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INSTITUT NATIONAL DE RECHERCHE EN INFORMATIQUE ET EN AUTOMATIQUE

*Univariate Algebraic Kernel and Application to  
Arrangements*

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## Univariate Algebraic Kernel and Application to Arrangements

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**Abstract:** Solving polynomials and performing operations with real algebraic numbers are critical issues in geometric computing, in particular when dealing with curved objects. Moreover, the real roots need to be computed in a certified way in order to avoid possible inconsistency in geometric algorithms. Developing efficient solutions for this problem is thus an important issue for the development of libraries of computational geometry algorithms and, in particular, for the state-of-the-art (open source) *CGAL* library. We present a *CGAL*-based univariate algebraic kernel, which provides certified real-root isolation of univariate polynomials with integer coefficients and standard functionalities such as basic arithmetic operations, greatest common divisor (gcd) and square-free factorization, as well as comparison and sign evaluations of real algebraic numbers.

We compare our kernel with other comparable kernels, demonstrating the efficiency of our approach. Our experiments are performed on large data sets including polynomials of high degree (up to 2 000) and with very large coefficients (up to 25 000 bits per coefficient).

We also address the problem of computing arrangements of  $x$ -monotone polynomial curves. We apply our kernel to this problem and demonstrate its efficiency compared to previous solutions available in *CGAL*. We also present the first bit-complexity analysis of the standard sweep-line algorithm for this problem.

**Key-words:** *CGAL*, algebraic kernel, RS, root isolation, algebraic number comparison

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Most part of this work was done while the author was at LORIA - INRIA Nancy-Grand Est.

## Noyau univarié et application aux arrangements

**Résumé :** Résoudre des systèmes polynomiaux et effectuer des opérations avec des nombres algébriques réels ont une importance critique en géométrie algorithmique, notamment dans la gestion d'objets courbes. De plus, les racines réelles nécessitent d'être isolées d'une manière certifiée pour éviter toute incohérence dans les algorithmes géométriques. La conception de solutions efficaces pour ce problème est donc une tâche importante pour le développement de bibliothèques d'algorithmes de géométrie algorithmique et, en particulier, pour la bibliothèque (open source) de référence CGAL. Nous présentons ici un noyau algébrique univarié répondant aux spécifications CGAL et permettant d'isoler et de comparer, de façon certifiée, les racines réelles de polynômes univariés à coefficients entiers. Ce noyau permet également le calcul de pgcd et l'élimination de facteurs carrés (square-free factorization), et il possède les fonctionnalités standards, comme les opérations arithmétiques.

Nous comparons notre noyau avec les autres noyaux similaires, démontrant ainsi l'efficacité de notre approche. Nous avons testé les noyaux sur de gros ensembles de données comprenant des polynômes de haut degré (jusqu'à 2 000) et avec de très grands coefficients (jusqu'à 25 000 bits par coefficient).

Nous considérons également le problème du calcul d'arrangements de courbes polynomiales monotones en  $x$ . Nous testons notre noyau sur ce problème et nous prouvons son efficacité par rapport aux solutions déjà existantes dans CGAL. Nous présentons enfin la première analyse de bit complexité de l'algorithme standard de balayage.

**Mots-clés :** CGAL, noyau algébrique, RS, isolation des racines, comparaison des nombres algébriques

## 1 Introduction

Implementing geometric algorithms robustly is known to be a difficult task for two main reasons. First, all degenerate situations have to be handled and second, algorithms often assume a real-RAM model (a random-access machine where each register can hold a real number and each arithmetic operation has unit cost) which is not realistic in practice. In recent years, the paradigm of exact geometric computing has arisen as a standard for robust implementations [27]. In this paradigm, geometric queries, also called predicates, such as “is a point inside, outside or on a circle?”, are made exactly using, usually, either (i) exact arithmetic combined, for efficiency, with interval arithmetic on doubles or (ii) interval arithmetic on arbitrary-fixed-precision floating-point numbers combined with separation bounds; on the other hand, geometric constructions, such as the circle through three points or points of intersection between two curves, may be approximated.

We address here one recurrent difficulty arising when implementing algorithms dealing, in particular, with curved objects. Such algorithms usually require evaluating, manipulating and solving systems of polynomial equations and comparing their roots. One of the most critical parts of dealing with polynomials or polynomial systems is the isolation of the real roots and their comparison.

We restrict here our attention to the case of univariate polynomials and address this problem in the context of CGAL, a C++ Computational Geometry Algorithms Library, which is an open source project and became a standard for the implementation of geometric algorithms [4].

