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On the Topology of Planar Algebraic Curves

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ABSTRACT

We revisit the problem of computing the topology and geometry of a real algebraic plane curve. The topology is of prime interest but geometric information, such as the position of singular and critical points, is also relevant. A challenge is to compute efficiently this information for the given coordinate system even if the curve is not in generic position.

Previous methods based on the cylindrical algebraic decomposition (CAD) use sub-resultant sequences and computations with polynomials with algebraic coefficients. A novelty of our approach is to replace these tools by Gröbner basis computations and isolation with rational univariate representations. This has the advantage of avoiding computations with polynomials with algebraic coefficients, even in non-generic positions. Our algorithm isolates critical points in boxes and computes a decomposition of the plane by rectangular boxes. This decomposition also induces a new approach for computing an arrangement of polylines isotopic to the input curve. We also present an analysis of the complexity of our algorithm. An implementation of our algorithm demonstrates its efficiency, in particular on high-degree non-generic curves.

Categories and Subject Descriptors

G.4 [Mathematical Software]: Algorithm design and analysis; I.1.4 [Symbolic and Algebraic Manipulation]: Applications; I.1.2 [Symbolic and Algebraic Manipulation]: Algorithms; I.3.5 [Computer Algebra and Geometry]: Computational Geometry and Object Modeling

General Terms

Algorithms, Experimentation.

Keywords

Effective Algebraic Geometry, Exact Geometric Computation, Topology of Algebraic Curves.

1. INTRODUCTION

We consider the problem of computing a geometric representation of a planar real algebraic curve, $C$, defined in a Cartesian coordinate system by a bivariate polynomial $f$ with rational coefficients, i.e., $C = \{(x, y) \in \mathbb{R}^2 \mid f(x, y) = 0\}$ with $f \in \mathbb{Q}[x, y]$. More precisely, we address the problem of computing a planar graph whose vertices are mapped to points in the plane (possibly at infinity) and such that drawing the arcs as line segments gives a drawing isotopic to the input curve (see Figure 1(a)).

In order to obtain an isotopic graph, the graph must contain vertices that correspond to self-intersections and isolated points of the curve. However, in order to avoid separating such relevant points from other singularities (e.g., cusps), all singular points of $C$, that is, points at which the tangent is not well defined, are usually mapped to vertices of the graph. The graph must also contain vertices that correspond to points at infinity for each of the branches of $C$ going to infinity. Finally, the graph should contain vertices corresponding to non-singular points on $C$ so that drawing the arcs as line segments gives a drawing isotopic to $C$.

While singular points are needed for computing the topology of a curve, the extreme points of a curve are also very important for representing its geometry. Precisely, the extreme points of $C$ for a particular direction, say the direction of the $x$-axis, are the non-singular points of $C$ at which the tangent line is vertical (i.e., parallel to the $y$-axis); the extreme points in the direction of the $x$-axis are called $x$-extreme. These extreme points are crucial for various applications and, in particular, for computing arrangements of curves by a standard sweep-line approach [17]. Of course, one can theoretically compute an arrangement of algebraic curves by computing the topology of their product. However, this approach is obviously highly inefficient compared to computing the topology of each input curve and, only then, computing the arrangement of all the curves with a sweep-line algorithm. Note that the $x$-extreme and singular points of $C$ form together the $x$-critical points of the curve.

Even more precisely we want the curve $C$ and the graph $G$ to be ambient isotopic in the real plane. In other words, we require the existence of a continuous map $F : \mathbb{R}^2 \times [0, 1] \to \mathbb{R}^2$, such that $F_t$ is an homeomorphism for any $t \in [0, 1]$, $F_0$ is the identity of $\mathbb{R}^2$ and $F(1) = G$. 


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(the $x$-coordinates of these points are exactly the positions of a vertical sweep line at which there may be a change in the number of intersection points with $C$).

It is thus useful to require that all the $x$-critical points of $C$ are mapped to vertices of the graph we want to compute. To our knowledge, almost all methods for computing the topology of a curve compute the critical points of the curve and associate corresponding vertices in the graph. (Refer to [1, 9] for recent subdivision methods that avoid the computation of non-singular critical points.) However, it should be stressed that almost all methods do not necessarily compute the critical points for the specified $x$-direction. Indeed, when the curve is not in generic position, that is, if two $x$-critical points have the same $x$-coordinate or if the curve admits a vertical asymptote, most algorithms shear the curve so that the resulting curve is in generic position. This is, however, an issue for several reasons. First, determining whether a curve is in generic position is not a trivial task and it is time consuming [23, 39]. Second, if one wants to compute arrangements of algebraic curves with a sweep-line approach, the extreme points of all the curves have to be computed for the same direction. Finally, if the coordinate system is sheared, the polynomial of the initial curve is transformed into a dense polynomial, which slows down, in practice, the computation of the critical points.

In this paper, given a curve $C$ which is not necessarily in generic position, we aim at computing efficiently its topology, including all the critical points for the specified $x$-direction. In terms of efficiency, our primary goal is the practical efficiency (rather than worst-case complexity). In particular, we want to avoid computations with non-rational algebraic numbers or, equivalently in this context, algebraic computations such as Sturm sequences, Sturm-Habicht sequences (which are a generalization of Sturm sequences, with better specialization properties [24]), and sub-resultant sequences.

We review below previous work on the problem and then present our contributions.

