An Elimination Method for Solving Bivariate Polynomial Systems: Eliminating the Usual Drawbacks

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Abstract
We present an exact and complete algorithm to isolate the real solutions of a zero-dimensional bivariate polynomial system. The proposed algorithm constitutes an elimination method which improves upon existing approaches in a number of points. First, the amount of purely symbolic operations is significantly reduced, that is, only resultant computation and square-free factorization is still needed. Second, our algorithm neither assumes generic position of the input system nor demands for any change of the coordinate system. The latter is due to a novel inclusion predicate to certify that a certain region is isolating for a solution. Our implementation exploits graphics hardware to expedite the resultant computation. Furthermore, we integrate a number of filtering techniques to improve the overall performance. Efficiency of the proposed method is proven by a comparison of our implementation with two state-of-the-art implementations, that is, LGP and Maple’s ISOLATE. For a series of challenging benchmark instances, experiments show that our implementation outperforms both contestants.

1 Introduction
Finding the real solutions of a bivariate polynomial system is a fundamental problem with numerous applications in computational geometry, computer graphics and computer aided geometric design. In particular, topology and arrangement computations for algebraic curves [14, 13, 7, 20] crucially rely on the computation of common intersection points of the given curves (and also the curves defined by their partial derivatives). For the design of robust and certified algorithms, we aim for exact methods to determine isolating regions for all solutions. Such methods should be capable of handling any input, that is, even systems with multiple solutions. The proposed algorithm BISOLVE constitutes such an exact and complete approach. Its input is a zero-dimensional (i.e., there exist only finitely many solutions) polynomial system \( f(x, y) = g(x, y) = 0 \) defined by two bivariate polynomials with integer coefficients. BISOLVE computes disjoint boxes \( B_1, \ldots, B_m \subset \mathbb{R}^2 \) for all real solutions, where each box \( B_i \) contains exactly one solution (i.e., \( B_i \) is isolating). In addition, the boxes can be refined to an arbitrary small size.

Main results. BISOLVE constitutes a classical elimination method which follows the same basic idea as the GRID method from [12] and the hybrid method proposed in [21]. More precisely, in a first step, the variables \( x \) and \( y \) are separately eliminated by means of a resultant computation. Then, in the second step, for each possible candidate (represented as pair of projected solutions in \( x \)- and \( y \)-direction), we check whether it actually constitutes a solution of the given system or not. The proposed method comes with a number of improvements compared to the aforementioned approaches and also to other existing elimination techniques [3, 25, 29, 13]. First, we tremendously reduced the amount of purely symbolic computations, namely, our method only demands for resultant computation and square-free factorization of univariate polynomials with integer coefficients. Second, our implementation profits from a novel approach [17, 16] to compute resultants exploiting the power of Graphics Processing Unite (GPUs). We
remark that, in comparison to the classical resultant computation on the CPU, the GPU implementation is typically more than 100-times faster. Our experiments show that, for the considered instances, the resultant computation is no longer a “global” bottleneck of an elimination approach. Third, the proposed method never uses any kind of a coordinate transformation, even for non-generic input. The latter is due to a novel inclusion predicate which combines information from the resultant computation and a homotopy argument to prove that a certain candidate box is isolating for a solution. Since we never apply a change of coordinates, our method particularly profits in the case where $f$ and $g$ are sparse or where we are only interested in “local” solutions within a given box. Finally, we integrated a series of additional filtering techniques which allow us to significantly speed up the computation for many instances.

We implemented our algorithm as a prototypical package of CGAL and ran our software on numerous challenging benchmark instances. For comparison, we considered two currently state-of-the-art implementations, that is, ISOLATE (based on Rs by Fabrice Rouillier with ideas from [29]) from Maple 13 and Lgp by Xiao-Shan Gao et al. [8]. Our experiments show that our method is efficient as it outperforms both contestants for most instances. More precisely, our method is comparable for all considered instances and typically between 5 and 10-times faster. For some instances, we even improve by a factor of 50 and more. Our filters apply to many input systems and crucially contribute to the overall performance. We further remark that the gain in performance is not solely due to the resultant computation on the GPU but rather due to the combination of the sparse use of purely symbolic computations and efficient (approximate) subroutines. We prove the latter fact by providing running times with and without fast GPU-resultant computation.

Related Work. Since polynomial root solving is such an important problem in several fields, plenty of distinct approaches exist and many textbooks are dedicated to this subject. In general, we distinguish between two kinds of methods.

The first comprises non-certified or non-complete methods which give, in contrast to our goal here, no guarantee on correctness or termination (e.g., if multiple roots exists). Representatives of this category are numerical (e.g. homotopy methods [33]) or subdivision methods (e.g., [27, 6, 2]). A major strength of these methods is that they are very efficient for most instances due to their use of approximate computations such as provided by IntBis, ALIAS, IntLab or MPFI.

The second category consists of certified and complete methods, to which ours is to be added. So far, only elimination methods based on (sparse) resultants, rational univariate representation, Groebner bases or eigenvalues have proven to be efficient representatives of this category; see, for instance, [28, 11, 35, 37] for introductions to such symbolic approaches. Common to all these methods is that they combine a projection and a lifting step similar to the proposed approach. Recent exact and complete implementations for computing the topology of algebraic curves and surfaces [14, 20, 5] also make use of such elimination techniques. However, already this low dimensional application shows the main drawback of elimination methods, that is, they tremendously suffer from costly symbolic computations. Furthermore, the given system might be in non-generic position which makes the lifting step non-trivial. In such “hard situations”, the existing approaches perform a coordinate transformation (or project in generic direction) which eventually increases the complexity of the input polynomials. In particular, if we are only interested in “local” solutions within a given box, such methods induce a huge overhead of purely symbolic computations. The proposed algorithm constitutes a contribution in two respects: The number of symbolic steps are crucially reduced and partially (resultant computation) outsourced to the GPU. In addition, generic and non-generic situations are treated in the same manner and, thus, a coordinate transformation which induces an overhead of symbolic computations is no longer needed.

\[1\] The system $f = g = 0$ is non-generic if there exist two solutions sharing a common coordinate.

\[2\] Subdivision methods can be made certifying and complete when considering worst case separation bounds for the solutions, an approach which has not shown effective in practice so far.
2 Setting

The input of our algorithm is the following polynomial system

\[ f(x, y) = \sum_{i,j \in \mathbb{N}: i+j \leq m} f_{ij} x^i y^j = 0, \quad g(x, y) = \sum_{i,j \in \mathbb{N}: i+j \leq n} g_{ij} x^i y^j = 0, \tag{2.1} \]

where \( f, g \in \mathbb{Z}[x, y] \) are polynomials of total degrees \( m \) and \( n \), respectively. We also write

\[ f(x, y) = \sum_{i=0}^{m} f_i(y) x^i \quad \text{and} \quad g(x, y) = \sum_{i=0}^{n} g_i(y) x^i, \]

where \( f_i(y), g_i(y) \in \mathbb{Z}[x], f_i(x), g_i(x) \in \mathbb{Z}[y] \) and \( m_x, n_x \) and \( m_y, n_y \) denote the degrees of \( f \) and \( g \) considered as polynomials in \( x \) and \( y \), respectively. Throughout the paper, it is assumed that \( f \) and \( g \) have no common factors.\(^\text{3}\)

Hence, the set \( V_C := \{(x, y) \in \mathbb{C}^2 | f(x, y) = g(x, y) = 0\} \) of (complex) solutions of \((2.1)\) is zero-dimensional and consists, by Bézout’s theorem, of at most \( m \cdot n \) distinct elements.