CGAL is designed in a modular fashion following the *paradigm of generic programming*. Algorithms are typically parameterized by a *traits* class which encapsulates the geometric objects, predicates and constructions used by the algorithm. Algorithms can thus typically be implemented independently of the type of input objects. For instance, the core of a line-sweep algorithm for computing arrangements of plane curves [7] can be implemented independently of whether the curves are lines, line segments, or general curves; on the other hand, the elementary operations that depend on the type of the objects (such as, comparing  $x$ -coordinates of points of intersection) are implemented separately in traits classes. Similarly, the model of computation, such as exact arbitrary-length integer arithmetic or approximate fixed-precision floating-point arithmetic, are encapsulated in the concept of *kernel*. An implementation is thus typically separated in three or four layers, (i) the geometric algorithm which relies on (ii) a traits class, which itself relies on (iii) a kernel for elementary (typically geometric) operations. CGAL provides several predefined Cartesian kernels, for instance allowing standard Cartesian geometric operations on inputs defined with doubles and providing approximate constructions (*i.e.*, defined with double) but exact predicates. However, a kernel can also rely on (iv) a number type which essentially encapsulates the type of number (such as, double, arbitrary-length integers, intervals) and the associated arithmetic operations. A choice of traits classes, kernels and number types is useful as it gives freedom to the users and it makes it easier to compare and improve the various building blocks of an implementation.

**Our Contributions.** We present in this paper a CGAL-compliant algebraic kernel that provides real-root isolation of univariate integer polynomials and basic operations, *i.e.* comparisons and sign evaluations, of real algebraic numbers. This open-source kernel follows the CGAL specifications for algebraic kernels [3]. The root isolation is based on the interval Descartes algorithm [5] and uses the library RS [22]. Moreover, our kernel provides various operations for polynomials, such as gcd, which are crucial for manipulating algebraic numbers.

We compare our kernel with other comparable kernels and demonstrate the efficiency of our approach. We perform experiments on large data sets including polynomials of high degree (up to 2 000) and with very large coefficients (up to 25 000 bits per coefficient).

Finally, we apply our kernel to the problem of computing arrangements of  $x$ -monotone polynomial curves and demonstrate its efficiency compared to previous solutions available in CGAL. We also present an output-sensitive bit-complexity analysis of the standard sweep-line algorithm for this problem. We establish a bound of  $\tilde{O}_B((n+k)d^3(\tau^2+s^2))$ , where  $n$  is the number of curves,  $k$  is the number of intersections,  $d$  bounds the degree of the polynomials,  $\tau$  bounds the bitsize of their coefficients, and  $s$  is the logarithm of the minimum distance between the (complex) roots of the difference of any two polynomials (that is roughly speaking the bitsize of this distance); the  $\tilde{O}_B(\cdot)$  notation ignores the logarithmic factors. If  $N = \max\{n, k, d, \tau, s\}$ , this bound is in  $\tilde{O}_B(N^6)$  which is, as expected, quadratic in the size of the input in the worst case; indeed, the input consists of  $n$  polynomials of degree  $d$  and with coefficients of bitsize  $\tau$  and is thus in  $\Theta(N^3)$  in the worst case. To the best of our knowledge, this is the first bit-complexity analysis of this algorithm.

**Related work.** Combining algebra and geometry for manipulating non-linear objects has been a long-standing challenge. Previous work includes, but it is not limited to, MAPC [17] a library for manipulating points that are defined algebraically and handling curves in the plane. More recently, the library EXACUS [2], which handles curves and surfaces in computational geometry and supports various algebraic operations, was developed and partially integrated into CGAL. The notion of algebraic kernel for CGAL was proposed in 2004 [12]; in this work, the underlying algebraic operations were based on the SYNAPS library [18]. Several methods and algebraic kernels have been developed since then.

One kernel was developed by Hemmer and Limbach [16] following the generic programming paradigm using the C++ template mechanism. This kernel is templated by the representation of algebraic numbers and by the real root isolation method, for which two classes have been developed; one is based on the Descartes method and the other on the Bitstream Descartes method [9]. This approach has the advantage to allow, in principle, using the best instances for both template arguments.

Another kernel developed at INRIA relies on the SYNAPS library [18]. In this kernel there are several approaches concerning real root isolation, *i.e.*, methods based on Sturm subdivisions, sleeves approximations, continued fractions, and a symbolic-numeric combination of the sleeve and continued fractions methods (see [11]). Moreover, there are specialized methods for polynomials of degree less or equal than four [24].

Emiris et al. [11] presented some benchmarks of these various approaches in these two kernels as well as some tests on the kernel we present here. The authors mention that our kernel based on interval Descartes performs similarly to one approach (refer to as NCF2) based on continued fractions [23] for coefficients with (very) large bitsize but NCF2 is more efficient for small bitsize. They conclude that, first, dedicated algorithms for polynomials of degree less than (or equal to) four is always the most efficient approach and, second, that NCF2 always perform the best except for low-degree and high-bitsize polynomials, in which case the kernel based on the Bitstream Descartes method performs the best. We moderate here these conclusions.

The rest of the paper is structured as follows. In the next section we describe our univariate algebraic kernel. In Section 3 we present various experiments concerning real root isolation and comparison of real algebraic numbers. Finally, in Section 4, we sketch our traits class for arrangements, we present experiments against the traits class that is currently available in CGAL, and we present the bit complexity analysis of the algorithm for computing the arrangement of curves defined by univariate polynomials.

