xi \subset \mathbb{F}^d\) and fixed term order \(\prec\), BM algorithm yields the reduced Gröbner basis for \(I(\Xi)\) w.r.t. \(\prec\) and a \(\prec\)-degree reducing interpolation Newton basis for \(d\)-variate Lagrange interpolation on \(\Xi\). The algorithm also produces the Gröbner éscalier of \(I(\Xi)\) w.r.t. \(\prec\) as a byproduct. Afterwards, in 1993, BM algorithm was applied in \([10]\) in order to solve the renowned FGLM-problem. In the same year, \([11]\) merged BM and FGLM algorithms into four variations that can solve more general zero-dimensional ideals therefore related ideal interpolation problems \([3]\). The algorithms are referred as MMM algorithms.

Although very important, BM algorithm (and MMM algorithms) has a very poor complexity that limits its applications. In this decade, many authors proposed new algorithms that can reduce the complexity but mostly suitable for special cases. \([12]\) presented a modular version of BM algorithm that is best suited to the computation over \(\mathbb{Q}\). \([13,14,15]\) presented algorithms for obtaining, with relatively little effort, the Gröbner éscalier of a vanishing ideal w.r.t. the (inverse) lexicographic order that can lead to an interpolation Newton basis or the reduced Gröbner basis for the vanishing ideal after solving a linear system.

For a fixed point set \(\Xi\) in \(\mathbb{F}^d\) and a term order \(\prec\), it is well known that there are \textit{two} factors that determine the Gröbner éscalier of \(I(\Xi)\) w.r.t. \(\prec\) thereby the reduced Gröbner basis for \(I(\Xi)\) and related degree reducing interpolation Newton bases (up to coefficients). One is apparently the cardinal of \(\Xi\). It is the unique determinate factor in univariate cases. Another one is the geometry (the distribution of the points) of \(\Xi\) that is dominating in multivariate cases but not taken into consideration by BM and MMM algorithms. Recent years, \([16,17,18]\) studied multivariate Lagrange interpolation on a special kind of point sets, cartesian point sets (aka lower point sets), and constructed the associated Gröbner éscalier and degree reducing interpolation Newton bases theoretically. We know from \([9,11]\) that, for a cartesian subset of \(\Xi\) (it always exists!), certain associated degree reducing interpoly-
tion Newton basis forms part of the output of BM algorithm w.r.t. some reordering of Ξ. Therefore, finding a large enough cartesian subset of Ξ with little enough effort will reduce the complexity of BM algorithm.

Following this idea, the paper proposes a preprocessing paradigm for BM algorithm with the organization as follows. The next section is devoted as a preparation for the paper. And then, the main results of us are presented in two sections. Section 3 will pursue the paradigm for two special term orders while Section 4 will set forth our solution for other more general cases. In the last section, Section 5, some implementation issues and experimental results will be illustrated.

2. Preliminary

In this section, we will introduce some notation and recall some basic facts for the reader’s convenience. For more details, we refer the reader to [19, 20].

We let \( \mathbb{N}_0 \) denote the monoid of nonnegative integers. A polynomial \( f \in \Pi^2 \) is of the form

\[
f = \sum_{\alpha \in \mathbb{N}_0^2} f_{\alpha} X^{\alpha}, \quad \#\{ \alpha \in \mathbb{N}_0^2 : 0 \neq f_{\alpha} \in \mathbb{F} \} < \infty,
\]

where \( \text{monomial } X^{\alpha} = x^{\alpha_1}y^{\alpha_2} \) with \( \alpha = (\alpha_1, \alpha_2) \). The set of bivariate monomials in \( \Pi^2 \) is denoted by \( \mathbb{T}^2 \).

Fix a term order \( \prec \) on \( \Pi^2 \) that may be lexicographical order \( \prec_{\text{lex}} \), inverse lexicographical order \( \prec_{\text{inlex}} \), or total degree inverse lexicographical order \( \prec_{\text{tdinlex}} \) etc. For all \( f \in \Pi^2 \), with \( f \neq 0 \), we may write

\[
f = f_{\gamma_1} X^{\gamma_1} + f_{\gamma_2} X^{\gamma_2} + \cdots + f_{\gamma_r} X^{\gamma_r},
\]

where \( 0 \neq f_{\gamma_i} \in \mathbb{F}, \gamma_i \in \mathbb{N}_0^2, i = 1, \ldots, r \), and \( X^{\gamma_1} \succ X^{\gamma_2} \succ \cdots \succ X^{\gamma_r} \). We shall call \( \text{LT}(f) := f_{\gamma_1} X^{\gamma_1} \) the leading term and \( \text{LM}(f) := X^{\gamma_1} \) the leading monomial of \( f \). Furthermore, for a non-empty subset \( F \subset \Pi^2 \), put

\[
\text{LT}(F) := \{ \text{LT}(f) : f \in F \}.
\]

As in [21], we define the \( \prec \)-degree of a polynomial \( f \in \Pi^2 \) to be the leading bidegree w.r.t. \( \prec \)

\[
\delta(f) := \gamma, \quad X^{\gamma} = \text{LM}(f),
\]
with \( \delta(0) \) undefined. Further, for any finite dimensional subset \( F \subset \Pi^2 \), define
\[
\delta(F) := \max_{f \in F} \delta(f).
\]
Finally, for any \( f, g \in \Pi^2 \), if \( \delta(f) \prec \delta(g) \) then we say that \( f \) is of lower degree than \( g \) and use the abbreviation
\[
f \prec g := \delta(f) \prec \delta(g).
\]
In addition, \( f \preceq g \) is interpreted as the degree of \( f \) is lower than or equal to that of \( g \).

Let \( \mathcal{A} \) be a finite subset of \( \mathbb{N}_{0}^2 \). \( \mathcal{A} \) is called a lower set if, for any \( \alpha = (\alpha_1, \alpha_2) \in \mathcal{A} \), we always have
\[
\mathcal{R}(\alpha) := \{ (\alpha'_1, \alpha'_2) \in \mathbb{N}_{0}^2 : 0 \leq \alpha'_i \leq \alpha_i, i = 1, 2 \} \subset \mathcal{A}.
\]
Especially, \( 0 \in \mathcal{A} \). Moreover, we set \( m_j = \max_{(h,j) \in \mathcal{A}} h, 0 \leq j \leq \nu \), with \( \nu = \max_{(0,k) \in \mathcal{A}} k \). Clearly, \( \mathcal{A} \) can be determined uniquely by the ordered \((\nu + 1)\)-tuple \((m_0, m_1, \ldots, m_\nu)\) hence represented as \( L_x(m_0, m_1, \ldots, m_\nu) \). Swapping the roles of \( x \) and \( y \), we can also represent \( \mathcal{A} \) as \( L_y(n_0, n_1, \ldots, n_m) \) with \( n_i = \max_{(i,k) \in \mathcal{A}} k, 0 \leq i \leq m_0 \). It should be noticed that \( \nu = n_0 \).

Given a set \( \Xi = \{ \xi^{(1)}, \ldots, \xi^{(\mu)} \} \subset \mathbb{R}^2 \) of \( \mu \) distinct points. For prescribed values \( f_i \in \mathbb{R}, i = 1, \ldots, \mu \), find all polynomials \( p \in \Pi^2 \) satisfying
\[
p(\xi^{(i)}) = f_i, \quad i = 1, \ldots, \mu. \tag{1}
\]
We call it the problem of bivariate Lagrange interpolation. Note that in most cases, especially from a numerical point of view, we are not interested in all such \( p \)'s but a “degree reducing” one, as in the univariate cases.

**Definition 1.** Fix term order \( \prec \). We call a subspace \( \mathcal{P} \subset \Pi^2 \) a degree reducing interpolation space w.r.t. \( \prec \) for the bivariate Lagrange interpolation \( \Pi \) if

**DR1.** \( \mathcal{P} \) is an interpolation space, i.e., for any \( f_i \in \mathbb{R}, i = 1, \ldots, \mu \), there is a unique \( p \in \mathcal{P} \) such that \( p \) satisfies \( \Pi \). In other words, the interpolation problem is regular w.r.t. \( \mathcal{P} \).