Our algorithm outputs disjoint boxes \( B_k \subset \mathbb{R}^2 \) such that the union of all \( B_k \) contains all real solutions of \((2.1)\) and each \( B_k \) is isolating, that is, it contains exactly one solution.

Notation. For an interval \( I = (a, b) \subset \mathbb{R} \), \( m_I := (a + b)/2 \) denotes the center and \( r_I := (b - a)/2 \) the radius of \( I \). For an arbitrary \( m \in \mathbb{C} \) and \( r \in \mathbb{R}^+ \), \( \Delta_r(m) \) denotes the disc with center \( m \) and radius \( r \).

3 The Algorithm

3.1 Resultants

Our algorithm is based on well known elimination techniques, namely, to consider the projections

\[ V_C^{(x)} := \{ x \in \mathbb{C} | \exists y \in \mathbb{C} \text{ with } f(x, y) = g(x, y) = 0 \}, \quad V_C^{(y)} := \{ y \in \mathbb{C} | \exists x \in \mathbb{C} \text{ with } f(x, y) = g(x, y) = 0 \} \]

of all complex solutions \( V_C \) onto the \( x \)- and \( y \)-coordinate. Resultant computation is a well studied tool to obtain an algebraic description of these projection sets, that is, polynomials whose roots are exactly the projections of the solution set \( V_C \). The resultant \( R^{(y)} = \text{res}(f, g, y) \) of \( f \) and \( g \) with respect to the variable \( y \) is the determinant of the \((m_y + n_y) \times (m_y + n_y)\) Sylvester matrix:

\[
S^{(y)}(f, g) := \begin{bmatrix}
 f^{(y)}_{m_y} & f^{(y)}_{m_y-1} & \cdots & f^{(y)}_{0} & 0 & \cdots & 0 \\
 \vdots & \ddots & \ddots & \vdots & \ddots & \ddots & \vdots \\
 0 & \cdots & 0 & f^{(y)}_{m_y} & f^{(y)}_{m_y-1} & \cdots & f^{(y)}_{0} \\
 g^{(y)}_{m_y} & g^{(y)}_{m_y-1} & \cdots & g^{(y)}_{0} & 0 & \cdots & 0 \\
 \vdots & \ddots & \ddots & \vdots & \ddots & \ddots & \vdots \\
 0 & \cdots & 0 & g^{(y)}_{m_y} & g^{(y)}_{m_y-1} & \cdots & g^{(y)}_{0}
\end{bmatrix}
\]

From the definition, it follows that \( R^{(y)} \in \mathbb{Z}[x] \) has degree less than or equal to \( m \cdot n \). The resultant \( R^{(x)} = \text{res}(f, g, x) \) of \( f \) and \( g \) with respect to \( x \) is defined in completely analogous manner by considering \( f \) and \( g \) as polynomials in \( x \) instead of \( y \). As mentioned above the resultant polynomials have the following important property (cf. \([4]\) for a proof):

**Theorem 1** The roots of \( R^{(y)} \) and \( R^{(x)} \) are exactly the projections of the solutions of \((2.1)\) onto the \( x \)- and \( y \)-coordinate, respectively. More precisely, \( V_C^{(x)} = \{ x \in \mathbb{C} | R^{(y)}(x) = 0 \} \) and \( V_C^{(y)} = \{ y \in \mathbb{C} | R^{(x)}(y) = 0 \} \). The multiplicity of a root \( \alpha \) of \( R^{(y)} \) \((R^{(x)})\) is the sum of the intersection multiplicities\(^\text{4}\) of all solutions of \((2.1)\) with \( x \)-coordinate \((y \text{-coordinate}) \) \( \alpha \).

\(^3\)Otherwise, \( f \) and \( g \) have to be decomposed into common and non-common factors (not part of our algorithm).

\(^4\)The multiplicity of a solution \((x_0, y_0)\) of \((2.1)\) is defined as the dimension of the localization of \( \mathbb{C}[x, y]/(f, g) \) at \((x_0, y_0)\) considered as \( \mathbb{C} \)-vector space (cf.\([4]\) p.148).
3.2 Isolating the Solutions: Project, Separate and Validate

We start with the following high level description of the proposed algorithm which decomposes into three subroutines: In the first step (PROJECT), we project the complex solutions $V_C$ of (2.1) onto the x- and onto the y-axis. More precisely, we compute the restrictions $V_R^{(x)} := V_C^{(x)} \cap \mathbb{R}$ and $V_R^{(y)} := V_C^{(y)} \cap \mathbb{R}$ of the complex projection sets $V_C^{(x)}$ and $V_C^{(y)}$ to the real axes and isolating intervals for their elements. Obviously, the real solutions $V_R$ are contained in the cross product $\mathcal{C} := V_R^{(x)} \times V_R^{(y)} \subset \mathbb{R}^2$. In the second step (SEPARATE), we compute isolating discs which well separate the projected solutions from each other. The latter prepares the third step (VALIDATE) in which candidates of $\mathcal{C}$ are either discarded or certified to be a solution of (2.1). Our main theoretical contribution is the introduction of a novel predicate to ensure that a certain candidate $(\alpha, \beta) \in \mathcal{C} \cap \mathbb{R}$ actually fulfills $f(\alpha, \beta) = g(\alpha, \beta) = 0$ (cf. Theorem 1). For all candidates $(\alpha, \beta) \in \mathcal{C} \setminus \mathbb{R}$, simple interval arithmetic suffices to exclude $(\alpha, \beta)$ as a solution of (2.1).

We remark that, in order to increase the efficiency of our implementation, we also introduce additional filtering techniques to eliminate many of the candidates in $\mathcal{C}$. However, for the sake of clarity, we refrain from integrating our filtering techniques in the following description of the three subroutines. Filtering techniques are covered separately in Section 4.2. Section 4.1 briefly discusses a highly parallel algorithm on the graphics hardware to accelerate computations of the resultants needed in the first step.

PROJECT: We compute the resultant $R := R^{(y)} = \text{res}(f, g, y) \in \mathbb{Z}[x]$ and a square-free factorization of $R$. More precisely, we determine square-free and pairwise coprime factors $r_i \in \mathbb{Z}[x]$, $i = 1, \ldots, \deg(R)$, such that $R(x) = \prod_{i=1}^{\deg(R)} (r_i(x))^i$. We remark that, for some $i \in \{1, \ldots, \deg(R)\}$, $r_i(x) = 1$. Yun’s algorithm [16, Alg. 14.21] constructs such a square-free factorization by essentially computing greatest common divisors of $R$ and its higher derivatives in an iterative way. Next, we isolate the real roots $\alpha_{i,j}$, $j = 1, \ldots, \ell_i$, of the polynomials $r_i$. That is, we determine disjoint isolating intervals $I(\alpha_{i,j}) \subset \mathbb{R}$ such that each interval $I(\alpha_{i,j})$ contains exactly one root (namely, $\alpha_{i,j}$) of $r_i$ and the union of all $I(\alpha_{i,j})$, $j = 1, \ldots, \ell_i$, covers all real roots of $r_i$. For the real root isolation, we consider the Descartes method [10, 30] as a suited algorithm. From the square-free factorization we know that $\alpha_{i,j}, j = 1, \ldots, \ell_i$, is a root of $R$ with multiplicity $i$.