## 2 Univariate algebraic kernel

We describe here our implementation of our univariate algebraic kernel. The two main requirements of the CGAL specifications, which we describe here, are the isolation of real roots and their comparison. We also describe our implementation of two operations, the gcd computation and the refinement of isolating intervals, that are both needed for comparing algebraic numbers.

**Preliminaries.** The kernel handles univariate polynomials and algebraic numbers. The polynomials have integer coefficients and are represented by arrays of GMP arbitrary-length integers [15]. We implemented in the kernel the basic functions for polynomials. An algebraic number that is a root of a polynomial  $F$  is represented by  $F$  and an isolating interval, that is an interval containing this root but no other root of  $F$ . We implemented intervals using the MPFI library [19], which represents intervals with two MPFR arbitrary-fixed-precision floating-point numbers [20]; note that MPFR is developed on top of the GMP library for multi-precision arithmetic [15].

**Root isolation.** For isolating the real roots of univariate polynomials with integer coefficients, we developed an interface with the library RS [22]. This library is written in C and is based on Descartes' rule for isolating the real roots of univariate polynomials with integer coefficients.

We briefly detail here the general design of the RS library; see [21] for details. RS is based on an algorithm known as *interval Descartes* [5]; namely, the coefficients of the polynomials obtained by changes of variable, sending intervals  $[a, b]$  onto  $[0, +\infty]$ , are only approximated using interval arithmetic when this is sufficient for determining their signs. Note that the order in which these transformations are performed in RS is important for memory consumption. The intervals and operations on them are handled by the MPFI library.



**Algebraic number comparison.** As mentioned above, one of the main requirements of the CGAL algebraic kernel specifications is to compare two algebraic numbers  $r_1$  and  $r_2$ . If we are lucky, their isolating intervals do not overlap and the comparison is straightforward. This is, of course, not always the case. If we knew that they were not equal, we could refine both isolating intervals until they are disjoint; see below for details on how we perform the refinements. Hence, the problem reduces to determining whether the algebraic numbers are equal or not.

To do so, we compute the square-free factorization of the gcd of the polynomials associated to the algebraic numbers (see below for details). The roots of this gcd are the common roots of both polynomials. We calculate the intersection,  $I$ , of the isolating intervals of  $r_1$  and  $r_2$ . The gcd has a root in this interval if and only if  $r_1 = r_2$ .

To determine whether the gcd has a root in interval  $I$ , it suffices to check the sign of the gcd at the endpoints of  $I$ : if they are different or one of them is zero, the gcd has a root in  $I$  and  $r_1 = r_2$ ; otherwise,  $r_1 \neq r_2$  and we can refine both intervals until they are disjoint.

**Gcd computations.** Computing greatest common divisors between two polynomials is not a difficult task, however, it is not trivial to do so efficiently. A naive implementation of the Euclidean algorithm works fine for small polynomials but the intermediate coefficients suffer an exponential growth in size, which is not manageable for medium to large size polynomials. We thus implemented a *modular* gcd function. We did not use some existing implementations mainly for efficiency because converting polynomials from one representation to another is substantially costly as soon as the degree and bitsize are large. Our function calculates the gcd of polynomials modulo some prime numbers and reconstructs later the result with the help of the *Chinese remainder theorem*. Details on these algorithms can be found in, e.g. [26]. Note that modular gcd is always more efficient than regular gcd and it is much more efficient when the two polynomials have no common roots.

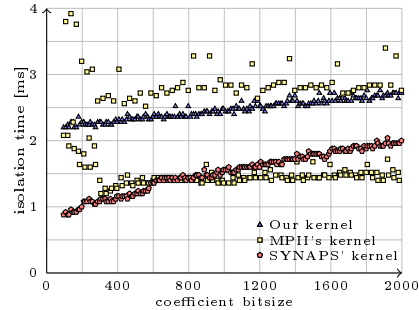
**Refining isolating intervals.** As we mentioned before, refining the interval representing an algebraic number is critical for comparing such numbers. We provide two approaches for refinement.

Both approaches require that the polynomial associated to the algebraic number is square free. The first step thus consists of computing the square-free part of the polynomial (by computing the gcd of the polynomial and its derivative).

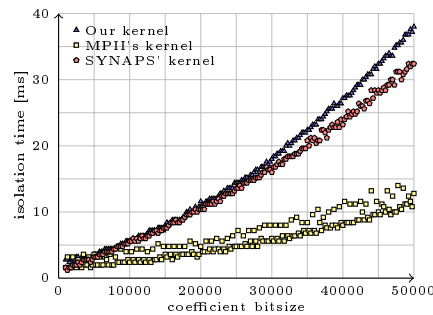
Our first approach is a simple bisection algorithm. It consists in calculating the sign of the polynomial associated to the algebraic number at the endpoints and midpoint of the interval. Depending on these signs, we refine the isolating interval to its left or right half.

Our second approach is a quadratic interval refinement [1]. Roughly speaking, this method splits the interval in many parts and, based on a linear interpolation, guesses in which one the root lies. If the guess is correct, the algorithm divides in the next refinement step the interval in more parts and, if not, in less.