**DR2.** \( \mathcal{P} \) is \( \prec \)-reducing, i.e., when \( L_{\mathcal{P}} \) denotes the Lagrange projector with range \( \mathcal{P} \), then the interpolation polynomial
\[
L_{\mathcal{P}} q \preceq q, \quad \forall q \in \Pi^2.
\]
For interpolation problem (1), a given interpolation space $P \subset \Pi^2$ will give rise to an interpolation scheme that is referred as $(\Xi, P)$, cf. [20]. Since (1) is regular w.r.t. $P$, we can also say that $(\Xi, P)$ is regular. Moreover, if $P$ is degree reducing w.r.t. $\prec$, a basis $\{p_1, \ldots, p_\mu\}$ for $P$ will be called a degree reducing interpolation basis w.r.t. $\prec$ for (1). Assume that $p_1 \prec p_2 \prec \cdots \prec p_\mu$.

If $p_j(\xi^{(i)}) = \delta_{ij}, \quad 1 \leq i \leq j \leq \mu,$

for some suitable reordering of $\Xi$, then we call $\{p_1, \ldots, p_\mu\}$ a degree reducing interpolation Newton basis (DRINB) w.r.t. $\prec$ for (1).

Let $G_\prec$ be the reduced Gröbner basis for the vanishing ideal $I(\Xi)$ w.r.t. $\prec$. The set

$$N_\prec(I(\Xi)) := \{x^\alpha \in T^2 : \text{LT}(g) \not| x^\alpha, \forall g \in G_\prec\}$$

is called the Gröbner escalier of $I(\Xi)$ w.r.t. $\prec$. From [2, 21], the interpolation space spanned by $N_\prec(I(\Xi))$, denoted by $P_\prec(\Xi)$, is canonical since it is the unique degree reducing interpolation space spanned by monomials w.r.t. $\prec$ for (1). Hence, we call $N_\prec(I(\Xi))$ the degree reducing interpolation monomial basis (DRIMB) w.r.t. $\prec$ for (1), with $\#N_\prec(I(\Xi)) = \mu$.

Let $N_\prec(\Xi) := \{\alpha : x^\alpha \in N_\prec(I(\Xi))\} \subset N_2^0$.

We can deduce easily that $N_\prec(\Xi)$ is a lower set and obviously has a one-to-one correspondence with $N_\prec(I(\Xi))$. Therefore, interpolation scheme $(\Xi, P_\prec(\Xi))$ can be equivalently represented as $(\Xi, N_\prec(\Xi))$.

According to [17], we can construct two particular lower sets from $\Xi$, denoted by $S_x(\Xi), S_y(\Xi)$, which reflect the geometry of $\Xi$ in certain sense.

Specifically, we cover the points in $\Xi$ by lines $l_i^x, l_i^y, \ldots, l_i^x$ parallel to the $x$-axis and assume that, without loss of generality, there are $m_j + 1$ points, say $u_{0j}^x, u_{1j}^x, \ldots, u_{m_jj}^x$, on $l_i^x$ with $m_0 \geq m_1 \geq \cdots \geq m_\nu \geq 0$ hence the ordinates of $u_{ij}^x$ and $u_{ij}^x, i \neq i'$, same. Now, we set

$$S_x(\Xi) := \{(i, j) : 0 \leq i \leq m_j, \ 0 \leq j \leq \nu\},$$

which apparently equals to $L_x(m_0, m_1, \ldots, m_\nu)$. We can also cover the points by lines $l_0^y, l_1^y, \ldots, l_\nu^y$ parallel to the $y$-axis and denote the points on line $l_i^y$ by $u_{0i}^y, u_{1i}^y, \ldots, u_{ni,m_i}^y$ with $n_0 \geq n_1 \geq \cdots \geq n_\lambda \geq 0$ hence the abscissae of $u_{ij}^y$ and $u_{ij}^y, j \neq j'$, same. Similarly, we put

$$S_y(\Xi) := \{(i, j) : 0 \leq i \leq \lambda, \ 0 \leq j \leq n_i\} = L_y(n_0, n_1, \ldots, n_\lambda).$$
In addition, we can also define the sets of abscissae and ordinates

\[ H_j(\Xi) := \{ \bar{x} : (\bar{x}, \bar{y}) \in l_j^x \cap \Xi \}, \quad 0 \leq j \leq \nu, \]
\[ V_i(\Xi) := \{ \bar{y} : (\bar{x}, \bar{y}) \in l_i^y \cap \Xi \}, \quad 0 \leq i \leq \lambda. \]

(2)

Definition 2. [17] We say that a set \( \Xi \) of distinct points in \( \mathbb{F}^2 \) is Cartesian if there exists a lower set \( \mathcal{A} \) such that \( \Xi \) can be written as

\[ \Xi = \{(x_i, y_j) : (i, j) \in \mathcal{A}\}, \]

where the \( x_i \)'s are distinct numbers, and similarly the \( y_j \)'s. We also say that \( \Xi \) is \( \mathcal{A} \)-Cartesian.

To the best of our knowledge, there are two criteria for determining whether a 2-dimensional point set is Cartesian.

Theorem 1. [17] A set of distinct points \( \Xi \subset \mathbb{F}^2 \) is Cartesian if and only if \( S_x(\Xi) = S_y(\Xi) \).

Theorem 2. [18] A set of distinct points \( \Xi \subset \mathbb{F}^2 \) is Cartesian if and only if

\[ H_0(\Xi) \supseteq H_1(\Xi) \supseteq \cdots \supseteq H_\nu(\Xi), \quad V_0(\Xi) \supseteq V_1(\Xi) \supseteq \cdots \supseteq V_\lambda(\Xi). \]

About the bivariate Lagrange interpolation on a Cartesian set, [17] proved the succeeding theorem.

Theorem 3. [17] Given a Cartesian set \( \Xi \subset \mathbb{F}^2 \), there exists a unique lower set \( \mathcal{A} \in \mathbb{N}_0^2 \) such that \( \Xi \) is \( \mathcal{A} \)-Cartesian and the Lagrange interpolation scheme \((\Xi, \mathcal{A})\) is regular.

Finally, we will redescribe the classical BM algorithm with the notation established above.

Algorithm 1. (BM Algorithm)

Input: A set of distinct points \( \Xi = \{\xi^{(i)} : i = 1, \ldots, \mu\} \subset \mathbb{F}^d \) and a fixed term order \( \prec \).

Output: The 3-tuple \((G, N, Q)\), where \( G \) is the reduced Gröbner basis for \( \mathcal{I}(\Xi) \) w.r.t. \( \prec \), \( N \) is the Gröbner éscalier of \( \mathcal{I}(\Xi) \) (the DRIMB for \( \Pi \) also) w.r.t. \( \prec \), and \( Q \) is a DRINB w.r.t. \( \prec \) for \( \Pi \).

BM1. Start with lists \( G = [\ ], N = [\ ], Q = [\ ], L = [1] \), and a matrix \( B = (b_{ij}) \) over \( \mathbb{F} \) with \( \mu \) columns and zero rows initially.
BM2. If \( L = [ ] \), return \((G, N, Q)\) and stop. Otherwise, choose the monomial \( t = \min_\prec L \), and delete \( t \) from \( L \).

BM3. Compute the evaluation vector \((t(\xi^{(1)}), \ldots, t(\xi^{(\mu)}))\), and reduce it against the rows of \( B \) to obtain

\[
(v_1, \ldots, v_\mu) = (t(\xi^{(1)}), \ldots, t(\xi^{(\mu)})) - \sum_i a_i(b_{i1}, \ldots, b_{i\mu}), \quad a_i \in \mathbb{F}.
\]

BM4. If \((v_1, \ldots, v_\mu) = (0, \ldots, 0)\), then append the polynomial \( t - \sum_i a_i q_i \) to the list \( G \), where \( q_i \) is the \( i \)-th element of \( Q \). Remove from \( L \) all the multiples of \( t \). Continue with BM2.

BM5. Otherwise \((v_1, \ldots, v_\mu) \neq (0, \ldots, 0)\), add \((v_1, \ldots, v_\mu)\) as a new row to \( B \) and \( t - \sum_i a_i q_i \) as a new element to \( Q \). Append the monomial \( t \) to \( N \), and add to \( L \) those elements of \( \{x_1, \ldots, x_dt\} \) that are neither multiples of an element of \( L \) nor of \( \text{LT}(G) \). Continue with BM2.