**Previous Work.** There have been many papers addressing the problem of computing the topology of algebraic plane curves (or closely related problems) defined by a bivariate polynomial with rational coefficients [1, 2, 11, 4, 14, 20, 23, 25, 27, 33, 38, 6, 3, 32, 41, 13]. Most of the algorithms assume generic position for the input curve. As mentioned above, this is without loss of generality since we can always shear a curve into generic position [39, 23] but this has a substantial negative impact on the time computation. All these algorithms perform the following phases. (1) Project the $x$-critical points of the curve on the $x$-axis, using sub-resultants sequences and isolate the real roots of the resulting univariate polynomial in $x$. This gives the $x$-coordinates of all the $x$-critical points. (2) For each such value $x_i$, compute the intersection points between the curve $C$ and the vertical line $x = x_i$. (3) Through each of these points, determine the number of branches of $C$ coming from the left and going to the right. (4) Connect all these points appropriately.

The main difficulty in all these algorithms is to compute efficiently all the critical points in Phase 2 because the $x$-critical values in Phase 1 are, a priori, non-rational thus computing the corresponding $y$-coordinates in Phase 2 amounts, in general, to solving a univariate polynomial with non-rational coefficients and at least a multiple root (corresponding to the critical point). To this end, most algorithms [2, 11, 4, 20, 23, 25, 27, 33, 38, 3, 41] rely heavily on computations with real algebraic numbers, Sturm sequences or sub-resultant sequences.

An approach using a variant of sub-resultant sequences for computing the critical points in Phase 2 was proposed by Hong [27]. He computed this way ($xy$-parallel) boxes with rational endpoints and separating the critical points. Counting the branches in Phase 3 can then be done by intersecting the boundary of the boxes with the curve which only involves univariate polynomials with rational coefficients. This approach, see also [3, 32, 6] and the software package Cad2D, does not assume that the curve is in generic position.

In a more recent paper, González-Vega and Necula [25] use Sturm-Habicht sequences which allow them to express the $y$-coordinate of the each critical point as a rational function of its $x$-coordinate. They implemented their algorithm in Maple with symbolic methods modulo the fact that the algebraic coefficients of the polynomials in Phase 2 are approximated in fixed-precision arithmetic. The algorithm takes as a parameter the initial precision for the floating-point arithmetic and recursively increases the precision. This approach is however not certified in the case where the curve is not in generic position because the algorithm checks for the equality of pairs of polynomials whose coefficients are evaluated (incorrect results have been reported in [40]). Note that there exists one variant of González-Vega and Necula algorithm that handles, without shearing, curves that are not in generic position [33]. This approach, however, requires substantial additional time-consuming symbolic computations such as computing Sturm sequences.

Recently, Seidel and Wolpert [40] presented an alternate approach for computing the critical points involving most costly algebraic computations but to the expense of computing several projections of the critical points. They project, in Phase 1, the $x$-critical points on both $x$ and $y$-axes and also on a third random axis. After isolating the roots on each axis, they can recover ($xy$-parallel) boxes with rational endpoints that contain each exactly one critical point. From there, all computations only involve rational numbers but they, however, still need to compute Sturm-Habicht sequences for refining the boxes containing the singular points until each box intereses only the branches of the curve incident to the singular point. Their approach assumes that the curve is in generic position by a pre-processing phase in which the curve is sheared if needed. They also present a Maple implementation, INSULATE, which is an implementation of a certified algorithm for curves in arbitrary positions. Note that their implementation does not report $x$-extreme points in the original system when the curved is sheared.

Even more recently, Eigenwillig et al. [14] (see also [28]) presented a variant of González-Vega and Necula approach, in which the roots of the polynomials with non-rational coefficients are efficiently isolated using an implementation of a variant of an interval-based Descartes algorithm [15]. This variant, as [25], does not assume that the curve is in generic position but detects such configurations online. More precisely, if the bit-stream Descartes algorithm is in, a sense, unlucky then, rather than refining down to a separation bound (e.g. [8, 31]), the algorithm shears the input curve and starts again. Note that this approach still computes Sturm-Habicht sequences for determining the polynomials.

\footnote{http://www.cs.usna.edu/~qepcad/B/user/cad2d.html}
appearing in Phase 2 and the multiplicity of its multiple root. Also, if the curve is sheared to a \((x', y')\)-coordinate system, they compute extreme points both for the \(x'\)-direction and the direction corresponding to the \(z\)-axis. This approach has been implemented in C++ and is an implementation of a certified algorithm that handles curves that are not necessarily in generic positions and that reports \(z\)-extreme points for the original coordinate system.

Note finally that another approach that avoids expensive algebraic computations is to compute the critical or singular points using subdivision methods [1, 9]. The major drawback of these methods is that, in order to certify the results, the subdivision has, in general, to reach a separation bound (certification can also be achieved by solving algebraic systems by other means). It follows that, if no certification is required, these methods are very fast in practice, however, they can become very slow on difficult instances, if certification is required. To our knowledge, no implementation of such certified algorithm is available.

As for the complexity of the problem, the best known bound so far is \(O(N^{1.5})\) [13], where \(N\) is essentially the maximum of the degree of \(f\) and of the maximum number of bits needed for representing the input coefficients and the notation \(\tilde{O}\) denotes that the poly-logarithmic factors are omitted. This assumes that the real algebraic numbers are represented by isolating intervals. This complexity is based on theoretically fast computations with Sturm-Habicht sequences and algorithms for univariate real root isolation.