Unfortunately, even with our careful implementation this approach turns out to be, on average, only just a bit faster than the bisection approach. Our



(a)



(b)

Figure 1: Running time for isolating all the real roots of degree 12 polynomials with 12 real roots in terms of the maximum bitsize of their coefficients.

experiments showed that the bottleneck of the refinement is the evaluation of polynomials.

### 3 Kernel benchmarks

In this section, we analyze the running time of the two main functions of our algebraic kernel, that (i) isolate the roots of a polynomial and (ii) compare two algebraic numbers that is, compare the roots of two polynomials. We also compare the performance of our kernel with the one based on the Bistream Descartes method [9] and developed by Hemmer and Limbach [16] (referred to as MPII's kernel)<sup>1</sup> and with a kernel based on continued fractions [23] and developed on top of the SYNAPS library [18] (referred to as SYNAPS' kernel).

All tests were ran on a single-core 3.2 GHz Intel Pentium 4 with 2 Gb of RAM and 2048 kb of cache memory, using 64-bit Linux.

**Root isolation.** We consider two suites of experiments in which we either fix the degree of the polynomials and vary the bitsize of the coefficients or the

<sup>1</sup> To compare both algebraic kernels with the same inputs, we parameterized MPII's kernel to use Bistream Descartes as root isolator, `algebraic_real_bfi_rep` as algebraic number representation and `CORE` integers and rationals to represent the coefficients of the polynomials and the isolation bounds of algebraic numbers, respectively. The choice of `CORE` (vs. `LEDA`) was induced by the need of testing the kernels in the same conditions, that is, relying on `GMP`.

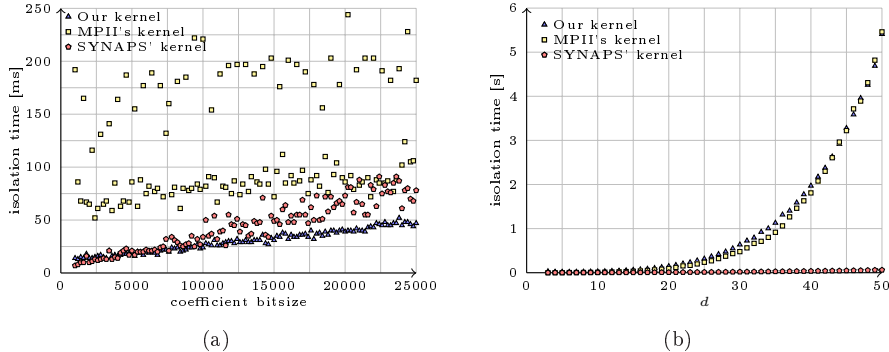


Figure 2: Running time for isolating all the real roots of (a) degree 100 polynomials in terms of the maximum bitsize of their coefficients and (b) Mignotte polynomials of the form  $f = x^d - 2(kx - 1)^2$  in terms of the degree  $d$ .

converse; see Figs. 1 and 2. In each experiment, we report the running time for isolating all the roots per polynomial, averaged over different trials, for our kernel, MPII's and SYNAPS' kernel.

**Varying bitsize.** We study here polynomials with rather low degree (12) but with no complex root and polynomials with reasonably large degree (100) with random coefficients (and thus with few real roots).

The first test sets comes from [16]. See Fig. 1. It consists of polynomials of degree 12, each one being the product of six degree-two polynomials with two roots, at least one of them in the interval  $[0, 1]$ ; every polynomial thus has 12 real roots. We vary the maximum bitsize of all the coefficients of the input polynomial from 100 to 50 000 and average each test over 250 trials.

Secondly, we consider random polynomials with constant degree 100 and coefficients with varying bitsize. See Fig. 2(a). Note that such random polynomials have few roots: the expected number of real roots of a polynomial of degree  $d$  with coefficients independently chosen from the standard normal distribution is  $\frac{2}{\pi} \ln(d) + C + \frac{2}{\pi d} + O(1/d^2)$  where  $C \approx 0.625735$  [8]; this gives, for degree 100 an average of about 3.6 roots (note that this bound matches extremely well experimental observations). We vary the maximum bitsize of all the coefficients from 2 000 to 25 000 and average each test over 100 trials.

**Varying degree.** We consider two sets of experiments in which we study random polynomials and Mignotte polynomials (which have two very close roots).

We first consider polynomials with random coefficients of fixed bitsize for various values between 32 and 1 000. We then vary the degree of the polynomials from 100 to 2 000 and average our experiments over 100 trials (see Fig. 3). Note that the above formula gives an expected number of roots varying from 3.6 to 5.5. We observe that the running time is almost independent of the bitsize in the considered range.