SEPARATE: We separate the real roots of $R = R^{(y)}$ from all other (complex) roots of $R$, a step which is crucial for the final validation. More precisely, let $\alpha = \alpha_{i_0,j_0}$ be the $j_0$-th real root of the polynomial $r_{i_0}$, where $i_0 \in \{1, \ldots, \deg(R)\}$ and $j_0 \in \{1, \ldots, \ell_{i_0}\}$ are arbitrary indices. We refine the corresponding isolating interval $I = (a, b) := I(\alpha)$ such that the disc $\Delta_{r_{i_0}}(m_{1})$ does not contain any root of $R^{(y)}$ except $\alpha$. For the refinement of $I$, we use quadratic interval refinement [1, 24] (QIR) which constitutes a highly efficient method because of its simple tests and the fact that it eventually achieves quadratic convergence.

In order to test whether the disc $\Delta_{r_{i_0}}(m_{1})$ isolates $\alpha$ from all other roots of $R$, we introduce a novel method based on the following test:

$$T_{K}^{p}(m, r) = |p(m)| - K \sum_{k \geq 1} \left| \frac{p^{(k)}(m)}{k!} \right| r^k > 0,$$

where $p \in \mathbb{R}[x]$ denotes an arbitrary polynomial and $m, r, K$ arbitrary real values. Then, the following theorem holds (cf. Appendix 5 for a proof):

Theorem 2 Consider a disk $\Delta = \Delta_{m}(r) \subset \mathbb{C}$ with center $m$ and radius $r$.

1. If $T_{K}^{p}(m, r)$ holds for some $K \geq 1$, then the closure $\overline{\Delta}$ of $\Delta$ contains no root of $p$.

2. If $T_{K}^{p}(m, r)$ holds for a $K \geq \sqrt{2}$, then $\overline{\Delta}$ contains at most one root of $p$.

Theorem 2 now directly applies to the above scenario. More precisely, $I$ is refined until $T_{3/2}^{(r_{i_0})'}(m_{1}, 8r_{i_1})$ and $T_{i}^{r_{i}}(m_{1}, 8r_{i_1})$ holds for all $i \neq i_0$. If the latter is fulfilled, $\Delta_{r_{i_0}}(m_{1})$ isolates $\alpha$ from all other roots of $R$. In this situation, we obtain a lower bound $LB(\alpha)$ for $|R(z)|$ on the boundary of $\Delta(\alpha) := \Delta_{2r_{i}}(m_{1})$:
Lemma 1 The disc $\Delta(\alpha) = \Delta_{2r_I}(m_I)$ isolates $\alpha$ from all other (complex) roots of $R$ and, for any $z$ on the boundary $\partial \Delta(\alpha)$ of $\Delta(\alpha)$, it holds that $|R(z)| > LB(\alpha) := 2^{-i\alpha-\deg(R)}|R(m_I - 2r_I)|$.

Proof: $\Delta(\alpha)$ is isolating as already $\Delta_{8r_I}(m_I)$ is isolating. Then, let $\beta \neq \alpha$ be an arbitrary root of $R$ and $d := |\beta - m_I| > 8r_I$ the distance between $\beta$ and $m_I$. Then, for any point $z \in \partial \Delta(\alpha)$, it holds that

$$\frac{|z - \beta|}{|m_I - 2r_I| - \beta} > \frac{d - 4r_I}{d + 2r_I} > \frac{1}{2} \quad \text{and} \quad \frac{|z - \alpha|}{|m_I - 2r_I| - \alpha} > \frac{r_I}{4r_I} > \frac{1}{4}.$$ 

Hence, it follows that

$$\frac{|R(z)|}{|R(m_I - 2r_I)|} > \prod_{\beta \neq \alpha; R(\beta) = 0} \frac{|z - \beta|}{|m_I - 2r_I| - \beta} \cdot \left(\frac{|z - \alpha|}{|m_I - 2r_I| - \alpha}\right)^{i\alpha} > 4^{-i\alpha}2^{-\deg(R)+i\alpha},$$

where each root $\beta$ occurs as many times in the above product as its multiplicity as a root of $R$.

We evaluate $LB(\alpha) = 2^{-i\alpha-\deg(R)}|R(m_I - 2r_I)|$ and store the interval $I(\alpha)$, the disc $\Delta(\alpha)$ and the lower bound $LB(\alpha)$ for $|R(z)|$ on the boundary $\partial \Delta(\alpha)$ of $\Delta(\alpha)$.

Proceeding in exactly the same manner for each real root $\alpha$ of $R(y)$, we get an isolating interval $I(\alpha)$, an isolating disc $\Delta(\alpha) = \Delta_{2r_I}(m_I)$ and a lower bound $LB(\alpha)$ for $|R(y)|$ on $\partial \Delta(\alpha)$. For the resultant polynomial $R(x)$, PROJECT and SEPARATE are processed in exactly the same manner: We compute $R(x)$ and a corresponding square-free factorization. Then, for each real root $\beta$ of $R(x)$, we compute a corresponding isolating interval $I(\beta)$, a disc $\Delta(\beta)$ and a lower bound $LB(\beta)$ for $|R(x)|$ on $\partial \Delta(\beta)$.

VALIDATE: We start with the following theorem:

Theorem 3 Let $\alpha$ and $\beta$ be arbitrary real roots of $R(y)$ and $R(x)$, respectively. Then,

1. the polydisc $\Delta(\alpha, \beta) := \Delta(\alpha) \times \Delta(\beta) \subset \mathbb{C}^2$ contains at most one (complex) solution of (2.1). If $\Delta(\alpha, \beta)$ contains a solution of (2.1), then this solution is real valued and equals $(\alpha, \beta)$.

2. For an arbitrary point $(z_1, z_2) \in \mathbb{C}^2$ on the boundary of $\Delta(\alpha, \beta)$, it holds that $|R(y)(z_1)| > LB(\alpha)$ if $z_1 \in \partial \Delta(\alpha)$, and $|R(x)(z_2)| > LB(\beta)$ if $z_2 \in \partial \Delta(\beta)$.

Proof: (1) is an easy consequence from the construction of the discs $\Delta(\alpha)$ and $\Delta(\beta)$. Namely, if $\Delta(\alpha, \beta)$ contains two distinct solutions of (2.1), then they would differ in at least one variable. Thus, one of the discs $\Delta(\alpha)$ or $\Delta(\beta)$ would contain two roots of $R(y)$ or $R(x)$. Since both discs are isolating for a root of the corresponding resultant polynomial, it follows that $\Delta(\alpha, \beta)$ contains at most one solution. In the case, where $\Delta(\alpha, \beta)$ contains a solution of (2.1), this solution must be real since, otherwise, $\Delta(\alpha, \beta)$ would also contain a corresponding complex conjugate solution ($f$ and $g$ have real valued coefficients). (2) follows directly from the definition of $\Delta(\alpha, \beta)$, the definition of $LB(\alpha)$, $LB(\beta)$ and Lemma 1.