**Our contributions.** We present an algorithm for computing the topology of an algebraic plane curve which is possibly in non-generic position. The algorithm handles curves in non-generic positions in the Cartesian coordinate system in which they are defined. In particular, the algorithm never shears the coordinate system, which avoids the associated extra costs discussed above.

The other originality of our approach is that we succeed to avoid, in all cases, the computation of sub-resultant sequences and computations with algebraic numbers; we use instead Gröbner bases and the rational univariate representation. One can argue that these are just two different tool sets, this is true from a theoretical point of view but we show in the experiments the benefit of our choice when computing with non-generic curves. Furthermore, the philosophy of our approach is to avoid, as much as possible, computations that are time consuming in practice. This leads to various algorithmic choices such as avoiding the computation of \(y\)-critical points and allowing the curve to intersect the top and bottom sides of boxes isolating \(z\)-extreme points (which avoids substantial subdivisions since the tangent at an extreme point is vertical).

The novelty of our algorithm mainly relies upon the use of three new ingredients for this problem. First, we use some of the state-of-the-art techniques to isolate the roots of bivariate systems, i.e., we use (i) Gröbner basis [19], (ii) Rational Univariate Representations (RUR) [35] which represent the roots of the system as rational functions of the roots of a univariate polynomial, and (iii) a subdivision technique based on Descartes’ rule of signs (and filtered with interval arithmetic for efficiency) for isolating the roots of the univariate polynomials [36, 16, 18]. Even though this approach is well known for system solving, it was not used before for computing the topology of algebraic curves.

Second, we compute and use the multiplicities in fibers (see Definition 1) of critical points to compute the topology at singular points and to do the connection at extreme points. For extreme points, we get these multiplicities by the RUR and a special case of a formula of Teissier [42]. For singular points, we solve the system of singular points with additional constraints. Note that the overall method to compute the topology at critical points is not new, it is described in full details in [40], see also for example [3, 27, 6, 32, 33, 1] for closely related approaches for curves in non-generic position. The novelty appears in the way we compute multiplicities in this context; once again we avoid computing sub-resultant sequences.

Third, we present a variant of the standard combinatorial part of the algorithm for computing the topology. We compute a decomposition of the plane (which is not a cylindrical algebraic decomposition) by rectangular boxes containing at most one critical point. Since we allow crossings on the top and bottom of extreme point boxes to avoid costly refinement of boxes, the connection is not always straightforward. To make the connection, we use the multiplicities in fibers of extreme points. Note that one advantage of this variant is that, even when the curve is in generic position, the algorithm does not require to refine these boxes until they do not overlap in \(x\).

We also analyze the worst-case bit complexity of our algorithm. To the best of our knowledge, this is, for the problem considered here, the first time that the complexity of a (certified) algorithm based on refinements and approximations is analyzed. Our technique, even though not novel, could presumably also be used to analyze the complexity of the algorithms in [25, 27, 40, 14].

Finally, we ran large scale benchmarks on over a thousand of curves during several weeks for comparing the different state-of-the-art available implementations.

The rest of the paper is organized as follows. First, we recall in Section 2 some basic material, in particular, on the multiplicity of the intersection of two curves and on Rational Univariate Representations of the roots of system of (algebraic) equations. We present our algorithm in Section 3 and its complexity in Section 4. Finally, we present in Section 5 extensive experiments comparing various implementations.

**2. PRELIMINARIES**

Let \(\mathcal{C}\), also denoted \(\mathcal{C}_f\), be a real algebraic plane curve defined by a bivariate polynomial \(f \in \mathbb{Q}[x, y]\). Since the geometry of the curve is not modified by taking the square-free part of \(f\), we can assume without loss of generality that \(f\) is square-free. Note that \(\mathcal{C}\) may consist of several algebraic components, that is, \(f\) is not necessarily irreducible in \(\mathbb{R}[x, y]\). The algebraic components of the curve that are vertical lines (i.e., lines parallel to the \(y\)-axis) can be easily computed since their abscessa correspond to the real roots of the polynomial in \(x\) obtained as the gcd of the coefficients of \(f\) seen as a polynomial in \(y\).

Partial derivatives are denoted with subscripts: for instance, \(f_x\) denotes the derivative of \(f\) with respect to \(x\) and \(f_y^k\) (sometimes also \(f_y\)) denotes the \(k\)th derivative with respect to \(y\). A point \(p = (\alpha, \beta) \in \mathbb{C}^2\) is called \(x\)-critical if \(f(p) = f_x(p) = 0\), singular if \(f(p) = f_x(p) = f_y(p) = 0\), and \(x\)-extreme if \(f(p) = f_y(p) = 0\) and \(f_x(p) \neq 0\) (i.e., it is \(x\)-critical and non-singular). Similarly are defined \(y\)-critical and \(y\)-extreme points. As \(x\)-critical and \(x\)-extreme points...
are more often used in the following, we often simply refer to them as critical and extreme points.

The ideal generated by polynomials $P_i, \ldots, P_t$ is denoted $\mathbb{I}(P_1, \ldots, P_t)$. In the following, we often identify the ideal and the system of equations $\{P_1 = 0, \ldots, P_t = 0\}$ (or any equivalent system induced by a set of generators of the ideal).

We consider, in particular, the ideals $I_x = \langle f, f_x, f_y \rangle$ and $I_y = \langle f, f_x, f_y \rangle$; their roots are, respectively, the $x$-critical and singular points of $C$.