Finally, we test Mignotte polynomials, that is polynomials of the form  $x^d - 2(kx - 1)^2$ . Such polynomials are known to be challenging for Descartes algorithms because two of their roots are very close to each other; the isolating

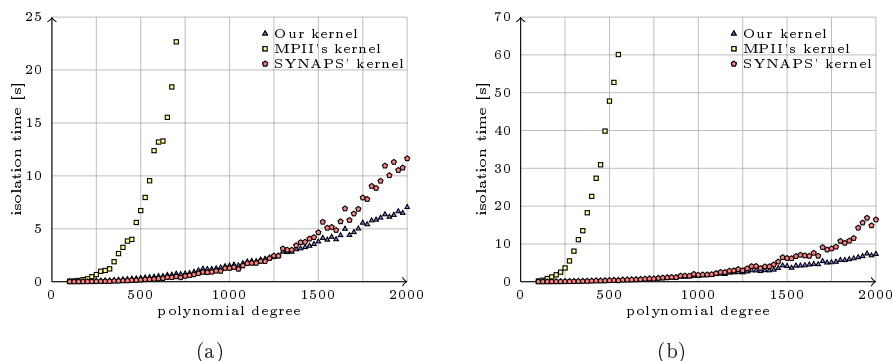


Figure 3: Running time for isolating all the real roots of random polynomials with coefficients of bitsize (a) 32 and (b) 1000, and depending on the degree.

intervals for these two roots are thus very small. For these tests, we used Mignotte polynomials with coefficients of bitsize 50, with varying degree  $d$  from 5 to 50. See Fig. 2(b). We averaged the running time over 5 trials for each degree. We observed essentially no difference between our kernel and MPII's one; they take roughly 0.2 and 5.5 seconds for Mignotte polynomials of degree 20 and 50, respectively. However, SYNAPS' kernel is much more efficient as the continued fractions algorithm is not so affected by the closeness of the roots.

**Discussion.** We observe (Fig. 1(a)) that SYNAPS' kernel is more efficient than both our and MPII's kernel in the case of polynomials of small degree (*e.g.*, twelve) and small to moderately large coefficients (up to 2 000 bits per coefficient). However, for extremely large coefficients MPII's kernel is substantially more efficient (by a factor of up to 3 for coefficients of up to 50 000 bits) than both our and SYNAPS' kernels, which perform similarly.

For polynomials of reasonable large degree, both our and SYNAPS' kernels are much more efficient than MPII's kernel; furthermore these two kernels behave similarly for degrees up to 1 500 and our kernel becomes more efficient for higher degrees (by a factor 2 for degree 2 000).

We also observe that the running time is *highly* dependent of the various settings. For instance, our kernel is up to 5 times slower when using approximate evaluation for high-degree and high-bitsize polynomials. Also, MPII's kernel is in some cases about 10 times slower when changing the arithmetic kernel to LEDA, the representation of algebraic numbers and some internal algorithms such as the refinement function. This explains why our benchmarks on both MPII's and SYNAPS' kernels are substantially better than in Emiris et al. experiments [11].

We also observe that the running time of MPII's kernel is unstable in our experiments (Figs. 1 and 2(a)); surprisingly, this instability occurs when the experiments are performed on a 64-bits architecture, but it is stable on 32-bits architecture as shown in previous experiments [11].

**Comparison of algebraic numbers.** We consider three suites of experiments for comparing algebraic numbers; see Fig. 4. Recall that an algebraic number  $\rho$  is here represented by a polynomial  $F$  that vanishes at  $\rho$  and an isolating interval containing  $\rho$  but no other root of  $F$ . Recall also that the comparison of

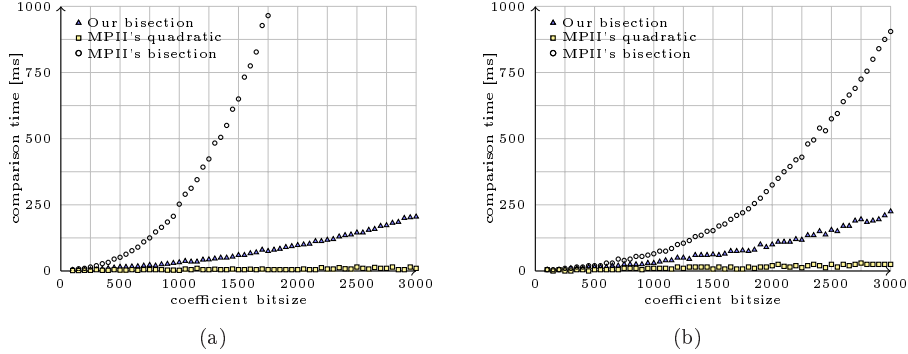


Figure 4: Running time for comparing two distinct close roots of two almost identical polynomials of degree 20 with (a) no common roots and (b) a common factor of degree 10.

two algebraic numbers is done by (i) testing whether the intervals are disjoint; if so, report the ordering, otherwise (ii) compute the gcd of the two polynomials and test whether the gcd vanishes in the intersection of the two intervals; if so, report the equality of the numbers, otherwise (iii) refine the intervals until they are disjoint.

First, we analyze the cost of trivial comparisons that is, when the two intervals representing the numbers are disjoint. For that we compare the roots of two random polynomials. We observe that, as expected, the comparison time is negligible and independent of both the degree of the polynomials and the bitsize of their coefficients.