We denote $B(\alpha, \beta) = I(\alpha) \times I(\beta) \subset \mathbb{R}^2$ a candidate box for a real solution of (2.1), where $\alpha$ and $\beta$ are real roots of $R(y)$ and $R(x)$, respectively. Due to Theorem 3, the corresponding “container polydisc” $\Delta(\alpha, \beta) \subset \mathbb{C}^2$ either contains no solution of (2.1) or $(\alpha, \beta)$ is the only solution contained in $\Delta(\alpha, \beta)$. Hence, for each candidate pair $(\alpha, \beta) \in \mathcal{C}$, it suffices to show that either $(\alpha, \beta)$ is no solution of (2.1) or the corresponding polydisc $\Delta(\alpha, \beta)$ contains at least one solution. In the following steps, we fix the polydiscs $\Delta(\alpha, \beta)$ whereas the boxes $B(\alpha, \beta)$ are further refined (by further refining the isolating intervals $I(\alpha)$ and $I(\beta)$). We also introduce exclusion and inclusion predicates such that, for sufficiently small $B(\alpha, \beta)$, either $(\alpha, \beta)$ can be discarded or certified as a solution of (2.1).
In order to exclude a candidate box, we use simple interval arithmetic. More precisely, we evaluate \( \Box f(B(\alpha, \beta)) \) and \( \Box g(B(\alpha, \beta)) \), where \( \Box f \) and \( \Box g \) constitute box functions for \( f \) and \( g \), respectively: If either \( \Box f(B(\alpha, \beta)) \) or \( \Box g(B(\alpha, \beta)) \) does not contain zero, then \((\alpha, \beta)\) cannot be a solution of \((2.1)\). Vice versa, if \((\alpha, \beta)\) is not a solution and \(B(\alpha, \beta)\) becomes sufficiently small, then either \(0 \notin \Box f(B(\alpha, \beta))\) or \(0 \notin \Box g(B(\alpha, \beta))\) and our exclusion predicate applies.

It remains to provide an inclusion predicate, that is, a method to ensure that a certain candidate \((\alpha, \beta) \in \mathcal{C}\) is actually a solution of \((2.1)\). We first rewrite the resultant polynomial \( R(y) \) as

\[
R(y)(x) = u(y)(x,y) \cdot f(x,y) + v(y)(x,y) \cdot g(x,y),
\]

where \(u(y), v(y) \in \mathbb{Z}[x,y]\). Furthermore, \(u(y)\) and \(v(y)\) can be expressed as determinants of “Sylvester-like” matrices \(U(y)\) and \(V(y)\). More precisely, \(U(y)\) and \(V(y)\) are obtained from \(S(y)(f,g)\) by replacing the last column with vectors \((g^m_{m-1} \ldots 0 \ldots 0)^T\) and \((0 \ldots 0 g^m_{m-1} \ldots 1)^T\) of appropriate size, respectively [19 p.287]. Both matrices have size \((n_y+m_y) \times (n_y+m_y)\) and univariate polynomials in \(x\) (the first \(n_y+m_y-1\) columns) or powers of \(y\) (only the last column) or zeros as entries. We now aim for upper bounds for \(|u(y)|\) and \(|v(y)|\) on the polydisc \(\Delta(\alpha, \beta)\). The polynomials \(u(y)\) and \(v(y)\) have huge coefficients and their computation, either via a signed remainder sequence or via determinant evaluation, is very costly. Hence, we directly derive such upper bounds from the corresponding matrix representations without computing \(u(y)\) and \(v(y)\): Due to Hadamard’s bound, \(|u(y)|\) is smaller than the product of the 2-norms of the column vectors of \(U(y)\). The absolute value of each of the entries of \(U(y)\) can be easily upper bounded by using interval arithmetic on a box in \(\mathbb{C}^2\) that contains the polydisc \(\Delta(\alpha, \beta)\). Hence, we get an upper bound on the 2–norm of each column vector and, thus, an upper bound \(UB(\alpha, \beta, u(y))\) for \(|u(y)|\) on \(\Delta(\alpha, \beta)\) by multiplying the bounds for the column vectors. In the same manner, we also derive an upper bound \(UB(\alpha, \beta, v(y))\) for \(|v(y)|\) on \(\Delta(\alpha, \beta)\). With respect to our second projection direction, we write \(R(x) = u(x) \cdot f + v(x) \cdot g\) with corresponding polynomials \(u(x), v(x) \in \mathbb{Z}[x,y]\). In exactly the same manner as done for \(R(y)\), we compute corresponding upper bounds \(UB(\alpha, \beta, u(x))\) and \(UB(\alpha, \beta, v(x))\) for \(|u(x)|\) and \(|v(x)|\) on \(\Delta(\alpha, \beta)\).

**Theorem 4** If there exists an \((x_0, y_0) \in \Delta(\alpha, \beta)\) with

\[
UB(\alpha, \beta, u(y)) \cdot |f(x_0, y_0)| + UB(\alpha, \beta, v(y)) \cdot |g(x_0, y_0)| < LB(\alpha) \quad \text{and} \quad (3.1)
\]

\[
UB(\alpha, \beta, u(x)) \cdot |f(x_0, y_0)| + UB(\alpha, \beta, v(x)) \cdot |g(x_0, y_0)| < LB(\beta), \quad (3.2)
\]

then \(\Delta(\alpha, \beta)\) contains a solution of \((2.1)\) and, thus, \(f(\alpha, \beta) = 0\).

**Proof:** The proof uses a homotopy argument. Namely, we consider the parameterized system

\[
f(x,y) - (1-t) \cdot f(x_0,y_0) = g(x,y) - (1-t) \cdot g(x_0,y_0) = 0, \quad (3.3)
\]

where \(t\) is an arbitrary real value in \([0,1]\). For \(t=1\), \((3.3)\) is equivalent to our initial system \((2.1)\). For \(t=0\), \((3.3)\) has a solution in \(\Delta(\alpha, \beta)\), namely, \((x_0,y_0)\). The complex solutions of \((3.3)\) continuously depend on the parameter \(t\). Hence, there exists a “solution path” \(\Gamma : [0,1] \mapsto \mathbb{C}^2\) which connects \(\Gamma(0) = (x_0, y_0)\) with a solution \(\Gamma(1) \in \mathbb{C}^2\) of \((2.1)\). We show that \(\Gamma(t)\) does not leave the polydisc \(\Delta(\alpha, \beta)\) and, thus, \((2.1)\) has a solution in \(\Delta(\alpha, \beta)\): Assume that the path \(\Gamma(t)\) leaves the polydisc, then there exists a \(t' \in [0,1]\) with \((x', y') = \Gamma(t') \in \partial\Delta(\alpha, \beta)\). We assume that \(x' \in \partial\Delta(\alpha)\) (the case \(y' \in \partial\Delta(\beta)\) is treated in analogous manner). Since \((x', y')\) is a solution of \((3.3)\) for \(t=t'\), we must have \(|f(x', y')| \leq |f(x_0, y_0)|\) and \(|g(x', y')| \leq |g(x_0, y_0)|\). Hence, it follows that

\[
|R(y)(x')| = |u(y)(x', y') f(x', y') + v(y)(x', y') g(x', y')| \leq |u(y)(x', y')| \cdot |f(x', y')| + |v(y)(x', y')| \cdot |g(x', y')| \\
\leq UB(\alpha, \beta, u(y)) \cdot |f(x_0, y_0)| + UB(\alpha, \beta, v(y)) \cdot |g(x_0, y_0)| < LB(\alpha).
\]
This contradicts the fact that $|R^{(y)}(x')|$ is lower bounded by $LB(\alpha)$. It follows that $\Delta(\alpha, \beta)$ contains a solution of (2.1) and, according to Theorem 3, this solution must be $(\alpha, \beta)$.