**Multiplicities.**

We now recall the notion of multiplicity of the roots of an ideal, then we state two lemmata using this notion for studying the local topology at critical points. Geometrically, the notion of multiplicity of intersection of two regular curves is intuitive. If the intersection is transverse, the multiplicity is one; otherwise, it is greater than one and it measures the level of degeneracy of the tangential contact between the curves. Defining the multiplicity of the intersection of two curves at a point that is singular for one of them (or possibly both) is more involved and an abstract and general concept of multiplicity in an ideal is needed.

**Definition 1** ([12, §4.2]). Let $I$ be an ideal of $\mathbb{Q}[x, y]$ and denote $\mathbb{Q}$ the algebraic closure of $\mathbb{Q}$. To each zero $(\alpha, \beta)$ of $I$ corresponds a local ring $(\mathbb{Q}[x, y]/I)_{(\alpha, \beta)}$ obtained by localizing the ring $\mathbb{Q}[x, y]/I$ at the maximal ideal $I(x - \alpha, y - \beta)$. When this local ring is finite dimensional as $\mathbb{Q}$-vector space, we say that $(\alpha, \beta)$ is an isolated zero of $I$ and this dimension is called the **multiplicity of $(\alpha, \beta)$ as a zero of $I$**.

Let $f, g \in \mathbb{Q}[x, y]$ be such that the intersection of $C_f$ and $C_g$ in $\mathbb{C}^2$ contains a zero-dimensional component equal to point $p = (\alpha, \beta)$. Then $(\alpha, \beta)$ is an isolated zero of $\{f, g\}$ and its multiplicity, denoted by $\text{Int}(f, g, p)$, is called the **multiplicity of $p$ in its fiber** and denote it as $\text{mult}(f, g, \beta)$.

The next lemma, due to Teissier [42], relates the multiplicity of a point in a fiber with the multiplicity in the critical ideal. We will use it to deduce the multiplicity in the fiber knowing multiplicity in the ideal. More precisely, we will use the multiplicity in fibers of extreme points during the connection step of our algorithm.

**Lemma 2** ([42, LEMMA D.3.4]). For an $x$-extreme point $p = (\alpha, \beta)$ of $f$, $\text{mult}(f, \alpha, \beta) = \text{Int}(f, f_x, f_y, p) + 1$.

To compute the local topology of the curve at a singular point, we aim at isolating the singular point in a box so that the intersection of its border and the curve determines the topology. Indeed for a small enough box, the topology is given by the connection of the singular point with all the intersections on the border. So the box shall avoid parts of the curve not connected to the singular point. Knowing the multiplicity of the singular point in the fiber enables to isolate the singular point from other crossings of the curve in this fiber. Requiring in addition that intersections with the curve only occur on the left or right sides of the box leads to the following.\(^3\)

\(^3\)The proof is based on a recursive application of the mean value theorem stating that the roots of the derivative of a polynomial $P$ lie between those of $P$.

**Univariate Root Isolation.**

We need to count and isolate the real roots of univariate polynomials, possibly in a given interval. This is, in particular, needed for computing the intersections between $C$ and the sides of the boxes isolating the critical points. Only polynomials with rational coefficients will be considered. The square-free part of the considered polynomials is first computed. The real roots are then isolated using recursive subdivision and the Descartes’ rule of signs (see [16, 36] for details and [36] for the way interval arithmetic can be used to speed up computations). In our implementation, we use the RS software [37].

**Rational Univariate Representation.**

In our algorithm, we need to represent solutions of zero-dimensional ideals depending on two variables by boxes containing them. We use the so-called Rational Univariate Representation (RUR) of the roots, which can be viewed as a univariate equivalent to the studied ideal. The key feature of this RUR is the ability to isolate solutions in easily refinable boxes and to compute multiplicities.

Given a zero-dimensional ideal $I = \mathbb{I}(P_1, \ldots, P_t)$ where the $P_i \in \mathbb{Q}[x_1, \ldots, x_n]$, a Rational Univariate Representation of the solutions $V(I)$ is given by $F(t) = 0, r_1 = \frac{G_1(t)}{t^1}, \ldots, r_n = \frac{G_n(t)}{t^n}$ where $F, G_1, \ldots, G_n$ are univariate polynomials in $\mathbb{Q}[t]$ (where $t$ is a new variable). All these univariate polynomials, and thus the RUR, are uniquely defined with respect to a given polynomial $\gamma \in \mathbb{Q}[x_1, \ldots, x_n]$ which is injective on $V(I)$; $\gamma$ is called the **separating polynomial** of the RUR.\(^4\) Note that a random degree-one polynomial in $x_1, \ldots, x_n$ is a separating polynomial with probability one. The RUR defines a bijection between the (complex and real) roots of the ideal $I$ and those of $F$. Furthermore, this bijection preserves the multiplicities and the real roots. Computing a box for a solution of the system is done by isolating the corresponding root of the univariate polynomial $F$ and evaluating the coordinate functions with interval arithmetic. To refine a box, one just needs to refine the corresponding root of $F$ and evaluate the coordinates again.

