

Real Algebraic Numbers: Complexity Analysis and Experimentation

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Abstract. We present algorithmic, complexity and implementation results concerning real root isolation of a polynomial of degree d , with integer coefficients of bit size $\leq \tau$, using Sturm (-Habicht) sequences and the Bernstein subdivision solver. In particular, we unify and simplify the analysis of both methods and we give an asymptotic complexity bound of $\tilde{O}_B(d^4\tau^2)$. This matches the best known bounds for binary subdivision solvers. Moreover, we generalize this to cover the non square-free polynomials and show that within the same complexity we can also compute the multiplicities of the roots. We also consider algorithms for sign evaluation, comparison of real algebraic numbers and simultaneous inequalities (SI) and we improve the known bounds at least by a factor of d . Finally, we present our C++ implementation in SYNAPS and some experimentations on various data sets.

1 Introduction

The representation and manipulation of shapes is important in many applications: CAGD, non linear computational geometry, robotics, molecular biology, ... The usual underlying models for these shapes are e.g. parameterized patches of rational surfaces, BSplines, natural quadrics, implicit algebraic curves or surfaces, ... Geometric processing on these objects, e.g. computing boundary representations, arrangements, Voronoi diagram of curved objects, etc [7,35,16,19], requires the intensive use of polynomial solvers and computations with algebraic numbers. In such applications, a geometric model may involve several thousands of algebraic primitives. Their manipulations involve the computation of intersection points of such primitives, of predicates on these intersection points (such as the comparison of coordinates), of the sign of polynomial expressions at these points (such as the sign of a polynomial which defines the boundary of an object), ... The coordinates of these intersection points, which are the solutions polynomial equations, are algebraic numbers that we need to manipulate efficiently.

The objective of this paper is to give of an overview of effective computations with real algebraic numbers, which unify, simplify and improve previous

approaches. Hereafter, we will tackle both complexity analysis and practical issues. We consider two approaches for real root isolation of univariate integer polynomials, one based on Sturm sequences and one based on Descartes' rule of sign. We will also analyse algorithms for sign evaluation, comparison of real algebraic numbers and the problem of simultaneous inequalities.

Our aim is to provide better insights on these algorithms and better bounds on their complexity. For the analysis we consider the bit complexity model which is more realistic than the arithmetic one in the problems we are interested in. Our algorithms are essentially output sensitive, since they depend not only on the input bit size, but also on the actual separation bound, as we will see.

Notation. In what follows \mathcal{O}_B means bit complexity and the $\tilde{\mathcal{O}}_B$ -notation means that we are ignoring logarithmic factors. For a polynomial $f \in \mathbb{Z}[X]$, $\deg(f)$ denotes its degree. By $\mathcal{L}(f)$ we denote an upper bound on the bit size of the coefficients of f (including a bit for the sign). For $\mathbf{a} \in \mathbb{Q}$, $\mathcal{L}(\mathbf{a})$ is the maximum bit size of the numerator and the denominator. Let $M(\tau)$ denote the bit complexity of multiplying two integers of bit size at most τ and $M(d, \tau)$ denote the bit complexity of multiplying two univariate polynomials of degrees bounded by d and coefficient bit size at most τ . Using FFT, $M(\tau) = \mathcal{O}_B(\tau \log^{c_1} \tau)$ and $M(d, \tau) = \mathcal{O}_B(d\tau \log^{c_2}(d\tau))$ for suitable constants c_1, c_2 .

Prior works. Various algorithms exist for polynomial real root isolation, but most of them focus on square-free polynomials. There is a huge bibliography on the problem and the references cited in this paper are only the tip of the iceberg of the existing bibliography.

Collins and Akritas [9] introduced a subdivision-based real root isolation algorithm that relies on Descartes' rule of sign (we call it *Descartes solver* from now on) and derive a complexity of $\tilde{\mathcal{O}}_B(d^6\tau^2)$. Johnson [24] improved the complexity of the algorithm to $\tilde{\mathcal{O}}_B(d^5\tau^2)$, without using fast Taylor shifts [44], and a gap in his proof was corrected by Krandick [26]. Rouillier and Zimmermann (c.f [40] and references therein) presented a unified approach with optimal memory management for various variants of the Descartes solver.

An algorithm (we call it *Bernstein solver* from now on) that is based on a combination of Descartes' rule and on the properties of Bernstein basis first introduced by Lane and Riesenfeld [28] and a bound on its complexity first obtained by Mourrain et al [36]. The interested reader may also refer to [34] for a variant with optimal memory management and the connection to Descartes solver. In the same context, Eigenwillig et al [14] proposed a randomized algorithm for square-free polynomials with bit stream coefficients. The complexity of all these algorithms is bounded by $\tilde{\mathcal{O}}_B(d^6\tau^2)$. Recently, the complexity bound was improved to $\tilde{\mathcal{O}}_B(d^4\tau^2)$ [15] for the square-free case.

If we restrict ourselves to real root isolation using Sturm (or Sturm-Habicht) sequences (we call it *Sturm solver* from now on) the first complete complexity analysis is probably due to Collins and Loos [10], that state a complexity of $\tilde{\mathcal{O}}_B(d^7\tau^3)$. Du et al [13] giving an amortized-like argument for the number of subdivisions, obtained a complexity of $\tilde{\mathcal{O}}_B(d^4\tau^2)$, for square-free polynomials.

