RECOGNIZING AND PARAMETRIZING CURVES WITHOUT AFFINE SINGULARITIES

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Abstract. Some time ago, Shpilrain and Yu reported an algorithm for deciding whether or not a polynomial \( p \in K[x, y] \) is a coordinate, or, equivalently, whether or not a plane curve \( p(x, y) = 0 \) is isomorphic to a line. Here \( K \) is any constructible field of characteristic 0. In this paper, we show that their algorithm requires \( O(n^2 \log^2 n) \) field operations, where \( n \) is the degree of a given polynomial. We also show how their algorithm can be used to find a polynomial parametrization of a plane curve \( p(x, y) = 0 \) which is isomorphic to a line (or, equivalently, has no affine singularities). This requires \( O(n^3) \) field operations.

1. Introduction

Let \( K[x, y] \) be the polynomial algebra in 2 variables over a field \( K \) of characteristic 0. We say that a polynomial \( p \in K[x, y] \) is a coordinate if there is an automorphism of \( K[x, y] \) that takes \( x \) to \( p \). By the famous theorem of Abhyankar and Moh [1], this is equivalent to saying that the curve \( p(x, y) = 0 \) in the affine plane \( K^2 \) is isomorphic to a line.

By the well known result of Jung [5] and van der Kulk [7], every coordinate of \( K[x, y] \) is tame, where \( K \) is any field. That means, every coordinate of \( K[x, y] \) can be obtained from \( x \) by applying a sequence of affine transformations and triangular automorphisms of the following types:

(i) \( (x, y) \longrightarrow (x + h(y), y) \) for some \( h(y) \in K[y], \deg_y h(y) \geq 2; \)

(ii) \( (x, y) \longrightarrow (x, y + h(x)) \) for some \( h(x) \in K[x], \deg_x h(x) \geq 2. \)

By now, many algorithms for recognizing coordinates of \( K[x, y] \) are known; see e.g. [4], [9], [10], or monographs [3], [8] and references thereto. With so many algorithms around, it becomes natural to ask for the one which is faster and/or consumes less resources than other ones.

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Recently, Gutierrez, Rubio and Schicho [4] have addressed the question of the time complexity of one of the algorithms for recognizing coordinates of $K[x, y]$ as well as of the auxiliary algorithm for finding a polynomial parametrization of a plane curve $p(x, y) = 0$ which is isomorphic to a line (or, equivalently, has no affine singularities). They showed that the time complexity of either algorithm is $O(n^3 \log_2 n)$, where $n$ is the degree of a given polynomial $p(x, y)$.

The purpose of the present paper is to estimate the time complexity of the algorithm for recognizing coordinates of $K[x, y]$ given by Shpilrain and Yu [9]. This complexity turns out to be $O(n^2 \log_2 n)$, which makes us believe that this algorithm is actually the fastest possible. It also requires very little computer memory since (1) the complexity of the output decreases with every recursion step of the algorithm, and (2) at every recursion step, the algorithm only utilizes the output of the immediately preceding step. A brief description of the algorithm is given in Section 2, and in Section 3, we estimate the time complexity of this and the auxiliary algorithm for finding a polynomial parametrization of a plane curve $p(x, y) = 0$ which is isomorphic to a line. As pointed out in [4], parametric representations of curves are important for geometric modeling; in particular, they are industrial standard in CAD systems. We show that one of the algorithms in [9] yields a parametrization algorithm of the time complexity $O(n^3)$. The most time-consuming part of the latter algorithm is evaluating the image of a given polynomial under a triangular automorphism of $K[x, y]$. It seems that this kind of computation cannot be avoided by any known algorithm, which makes us believe that the $O(n^3)$ estimate cannot be improved.

Finally, we note that the degree of a polynomial $p$ does not seem to be the most adequate measure of its complexity as far as most real-life applications are concerned. A more adequate measure appears to be the number of monomials that occur in $p$ with non-zero coefficients, together with the set of exponents and the set of coefficients; this more accurately reflects the amount of information one has to input in order to describe a polynomial. For example, to describe a polynomial of the form $x^N$, we only need $\log_2 N$ bits of information, not $N$. This kind of “informational complexity” was introduced by Kolmogorov and is now called Kolmogorov complexity, see [6].
2. The algorithms

Let $>_{\text{lex}}$ be the lexicographic ordering on the set of monomials in $x$ and $y$, with $x > y$. For any $p \in K[x, y]$, we use $\text{lm}(p)$ to denote the leading monomial of $p$ with respect to $>_{\text{lex}}$.