Theorem 4 now directly applies as an inclusion predicate. Namely, in each refinement of $B(\alpha, \beta)$, we choose an arbitrary $(x_0, y_0) \in B(\alpha, \beta)$ (e.g., the center $(m_{I(\alpha)}, m_{I(\beta)})$) of the candidate box $B(\alpha, \beta)$ and check whether both inequalities (3.1) and (3.2) are fulfilled. If $(\alpha, \beta)$ is a solution of (2.1), then both inequalities eventually hold and, thus, we have shown that $(\alpha, \beta)$ is a solution.

We remark that the upper bounds $UB(\alpha, \beta, u^{(y)})$, $UB(\alpha, \beta, v^{(y)})$, $UB(\alpha, \beta, u^{(x)})$ and $UB(\alpha, \beta, v^{(y)})$ are far from being optimal. Nevertheless, our inclusion predicate is still efficient since we can approximate the potential solution $(\alpha, \beta)$ with quadratic convergence due to QIR. Hence, the values $f(x_0, y_0)$ and $g(x_0, y_0)$ become very small after a few iterations. In order to improve the above bounds, we propose to consider more sophisticated methods from numerical analysis and matrix perturbation theory [22, 31]. Finally, we would like to emphasize that our method applies particularly well to the situation where we are only interested in the solutions of (2.1) within a given box $B$. We briefly discuss the algorithm; we refer the reader to [17, 16] for details.

Our approach is based on the classical “divide-conquer-combine” modular algorithm by Collins [9]. The algorithm can be summarized in the following steps. 1. Apply modular and evaluation homomorphisms to map the problem to computing a large set of problems over a simple domain. 2. Compute a set of univariate resultants over a prime field. 3. Recover the resultant through polynomial interpolation and Chinese remaindering.

Unfortunately, Collins’ algorithm in its original form is not suitable for a realization on the GPU. This is because the amount of parallelism exposed by the modular approach is far too low to satisfy the needs of massively-threaded architectures. We deal with this issue by reducing the problem to computations with structured matrices because matrix operations typically map very well to the GPU’s threading model. As a result, all steps of the algorithm except the initial modular reduction and partly the Chinese remaindering are run on the graphics hardware, thereby minimizing the amount of work to be done on the CPU. For expository purposes, we outline here the computation of univariate resultants in more detail.

Suppose, $f$ and $g$ are polynomials in $\mathbb{Z}[x]$ of degrees $m$ and $n$ respectively. It is clear that the resultant of $f$ and $g$ reduces to the triangular factorization of the Sylvester matrix $S$ (see Section 3.1). The matrix $S \in \mathbb{Z}^{r \times r}$ ($r = m + n$) is structured as it satisfies the displacement equation [23]:

$$S - Z_s S A^T = G B^T,$$

with $A = Z_m \oplus Z_n$ and $G, B \in \mathbb{Z}^{r \times 2}$.

Here $Z_s \in \mathbb{Z}^{s \times s}$ is a down-shift matrix zeroed everywhere except for 1’s on the first subdiagonal. Accordingly, the generators $G, B$ are matrices whose entries can be deduced from the matrix $S$ by inspection. Hence, we can apply the generalized Schur algorithm which operates on the matrix generators to compute the matrix factorization in $O(r^2)$ time, see [23, p. 323].

In short, the Schur algorithm is an iterative procedure: In each step, it brings the matrix generators to a “special form” from which triangular factors can easily be deduced based on the displacement equation.
Figure 4.1: (a) Intervals containing the roots of \( f(\alpha, y) \) and \( g(\alpha, y) \) are refined until they either do not overlap or are fully included in candidate boxes. In the former case, the boxes can be discarded. (b) Unvalidated candidates are passed to bidirectional filter which runs bitstream isolation in another direction.

Using division-free modifications this procedure can be efficiently performed in a finite field giving rise to the factorization algorithm running in \( O(r) \) time using \( r \) processors.

Suppose that we have evaluated the polynomials \( f, g \in \mathbb{Z}[x, y] \) as defined in (2.1) at a number of points \( x_i \in \mathbb{Z}_p \) and computed a set of univariate resultants over a prime field \( \mathbb{Z}_p \), that is, \( z_i^{(p)} = \text{res}(f(x_i, y), g(x_i, y), y) \in \mathbb{Z}_p \). Then, the resultant polynomial \( R^{(y)}(x) \) is interpolated over the prime field \( \mathbb{Z}_p \) and eventually lifted to an integer solution via Chinese remaindering. We remark that polynomial interpolation corresponds to solving the Vandermonde system. Again, exploiting the structure of Vandermonde matrix we can use the Schur algorithm to solve the system in a small parallel time.

4.2 Filters

Besides the parallel resultant computation, our algorithm elaborates some filtering techniques to early validate a majority of the candidates.

As first step, we group candidates along the same vertical line (a fiber) at an \( x \)-coordinate \( \alpha \) (a root of \( R^{(y)} \)) to process them together. This allows us to use extra information on the real roots of \( f(\alpha, y) \in \mathbb{R}[y] \) and \( g(\alpha, y) \in \mathbb{R}[y] \) for candidate validation. We replace the tests based on interval evaluation (see page 5) by a test based on the bitstream Descartes isolator [15] (Bdc for short). This method allows us to isolate the real roots of a polynomial with “bitstream” coefficients, that is, coefficients that can be approximated to arbitrary precision. Bdc starts from an interval guaranteed to contain all real roots of a polynomial, and proceeds with interval subdivisions giving rise to a subdivision tree. Accordingly, the approximation precision for coefficients is increased in each step of the algorithm. Each leaf of the tree is associated with an interval and stores an upper and a lower bound on the number of real roots within this interval based on Descartes’ Rule of Signs. An interval is not further subdivided when both bounds equal 0, where the interval is discarded, or 1, where we have found an isolating interval. Isolating intervals can be refined to arbitrary precision. We remark that Bdc terminates if all real roots are simple. Otherwise, intervals which contain a multiple root are further refined but never certified to contain a root.