3. Special cases

In this section, we will focus on \( \prec_{\text{lex}} \) and \( \prec_{\text{inlex}} \) that may be the most talked about term orders. For these special cases, our preprocessing paradigm will first provide exact \( N, Q \) of the 3-tuple output \((G, N, Q)\) to BM algorithm directly and effortlessly. And then, \( G \) can be obtained by BM algorithm easily. Note that we will continue with all the notation that we established for \( S_x(\Xi) \) and \( S_y(\Xi) \) in the previous section.

Proposition 4. Let \( \Xi \) be a set of \( \mu \) distinct points \( u_{mn}^x = (x_{mn}, y_{mn}) \in \mathbb{F}^2, (m, n) \in S_x(\Xi) \). The points give rise to polynomials

\[
\phi_{ij}^x = \varphi_{ij}^x \prod_{t=0}^{j-1} (y - y_{it}) \prod_{s=0}^{i-1} (x - x_{sj}), \quad (i, j) \in S_x(\Xi),
\]

where \( \varphi_{ij}^x = 1/\prod_{t=0}^{j-1} (y_{0t} - y_{jt}) \prod_{s=0}^{i-1} (x_{is} - x_{sj}) \in \mathbb{F} \), and the empty products are taken as 1. Then we have

\[
\phi_{ij}^x(u_{mn}^x) = \delta_{(i,j),(m,n)}, \quad (i, j) \succ_{\text{inlex}} (m, n).
\]

Proof. Fix \((i, j) \in S_x(\Xi)\). Recalling the definition of \( u_{ij}^x \), we have \( y_{0j} = y_{ij} \).
If \((i, j) = (m, n)\), by \( y_{00} \neq y_{01} \neq \cdots \neq y_{0j} \) and \( x_{0j} \neq x_{1j} \neq \cdots \neq x_{ij} \), we have

\[
\phi_{ij}^x(u_{ij}^x) = \varphi_{ij}^x \prod_{t=0}^{j-1} (y_{ij} - y_{0t}) \prod_{s=0}^{i-1} (x_{ij} - x_{sj}) = \varphi_{ij}^x \prod_{t=0}^{j-1} (y_{0t} - y_{00}) \prod_{s=0}^{i-1} (x_{0t} - x_{0s}),
\]

\[
= \varphi_{ij}^x \prod_{t=0}^{j-1} (y_{0t} - y_{00}) \prod_{s=0}^{i-1} (x_{0t} - x_{0s}) = \delta_{(0,0), (0,0)} = 1,
\]

where the empty products are taken as 1. If \((i, j) \neq (m, n)\), we have \( y_{0j} = y_{ij} \) and \( x_{0j} = x_{ij} \). Then \( \phi_{ij}^x(u_{ij}^x) = \varphi_{ij}^x \prod_{t=0}^{j-1} (y_{ij} - y_{0t}) \prod_{s=0}^{i-1} (x_{ij} - x_{sj}) = \varphi_{ij}^x \prod_{t=0}^{j-1} (y_{0t} - y_{00}) \prod_{s=0}^{i-1} (x_{0t} - x_{0s}),
\]

and the empty products are taken as 1. Therefore, \( \phi_{ij}^x(u_{ij}^x) = \delta_{(i,j),(m,n)}, \quad (i, j) \succ_{\text{inlex}} (m, n)\).
which implies $\phi^x_{ij}(u^x_{mn}) = 1$.

Otherwise, if $(i, j) \succ_{\text{inlex}} (m, n)$, we have $j > n$, or $j = n, i > m$. When $j > n$, we have

$$
\phi^x_{ij}(u^x_{mn}) = \phi^x_{ij}(y_{mn} - y_{00}) \cdots (y_{mn} - y_{0n}) \cdots (y_{mn} - y_{0,j-1}) \prod_{s=0}^{i-1} (x_{mn} - x_{sj})
$$

$$
= \phi^x_{ij}(y_{0n} - y_{00}) \cdots (y_{0n} - y_{0n}) \cdots (y_{0n} - y_{0,j-1}) \prod_{s=0}^{i-1} (x_{mn} - x_{sj})
$$

$$
= 0,
$$

and when $j = n, i > m$,

$$
\phi^x_{ij}(u^x_{mn}) = \phi^x_{ij} \prod_{t=0}^{j-1} (y_{mn} - y_{0t})(x_{mn} - x_{0j}) \cdots (x_{mn} - x_{mj}) \cdots (x_{mn} - x_{i-1,j})
$$

$$
= \phi^x_{ij} \prod_{t=0}^{n-1} (y_{mn} - y_{0t})(x_{mn} - x_{0n}) \cdots (x_{mn} - x_{mn}) \cdots (x_{mn} - x_{i-1,n})
$$

$$
= 0,
$$

which leads to

$$
\phi^x_{ij}(u^x_{mn}) = 0, \quad (i, j) \succ_{\text{inlex}} (m, n).
$$

Similarly, we can prove the following proposition:

**Proposition 5.** Let $\Xi$ be a set of $\mu$ distinct points $u^y_{mn} = (x_{mn}, y_{mn}) \in \mathbb{F}^2, (m, n) \in S_y(\Xi)$. We define the polynomials

$$
\phi^y_{ij} = \varphi^y_{ij} \prod_{s=0}^{i-1} (x - x_{s0}) \prod_{t=0}^{j-1} (y - y_{it}), \quad (i, j) \in S_y(\Xi),
$$

(4)

where $\varphi^y_{ij} = 1/\prod_{s=0}^{i-1} (x_{i0} - x_{s0}) \prod_{t=0}^{j-1} (y_{ij} - y_{it}) \in \mathbb{F}$. The empty products are taken as 1. Then,

$$
\phi^y_{ij}(u^y_{mn}) = \delta_{(i,j),(m,n)}, \quad (i, j) \succeq_{\text{lex}} (m, n).
$$

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In 2004, [17] proved that the Lagrange interpolation schemes $(\Xi, S_x(\Xi))$ and $(\Xi, S_y(\Xi))$ are both regular. Here we reprove the regularities in another way for the purpose of presenting the degree reducing interpolation bases theoretically.

**Theorem 6.** Resume the notation in Proposition 4 and 5. Then the Lagrange interpolation schemes $(\Xi, S_x(\Xi))$ and $(\Xi, S_y(\Xi))$ are regular. Furthermore,

(i) the set $N_x := \{x^iy^j : (i, j) \in S_x(\Xi)\}$ is the DRIMB as well as $Q_x := \{\phi^x_{ij} : (i, j) \in S_x(\Xi)\}$ is a DRINB w.r.t. $\prec_{\text{lex}}$ for the interpolation problem (II).

(ii) the set $N_y := \{x^iy^j : (i, j) \in S_y(\Xi)\}$ is the DRIMB as well as $Q_y := \{\phi^y_{ij} : (i, j) \in S_y(\Xi)\}$ is a DRINB w.r.t. $\prec_{\text{inlex}}$ for (II).

**Proof.** We only give the proof for $S_x(\Xi)$. The statements about $S_y(\Xi)$ can be proved likewise.

First, we will show the regularity of the interpolation scheme $(\Xi, S_x(\Xi))$.

Let $P_x := \text{Span}_xN_x \subset \Pi^2$ with $\dim P_x = \#\Xi = \mu$. Obviously, $N_x$ is the monomial basis for it. By (3), we can check easily that $\text{Span}_xQ_x \subseteq P_x$.

Construct a square matrix $B_{\mu \times \mu}$ whose $(h, k)$ entry is $\phi^x_h(u^x_k)$ where $\phi^x_h, u^x_k$ are $h$th and $k$th elements of $Q_x$ and $\Xi = \{u^x_{mn} : (m, n) \in S_x(\Xi)\}$ w.r.t. the increasing $\prec_{\text{inlex}}$ on $(i, j)$ and $(m, n)$ respectively. From Proposition 3, $B_{\mu \times \mu}$ is upper unitriangular which implies that $\text{Span}_xQ_x = P_x$ and $Q_x$ forms a Newton basis for $P_x$. It follows that $P_x$ is an interpolation space for Lagrange interpolation (II) therefore the scheme $(\Xi, P_x)$ is regular. Since $(\Xi, S_x(\Xi)) = (\Xi, P_x)$, according to Section 2, $(\Xi, S_x(\Xi))$ is regular.