Second, we analyze the cost of comparing roots that are very close to each other but whose associate polynomials have no common root. This case is expensive because we need to refine the intervals until they do not overlap; this is, however, not the worst situation because the gcd of the two polynomials is 1 which is tested efficiently with a modular gcd. We perform these experiments as follows. We generate pairs of polynomials, one with random coefficients and the other by only adding 1 to one of the coefficients of the first polynomial. Such polynomials are such that the  $i$ -th roots of both polynomials are very close to each other. We generate such pairs of polynomials with constant degree (equal to 20) and vary the maximum bitsize of the coefficients. As the bitsize increases, the pairs of roots that are close become even closer and thus the comparison time increases. The results of these experiments are presented in Fig. 4(a), which reports the average running time for comparing two close roots. We show in this figure three curves, one corresponding to our bisection algorithm, and two corresponding the two refinement methods implemented in the MPII's kernel: the usual bisection and a quadratic refinement algorithm.

Third, we consider the, a priori, most expensive scenario in which we compare roots that are either equal or very close to each others and such that their associate polynomials have some roots in common. In this case, we accumulate the cost of computing a non-trivial gcd of the two polynomials with the cost of refining intervals when comparing two non-equal roots. In practice, we generate pairs of degree-20 polynomials each defined as the product of two degree-10 terms; one of these factors is random and common to the two polynomials; the

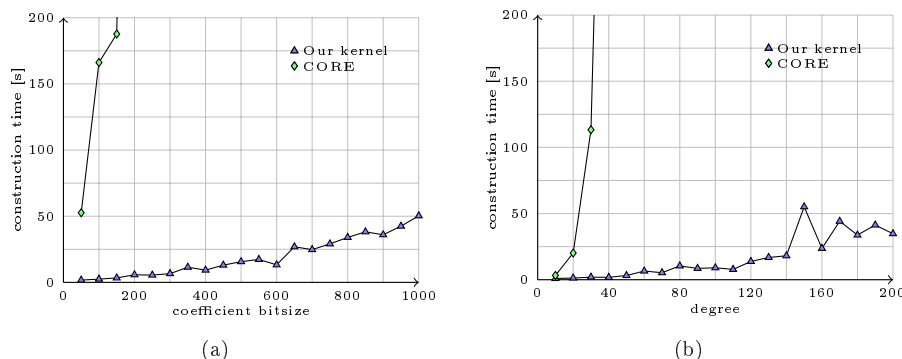


Figure 5: Arrangements of five polynomials, shifted four times each, (a) of degree 20 and varying bitsize and (b) of bitsize 32 and varying degree.

other factor is random in one of the polynomials and slightly modified in the other polynomial where, slightly modified means, as above, that we add 1 to one of the coefficients. We then vary the maximum bitsize of the coefficients.

**Discussion.** We see in Fig. 4 that the MPII’s quadratic refinement algorithm largely outperforms the two bisection methods. However, our bisection method is faster than MPII’s one, by a factor up to 10. We also observed that the running time for comparing equal roots is negligible compared to the cost of comparing close but distinct roots. (The running time reported in Fig. 4(b) is actually the total time for comparing all pairs of roots divided by the number of comparisons of close but distinct roots.) This explains why our kernel behaves similarly in Figs. 4(a) and 4(b). Overall, it appears that comparing algebraic numbers that are very close is fairly time consuming and that the most time-consuming part of the comparison is the evaluation of polynomials performed during the interval refinements.

## 4 Arrangements

As an example of possible benefit of having efficient algebraic kernels in CGAL, we used our implementation to construct arrangements of polynomial functions. Wein and Fogel [25] provided a CGAL package for calculating arrangements of general curves which requires as parameter a *traits class* containing the data structures to store the curves and various primitive operations, such as comparing the relative positions of points of intersection. We implemented a traits class which uses the functions of our algebraic kernel and compared its performance with another traits classes which comes with CGAL’s arrangement package and uses the CORE library [6].

In order to generate challenging data sets we proceed as follows. First we generate  $n$  random polynomials. To each of them we add 1 to the constant coefficient,  $m$  times, thus producing a data set of  $n(m+1)$  univariate polynomials. Notice that the arrangement of the graphs of these polynomials is guaranteed to be degenerate, *i.e.*, there are intersections with the same  $x$ -coordinate. The arrangements generated this way have four parameters: the number  $n$  of initial

polynomials, the number  $m$  of “shifts” that we perform, the degree  $d$  of the polynomials, and the bitsize  $\tau$  of their coefficients. We ran experiments varying the values of the last three of these parameters and setting  $n = 5$ .

Fig. 5(a) shows the running time in terms of the bitsize  $\tau$  for a data set where  $d = 20$  and  $m = 4$  (giving 25 polynomials). Fig. 5(b) shows the running time in terms of the degree  $d$  for a second data set where  $\tau = 32$  and  $m = 4$ . We see from these experiments that running time using CORE is considerably higher than when using our kernel. We also make the following observations.