We start by reproducing the algorithm for recognizing coordinate of $K[x, y]$ given by Shpilrain and Yu in [9].

**Algorithm 2.1.** Let $p := p(x, y)$ be a polynomial of $K[x, y]$.

**Step 1:** Take $q_1 := \frac{\partial p}{\partial x}$ and $q_2 := \frac{\partial p}{\partial y}$.

**Step 2:** If $\text{lm}(q_1)$ is not divisible by $\text{lm}(q_2)$ or vice versa, then $p$ is not a coordinate. If $\text{lm}(q_1) = h \cdot \text{lm}(q_2)$ (respectively $\text{lm}(q_2) = h \cdot \text{lm}(q_1)$) for a monomial $h$, then go to Step 3.

**Step 3:** Set $q_1' = q_1 - h \cdot q_2$ (respectively $q_2' = q_2 - h \cdot q_1$). If $\text{lm}(q_1')$ (respectively $\text{lm}(q_2')$) is in $K^*$, then $p$ is a coordinate. If $\text{lm}(q_1') = 0$ (respectively $\text{lm}(q_2') = 0$), then $p$ is a coordinate if and only if $\text{lm}(q_2) = 1$ (respectively $\text{lm}(q_1) = 1$). If $\text{lm}(q_1') \notin K$, replace $(q_1, q_2)$ by $(q_1', q_2)$. (Respectively, if $\text{lm}(q_2') \notin K$, replace $(q_1, q_2)$ by $(q_1, q_2')$). Then go to Step 2.

We note that if $p = p(x, y)$ is a coordinate polynomial, then, by [2, Theorem 6.8.5] and [9, Theorem 1.4], at Step 3 of Algorithm 2.1 one can choose a polynomial (not necessarily a monomial) $h$ such that either $h = h(y) \in K[y]$ and $\text{lm}(q_1 - h(y) \cdot q_2) < \text{lm}(q_2)$ or $h = h(x) \in K[x]$ and $\text{lm}(q_2 - h(x) \cdot q_1) < \text{lm}(q_1)$. Thus, Algorithm 2.1 is very similar to the familiar Euclidean algorithm for finding the g.c.d.

The next algorithm applies to a given coordinate polynomial $p = p(x, y)$ and outputs a sequence of elementary (i.e., triangular or affine) automorphisms that takes $p$ to $x$. This algorithm can be also considered a refinement of Algorithm 2.1, i.e., it can be also used for recognizing coordinates because if the input is not a coordinate, then Step 2 will eventually fail.

**Algorithm 2.2.** **Input.** $p := p(x, y)$, a coordinate of $K[x, y]$. Let $\varphi := (x, y)$ (i.e., the identical automorphism).

**Output.** A sequence of elementary automorphisms of $K[x, y]$ that takes $p(x, y)$ to $x$.

**Step 1 (Initialization):** Let $(q_1, q_2) := (p_x, p_y)$. Let $S = \emptyset$ (this is going to be a collector of elementary automorphisms).
Suppose there is either \( h = h(y) \in K[y] \) or \( h = h(x) \in K[x] \) such that \( \text{lm}(q_1 - h(y) \cdot q_2) < \text{lm}(q_2) \) (respectively \( \text{lm}(q_2 - h(x) \cdot q_1) < \text{lm}(q_1) \)). Find such \( h \) (see the observation (3) below), then go to Step 3.

**Step 3:** Define the automorphism \( \psi := (x - \int h(y)dy, y) \) (respectively, \( \psi := (x, y - \int h(x)dx) \)). Set \( q'_1 = q_1 - h(y) \cdot q_2 \) (respectively, \( q'_2 = q_2 - h(x) \cdot q_1 \)). If \( \text{lm}(q'_1) \) (respectively, \( \text{lm}(q'_2) \)) is in \( K^* \), then \( \psi(p) = x \). If \( \text{lm}(q'_1) = 0 \) (respectively \( \text{lm}(q'_2) = 0 \)), then \( \pi\psi\phi(p) = x \) where \( \pi = (y, x) \), the permutation automorphism. Otherwise, replace \( p = p(x, y) \) by \( p = p(x - \int h(y)dy, y) \) (respectively, by \( p = p(x, y - \int h(x)dx) \)), denote \( \psi\phi \) by \( \phi \), and add \( \psi \) into \( S \). Then go to Step 2.