In our algorithm, we apply Bdc to the polynomials \( f(\alpha, y) \) and \( g(\alpha, y) \). Eventually, intervals that do not share a common root of both polynomials will be discarded. This property is essential for our “filtered” algorithm: a candidate box \( B(\alpha, \beta) \) can be rejected as soon as the associated \( y \)-interval \( I(\beta) \) does not overlap with at least one of the isolating intervals associated with \( f(\alpha, y) \) or \( g(\alpha, y) \); see Figure 4.1 (a).

Grouping candidates along a fiber \( x = \alpha \) also enables us to use combinatorial tests to discard or to certify them. First, when the number of certified solutions reaches \( \text{mult}(\alpha) \), the remaining candidates are automatically discarded because each real solution contributes at least once to \( \alpha \)'s multiplicity as a

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5Here we are not concerned with the fact that Vandermonde systems are notoriously ill-conditioned since all operations are performed in a finite field.
Table 1: Description of the curves used in the first part of experiments. In case only a single curve given, the second curve is taken to be the first derivative w.r.t. $y$-variable.

| Instance  | Description                                      | Instance   | Description                                      |
|-----------|--------------------------------------------------|------------|--------------------------------------------------|
| L4_circles| circles w.r.t. L4-norm, clustered solutions       | SA_4.4_psp| singular points with high tangencies, displaced  |
| curve_issac | a curve appeared in [8]                           | FTT_5_4.4  | many non-rational singularities                  |
| tryme     | covertical solutions, many candidates to check    | dfold_10.6 | a curve with many half-branches                   |
| large_curves | large number of solutions                           | cov_sol_20 | covertical solutions                             |
| degree_6_surf | silhouette of an algebraic surface, covertical solutions in both directions | mignote_xy | a product of $x/y$- Mignotte polynomials, displaced; many clustered solutions |
| challenge_12 | many candidate solutions to be checked             | spider     | degenerate curve, many clustered solutions        |

* These curves were taken from [20].

root of $R^{(y)}$ (cf. Theorem 1). Second, if $\text{mult}(\alpha)$ is odd and all except one candidate along the fiber are discarded, then the remaining candidate must be a real solution. This is because complex roots come in conjugate pairs and, thus, do not change the parity of $\text{mult}(\alpha)$. We remark that, in case where the system (2.1) is in \textit{generic position} and the multiplicities of all roots of $R$ are odd, the combinatorial test already suffices to certify all solutions without the need to apply our inclusion predicate.

Now, suppose that, after the combinatorial test, there are several candidates left along a fiber. For instance, the latter can indicate the presence of \textit{covertical} solutions. In this case, before using the inclusion predicate, we can apply the aforementioned filters in \textit{horizontal} direction. More precisely, we construct the lists of unvalidated candidates sharing the same $y$-coordinate $\beta$ and process them along a horizontal fiber. For this step, we initialize the bitstream trees for $f(x, \beta)$ and $g(x, \beta)$ and proceed in exactly the same way as done for vertical fibers; see Figure 4.1 (b). Candidates that still remain undecided after all tests are processed by considering our inclusion predicate. In Section 5, where we next examine the efficiency of our filters, we will refer to this procedure as the \textit{bidirectional} filter.

5 Implementation & Experiments

We have implemented our algorithm as a prototypical package of CGAL.\(^6\) As throughout the library we follow a \textit{generic programming paradigm} that, for instance, enables us to easily exchange the number types used or the method to isolate the roots of a polynomial without altering the main structure of the implementation.

In our experiments, we have used the number types provided by GMP 4.3.1 and fast polynomial GCD from NTL 5.5 library.\(^7\) All experiments have been run on 2.8GHz 8-Core Intel Xeon W3530 with 8 MB of L2 cache under Linux platform. For the GPU-part of the algorithm, we have used the GeForce GTX480 graphics processor (Fermi Core). We compared our approach to the bivariate version of ISOLATE (based on RS by Fabrice Rouillier\(^8\)) and Lgp by Xiao-Shan Gao et al.\(^9\) We remark that, for the important substep of isolating the real roots of the elimination polynomial, all three contestants (including our implementation) use the highly efficient implementation provided by RS.

Our tests consist of two parts: In the first part, we consider “special” curves (and their derivative w.r.t. $y$-variable) selected in the aim of challenging different parts of the algorithm and showing the efficiency of the filtering techniques given in Section 4.2. These curves, for instance, have many singularities or high-curvature points which requires many candidates to be tested along each vertical line, or prohibit the use of special filters. Descriptions of the considered curves and corresponding timings are listed in Table 1 and Table 2 respectively. In the second part of our experiments, we study the performance of

\(^6\)Computational Geometry Algorithms Library, \url{www.cgal.org}

\(^7\)GMP: \url{http://gmplib.org}, NTL: \url{http://www.shoup.net/ntl}

\(^8\)RS: \url{http://www.loria.fr/equipes/vegas/rs}

\(^9\)The software is available at \url{http://www.mmrc.iss.ac.cn/~xgao/software.html}
the Bisolve on random polynomials with increasing total degrees and coefficient bit-lengths. We refer the reader to Table 3 for the corresponding timings. Appendix B features further experiments.

In columns 4–8, the experiments for our algorithm are given with all filters set on (BS+ALLFILTERS), with bitstream and combinatorial filter (BS+BSTR+COMB), with bitstream filter only (BS+BSTR) and with all filters set off (BS). For Bisolve, we report timings respectively with and without GPU resultant algorithm. For the remaining configurations we show only the timings using GPU resultants. CPU-based timings can easily be obtained by taking the difference between Bisolve-columns.

One can observe that our algorithm is generally superior to Isolate and Lgp even if the filters are not used. By comparing columns 5–8 in the table, one can see that filtering sometimes results in a significant performance improvement. The combinatorial test is particularly useful when the defining polynomials of the system (2.1) have large degrees and/or large coefficient bit-length while at the same time the number of covertical or singular solutions is small compared to the total number of candidates being checked. The bidirectional filter is advantageous when the system has covertical solutions in one direction (say along the $y$-axis) which are not cohorizontal. This is essentially the case for challenge$_{12}$, cov$_{sol,20}$ and spider.

Another strength of our approach relates to the fact that the amount of symbolic operations is crucially reduced. Hence, when the time for computing resultants is dominating, the GPU-based algorithm offers a speed-up by the factor of 2–5 over the version with default resultant implementation. It is also worth mentioning that both Isolate and Lgp benefit from the fast resultant computation available in Maple 13 while CGAL’s default resultant computation is generally much slower than that of Maple. As a result, there is a large discrepancy columns 4 and 5 for Bisolve.