There exists several ways for computing a RUR. One can use the strategy from [35] which consists of computing a Gröbner basis of $I$ and then to perform linear algebra operations to compute a separating element as well as the full expression of the RUR. The Gröbner basis computation can also be replaced by the generalized normal form from [34]. There exists more or less certified alternatives such as the Geometrical resolution [22] (it is probabilistic since the separating element is randomly chosen and its validity is not checked, one also loses the multiplicities of the roots) or re-

\(^4\)The polynomial $F$ is the characteristic polynomial of $m_\gamma$, the multiplication operator by the polynomial $\gamma$, in $\mathbb{Q}[x_1, \ldots, x_n]/I$. 
3. ALGORITHM

As discussed in Section 2, we consider, as input, a curve $C$ without vertical components and defined by a square-free polynomial $f \in \mathbb{Q}[x,y]$ (note that our implementation does not require these assumptions, but the processing of vertical components is rather technical and we choose not to include all these details here). In a few words, the algorithm first focuses on critical points, their rational univariate representations enabling to compute multiplicities and boxes isolating each point with known topology inside the box. Then a sweep method computes a rectangular decomposition of the plane induced by the boxes of critical points. Eventually the connection is processed in all rectangles with a greedy method using multiplicities in fibers for extreme points. We describe more precisely our algorithm in six steps.

Step 1. Isolating boxes of the singular points and of the $z$-extreme points. As a general practical rule, the smaller the number of solutions of a system, the easier it is to work with. Hence we split the system of critical points into the system of singular points and the system of extreme points. The system of singular points is the one of critical with in addition the equation $f_x = 0$. The system of extreme points, denoted $I_e$, is computed by saturation. Indeed, the extreme points are critical for which $f_x \neq 0$, thus we add to the critical system the equation $1 - u f_x = 0$ with a new variable $u$ that we eliminate afterwards. We then compute the RURs of these systems $I_s$ and $I_e$ and isolating boxes for the solutions. We may need to refine the boxes of extreme and singular points to avoid overlaps.

Step 2. Multiplicities of critical points in fibers. For extreme points, we use the Teissier formula: the multiplicity of an extreme point in $I_s$ is the same than in $I_e$ because precisely $f_x$ does not vanish at these points. The multiplicity in $I_s$ is given by the RUR, and hence the multiplicity of an extreme point in its fiber is this number plus one according to the equation of Lemma 2.

For singular points, we use the definition of univariate multiplicity, namely the smallest integer $k$ such that the $k^{th}$ derivative no longer vanishes. Let $I_{s,k}$ be the system of singular points with in addition the equations $f_{xy} = 0$ for $i$ from 2 to $k$. Hence we solve, for $k$ increasing from 2, the systems $I_{s,k}$ until it has no solutions. At each step, a singular point which is no longer solution of $I_{s,k}$ has its multiplicity in fiber equal to $k$.

Step 3. Refinement of the isolating boxes of the $z$-extreme points. Consider each such box, $B$, in turn. For each vertical or horizontal side of $B$, isolate its intersections with $C$ and refine the box until there are two intersection points. We further refine until there is at most one crossing on the top (resp. bottom) side of $B$. Note that, unlike comparable algorithms, we do not require that $C$ intersects the boundary of $B$ on its vertical sides. This is important in practice because, since the curve has a vertical tangent at an $z$-extreme point, refining until the curve intersects the vertical side is time consuming.

Step 4. Refinement of the isolating boxes of the singular points. We refine these boxes exactly as in [40] (see Lemma 3) except for the way the multiplicity $k$ of each singular point in its fiber is computed. In [40], $k$ is computed using Sturm-Habicht sequences under the assumption of generic position while we deduce $k$ as explained in Step 2. Then, as in [40], every box is refined until the interval evaluation of $f_{xy}$ does not contain 0. Further refine the $x$-coordinates of the box until $C$ only intersects the vertical boundary of the box.

Step 5. Vertical asymptotes. To deal with an asymptote $x = \alpha$, the idea is, informally, to isolate the point $(\alpha, \infty)$ in a box $[a,b] \times [M, \ldots, \infty]$, whose vertical sides do not intersect the curve $C$. The number of crossings with the horizontal sides then gives the number of branches going to $\pm \infty$ with this asymptote (see Figure 1(b)). First, compute an upper bound $M_e$ on the absolute value of the $y$-coordinates of the $y$-critical points (this is of course done without computing these critical points, but only the discriminant with respect to $y$ and an upper bound of the roots of this univariate polynomial). Compute also a bound $M_e$ on the absolute value of the $y$-coordinates of the $z$-critical points (for which we have already computed boxes). Isolate the roots of the polynomial in $y$ obtained as the leading coefficients of $f$ seen as a polynomial in $y$. For each root $\alpha$ we have an isolating...
drawings are two possible connections in the middle rectangle. The most one vertex on the top or bottom of any rectangle. The assume here for simplicity that the curve has no vertical in a rectangle below a critical box are computed once the rectangles starting from the top such that the connections for different parities of the multiplicities. To distinguish the information we may have on the crossings for two ex-
ceptually simplest is to refine boxes of extreme points so as to avoid top and bottom crossings; then, the number of left and right crossings in rectangles always match and the connection is one-to-one. Since we allow top/bottom crossings for efficiency (see above), this straightforward method does not apply. Another approach (see [1, 38]) is to compute the sign of the slope of the tangent to the curve at the top/bottom crossings (this yields whether the top/bottom crossing should be connected to vertex to the right or to the left of its rectangle). We however want to avoid such additional computations.