Next, we shall verify that the statements in (i), which is equivalent to the statement that $P_x$ is a degree reducing interpolation space w.r.t. $\prec_{\text{lex}}$ for (II) that coincides with $P_{\prec_{\text{lex}}}$. Since the arguments above have proved that $P_x$ satisfies the DR1 condition in Definition 1, what is left for us is to check the DR2 condition. From [21], we only need to check it for monomials.

Take a monomial $x^iy^j \in T^2$. We shall prove that

$$L_{P_x}x^iy^j \preceq_{\text{lex}} x^iy^j.$$  

Since $P_x$ satisfies DR1, $L_{P_x}x^iy^j$ is the unique polynomial in $P_x$ that matches $x^iy^j$ on $\Xi$. Therefore, when $x^iy^j \in N_x$, we have $L_{P_x}x^iy^j = x^iy^j$, 

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has a unique solution. Assume that
\[ S_x(\Xi) = L_x(m_0, \ldots, m_{n_0}) = L_y(n_0, \ldots, n_{n_0}). \]
It is easy to see that \( \delta(\mathcal{P}_x) = (m_0, n_{n_0}) \). If \( x^{m_0} y^{n_{n_0}} \prec_{\text{lex}} x^{i_0} y^{j_0} \) then \( \delta(L_{\mathcal{P}_x} x^{i_0} y^{j_0}) \prec_{\text{lex}} \delta(\mathcal{P}_x) = (m_0, n_{n_0}) \prec_{\text{lex}} (i_0, j_0) = \delta(x^{i_0} y^{j_0}) \) that leads to (1) for the case.

Thus, what remains for us is to check (3) for \( x^{i_0} y^{j_0} \not\in N_x \) with \( (i_0, j_0) \prec_{\text{lex}} (m_0, n_{n_0}) \) that implies \( 0 \leq i_0 < m_0, j_0 > n_{n_0} \). For this, we only need to verify that
\[ L_{\mathcal{P}_x} x^{i_0} y^{j_0} \in \text{Span}_F \{ x^i y^j : (i, j) \in F_{i_0} \}, \]
where \( F_{i_0} = \{ (i, j) \in S_x(\Xi) : (i, j) \prec_{\text{lex}} (i_0, j_0) \} \subset S_x(\Xi) \). If \( x^{i_0} y^{j_0} \in \mathcal{I}(\Xi) \), then \( L_{\mathcal{P}_x} x^{i_0} y^{j_0} = 0 \prec_{\text{lex}} x^{i_0} y^{j_0} \). The statement (6) becomes trivial in this case. Otherwise, if we can find a polynomial \( p \in \Pi^2 \) such that
\[ p = x^{i_0} y^{j_0} - \sum_{(i,j) \in F_{i_0}} a_{ij} x^i y^j \in \mathcal{I}(\Xi), \]
where \( a_{ij} \in \mathbb{F} \) are not all zero, then (6) follows.

According to Section 2, our point set \( \Xi = \{ u_{ij}^x = (x_{ij}, y_{ij}) : (i, j) \in S_x(\Xi) \} \). Let \( \Xi' = \{ u_{mn}^x \in \Xi : (m, n) \in F_{i_0} \} \subset \Xi \). Now, we claim that there exists a unique polynomial \( p \) of the form (7) such that \( p \in \mathcal{I}(\Xi') \), which is equivalent to the statement that the linear system
\[ \sum_{(i,j) \in F_{i_0}} a_{ij} x^i_{mn} y^j_{mn} = x^{i_0}_{mn} y^{j_0}_{mn}, \quad u_{mn}^x \in \Xi', \]
has a unique solution.

Note that \( \text{Span}_F \{ x^i y^j : (i, j) \in F_{i_0} \} = \text{Span}_F \{ \phi_{ij}^x : (i, j) \in F_{i_0} \} \). We can conclude that the rank of the coefficient matrix of (8) is equal that of the matrix \( B'_{\#F_{i_0} \times \#F_{i_0}} \), which is a submatrix of \( B \) whose \((h, k)\) entry is \( \phi_{hk}^x(u_{ij}^x) \) where \( \phi_{hk}^x, u_{ij}^x \) are \( h \)th and \( k \)th elements of \( \{ \phi_{ij}^x : (i, j) \in F_{i_0} \} \) and \( \Xi' = \{ u_{mn}^x \} \) w.r.t. the increasing \( \prec_{\text{lex}} \) on \((i, j)\) and \((m, n)\) respectively. By (3), we see easily that \( B' \) is upper unitriangular which implies that the coefficient matrix of (8) is of full rank. Accordingly, there is a unique polynomial \( p \in \mathcal{I}(\Xi') \) that has the form (7).
Now we shall verify that $p(u^x_{ij}) = 0$, $u^x_{ij} \in \Xi \setminus \Xi'$. By the definition of $\Xi'$, we know that $i > i_0$ here. Let

$$q(x) := p(x, y_{ij}) = \sum_{s=0}^{i_0} b_s x^s \in \Pi^1, \quad b_s \in \mathbb{F}.$$ 

Since $y_{0j} = y_{1j} = \cdots = y_{i_0,j} = y_{ij}$ and $u^x_{0j}, u^x_{1j}, \ldots, u^x_{i_0,j} \in \Xi'$, it follows that

$$q(x_s) = p(x_s, y_{ij}) = p(x_s, y_{sj}) = p(u^x_{sj}) = 0, \quad s = 0, \ldots, i_0,$$

namely $q(x)$ has $i_0 + 1$ zero points that clearly implies $q(x) \equiv 0$. Since $p(u^x_{ij}) = q(x_{ij}) = 0$, we have $p \in \mathcal{I}(\Xi)$. By (6), (5) is true in this case. As a result, for any $f \in \Pi^2$, we have

$$L_{\mathcal{P}_x} f \geq_{\text{lex}} f,$$

that is to say $\mathcal{P}_x$ satisfies DR2.

Consequently, by Definition 1, $\mathcal{P}_x$ is a degree reducing interpolation space w.r.t. $\prec_{\text{lex}}$ for Lagrange interpolation (11). Hence $N_x$ is the DRIMB and $Q_x$ is a Newton basis w.r.t. $\prec_{\text{inlex}}$ for (11).

Note that $\mathcal{P}_{\prec_{\text{inlex}}}(\Xi)$ is the unique degree reducing interpolation space spanned by monomials w.r.t. $\prec_{\text{lex}}$, thus we have $\mathcal{P}_x = \mathcal{P}_{\prec_{\text{inlex}}}(\Xi)$. Therefore, $N_x = N_{\prec_{\text{inlex}}}(\mathcal{I}(\Xi))$ holds, which means that $N_x$ is also the Gröbner escalier of $\mathcal{I}(\Xi)$ w.r.t. $\prec_{\text{lex}}$.

Corollary 7. If $\Xi \subset \mathbb{F}^2$ is an $A$-cartesian set, then $A = S_x(\Xi) = S_y(\Xi)$.

Proof. Since $\Xi$ is cartesian, by Theorem 1 and 6, we have $S_x(\Xi) = S_y(\Xi)$ hence $(\Xi, S_x(\Xi)) = (\Xi, S_y(\Xi))$ are both regular. But from Theorem 3, only $A$ can make $(\Xi, A)$ regular, therefore $A = S_x(\Xi) = S_y(\Xi)$.

From Algorithm 1 we know that $G, N, Q$ are essential elements of BM algorithm and compose its output. For $\prec_{\text{lex}}$ and $\prec_{\text{inlex}}$ cases, Theorem 6 presents us $N$ and $Q$ theoretically hence we can obtain them with little effort. According to [11], the leading terms of $G$ are contained in the border set of $N$. Therefore, we can get $G$ faster than compute $G$ directly with BM algorithm. Now is our algorithm.
Algorithm 2. (SPBM)

Input: A set of distinct affine points \( \Xi \subset \mathbb{F}^2 \) and fixed \( \prec_{\text{lex}} \) or \( \prec_{\text{inlex}} \).

Output: The 3-tuple \((G, N, Q)\), where \( G \) is the reduced Gröbner basis of \( I(\Xi) \), \( N \) is the Gröbner escalier \( N(I(\Xi)) \), and \( Q \) is a DRINB for the Lagrange interpolation on \( \Xi \).