The output of Algorithm 2.2 (i.e., the sequence of elementary automorphisms that takes \( p(x, y) \) to \( x \)) can be used to obtain a parametrization of the curve \( p(x, y) = 0 \) if we apply this sequence to the pair \((0, t)\) (which is a parametrization of the line \( x = 0 \)) instead of the pair \((x, y)\). For example, a parametrization of the curve \( x + y^2 = 0 \) would be \((0 - t^2, t) = (-t^2, t)\). It corresponds to the automorphism \( \varphi : x \to x - y^2, \ y \to y \) that takes \( x + y^2 \) to \( x \).

We now make three observations relevant to Algorithm 2.2.

1. \( h(y) \) may be a constant only if \( p(x, y) \) is linear. Therefore, at all recursion steps of Algorithm 2.2 except, perhaps, the last one, we have \( \deg_y h(y) \geq 1 \).
2. Suppose \( h(y) \) is in \( K[y] \) with \( \deg_y h(y) = m \). To evaluate \( \int h(y)dy \), it takes \( m + 1 \) divisions. More specifically, if \( h(y) = a_m y^m + \cdots + a_0 \), we may denote by \( h(y) \) the \( (m + 1) \)-tuple \( h = (a_m, \cdots, a_0) \). Then integration of \( h(y) \) is equivalent to the following transformation of tuples:

\[
(a_m, \ldots, a_i, \ldots, a_0) \to \left(\frac{a_m}{m}, \ldots, \frac{a_i}{i}, \ldots, \frac{a_0}{1}, 0\right).
\]

3. Suppose there is \( h(y) = a_0 y^m + \cdots + a_m \in K[y] \) such that \( \text{lm}(p_2(x, y) - h(y)p_2(x, y)) < \text{lm}(p_2(x, y)) \). To find \( h(y) \), consider the process of division of \( p_2(0, y) \) by \( p_2(0, y) \).

Suppose \( p(x, y) \) is a coordinate with \( \deg(p) = n \). Then \( p_2(0, y) \) has at most \( n - 1 \) monomials. To find \( a_0 \), we need a division. Once \( a_0 \) is found, we can evaluate \( a_1 \) by comparing the leading terms of \( p_2(0, y) - a_0 y^m p_2(0, y) \) and \( p_2(0, y) \). It requires at most \( n - 1 \) multiplications, \( n - 1 \) subtractions, and a division. Inductively, suppose \( a_0, \ldots, a_i \) are found. To find \( a_{i+1} \), we need at most \( n - 1 \) multiplications, \( n - 1 \) subtractions, and
a division. Hence, at each step, there are at most \( n - 1 \) multiplications, \( n - 1 \) subtractions, and a division. The number of steps is at most \( m + 1 \). Therefore, to find \( h(y) \), one needs at most \( (m + 1)(1 + 2(n - 1)) \) field operations.

3. THE COMPLEXITY OF RECOGNIZING AND PARAMETRIZING COORDINATES

In this section, we estimate the time complexity of Algorithm 2.2 for finding an automorphism \( \varphi \) (if it exists) of \( \mathbb{K}[x,y] \) such that \( \varphi(p) = x \). As we have mentioned in the previous section, finding such an automorphism immediately leads to a parametrization of the curve \( p(x,y) = 0 \). The estimate will be given with respect to the degree \( \deg p \) of a given polynomial \( p \). First, we will find an upper bound for the number of field operations at each recursion step of Algorithm 2.2. We need some preliminary observations.

**Lemma 3.1.** Let \( \varphi = \sigma^d_1 \tau_1 \cdots \tau_s \) be an automorphism of \( \mathbb{K}[x,y] \), where \( \delta_1 = 0 \) or \( 1 \), \( \sigma \) and \( \tau_i \) are affine and triangular automorphisms respectively. Let \( \tau_i = (x + f_i(y), y) \) ("type 1") or \( (x, y + f_i(x)) \) ("type 2") with \( d_i = \deg f_i \geq 2 \), and suppose that for every \( i \), the automorphisms \( \tau_i \) and \( \tau_{i+1} \) are not of the same type. Then

\[
\deg(\varphi) = d_1 \cdot d_2 \cdot \ldots \cdot d_s.
\]