For computing the connections in the non-critical rectangles of the decomposition, we use the multiplicities in fibers of the extreme points and a greedy algorithm. The geometric meaning of the parity of the multiplicity is as follows: if it is even, the curve makes a U-turn at the extreme point; if it is odd and the curve is x-monotone in the neighborhood of the extreme point. Still, there are some difficulties for connecting the vertices, as illustrated on Figure 1(c): on the left is the information we may have on the crossings for two extreme points with x-overlapping boxes; the second and third drawings are two possible connections in the middle rectangle for different parities of the multiplicities. To distinguish between these two situations we compute the connections in rectangles starting from the top such that the connections in a rectangle below a critical box are computed once the connections in all the rectangles above the box are done.

The connections can easily be computed as follows. We assume here for simplicity that the curve has no vertical asymptote (similar arguments can be used otherwise). Then, due to requirements on boxes of extreme point, there is at most one vertex on the top or bottom of any rectangle. The

\[ x = a \] (resp. \( x = b \)) in \( f \) and deduce an upper bound, \( M_f \), on the absolute value of the \( y \)-coordinates of the intersection of \( C \) and \( x = a \) (resp. \( x = b \)). Set \( M = \max(M_f, M_\ell, M_r) \). Then, a branch crossing the segment \([a, b] \times M \) (resp. \([a, b] \times -M \)) goes to \(+\infty \) (resp. \(-\infty \)) with asymptote \( x = a \). Finally, we determine whether a given branch is to the left or to the right of the asymptote by comparing the \( x \)-coordinates of the asymptote and the crossing point.

**Step 6. Connections.** For simplicity, all the boxes computed above are called critical boxes and the points at infinity on vertical asymptotes are also called critical. First compute, with a sweep-line algorithm, the vertical rectangular decomposition obtained by extending the vertical sides of the critical boxes either to infinity or to the first encountered critical box (see Figure 1(b)). On each of the edges of the decomposition, isolate the intersections with \( C \). Create vertices in the graph corresponding to these intersection points and to the critical points. For describing the arcs connecting these vertices in the graph, we assimilate, for simplicity, the points and the graph vertices. In each critical box the topology is simple: the critical point is connected to each of the intersection points on the boundary of the box.

There are several approaches to do the connections in the other rectangles of the decomposition. The usual and conceptually simplest is to refine boxes of extreme points so as to avoid top and bottom crossings; then, the number of left and right crossings in rectangles always match and the connection is one-to-one. Since we allow top/bottom crossings for efficiency (see above), this straightforward method does not apply. Another approach (see [1, 38]) is to compute the sign of the slope of the tangent to the curve at the top/bottom crossings (this yields whether the top/bottom crossing should be connected to vertex to the right or to the left of its rectangle). We however want to avoid such additional computations.

For computing the connections in the non-critical rectangles of the decomposition, we use the multiplicities in fibers of the extreme points and a greedy algorithm. The geometric meaning of the parity of the multiplicity is as follows: if it is even, the curve makes a U-turn at the extreme point; if it is odd and the curve is x-monotone in the neighborhood of the extreme point. Still, there are some difficulties for connecting the vertices, as illustrated on Figure 1(c): on the left is the information we may have on the crossings for two extreme points with x-overlapping boxes; the second and third drawings are two possible connections in the middle rectangle for different parities of the multiplicities. To distinguish between these two situations we compute the connections in rectangles starting from the top such that the connections in a rectangle below a critical box are computed once the connections in all the rectangles above the box are done.

The connections can easily be computed as follows. We assume here for simplicity that the curve has no vertical asymptote (similar arguments can be used otherwise). Then, due to requirements on boxes of extreme point, there is at most one vertex on the top or bottom of any rectangle. The problem is to determine if such a vertex should be connected to another on the right or left side in the rectangle. First, the connections in the unbounded rectangles above critical boxes are straightforward: the connections between the vertices on the two vertical sides are one-to-one starting from infinity and if a vertex remains on a vertical side, there is a vertex on the horizontal side which it has to be connected with. Now, once all the connections have been computed in the rectangle(s) above the box of an extreme point, these connections and the multiplicity of the extreme point yield the connections in the rectangle(s) below. Indeed, if there is a vertex on the bottom side of the critical box, it lies on the top side of a rectangle and, inside this rectangle, the vertex is connected to the topmost vertex on the left or right side, depending on the multiplicity of the extreme point and on the side of the connection of the branch above the extreme point; the other connections in this rectangle and in the possible other rectangles below the critical box are performed similarly as for unbounded rectangles. Note that the two unbounded rectangles (the leftmost and the rightmost) that are vertical half-planes are treated separately in a straightforward manner: for each vertex on the vertical side is associated an arc going to infinity.

**Output.** A graph isotopic to the curve is output. In addition, x-extreme points, singular points and vertical asymptotes are identified and their position is approximated by boxes.

**4. COMPLEXITY ANALYSIS**

We consider the Turing machine model of computation and \( \hat{O}_B \) denotes the bit complexity where polylogarithmic factors are omitted. Let an algebraic curve \( C \) be given by a square-free polynomial \( f \in \mathbb{Z}[x, y] \) of total degree bounded by \( d \) and coefficients of bitsize bounded by \( \tau \). Let \( R \) be the number of critical points of the curve. We refer to the full version of the paper for a proof of the following theorem.

**Theorem 4.** The bit complexity of our algorithm for the computation of the topology of the curve \( C \) is \( \hat{O}_B(N^{22}) \), which is \( \hat{O}_B(N^{26}) \), where \( N = \max\{d, \tau\} \).