SPBM1. Construct lower set \( S_x(\Xi) \) or \( S_y(\Xi) \) according to Section 2.

SPBM2. Compute the sets \( N \) and \( Q \) by Theorem 6.

SPBM3. Construct the border set \( L := \{x \cdot t : t \in N\} \cup \{y \cdot t : t \in N\} \setminus N \) and the matrix \( B \) that is same to the \( B_{\mu \times \mu} \) in the proof of Theorem 6.

SPBM4. Goto BM2 of BM algorithm for the reduced Gröbner basis \( G \).

Example 1. Let

\[ \Xi = \{(0, 1), (0, 3), (1, 0), (1, 2), (1, 3), (1, 4), (2, 1), (2, 2), (3, 1)\} \subset \mathbb{Q}^2. \]

First, we choose lines \( x = 1, x = 0, x = 2, x = 3 \) as \( l_0^y, l_1^y, l_2^y, l_3^y \) respectively (Shown in (a) of Figure 1), therefore we have

\[ S_y = \{(0, 0), (0, 1), (0, 2), (0, 3), (1, 0), (1, 1), (2, 0), (2, 1), (3, 0)\}, \]

which is illustrated in (b) of Figure 1.

![Figure 1: The point set and related \( S_y \) of Example 1.](image-url)
Thus, by Theorem 6, we have
\[ N = \{1, y, y^2, y^3, x, xy, x^2y, x^3\}; \]
\[ Q = \left\{ \frac{1}{2}y - \frac{2}{3}y^2 - \frac{2}{3}, \frac{1}{8}y^3 - \frac{5}{4}y^2 + x + 1, -\frac{1}{2}xy + \frac{1}{2}y + \frac{1}{2}x - \frac{1}{2}, \right. \]
\[ \frac{1}{2}x^2 - \frac{1}{2}x, -\frac{1}{2}x^2y - \frac{1}{2}x^2 + \frac{1}{2}x, \frac{1}{6}x^3 - \frac{1}{2}x^2 + \frac{1}{3}, \}

Next, from SPBM3, the border set \( L = \{y^4, xy^2, x^2y^3, x^2y, x^3y, x^4\} \) and the matrix
\[
B = \begin{pmatrix}
1 & 1 & 1 & \cdots \\
0 & 1 & 3/2 & \cdots \\
0 & 0 & 1 & \cdots \\
\vdots & \vdots & \vdots & \ddots 
\end{pmatrix}.
\]

Finally, turn to BM2 with these \( N, Q, L, B \), we can get the reduced Gröbner basis
\[
G = \{x^4 - 6x^3 + 11x^2 - 6x, x^3y - 3x^2y + 2xy - x^3 + 3x^2 - 2x, \]
\[
xy^2 - y^2 + \frac{1}{2}x^2y - \frac{9}{2}xy + 4y - \frac{1}{2}x^2 + \frac{7}{2}x - 3, \\
y^4 - 9y^3 + 26y^2 - \frac{9}{2}x^2y - \frac{15}{2}xy - 27y - 3x^3 + \frac{39}{2}x^2 - \frac{51}{2}x + 9\}.
\]

for \( \mathcal{I}(\Xi) \) w.r.t. \( \prec_{\text{inlex}} \).

**Example 2.** Given a bivariate point set
\[
\Xi = \{(0, 0), (0, 2), (0, 3), (1, 1), (\frac{5}{2}, 0), (\frac{5}{2}, 1), (\frac{5}{2}, 2), (4, 0), (4, 2)\} \subset \mathbb{Q}^2.
\]

We choose lines \( y = 0, y = 2, y = 1, y = 3 \) as \( l_0^\varepsilon, l_1^\varepsilon, l_2^\varepsilon, l_3^\varepsilon \) respectively (Illustrated in (a) of Figure 2), which follows that
\[
S_x = \{(0, 0), (1, 0), (2, 0), (0, 1), (1, 1), (2, 1), (0, 2), (1, 2), (0, 3)\}.
\]
Thus, with SPBM algorithm, we have

\[ N = \{1, x, x^2, y, xy, x^2y, y^2, xy^2, y^3\} , \]

\[ Q = \{1, \frac{1}{4} x, -\frac{4}{15} x^2 + \frac{16}{15} x, \frac{1}{2} y, \frac{1}{8} xy, -\frac{2}{15} x^2y + \frac{8}{15} xy, -y^2 + 2y, \]

\[ -\frac{2}{3} xy^2 + \frac{2}{3} y^2 + \frac{4}{3} xy - \frac{4}{3} y, \frac{1}{6} y^3 - \frac{1}{2} y^2 + \frac{1}{3} y \} , \]

\[ G = \{y^4 - 6y^3 + 11y^2 - 6y, xy^3 - 3xy^2 + 2xy, x^2y^2 - 2x^2y - \frac{7}{2} xy^2 \]

\[ + 7xy - \frac{5}{4} y^3 + \frac{25}{4} y^2 - \frac{15}{2} y, \frac{15}{2} x^3 - \frac{13}{2} x^2 - 3xy^2 + 6xy \]

\[ + 10x - \frac{15}{4} y^3 + \frac{75}{4} y^2 - \frac{45}{2} y \} . \]

4. General cases

Next, we will discuss how to accelerate BM algorithm with respect to term orders other than \( \prec_{\text{lex}} \) or \( \prec_{\text{inlex}} \). In [17], the author proposed that if the set of points \( \Xi \) is cartesian, then we can obtain the interpolation basis without any difficulty, see Theorem 3. But in general \( \Xi \) may not be cartesian. However, we have the following proposition.

**Proposition 8.** There must exist at least one cartesian subset for any non-empty set of points in \( \mathbb{R}^2 \).
Proof. Let $\Xi$ be a non-empty set of points. Hence, there exists at least one point $\xi \in \Xi$. But $\xi$ itself can construct a cartesian subset $\{\xi\} \subset \Xi$. 

Definition 3. Let $\Xi$ be a set of points in $\mathbb{F}^2$ and $\Xi'$ be a cartesian subset of $\Xi$. We say that $\Xi'$ is a maximal cartesian subset of $\Xi$ if any cartesian proper subset $\Xi''$ of $\Xi$ containing $\Xi'$ is such that $\Xi'' = \Xi'$. In addition, a maximal row subset of $\Xi$ is a non-empty subset that equals the intersection of $\Xi$ and a horizontal line.

From Proposition 8 we know that, for a set of given points, we can surely find a maximal cartesian subset of it. Is it unique? Unfortunately, the answer is often false.

Example 3. Recall Example 2, let

$$\Xi_1' = \{(0,0), (0,2), (\frac{5}{2}, 0), (\frac{5}{2}, 1), (\frac{5}{2}, 2), (4,0), (4,2)\},$$

$$\Xi_2' = \{(0,0), (0,2), (0,3), (\frac{5}{2}, 0), (\frac{5}{2}, 2), (4,0), (4,2)\},$$

$$\Xi_3' = \{(1,1), (\frac{5}{2}, 0), (\frac{5}{2}, 1), (\frac{5}{2}, 2)\}.$$

We can check easily that $\Xi_1', \Xi_2', \Xi_3'$ are all maximal cartesian subsets of $\Xi$ (Illustrated in Figure 3).