**Proof.** We use induction by \( s \). The basis of induction is quite obvious, so we proceed with the induction step. Let \( \varphi_i = \sigma^d_i \tau_1 \tau_2 \cdots \tau_{i-1} \). Compose \( \varphi_i \) with \( \tau_i \), i.e., consider the automorphism \( \varphi_i \tau_i = \varphi_i(\tau_i) \). In the polynomials \( \varphi_i(x), \varphi_i(y) \) pick a monomial \( m_i \) whose degree equals \( \deg(\varphi_i) \). It is well known (see e.g. [2, Theorem 6.8.5]) that \( m_i = x^k \) or \( y^k \) for some \( k \). More precisely, if \( \tau_{i-1} \) is of type 1, then \( m_i = y^k \), and if \( \tau_{i-1} \) is of type 2, then \( m_i = x^k \). Assume, without loss of generality, that \( \tau_{i-1} \) is of type 1. Then \( \tau_i \) must be of type 2; therefore, the degree of \( \varphi_i \tau_i \) equals \( \deg(y + f_i(x))^k = d_i \cdot k = d_i \cdot \deg(\varphi_i) \). The result follows. \( \square \)

The next lemma is obvious.

**Lemma 3.2.** Let \( p \) be a polynomial of \( \mathbb{K}[x,y] \) with \( \deg p = n \). Then both \( p_x \) and \( p_y \) have at most \( \frac{n(n+1)}{2} \) monomials.

We are now ready to prove the following

**Lemma 3.3.** Suppose \( p \) is a coordinate of \( \mathbb{K}[x,y] \) with \( \deg p = n \). In Algorithm 2.2, the time complexity of finding an elementary automorphism \( \varphi := (x - \int h(y)dy, y) \) (or \( \varphi := (x, y - \int h(x)dx) \)) is \( O(n^2) \).
Proof. Without loss of generality, we may assume there is $h(y)$ in $K[y]$ such that $\text{lm}(p_x(x,y) - h(y)p_y(x,y)) < \text{lm}(p_y(x,y))$. Obviously, $m = \deg h(y) \leq \deg p_x = n-1$. By Observation 3, the number of field operations for finding $h(y)$ is bounded by

\[(1) \quad (m + 1)(1 + 2(n - 1)) \leq n(1 + 2(n - 1)).\]

Furthermore, to construct an elementary automorphism, integration of $h(y)$ is required. By Observation 2, this takes $m + 1 \leq n$ field operations. Therefore, the total time complexity of finding the automorphism is $O(n^2)$. □

Lemma 3.4. Let the polynomial $p(x,y)$ at Step 3 of Algorithm 2.2 have degree $n$. Then the number of field operations required for evaluating $p(x - \int h(y)dy, y)$ at Step 3 is bounded by $C \cdot n^3$ for some constant $C$ independent of $p(x,y)$.

Proof. It is sufficient to show that the number of field operations required for evaluating $p(x - \mu \cdot y^k, y), \mu \in K^*, 1 \leq k \leq n,$ is bounded by $C \cdot n^2$.

First we precompute $n!$ and therefore also all $m!$ for $m \leq n$. This takes at most $n$ multiplications. Then we precompute $\mu^1, 1 \leq i \leq n$. This, again, takes at most $n$ multiplications. Finally, we precompute all binomial coefficients of the form $\binom{n}{ij}, 1 \leq i, j \leq n$. This will take at most $2n^2$ multiplications.

Now to evaluate $(x - \mu \cdot y^k)^j$, we only need $j$ multiplications. Therefore, to evaluate all $(x - \mu \cdot y^k)^j, 1 \leq j \leq n$, we need at most $n^2$ multiplications.

Rewrite now our polynomial in the form $p(x,y) = \sum_{i=0}^{n} f_i(x) y^i$ (this rewriting will take linear time in $n$). Evaluating each $f_i(x - \mu \cdot y^k)$ now takes at most $n$ multiplications, and so does evaluating $f_i(x - \mu \cdot y^k) y^i$. Therefore, evaluating $p(x - \mu \cdot y^k, y) = \sum_{i=0}^{n} f_i(x - \mu \cdot y^k) y^i$ will take at most $2n^2$ field operations. This completes the proof. □

Lemma 3.5. The number of recursion steps in Algorithm 2.2 is $O(\log_2 n)$.

Proof. Each recursion step in Algorithm 2.2 corresponds to a triangular automorphism either of type 1 (i.e., of the form $(x + f_i(y), y)$) or of type 2 (i.e., of the form $(x, y + f_i(x))$), where $\deg f_i \geq 2$ (because $\deg h(y) \geq 1$, see Observation (1) in the end of the previous section), and no two successive recursion steps correspond to triangular automorphisms of the same type.