**5. EXPERIMENTS**

We implemented our algorithm, ISOTOP, in MAPLE using the Gb/Rs maple package [21, 37], implemented in C, for computing Gröbner bases, RURs and isolating roots. We believe that comparing MAPLE and C/C++ implementations is fair for our problem when the running time is not too small because then, most of the time is usually spent on algebraic computations which are coded in C/C++ (possibly in the kernel of MAPLE). When the running time is too small, the MAPLE part of the code is not negligible and comparing MAPLE and C/C++ implementations becomes meaningless. This is why we focused our tests on examples for which the running time exceeds 1 second. We measure the running time for computing the isotopic graph, but not the drawing. All the experiments were performed using 2.6 GHz single-core Pentium 4 with 1.5Gb of RAM and 512kb of cache, running 32-bit linux.

We compared our code, ISOTOP, with two C++ implementations Alcix [14] and Cad2d [7] and two MAPLE implementations, Top [25] and Insulate [40]. Another promising
of interesting cases such as isolated points, high multiplicity of tangency at singularities, large number of branches at singularities or many singularities. This set contains curves of degree up 90 that are both in generic and non-generic position. As suggested in [7], typical curves in generic position can be generated (i) as a random bivariate polynomial (which usually do not have singular points) or (ii) as resultants of two random trivariate polynomials (which usually have singular points, including isolated points). In both cases, we considered random polynomials with 50% non-zero coefficients of bitsize 32 in Case (i) and initial bitsize 8 in Case (ii). We generated such curves with degrees up to 25. We also generated classes of curves in non-generic position in two different ways. First, we considered products of a curve with one or several of its vertical translates. Second, we considered curves of the type \( g = f^2(x, y) + f^2(x, -y) \); such curves are usually irreducible and consist of isolated points which are the intersections of the curve \( C_f \) with its symmetric with respect to the \( x \)-axis. We generated such curves with degrees up to 24.

We set in our experiments a limit of 30 minutes for the computation of the topology of one curve. We report as time out instances that exceed this running time. Also, Cad2d which uses Singular for modular arithmetic, often reports on difficult instances that the table of primes has been exhausted, which results in an interruption of computation; this is reported in the tables as aborted.

In summary, we ran our benchmarks on a total of 1500 curves. As mentioned above, it is not significant to compare C++ and Maple implementations when the running time is too small. We thus only report experiments on 650 curves whose running times exceeded 1 second for ISOTOP. The distribution of degrees and number of critical points of these 650 curves is shown in Figure 2.

Figure 3 shows the ratio of running times between each of the competing implementations and ISOTOP over our 650 curves. It appears difficult to analyze the benchmarks globally because there are always particular examples that are processed faster by a given implementation. We note, however, that Insulate is almost always slower than ISOTOP, except for random curves with no singular points. In addition, Insulate and Top reached the time limit on more than half of the examples and, in particular, on difficult examples. We can, nevertheless, comment on the general behavior of the different approaches depending on the classes of examples.

To illustrate the behavior on curves in generic positions, we report the running times for random curves in Table 1 and for resultants of surfaces in Table 2. Random curves have no singular points and few extreme points. In this case, we observe that ISOTOP is the least efficient implementation. This can be explained by the fact that ISOTOP computes the Gröbner basis of a large system without multiplicities, which is the worst case in practice. On the other hand, the other implementations benefit from interval arithmetic filters in the lifting phase, which speed up computations by avoiding expensive symbolic computations, see for example [7]. Generic curves generated as resultants have many singularities and extreme points. ISOTOP benefits from splitting the critical system in two smaller (singular and extreme) systems and hence it performs relatively better than in the completely random case. We observe that ISOTOP is typically a bit slower than Alcix but faster than Top, and that Cad2d aborts.

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6Following the recommendation of M. Kerber, we ran two versions of the code with the flag CGAL\_ACK\_RESULTANT\_FIRST\_STRATEGY set to 1 and 0. One being optimized for generic cases, while the other is optimized for singular curves. We always compare to the better running time.

To illustrate the behavior on curves in non-generic position, we consider different classes of curves. The first class of non-generic curves are constructed with one curve multiplied by one or several of its vertical translates. The initial curve is taken either randomly, in Table 3, or it is a resultant of two surfaces, in Table 4. Table 5 reports results on the second class of non-generic curves of the type $f^2(x, y) + f^2(x, -y)$ for random polynomials $f$. For these non-generic curves, isotop is typically faster than other implementations.

As a general rule, we observe that, except for random curves, that is, curves in generic position and without singular point, the ratio of the running times between other implementations and isotop is increasing with the degree of the curve. In other words, except for random curves, isotop tends to perform better, compared to others, when the degree increases.

6. CONCLUSION

We presented a new algorithm and implementation for computing the topology of plane algebraic curves. Instead of a cylindrical algebraic decomposition based method, our algorithm relies upon Gröbner bases, Rational Univariate Representations and hence avoids computations with algebraic numbers even in non-generic cases. A strength of our approach is to be insensitive to the non-genericity of the curve. As demonstrated by the experiments, our implementation is competitive with state-of-the-art C/C++ implementations in the case of generic curves and faster for high-degree non-generic ones. Future work includes taking advantage of the possible decomposition of the curve into factors. We already decompose the system of critical points into systems of extreme and singular points. One natural step further would be to consider the primary decompositions of the ideals and thus work with systems of lower complexity.

Acknowledgments

We would like to thank the authors of Alcix, Cad2d, Insulate, and Top for supplying their code and O. Labs for his list of curve challenges.