Lemma 9. Let $\Xi$ be a set of distinct points in $\mathbb{F}^2$ and $\prec$ a fixed term order. If $\Xi'$ is an $\mathcal{A}'$-cartesian subset of $\Xi$, then

$$\mathcal{A}' = N_\prec(\Xi') \subset N_\prec(\Xi),$$

or equivalently,

$$\{x^iy^j : (i, j) \in \mathcal{A}'\} = N_\prec(\mathcal{I}(\Xi')) \subset N_\prec(\mathcal{I}(\Xi)).$$

Proof. From Section 2 the Gröbner écalier $N_\prec(\mathcal{I}(\Xi'))$ is the DRIMB w.r.t. $\prec$ for the bivariate Lagrange interpolation on $\Xi'$ hence the interpolation scheme $(\Xi', N_\prec(\Xi'))$ is regular. Since $\mathcal{A}' \subset N_0^2$ is lower and $\Xi'$ is $\mathcal{A}'$-cartesian, according to Theorem 3 $\mathcal{A}'$ is the unique lower set making the bivariate Lagrange interpolation on $\Xi'$ regular. This gives

$$\mathcal{A}' = N_\prec(\Xi').$$
Figure 3: Maximal cartesian subsets of $\Xi$, where $\bullet$ denotes the points in $\Xi'_i$, $i = 1, 2, 3$, while $\circ$ denotes the points in $\Xi \setminus \Xi'_i$. 
Since $\Xi' \subset \Xi$, from [19], we know that the vanishing ideals satisfy $I(\Xi') \supset I(\Xi)$. Denote by $G', G$ the reduced Gröbner bases for $I(\Xi')$ and $I(\Xi)$ w.r.t. $\prec$ respectively. We will prove $N_\prec(I(\Xi')) \subset N_\prec(I(\Xi))$ by contradiction. For any $x^iy^j \in N_\prec(I(\Xi'))$, we suppose there were some $g \in G$ such that $\text{LT}(g)|x^iy^j$. By [19],

$$\langle \text{LT}(G') \rangle = \langle \text{LT}(I(\Xi')) \rangle \supset \langle \text{LT}(I(\Xi)) \rangle \supset \langle \text{LT}(G) \rangle.$$ 

Therefore, $\text{LT}(g) \in \text{LT}(G) \subset \langle \text{LT}(G') \rangle$ implies that there exists some $g' \in G'$ such that $\text{LT}(g')|\text{LT}(g)$. Since $\text{LT}(g')|x^iy^j$, we have $\text{LT}(g')|x^iy^j$ that contradicts our assumption on $x^iy^j$, which proves that $N_\prec(I(\Xi')) \subset N_\prec(I(\Xi))$ due to the definition of $N_\prec(I(\Xi))$. Finally, $N_\prec(\Xi') \cong N_\prec(I(\Xi'))$ and $N_\prec(\Xi) \cong N_\prec(I(\Xi))$ complete the proof. □

Remark 1. For any $A$-cartesian set $\Xi$, by Corollary [17], we have $A = S_x(\Xi) = S_y(\Xi)$ that obviously leads to $A = S_x(\Xi) = S_y(\Xi) = N_\prec(\Xi)$, according to the Lemma above, where term order $\prec$ is arbitrary.

Now comes an algorithm for constructing a maximal cartesian subset of a given point set in $\mathbb{F}^2$.

Algorithm 3. (Maximal Cartesian Subset Construction Algorithm)

**Input**: A set of distinct points $\Xi = \{\xi^{(i)} : i = 1, \ldots, \mu\} \subset \mathbb{F}^2$.

**Output**: A maximal cartesian subset $\Xi'$ of $\Xi$.

**MCS1**. Start with an empty list $\Xi' = [\ ]$.

**MCS2**. If $\Xi = [\ ]$, return the set $\Xi'$ and stop. Otherwise, compute lower sets $S_x(\Xi)$ and $S_y(\Xi)$.

**MCS3**. If $S_x(\Xi) = S_y(\Xi)$, then replace $\Xi'$ by $\Xi' \cup \Xi$, return the set $\Xi'$ and stop.

**MCS4**. Otherwise, we first choose a maximal row subset of $\Xi$ with maximal cardinal number, denoted by $A$. Next, delete from $\Xi$ the points either in $A$ or have different abscissae from the points in $A$. Finally, replace $\Xi'$ by $\Xi' \cup A$ and continue with **MCS2**.

The following theorem ensure that this algorithm will terminate in finite steps with a maximal cartesian subset as its output.

**Theorem 10**. The algorithm described above will stop in a finite number of loops. Furthermore, the set $\Xi'$ returned by the algorithm is a maximal cartesian subset.
PROOF. As input data of Algorithm\[3\] point set $\Xi$ is finite. Observing that $\#\Xi$ decreases actually in every loop, the algorithm will terminate in a finite number, say $M$, of loops for sure. We assume that $M > 1$ since $M = 1$ is trivial.

$\Xi'_\text{in}$ and $\Xi'_\text{out}$ signify the input and output $\Xi'$ of MCS4 in some loop respectively. Next, we will prove by induction on $1 \leq r \leq M - 1$ that in the $r$th loop $\Xi'_\text{out}$ is a cartesian set. The case $r = 1$ is obvious since $\Xi'_\text{in} = \emptyset$ and $\Xi'_\text{out}$ is clearly cartesian as a maximal row subset of $\Xi$. Assume the statement is true for $r = l < M - 1$. When $r = l + 1$, by the induction hypothesis, $\Xi'_\text{in}$ is cartesian. Therefore, by Corollary\[7\] we assume that

$$\Xi'_\text{in} = \{(x_i, y_j) : (i, j) \in S_x(\Xi'_\text{in})\},$$

where $S_x(\Xi'_\text{in}) = L_x(m_0, \ldots, m_{m_0}) = L_y(n_0, \ldots, n_{m_0})$. Observing the construction process of $\Xi'$ in the algorithm, we see easily that $n_0 = n_1 = \cdots = n_{m_0}$. Let the maximal row subset of $\Xi$ we choose at this moment be $A = \{(\overline{x}(0), \overline{y}), (\overline{x}(1), \overline{y}), \ldots, (\overline{x}(k), \overline{y})\}$. Due to the nature of $A$, we have $k \leq m_0$ and $\overline{y} \neq y_j$, $j = 0, \ldots, n_0$.

We claim that the set $\Xi'_\text{in} \cup A$ is cartesian. In fact, we will focus on the horizontal parallel lines $l^x_i : y = y_j, j = 0, \ldots, n_0$, and $l^x_{n_0+1} : y = \overline{y}$. Resume the notation in\[2\]. $H_j(\Xi'_\text{in} \cup A) = H_j(\Xi'_\text{in}) = \{x_i : 0 \leq i \leq m_j\}, j = 0, \ldots, n_0$, and $H_{n_0+1}(\Xi'_\text{in} \cup A) = \{\overline{x}(i) : 0 \leq i \leq k\}$. Since $\Xi'_\text{in}$ is $S_x(\Xi'_\text{in})$-cartesian, by Theorem\[2\] the relation $H_0(\Xi'_\text{in} \cup A) \supseteq H_1(\Xi'_\text{in} \cup A) \supseteq \cdots \supseteq H_{n_0}(\Xi'_\text{in} \cup A)$ holds. From the description of MCS4, we can deduce that $H_{n_0}(\Xi'_\text{in} \cup A) \supseteq H_{n_0+1}(\Xi'_\text{in} \cup A)$, which leads to

$$H_0(\Xi'_\text{in} \cup A) \supseteq H_1(\Xi'_\text{in} \cup A) \supseteq \cdots \supseteq H_{n_0+1}(\Xi'_\text{in} \cup A).$$

Note that for any $\overline{x}(i), 0 \leq i \leq k$, there exists $h_i \in \{0, 1, \ldots, m_0\}$ such that $\overline{x}(i) = x_{h_i}$. Therefore, we could find a permutation $\sigma$ of $\{0, 1, \ldots, m_0\}$ satisfying $\sigma(i) = h_i, i = 0, \ldots, k$, and $\sigma(i) = i, i = m_{m_0} + 1, \ldots, m_0$. Choose lines $l^x_i : x = x_{\sigma(i)}, i = 0, \ldots, m_0$, that give rise to $V_i(\Xi'_\text{in}) = \{y_j : 0 \leq j \leq n_{\sigma(i)}, i = 0, \ldots, m_0\}$. Since $n_0 = n_1 = \cdots = n_{m_0}$, the relation $V_0(\Xi'_\text{in}) = V_1(\Xi'_\text{in}) = \cdots = V_{m_{m_0}}(\Xi'_\text{in}) \supseteq V_{m_{m_0}+1}(\Xi'_\text{in}) \supseteq \cdots \supseteq V_{m_0}(\Xi'_\text{in})$ holds. Observing that $V_i(\Xi'_\text{in} \cup A) = V_i(\Xi'_\text{in} \cup \{\overline{y}\}), i = 0, \ldots, k$, and $V_k(\Xi'_\text{in} \cup A) = V_k(\Xi'_\text{in})$, $i = k + 1, \ldots, m_0$, it is easy to get

$$V_0(\Xi'_\text{in} \cup A) \supseteq \cdots \supseteq V_k(\Xi'_\text{in} \cup A) \supseteq V_{k+1}(\Xi'_\text{in} \cup A) \supseteq \cdots \supseteq V_{m_0}(\Xi'_\text{in} \cup A).$$

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Thus together with (9), $\Xi'_{\text{out}} = \Xi'_{\text{in}} \cup A$ is cartesian due to Theorem 2, hence our statement is true.