Table 3 lists timings for experiments with random curves. Each instance consists of five curves of the same degree (9 or 15, dense or sparse) and we report the average time to compute the solutions for one of all ten pairs of curves. In order to analyze the influence of the coefficients’ bit-lengths, we multiplied each curve by $2^k$ with $k \in \{128, 512, 2048\}$ and increased the constant coefficient by one. Since the latter operation constitutes only a small perturbation of the vanishing set of the input system, the number of solutions remains constant while the content of the polynomials’ coefficients also stays trivial. We see that the bidirectional filtering is not of any advantage because the system defined by random polynomials is unlikely to have covertical solutions. However, in this case, most candidates are rejected by the combinatorial check, thereby omitting (a more expensive) test based on Theorem 4. This results

\[timeout: \text{algorithm timed out (> 1500 sec)}\]

### Table 2: Experiments for the curves listed in Table 1. Execution times are in seconds, including resultant computations. Bisolve-GPU: our approach with GPU-resultants; Bisolve-CPU: our approach with CGAL’s CPU-resultants; Isolate and Lgp use Maple’s implementation for the resultant computation. Bold face indicates default setup for Bisolve.

| Instance | $y$-degree | #sols | BS+ALLFILTERS | BS+BSTR+COMB | BS+BSTR | BS | Isolate | Lgp |
|----------|------------|-------|---------------|--------------|---------|----|---------|-----|
| L4,circles | 16 | 17 | 2.74 | 1.68 | 1.52 | 1.71 | 0.68 | 1.20 | 7.40 |
| curve_issac | 15 | 18 | 4.30 | 3.21 | 2.70 | 3.47 | 1.84 | 70.91 | 3.54 |
| tryme | 24, 34 | 20 | 98.56 | 29.31 | 31.63 | 89.83 | 89.86 | 167.81 | 176.86 |
| large_curves | 24, 19 | 137 | 110.20 | 91.15 | 90.82 | 376.71 | 377.55 | 501.52 | 138.35 |
| degree_6_surf | 42 | 13 | 149.33 | 17.46 | 16.50 | 18.63 | 62.18 | timeout | 133.77 |
| challenge$_{12}$ | 40 | 99 | 108.74 | 23.07 | 27.66 | 27.20 | 195.76 | 41.13 | 40.86 |
| SA4.4.ep | 33 | 2 | 155.92 | 2.83 | 2.85 | 3.81 | 8.57 | 296.02 | 56.30 |
| FTT5.4_4 | 40 | 62 | 73.89 | 17.58 | 20.92 | 20.99 | 111.22 | timeout | 199.92 |
| dfold10.6 | 32 | 21 | 25.20 | 4.80 | 3.12 | 3.19 | 3.54 | 3.14 | 3.84 |
| cov$_{sol,20}$ | 20 | 8 | 27.25 | 12.36 | 36.10 | 42.81 | 52.16 | 762.80 | 175.85 |
| mignotte_xy | 32 | 30 | 545.88 | 438.38 | 440.64 | 986.68 | 1021.50 | timeout | timeout |
| spider | 28 | 38 | 389.06 | 81.63 | 87.44 | 135.15 | 314.56 | timeout | timeout |

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10 Authors are indebted to CGAL developers working on resultants.
Table 3: Averaged running times for 10 pairs of curves defined by random polynomials of degree 9 and 15 with increasing bit-lengths (given by \textit{shift} parameter). For description of configurations, see Table 2.

| Density of polynomials | $y$-degree | shift | avg. #sols | BS+allfilters CPU | BS+allfilters GPU | BS+bstr+comb CPU | BS+bstr+comb GPU | BS+bstr CPU | BS+bstr GPU | Isolate Maple | Lgp Maple |
|------------------------|------------|-------|------------|-------------------|-------------------|-------------------|-------------------|-------------|-------------|--------------|------------|
| dense                  | 9,9        | -     | 5.6        | 0.30              | 0.21              | 0.22              | 0.21              | 0.19        | 0.19        | 0.33         | 0.24       |
|                        |            | 128   |            | 0.48              | 0.15              | 0.16              | 0.15              | 0.49        | 0.66        | 0.66         | 0.93       |
|                        |            | 512   |            | 2.22              | 0.31              | 0.31              | 0.31              | 2.03        | 1.51        | 1.51         | 3.33       |
|                        |            | 2048  |            | 16.47             | 2.07              | 2.06              | 2.07              | 13.86       | 7.48        | 7.48         | 102.37     |
| dense                  | 15,15      | -     | 5.0        | 1.82              | 0.71              | 0.70              | 0.71              | 1.64        | 6.85        | 3.88         | 3.88       |
|                        |            | 128   |            | 6.02              | 0.69              | 0.67              | 0.67              | 3.74        | 14.66       | 8.31         | 8.31       |
|                        |            | 512   |            | 32.18             | 1.48              | 1.45              | 1.48              | 14.35       | 38.27       | 22.36        | 22.36      |
|                        |            | 2048  |            | 251.07            | 8.97              | 8.94              | 8.97              | 94.09       | 141.69      | 102.36       | 102.36     |
| sparse                 | 9,9        | -     | 4.5        | 0.10              | 0.07              | 0.07              | 0.07              | 0.09        | 0.07        | 0.07         | 0.22       |
|                        |            | 128   |            | 0.14              | 0.08              | 0.08              | 0.08              | 0.21        | 0.20        | 0.20         | 0.57       |
|                        |            | 512   |            | 0.46              | 0.16              | 0.15              | 0.15              | 0.85        | 0.70        | 0.70         | 1.74       |
|                        |            | 2048  |            | 3.11              | 0.84              | 0.84              | 0.84              | 5.86        | 5.40        | 5.40         | 7.38       |
| sparse                 | 15,15      | -     | 3.8        | 0.65              | 0.36              | 0.36              | 0.36              | 0.66        | 0.99        | 0.99         | 1.24       |
|                        |            | 128   |            | 1.55              | 0.46              | 0.47              | 0.46              | 1.50        | 4.68        | 4.68         | 3.51       |
|                        |            | 512   |            | 7.70              | 1.55              | 1.55              | 1.55              | 6.80        | 16.01       | 11.93        | 11.93      |
|                        |            | 2048  |            | 58.97             | 13.45             | 13.38             | 13.46             | 51.14       | 132.76      | 74.03        | 74.03      |

in a clear speed-up over a “non-filtered” version. Also, observe that GPU-BISOLVE is not vulnerable to increasing the bit-length of coefficients while this becomes critical for ISOLATE’s and Lgp’s performance. We have also observed that, for our filtered versions, the time for the validation step is almost independent of the bit-lengths.

We omit experiments to refine the solution boxes to certain precision as this matches the efficiency of QIR due to the fact that we have algebraic descriptions for solutions’ $x$- and $y$-coordinates.

6 Summary and Outlook

We propose an exact and complete method to isolate the real solutions of a bivariate polynomial system. Our algorithm is designed to reduce the number of purely symbolic operations as much as possible. Eventually, only resultant computation and square-free factorization of the resultant polynomial are still needed. By transferring the resultant computation to the GPU, we are able to remove a major bottleneck of elimination approaches. In order to further improve our implementation, we aim to outsource the square-free factorization to the GPU as well, a step which seems to be feasible since factorization is also well suited for a ”divide-conquer-combine” modular approach. Since our initial motivation was to speed up the topology and arrangement computation for algebraic curves and surfaces, we plan to extend our method towards this direction. Furthermore, it would be interesting to extend our algorithm to handle higher dimensional systems or complex solutions. Finally, we would like to investigate in hybrid methods such as the combination of a numerical complex root solver and an exact post certification method serving as an additional filter in the validation step (in the spirit of [34, 32]). We are convinced that most of the candidate boxes could be treated even more efficiently by the use of such methods. We claim that, eventually, the total costs for solving a bivariate system should only be dominated by those of the root isolation step for the elimination polynomial. For many instances, our experiments already hint to the latter claim. We aim to further improve our implementation to show this behavior for all instances and to provide a proof in terms of complexity as well.
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\section*{A Proofs}

\textbf{Theorem 5} Consider a disk $\Delta = \Delta_m(r) \subset \mathbb{C}$ with center $m$ and radius $r$.