7. REFERENCES

### Table 1: Running times in seconds (averaged over 5 runs) for random bivariate polynomials with 50% non-zero coefficients of bitsize 32.

<table>
<thead>
<tr>
<th>degree</th>
<th>bitsize</th>
<th>ISOTOP</th>
<th>Alcix</th>
<th>Cad2d</th>
<th>Top</th>
<th>Insulate</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>32</td>
<td>3.8</td>
<td>0.42</td>
<td>1.6</td>
<td>1.4</td>
<td>4.3</td>
</tr>
<tr>
<td>12</td>
<td>32</td>
<td>9.7</td>
<td>2.2</td>
<td>2.3</td>
<td>4.8</td>
<td>8.1</td>
</tr>
<tr>
<td>14</td>
<td>32</td>
<td>22</td>
<td>3.7</td>
<td>4.3</td>
<td>15</td>
<td>26</td>
</tr>
<tr>
<td>16</td>
<td>32</td>
<td>72</td>
<td>4.5</td>
<td>6.1</td>
<td>38</td>
<td>39</td>
</tr>
<tr>
<td>18</td>
<td>32</td>
<td>160</td>
<td>16</td>
<td>12</td>
<td>94</td>
<td>98</td>
</tr>
<tr>
<td>20</td>
<td>32</td>
<td>320</td>
<td>31</td>
<td>25</td>
<td>210</td>
<td>120</td>
</tr>
</tbody>
</table>

### Table 2: Running times (averaged over five runs) in seconds for resultants of two random trivariate polynomials, both of total degree 4 or 5, with 50% non-zero coefficients of bitsize 8.

<table>
<thead>
<tr>
<th>degree</th>
<th>τ</th>
<th>ISOTOP</th>
<th>Alcix</th>
<th>Cad2d</th>
<th>Top</th>
<th>Insulate</th>
</tr>
</thead>
<tbody>
<tr>
<td>12</td>
<td>96</td>
<td>4.1</td>
<td>6.5</td>
<td>1.6</td>
<td>9 or &gt;600</td>
<td>&gt;600</td>
</tr>
<tr>
<td>15</td>
<td>96</td>
<td>15</td>
<td>38</td>
<td>10</td>
<td>40 or &gt;600</td>
<td>&gt;600</td>
</tr>
<tr>
<td>18</td>
<td>96</td>
<td>49</td>
<td>120</td>
<td>66</td>
<td>&gt;600</td>
<td>&gt;600</td>
</tr>
<tr>
<td>21</td>
<td>96</td>
<td>140</td>
<td>510</td>
<td>510</td>
<td>&gt;1200</td>
<td>&gt;1200</td>
</tr>
</tbody>
</table>

### Table 3: Running times (averaged over five runs) in seconds for non-generic curves generated by the product of a curve $f$ with two of its translates $f(x, y + 1)$ and $f(x, y + 2)$. The curve $f$ is chosen randomly with degree between 4 and 7 and 50% non-zero coefficients of bitsize 32, resulting in curves of bitsize $\tau$. The running-time discrepancy of Top is large and is not averaged.

<table>
<thead>
<tr>
<th>$d_1$</th>
<th>$d_2$</th>
<th>degree</th>
<th>$\tau$</th>
<th>ISOTOP</th>
<th>Alcix</th>
<th>Cad2d</th>
<th>Top</th>
<th>Insulate</th>
</tr>
</thead>
<tbody>
<tr>
<td>3</td>
<td>3</td>
<td>18</td>
<td>96</td>
<td>26</td>
<td>93</td>
<td>160</td>
<td>&gt;600</td>
<td>&gt;600</td>
</tr>
<tr>
<td>3</td>
<td>4</td>
<td>24</td>
<td>112</td>
<td>250</td>
<td>510 or &gt;1800</td>
<td>aborted</td>
<td>&gt;1800</td>
<td>&gt;1800</td>
</tr>
</tbody>
</table>

### Table 4: Running times (averaged over five runs) in seconds for non-generic curves generated by the product of a curve $f$ with its translate $f(x, y + 1)$. The curve $f$ is the resultant of two random trivariate polynomials of total degree $d_1$ and $d_2$ and 50% non-zero coefficients of bitsize 8. The running-time discrepancy of Alcix is large for degree 24 and is not averaged.

<table>
<thead>
<tr>
<th>$d_1$</th>
<th>$d_2$</th>
<th>degree</th>
<th>$\tau$</th>
<th>ISOTOP</th>
<th>Alcix</th>
<th>Cad2d</th>
<th>Top</th>
<th>Insulate</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>64</td>
<td>39</td>
<td>180</td>
<td>25</td>
<td>&gt;600</td>
<td>&gt;600</td>
<td>4.5</td>
<td>0.6</td>
</tr>
<tr>
<td>12</td>
<td>64</td>
<td>240</td>
<td>300</td>
<td>aborted</td>
<td>&gt;1800</td>
<td>&gt;1800</td>
<td>1.2</td>
<td>-</td>
</tr>
<tr>
<td>14</td>
<td>64</td>
<td>350</td>
<td>&gt;1800</td>
<td>aborted</td>
<td>&gt;1800</td>
<td>&gt;1800</td>
<td>&gt;5</td>
<td>-</td>
</tr>
</tbody>
</table>

### Table 5: Running times (averaged over five runs) in seconds for polynomials of the form $g = f^2(x, y) + f^2(x, -y)$. The random polynomials $f$ have bitsize 32, term density 50% and degrees varying between 5 and 7.