For the $M$th loop, if $\Xi = [],$ then $\Xi'$ here equals to the $\Xi'_{\text{out}}$ of the MCS4 of the $(M - 1)$th loop that is cartesian due to the statement above. Otherwise, since the algorithm stops in MCS3 of this loop, $\Xi$ is a non-empty cartesian set. Similar to the arguments above, we can prove that $\Xi' = \Xi'_{\text{out}} \cup \Xi$ is also cartesian.

Finally, we should verify that the output $\Xi'$ of the algorithm is maximal. Otherwise, there must exist a maximal $S_x(\Xi'')$-cartesian subset $\Xi''$ of $\Xi$ satisfying $\Xi'' \supsetneq \Xi'$. Take a point $\xi_0 = (x_{i_0}, y_{j_0})$ with $(i_0, j_0) = \min_{\text{inlex}} \{(i, j) \in S_x(\Xi'') : (x_i, y_j) \in \Xi'' \setminus \Xi\}$. Suppose there exists a point in $\Xi'$ sharing the ordinate with $\xi_0$. If it is chosen as a point in the maximal row subset in MCS4 of some loop, by the definition of $\xi_0$, we know that $\xi_0$ is surely contained in the set $\Xi$ of that step, which contradicts the definition of the maximal row subset. Otherwise, it must appear in the cartesian set $\Xi$ in MCS3 in the final loop. Then, by the definition of $\xi_0$, it should be contained in $\Xi$ hence the output set $\Xi'$, which introduces a contradiction. If there does not exist a point in $\Xi'$ sharing the ordinate with $\xi_0$, since $\Xi''$ is also cartesian, by Theorem 2, it is easily to see that $\xi_0$ must remain in $\Xi$ in every loop, which contradicts the termination condition. As a result, the output of the Algorithm 3 is a maximal cartesian subset.

Let us continue with the setup and notation in Algorithm 3 and assume that the final output of it is $\Xi'$ who is $S_x(\Xi')$-cartesian. We now discuss how to preprocess the BM algorithm with the help of $\Xi'$.

Define an order $\prec_{\Xi}$ on the set $\Xi$. Let $\xi^{(1)}, \xi^{(2)} \in \Xi$. We say that $\xi^{(1)} \prec_{\Xi} \xi^{(2)}$ if one of the following conditions holds:

1. $\xi^{(1)} \in \Xi'$ and $\xi^{(2)} \in \Xi \setminus \Xi'$.
2. $\xi^{(1)} = (x_{i_1}, y_{j_1}), \xi^{(2)} = (x_{i_2}, y_{j_2}) \in \Xi'$ and $(i_1, j_1) \prec_{\text{inlex}} (i_2, j_2)$ with $(i_k, j_k) \in S_x(\Xi'), k = 1, 2$.

It should be noticed that the order is not total. For the points in $\Xi \setminus \Xi'$, any order of them can be interpreted as increasing. Hereafter, we will suppose that the points in $\Xi = \{\xi^{(1)}, \ldots, \xi^{(#\Xi)}\}$ have been ordered increasingly w.r.t. $\prec_{\Xi}$, namely $\xi^{(i)} \prec_{\Xi} \xi^{(j)}, 0 \leq i < j \leq #\Xi$. By the definition of $\prec_{\Xi}$, we have $\Xi' = \{\xi^{(1)}, \ldots, \xi^{(#\Xi')}\}$. 

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According to Lemma 9, \( N' = \{ x^i y^j : (i, j) \in S_x(\Xi') \} \subset N \), with \( N \) as a member of the 3-tuple output of BM algorithm. Thus the other monomials of \( N \) are obviously contained in \( T_2 \setminus N' \). Notice that the generators of \( T_2 \setminus N' \) are located in the border of \( N' \), denoted by \( L \), we can continue to spot the elements in \( L \) by BM algorithm to complete \( N \).

Next, we will pay attention to the computation of the Newton basis. Since \( \Xi' \) is cartesian, recalling Proposition 4, we can construct the polynomials \( \phi_{ij} \) w.r.t. \( S_x(\Xi') \). Order \( \phi_{ij} \), \((i, j) \in S_x(\Xi') \), increasingly w.r.t. \((i, j)\) under \( \prec_{\text{inlex}} \), denoted them as \( q_1, q_2, \ldots, q_{#\Xi'} \). Set the matrix

\[
B = \begin{pmatrix}
q_1(\xi^{(1)}) & q_1(\xi^{(2)}) & \cdots & q_1(\xi^{(\#\Xi')}) \\
q_2(\xi^{(1)}) & q_2(\xi^{(2)}) & \cdots & q_2(\xi^{(\#\Xi')}) \\
\vdots & \vdots & \ddots & \vdots \\
q_{\#\Xi'}(\xi^{(1)}) & q_{\#\Xi'}(\xi^{(2)}) & \cdots & q_{\#\Xi'}(\xi^{(\#\Xi')})
\end{pmatrix}.
\tag{10}
\]

By Proposition 4, \( B \) is obviously upper unitriangular which implies that the polynomials \( q_1, q_2, \ldots, q_{\#\Xi'} \) constitute a Newton basis for \( P_{\prec}(\Xi') = \text{Span}_F N' \).

All in all, with the notation above, we get our preprocessing procedure for BM algorithm.

**Algorithm 4.** (GPBM)

**Input:** A set of distinct points \( \Xi \subset \mathbb{F}^2 \) and a term order \( \prec \).

**Output:** The 3-tuple \( (G, N, Q) \).

- **GPBM1:** Get a maximal cartesian subset \( \Xi' \) of \( \Xi \) by Algorithm 3.
- **GPBM2:** Compute the lower set \( S_x(\Xi') \) w.r.t. \( \Xi' \), the set \( N := \{ x^i y^j : (i, j) \in S_x(\Xi') \} \), and the set \( Q := \{ q_1, q_2, \ldots, q_{\#\Xi} \} \) where the \( q_i \)'s are as in (10).
- **GPBM3:** Construct \( L := \{ x \cdot t : t \in N \} \cup \{ y \cdot t : t \in N \} \setminus N \) and the matrix \( B \) that is same to (10).
- **GPBM4:** Goto BM2 of the BM algorithm to complete the computation and get the whole output.

5. Implementation and Timings

From the above section, we can see easily that our preprocessing paradigm is more suitable to the cases where the constructed maximal cartesian subset \( \Xi' \) forms a relatively large proposition in \( \Xi \). Especially, when the field \( \mathbb{F} \) is finite, our preprocessing will play a more important role in consideration of
the nature of finite fields. In this section, we will present some experimental results to compare the effectiveness of our paradigm with the classical BM. First see an example with point set of small size.