Fig. 5(a) shows that the running time depends on the bitsize. When we change the bitsize of the coefficients of the random polynomials, the size of the arrangement does not change; that means that the number of comparisons and root isolations the kernel must perform is roughly the same in all the arrangements of the test suite. The isolation time for random polynomials does not depend much on the bitsize (as shown in Fig. 2(a)), but the comparison time does. It follows that the running time increases with the bitsize.

Fig. 5(b) shows that the running time depends also on the degree of the input polynomials. As we saw in Section 3, the expected number of real roots of a random polynomial depends on its degree. The size of the arrangement thus increases with the degree of the input polynomials: each vertex is the root of the difference between two input polynomials, therefore there will be more vertices. Thus, when we increment the degree of the inputs, the number of comparisons and isolations increases; furthermore, the running time for each of these operations increases with the degree of the input.

We ran additional tests to see the impact of the input shifts in the calculation time. We generated five random polynomials of bitsize 1000 and degree 20. We calculated arrangements, then, varying the number of shifts we perform to each polynomial. As Fig. 6 shows, we were only able to solve, using CORE, the first arrangement, generated without shifts (note the point on the vertical axis). We note that the running time increases fast with the number of shifts. This is reasonable since, each time we increase by 1 the number of shifts, we add to the arrangement  $n$  polynomials, hence increasing the number of vertices of the arrangement. Since the root isolation and comparison time remains the same (because the degree and the bitsize are constant), the running time increases with the number of these operations.

## 4.1 Complexity analysis

In this section we present an output-sensitive analysis of the bit complexity of the standard line-sweep algorithm for computing arrangements of graphs of univariate polynomial functions. The same analysis, and thus the same complexity bound, applies also in the case of rational univariate functions.

In what follows, combinatorial and bit complexities will be denoted by  $\mathcal{O}$  and  $\mathcal{O}_B$ , respectively. The  $\tilde{\mathcal{O}}$  and  $\tilde{\mathcal{O}}_B$  notations refer to complexities in which we ignore (poly-)logarithmic factors. We also refer to the *separation bound* of a univariate polynomial as to the minimum distance between any two (possibly complex) roots of the polynomial. The bitsize of the separation bound is the number of bits,  $s$ , needed to represent the largest lower bound of the form  $2^{-s}$  that is smaller than the separation bound.

First, we note that the results of [10, 13] can easily be generalized to express the complexities of isolating the real roots of a polynomial in terms of the

separation bounds of the considered instances of polynomials rather than in terms of worst-case separation bounds.

**Proposition 1.** *The real roots of a univariate polynomial of degree  $d$  with integer coefficients of bitsize at most  $\tau$  and separation bound of bitsize  $s$  can be isolated, with their multiplicities, in  $\tilde{O}_B(d^3(\tau^2 + s^2))$  time. The bitsize of the endpoints of the isolating intervals is in  $\mathcal{O}(s)$ .*

*Proof.* In [10] it was proven that the worst case complexity is  $\tilde{O}_B(d^4 + d^4\tau^2)$ . The result is based on the fact that the bitsize of the worst case separation bound is  $\mathcal{O}(d\tau)$ . To derive an output sensitive result we replace this by  $s$ .

Then, to isolate a root we need to perform at most  $\tilde{O}(\tau + s)$  step. At each step we perform a polynomial shift, in the case of Descartes' solver, or an evaluation of a Sturm sequence, with a number of bitsize  $\tau + s$ , resulting a cost, in both cases, of  $\tilde{O}_B(d^2\tau^2 + d^2\tau s + d^2s^2)$  for each root. The result follows if we multiply by the number of roots,  $d$ .  $\square$

We also recall an output-sensitive result on the complexity of comparing the roots of two polynomials.

**Proposition 2** ([13]). *Two real algebraic numbers defined as roots of polynomials of degree at most  $d$  with integer coefficients of bitsize at most  $\tau$  and separation bounds of bitsize at most  $s$  can be compared in  $\tilde{O}_B(d^2(\tau + s))$  time.*

These complexity results yield, almost directly, the following output-sensitive bit complexity of the standard line-sweep algorithm for computing arrangements of graphs of univariate polynomial functions

**Theorem 3.** *The arrangement of  $n$  curves, defined by univariate polynomials of degree at most  $d$ , with integer coefficients of bitsize at most  $\tau$ , and separation bound of bitsize at most  $s$  can be computed in time  $\tilde{O}_B((n + k)d^3(\tau^2 + s^2))$ , where  $k$  is the number of intersection points between the curves.*

*Proof.* Recall first that the combinatorial complexity of the standard line-sweep algorithm for computing arrangements of  $n$  algebraic curves of bounded degree is  $\mathcal{O}((n + k) \log n)$ , where  $k$  is the number of intersection points (see, e.g., [7, 14]). To evaluate the bit complexity of the algorithm, we split the analysis in two parts. In the first part, we consider the complexity of the construction of the intersection points of the curves. In the second part, we consider the cost of comparing the  $x$ -coordinates of the intersection points.