Thus, the output of Algorithm 2.2 is an automorphism $\varphi$ of the form $\sigma^{\delta_1} \tau_1 \tau_2 \cdots \tau_s(x)$, where $\delta_1 = 0$ or 1, $\sigma$ and $\tau_i$ are affine and triangular automorphisms respectively, and for every $i$, $\tau_i$ and $\tau_{i+1}$ are not of the same type. Then, by Lemma 3.1,
\[ n = \deg(\varphi(x)) = d_1 \cdot d_2 \cdot \ldots \cdot d_s \geq 2^s \] since \( d_i = \deg \tau_i \geq 2 \). Therefore, \( \log_2 n \geq s \), hence the number of recursion steps is \( O(\log_2 n) \). \qed

**Theorem 3.6.** The time complexity of Algorithm 2.1 for recognizing coordinates is \( O(n^2 \log_2 n) \), where \( n \) is the degree of the given polynomial \( p(x, y) \), whereas the complexity of Algorithm 2.2 for finding a parametrization of a given curve \( p(x, y) = 0 \) isomorphic to a line, is \( O(n^3) \).

**Proof.** The statement about Algorithm 2.1 is fairly obvious, so we focus on the complexity of Algorithm 2.2. As we see from Lemmas 3.3 and 3.4, the most time-consuming part of each recursion step of Algorithm 2.2 is evaluating \( p(x - \int h(y)dy, y) \). For a polynomial \( p(x, y) \) of degree \( n \), this takes at most \( C \cdot n^3 \) field operations. By Lemma 3.5, the number of recursion steps in Algorithm 2.2 is \( O(\log_2 n) \). This may make it seem that the complexity of the algorithm is \( O(n^3 \log_2 n) \). However, only at the first step of the recursion is the degree of \( p(x, y) \) equal to \( n \); at each subsequent step (except, perhaps, the last one), the degree of \( p(x, y) \) is reduced by at least the factor of 2, by Lemma 3.1. Therefore, by Lemma 3.4, the complexity of the algorithm is actually bounded by \( C \cdot (n^3 + (\frac{n}{2})^3 + (\frac{n}{4})^3 + \ldots) = O(n^3) \). This completes the proof. \qed

**Example 3.7.** In [4], two recursion steps are required for finding a parametrization of \( p(x, y) = 4x^2 + 8xy^3 - 8xy - 2x + 4y^6 - 8y^4 - 2y^3 + 4y^2 + 3y - 1 \). The parametrization itself is:

\[
\begin{align*}
  x &= t - \frac{29}{64} + \frac{59}{4} t^2 - 60 t^4 + 64 t^6 \\
  y &= 5 - 4 t^2
\end{align*}
\]

Since \( p_x = 8x + 8y^3 - 8y - 2 \) and \( p_y = 24xy^2 - 8x + 24y^5 - 32y^3 - 6y^2 + 8y + 3 \), we have \( p_y - (3y^2 - 1)p_x = 1 \). After integrating \( h(y) = 3y^2 - 1 \), we obtain the automorphism \( \varphi := (x - y^3 + y, y) \). Then evaluate: \( \varphi(p(x, y)) = -1 + 2x + 4x^3 + y \). This latter polynomial (more formally, the corresponding curve) can already be parametrized, just “by inspection”, as \( x = t \), \( y = 1 + 2t - 4t^2 \). Then we compute:

\[
(t, 1 + 2t - 4t^2)\varphi = (t - (1 + 2t - 4t^2)^3 + (1 + 2t - 4t^2), 1 + 2t - 4t^2)
\]

\[
= (-3t - 4t^2 + 40t^3 - 96t^5 + 64t^6, 1 + 2t - 4t^2). 
\]
Thus, our parametrization of $p(x, y) = 0$ is:

$$x = -3t - 4t^2 + 40t^3 - 96t^5 + 64t^6$$

$$y = 1 + 2t - 4t^2$$

We see therefore that our parametrization is different from that of [4]. Both parametrizations have the same degree, but our parametrization has smaller space complexity (at least, in this particular example). Indeed, the collection of coefficients that corresponds to our parametrization is $\{-3, -4, 40, -96, 64, 1, 2, -4\}$, which takes up approximately 30 bits of memory, whereas the collection corresponding to the parametrization in [4] is $\{1, -29, 64, 59, 4, -60, 64, 5, 4, -4\}$, which takes up approximately 45 bits of memory.

References