1. If $T_K^p(m,r)$ holds for some $K \geq 1$, then the closure $\overline{\Delta}$ of $\Delta$ contains no root of $p$.

2. If $T_K^p(m,r)$ holds for a $K \geq \sqrt{2}$, then $\overline{\Delta}$ contains at most one root of $p$.

\textbf{Proof:} (1) follows from a straightforward computation: For each $z \in \overline{\Delta}$, we have

$$p(z) = p(m + (z - m)) = p(m) + \sum_{k \geq 1} \frac{p^{(k)}(m)}{k!} (z - m)^k$$

and, thus,

$$\frac{|p(z)|}{|p(m)|} \geq 1 - \frac{1}{|p(m)|} \sum_{k \geq 1} \frac{|p^{(k)}(m)|}{k!} |z - m|^k > \left(1 - \frac{1}{K}\right)$$

since $|z - m| \leq r$ and $T_K^p(m,r)$ holds. In particular, for $K \geq 1$, the above inequality implies $|p(z)| > 0$ and, thus, $p$ has no root in $\overline{\Delta}$.

It remains to show (2): If $T_K^p(m,r)$ holds, then, for any point $z \in \overline{\Delta}$, the derivative $p'(z)$ differs from $p'(m)$ by a complex number of absolute value less than $|p'(m)|/K$. Consider the triangle spanned by the points 0, $p'(m)$ and $p'(z)$, and let $\alpha$ and $\beta$ denote the angles at the points 0 and $p'(z)$, respectively. From the Sine Theorem, it follows that

$$|\sin \alpha| = |p'(m) - p'(z)| \cdot \frac{|\sin \gamma|}{|p'(m)|} < \frac{1}{K}.$$ 

Thus, the arguments of $p'(m)$ and $p'(z)$ differ by less than $\arcsin(1/K)$ which is smaller than or equal to $\pi/4$ for $K \geq \sqrt{2}$. Assume that there exist two roots $a, b \in \overline{\Delta}$ of $p$. Since $a = b$ implies $p'(a) = 0$, which is not possible as $T_1^p(m,r)$ holds, we can assume that $a \neq b$. We split $f$ into its real and imaginary part, that is, we consider $p(x + iy) = u(x,y) + iv(x,y)$ where $u, v : \mathbb{R}^2 \to \mathbb{R}$ are two bivariate polynomials. Then, $p(a) = p(b) = 0$ and so $u(a) = v(a) = u(b) = v(b) = 0$. But $u(a) = u(b) = 0$ implies, due to the Mean Value Theorem in several real variables [], that there exists a $\phi \in [a, b]$ such that

$$\nabla u(\phi) \perp (b - a).$$

Similarly, $v(a) = v(b) = 0$ implies that there exists a $\xi \in [a, b]$ such that $\nabla v(\xi) \perp (b - a)$. But $\nabla v(\xi) = (v_x(\xi), v_y(\xi)) = (-u_y(\xi), u_x(\xi))$, thus, it follows that $\nabla u(\xi) \parallel (b - a)$. Therefore, $\nabla u(\psi)$ and $\nabla u(\xi)$ must be perpendicular. Since $p' = u_x + iv_x = u_x - iv_y$, the arguments of $p'(\psi)$ and $p'(\xi)$ must differ by $\pi/2$. This contradicts our above result that both differ from the argument of $p'(m)$ by less than $\pi/4$, thus, (2) follows. \qed
## B  Further Experiments

Table 4: Description of the curves used in experiments. In case only a single curve is given, the second curve is taken to be the first derivative w.r.t. $y$-variable.

| Instance       | Description                                           | Instance           | Description                                           |
|----------------|-------------------------------------------------------|--------------------|-------------------------------------------------------|
| hard_one       | vertical lines as component of one curve, many candidates to test | compact_surf       | silhouette of an algebraic surface, many singularities, isolated solutions |
| grid_deg_10    | large coefficients, curve in generic position         | 13_sings_9         | large coefficients, high-curvature points              |
| huge_cusp      | large coefficients, high-curvature points             | swimmerston_dyer   | coterminal solutions in both directions                 |
| cusps_and_flexes| high-curvature points                                 | challenge_12_1     | many candidate solutions to be checked                 |
| L6_circles     | 4 circles w.r.t. $L_6$-norm, clustered solutions      | SA_2_4_eps         | singular points with high tangencies, displaced         |
| ten_circles    | set of 10 random circles multiplied together, rational solutions | spiral29_24        | taylor expansion of a spiral intersecting a curve with many branches, many candidates to check |
| curve24        | curvature of degree 8 curve, many singularities       |                    |                                                        |

* These curves were taken from [26]

Table 5: Results for the curves listed in Table 4. We used the same configurations as in Table 2.

| Instance     | $y$-degree | #sols | BS+allfilters | BS+bstr+comb | BS+bstr | BS | ISOLATE | LGP |
|--------------|------------|-------|---------------|--------------|---------|----|---------|-----|
|              |            |       | CPU           | GPU          |         |    | Maple   |     |
| hard_one     | 27, 6      | 46    | 8.17          | 6.95         | 6.96    | 12.44 | 12.09   | 25.20|
| grid_deg_10  | 10, 20     |       | 4.05          | 1.63         | 1.64    | 3.22  | 3.01    | 106.95|
| huge_cusp    | 8, 24      |       | 33.24         | 21.43        | 21.15   | 26.97 | 26.47   | 768.56|
| cusps_and_flexes | 9, 20 |       | 2.31          | 1.37         | 1.28    | 1.70  | 1.38    | 28.42 |
| L6_circles   | 24, 18     |       | 25.00         | 6.21         | 5.68    | 6.88  | 5.08    | 46.61 |
| curve24      | 24, 28     |       | 41.26         | 16.61        | 16.66   | 118.91| 115.62  | 49.69 |
| ten_circles  | 20, 45     |       | 10.51         | 7.64         | 4.63    | 4.93  | 2.57    | 5.22  |
| compact_surf | 18, 57     |       | 19.01         | 6.53         | 5.98    | 5.85  | 23.56   | timeout|
| 13_sings_9   | 9, 35      |       | 3.38          | 2.39         | 2.23    | 2.98  | 2.41    | 28.60 |
| swimmerston_dyer | 40, 63 |       | 50.32         | 22.60        | 22.53   | 22.33 | 56.46   | 71.00 |
| challenge_12_1| 30, 99    |       | 24.67         | 9.17         | 9.89    | 9.44  | 41.84   | 41.13 |
| SA_2_4_eps   | 17, 6      |       | 6.71          | 0.56         | 0.59    | 0.70  | 1.66    | 7.83  |
| spiral29_24  | 29, 24     |       | 80.37         | 35.34        | 35.13   | 290.35| 286.79  | 144.79|

*timeout: algorithm timed out (> 1500 sec)*