In order to compute (that is, to isolate) the intersection points of two curves  $y = f_1(x)$  and  $y = f_2(x)$ , represented by polynomials  $f_1, f_2 \in \mathbb{Z}[x]$  of degree at most  $d$  with integer coefficients of bitsize at most  $\tau$  and separation bound of bitsize at most  $s$ , we can first isolate the real roots of the polynomial  $f(x) = f_1(x) - f_2(x)$  in time  $\tilde{O}_B(d^3(\tau^2 + s^2))$  (by Proposition 1). We can then compute the image by  $f$  of these intervals of in time  $\tilde{O}_B(d(\tau + ds))$  (by Horner's rule).

To begin the algorithm we need to compute a vertical line that is to the left all the intersection points between the curves. The cost of computing such a line is  $\tilde{O}_B(nd\tau)$  and is dominated by the other steps of the algorithm. We then compute the intersection points of this line with all the curves, so that to order the curves along the sweep line. We then compute the intersection between the



$n - 1$  pairs of adjacent curves along the sweep line. Thus, we initially perform  $\mathcal{O}(n)$  intersections between pairs of curves.

Then, during the sweep, every time an intersection point is encountered by the sweep line, we exchange two curves in the list of curves intersected by the sweep line and we compute the intersection between (at most) two new pairs of adjacent curves in this list. Hence, we perform, in total,  $\mathcal{O}(n + k)$  intersections between pairs of curves in  $\tilde{\mathcal{O}}_B((n + k)d^3(\tau^2 + s^2))$  time.

We now consider the cost of comparing the  $x$ -coordinates of the intersection points when updating the event list. Every time two curves become adjacent along the vertical line of sweep, we insert their first intersection point that is to the right of the line. Since we only insert one intersection point (rather than  $d$ ), this requires in total  $\mathcal{O}((n + k) \log n)$  comparisons which can be done in  $\tilde{\mathcal{O}}_B((n + k)d^2(\tau + s))$  time by Proposition 2.  $\square$

Note that, as mentioned in section 1, if  $N = \max\{n, k, d, \tau, s\}$ , this bound is in  $\tilde{\mathcal{O}}_B(N^6)$  which is, as expected, quadratic in the size of the input in the worst case.

## 5 Conclusion

We presented a new CGAL-compliant algebraic kernel that provides certified real-root isolation of univariate polynomials with integer coefficients based on the interval Descartes algorithm. This kernel also provides the comparison of algebraic numbers and other standard functionalities.

We compared our kernel with other comparable kernels on large data sets including, for the first time, polynomials of high degree (up to 2000) and with extremely large coefficients (up to 25 000 bits per coefficient). We demonstrated the efficiency of our approach and showed that it performs similarly, in most cases, with one kernel based on the SYNAPS library; more precisely, our kernel is more efficient for polynomials of very large degree (greater than 1 800) and less efficient for polynomials of very small degree and with small to moderate size coefficients. Also, our kernel is a lot more efficient than the kernel developed at MPII for polynomials of large degree (greater than 200); it is however less efficient for polynomials of small degree and with extremely large coefficients.

Our tests indicate that the kernel developed at MPII appears to be less efficient than the other two for polynomials of large degree. However it should be stressed that this kernel is the only one among the three that is templated by the number type of the coefficients. Of course this does not imply that efficiency is necessarily lost by following the generic programming paradigm, but it does imply that, from the user point of view, some substantial gain of efficiency can sometimes be made by using a kernel that does not follow this paradigm.

We also compared the performance of the kernels on the comparison of algebraic numbers. We observed in these tests that the bisection algorithm runs much faster when it is specialized on a number type since it allows for low level optimizations, confirming thus the assertion in the previous paragraph. On the other hand, it becomes evident that the bisection method is not the most efficient algorithm when a large number of refinements is needed, and MPII's quadratic refinement is the fastest method by far.

A fairly large choice of algebraic kernels and, in particular, of methods for isolating the real roots of polynomials, is now available in CGAL. This allows, in particular, to compare and improve the various methods. It appears that between the two big classes of methods, based on continued fractions and Descartes algorithms, neither is clearly much better than the other. However, some substantial differences appear between the various implementations, but, of course, it is always very difficult to benchmark implementations. For instance, we observed here that the running times are highly dependent of the various settings and architectures.

Finally, we also address the problem of computing arrangements of  $x$ -monotone polynomial curves. We apply our kernel to this problem and demonstrate its efficiency compared to previous solutions available in CGAL. We also present the first bit-complexity analysis of the standard sweep-line algorithm for this problem.

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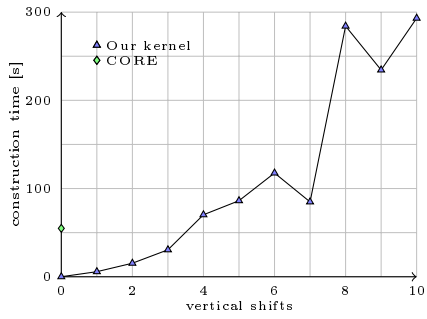
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