**Example 4.** We choose the field \( \mathbb{F}_7 \), and let
\[
\Xi = \{(0, 0), (0, 1), (0, 4), (0, 5), (1, 0),
\quad (1, 1), (1, 4), (1, 6), (2, 1), (2, 2),
\quad (2, 6), (3, 2), (4, 2), (4, 5), (4, 6),
\quad (5, 1), (5, 5), (5, 6), (6, 0), (6, 2)\}.
\]

By Algorithm 3, we can construct the maximal cartesian subset
\[
\Xi' = \{(0, 1), (1, 1), (2, 1), (5, 1), (1, 6), (2, 6), (5, 6), (1, 0), (1, 4)\}
\]
hence get
\[
N = \{1, x, x^2, x^3, y, xy, x^2 y, y^2\},
Q = \{1, x, 4x^2 + 3x, 2x^3 + x^2 + 4x, 3y + 4, 3xy + 4x + 4y + 3,
\quad 2x^2 y + 5x^2 + xy + 6x + 4y + 3, 6y^2 + 1, 2y^3 + 5y\},
L = \{y^4, xy^2, x^3y^2, x^3y, x^4\},
B = \begin{pmatrix}
1 & 1 & 1 & \cdots \\
0 & 1 & 2 & \cdots \\
0 & 0 & 1 & \cdots \\
\vdots & \vdots & \vdots & \ddots
\end{pmatrix}.
\]

Put these \( N, Q, L, B \) into BM algorithm, we can get the final output
\[
N = \{1, x, x^2, x^3, y, xy, x^2 y, y^2, y^3, xy^2, y^4, x^2y^2, x^3 y, x^4, y^5, xy^4, x^2y^3,
\quad x^3 y^2, x^4 y\},
Q = \{1, x, 4x^2 + 3x, 2x^3 + x^2 + 4x, 3y + 4, 3xy + 4x + 4y + 3,
\quad 2x^2 y + 5x^2 + xy + 6x + 4y + 3, 6y^2 + 1, 2y^3 + 5y, xy^2 + 6y^2 + 6x + 1,
\quad y^3 + 3y^3 + 6y^2 + 4y, 5xy^3 + 5y^4 + 3y^3 + 2xy + 2y^2 + 4y,
\quad 6x^2 y^2 + xy^2 + x^2 + 6x, \ldots\}.
\]
\[ G = \{y^6 + 3y^5 + 2y^4 + 6y^3 + 4y^2 + 5y, \\
xy^5 + x^4y + 6x^3y^2 + x^2y^3 + 5xy^4 + 6y^5 + 6x^4 + 2x^3y + 6x^2y^2 + 3xy^3 + \\
3y^4 + 6x^3 + 6x^2y + 2xy^2 + 6y^3 + x^2 + 2xy + 6y^2 + x, \\
x^2y^4 + x^4y + 3x^2y^3 + 3xy^4 + 5y^5 + x^4 + 6x^3y + 3x^2y^3 + 2xy^3 + 4y^4 + \\
6x^3 + 4y^3 + 6x^2 + 2xy + 3y^2 + x + 5y, \ldots \}. \]

In the following, several tables show the timings for the computations of BM-problems on sets of distinct random points w.r.t. the term order \( \prec_{\text{lex}} \) or \( \prec_{\text{tdinlex}} \). The algorithms presented in the paper were implemented on Maple 12 installed on a laptop with 2 Gb RAM and 1.8 GHz CPU.

Take the field \( \mathbb{F}_{23} \), we have

| \#\Xi | 200  | 300  | 400  | 500  |
|-------|------|------|------|------|
| BM    | 4.968 s | 15.359 s | 34.609 s | 61.172 s |
| SPBM  | 1.438 s | 3.766 s | 7.141 s | 7.969 s |

For \( \mathbb{F}_{37} \), we have

| \#\Xi | 300  | 600  | 900  | 1200 |
|-------|------|------|------|------|
| BM    | 16.265 s | 121.766 s | 420.219 s | 1060.203 s |
| SPBM  | 4.172 s | 25.125 s | 82.000 s | 132.719 s |

For \( \mathbb{F}_{17} \), we have

| \#\Xi | 100  | 150  | 200  | 250  |
|-------|------|------|------|------|
| BM    | 0.875 s | 2.421 s | 4.953 s | 8.188 s |
| GPBM  | 0.797 s | 2.125 s | 4.250 s | 5.641 s |
| Preprocessing | 0.015 s | 0.094 s | 0.172 s | 0.391 s |
| \#\Xi'/\#\Xi | 0.310 | 0.393 | 0.430 | 0.616 |

Take the field \( \mathbb{F}_{29} \), we have

| \#\Xi | 200  | 400  | 600  | 800  |
|-------|------|------|------|------|
| BM    | 5.672 s | 38.063 s | 112.156 s | 235.813 s |
| GPBM  | 5.562 s | 36.906 s | 105.828 s | 135.609 s |
| Preprocessing | 0.046 s | 0.313 s | 1.671 s | 8.125 s |
| \#\Xi'/\#\Xi | 0.125 | 0.178 | 0.328 | 0.711 |
References

[1] D. A. Cox, J. Little, D. O'Shea, Using Algebraic Geometry, 2nd Edition, Vol. 185 of Graduate Texts in Mathematics, Springer, New York, 2005.

[2] T. Sauer, Polynomial interpolation in several variables: Lattices, differences, and ideals, in: K. Jetter, M. Buhmann, W. Haussmann, R. Schaback, J. Stöckler (Eds.), Topics in Multivariate Approximation and Interpolation, Vol. 12 of Studies in Computational Mathematics, Elsevier, Amsterdam, 2006, pp. 191–230.

[3] C. de Boor, Ideal interpolation, in: C. K. Chui, M. Neamtu, L. L. Schumaker (Eds.), Approximation Theory XI: Gatlinburg 2004, Nashboro Press, Brentwood TN, 2005, pp. 59–91.

[4] S. Sakata, Gröbner bases and coding theory, in: B. Buchberger, F. Winkler (Eds.), Gröbner Bases and Applications, Vol. 251 of London Mathematical Society Lecture Notes Series, Cambridge University Press, New York, 1998, pp. 470–485.

[5] M. Sala, Gröbner bases, coding, and cryptography: a guide to the state-of-art, in: M. Sala, T. Mora, L. Perret, S. Sakata, C. Traverso (Eds.), Gröbner Bases, Coding, and Cryptography, Springer, Berlin, 2009, pp. 1–8.

[6] L. Robbiano, Gröbner bases and statistics, in: B. Buchberger, F. Winkler (Eds.), Gröbner Bases and Applications, Vol. 251 of London Mathematical Society Lecture Notes Series, Cambridge University Press, New York, 1998, pp. 179–204.

[7] R. Laubenbacher, B. Stigler, A computational algebra approach to the reverse engineering of gene regulatory networks, J. Theoret. Biol. 229 (4) (2004) 523–537.

[8] W. Just, B. Stigler, Computing Gröbner bases of ideals of few points in high dimensions, ACM Commun. Comput. Algebra 40 (3-4) (2006) 67–78.

[9] H. Möller, B. Buchberger, The construction of multivariate polynomials with preassigned zeros, in: J. Calmet (Ed.), Computer Algebra: EUROCAM ’82, Vol. 144 of Lecture Notes in Computer Science, Springer, Berlin, 1982, pp. 24–31.
[10] J. C. Faugère, P. Gianni, D. Lazard, T. Mora, Efficient computation of zero-dimensional Gröbner bases by change of ordering, J. Symbolic Comput. 16 (4) (1993) 329–344.

[11] M. G. Marinari, H. M. Möller, T. Mora, Gröbner bases of ideals defined by functionals with an application to ideals of projective points, Appl. Algebra Engrg. Comm. Comput. 4 (2) (1993) 103–145.

[12] J. Abbott, A. Bigatti, M. Kreuzer, L. Robbiano, Computing ideals of points, J. Symbolic Comput. 30 (2000) 341–356.

[13] L. Cerlienco, M. Mureddu, From algebraic sets to monomial linear bases by means of combinatorial algorithms, Discrete Math. 139 (1-3) (1995) 73–87.

[14] S. Gao, V. M. Rodrigues, J. Stroomer, Gröbner basis structure of finite sets of points (Preprint).

[15] B. Felszeghy, B. Ráth, L. Rónyai, The lex game and some applications, J. Symbolic Comput. 41 (6) (2006) 663–681.

[16] T. Sauer, Lagrange interpolation on subgrids of tensor product grids, Math. Comp. 73 (245) (2004) 181–190.

[17] N. Crainic, Multivariate Birkhoff-Lagrange interpolation schemes and cartesian sets of nodes, Acta Math. Univ. Comenian.(N.S.) LXXIII (2) (2004) 217–221.

[18] T. Chen, T. Dong, S. Zhang, The Newton interpolation bases on lower sets, J. Inf. Comput. Sci. 3 (3) (2006) 385–394.

[19] T. Becker, V. Weispfenning, Gröbner Bases, Vol. 141 of Graduate Texts in Mathematics, Springer-Verlag, New York, 1993.

[20] R. Lorentz, Multivariate Birkhoff Interpolation, Vol. 1516 of Lecture Notes in Mathematics, Springer, Heidelberg, 1992.

[21] C. de Boor, Interpolation from spaces spanned by monomials, Adv. Comput. Math. 26 (1) (2007) 63–70.