Another family of solvers (that we call numerical), compute an approximation of all the roots (real and complex) of a polynomial up to a desired accuracy (see e.g [42,37]). They are based on the construction of balanced splitting circles in the complex plane and achieve the quasi-optimal complexity bound $\tilde{O}_B(d^3\tau)$, if we want to isolate the roots. However, performance in practice does not always agree with that predicted by asymptotic analysis. Let us also mention the Aberth solver [4,5], which has unknown (bit) complexity, but is efficient in practice.

For sign evaluation and comparison as well as computations with real algebraic numbers the reader may refer to [39]. In [17] for degree ≤ 4 , it is proved that these operations can be performed in $\mathcal{O}(1)$, or $\tilde{O}_B(\tau)$. For the problem of simultaneous inequalities (SI), we are interested in computing the (number of) real roots of a polynomial f , such that n other polynomials achieve specific sign conditions, where the degree of all the polynomials is bounded by d and their bit size by τ . Ben-Or, Kozen and Reif [2] presented the BKR algorithm for SI and Canny [6] improved it in the univariate case (by a factor) achieving $\mathcal{O}(n(m d \log(m) \log^2(d) + m^{2.376}))$ arithmetic complexity, where m is the number of real roots of f . Coste and Roy [11] introduced Thom's encoding for the real roots of a polynomial and SI in this encoding (see also [41]). Their approach is purely symbolic and works over arbitrary real closed fields. They state a complexity of $\tilde{O}_B(N^8)$, using fast multiplication algorithms but not fast computations and evaluation of polynomial sequences, where $N \geq n, d, \tau$. In [1] an algorithm for SI is presented where the real algebraic numbers are in isolating interval representation, with complexity $\tilde{O}_B(nd^6\tau^2)$ or $\tilde{O}_B(N^9)$, that uses repeated refinements of the isolating intervals and does not assume fast multiplication algorithms.

Results. For the problem of real root isolation of a univariate polynomial, using the Sturm solver we present an algorithm with complexity $\tilde{O}_B(d^4\tau^2)$, that improves the result of [13], by extending it to non square-free polynomials. We also simplify significantly the proof (Th. 7) and unify it with the Bernstein approach. We also show that computing the multiplicities of the roots can be achieved within the same complexity bound.

For the Bernstein solver, we simplify the proof from [15,43] for the number of subdivisions by considering the subdivision tree at an earlier level and by using Th. 6 exactly as stated in [24,27]. Thus, we arrive at the same bound for the Bernstein subdivision method (Th. 7) as in [15], but for polynomials which are not necessarily square-free.

The analysis that we present applies to both solvers and simplifies significantly the previous approaches. Moreover our analysis applies also to Descartes solves, since the subdivision tree, i.e the number of steps that the algorithm performs, is the same as in the case of Bernstein solver.

Real root isolation is an important ingredient for the construction of algebraic numbers. We also analyze the complexity of comparison, sign evaluation and simultaneous inequalities (Sec. 6). Even though the algorithms for these operations are not new [1,17,39,47], the results from real solving and optimal algorithms for polynomial remainder sequences, allow us to improve the com-

plexity of all the algorithms, at least by a factor d (Cor. 2, 3). For SI we prove a bound (Cor. 4) of $\tilde{O}_B(d^4 \tau \max\{n, \tau\})$, or $\tilde{O}_B(N^6)$ under the notation of [41].

These algebraic operations ought to have efficient and generic implementations so that they can be used by other scientific communities. We present a package of SYNAPS [33] that provides these functionalities on real algebraic numbers and exploits various algorithmic and implementation techniques. Experimental results (Sec. 7) illustrate the behavior of the software.

Our results extend directly to the bivariate case, i.e real solving of polynomial system, sign evaluation of a bivariate polynomial evaluated over two algebraic numbers, SI etc. However due to reasons of space, we cannot present these results here. The reader may refer to [20,18].

Outline. In Sec. 2, we recall the main ingredient of the Sturm solver and analyse them in detail. Sec. 3 presents the ingredients of the Bernstein solver and their complexity. In Sec. 4, we present the general scheme for two algorithms based on Sturm-Habicht sequences and on Bernstein basis representation, for real root isolation and computation of the multiplicities. The following section is devoted to the complexity analysis of both methods. Sec. 6 is devoted to operations with real algebraic numbers, i.e. comparison, sign evaluation and SI. Sec. 7 illustrates our implementation in SYNAPS and experiments on various data sets (cf also the Appendix). Finally, we sketch our current and future work in Sec. 8.

2 Preliminaries for Sturm–Habicht Sequences

We recall here the main ingredients related to Sturm sequence computations and their bit complexity.

Let $f = \sum_{k=0}^p f_k x^k, g = \sum_{k=0}^q g_k x^k \in \mathbb{Z}[x]$ where $\deg(f) = p \geq q = \deg(g)$ and $\mathcal{L}(f) = \mathcal{L}(g) = \tau$. We denote by $\mathbf{rem}(f, g)$ and $\mathbf{quo}(f, g)$ the remainder and the quotient, respectively, of the Euclidean division of f by g , in $\mathbb{Q}[x]$.

Definition 1. [29] *The signed polynomial remainder sequence of f and g , denoted by $\mathbf{SPRS}(f, g)$, is the polynomial sequence*

$$R_0 = f, R_1 = g, R_2 = -\mathbf{rem}(f, g), \dots, R_k = -\mathbf{rem}(R_{k-2}, R_{k-1})$$

where $\mathbf{rem}(R_{k-1}, R_k) = 0$. The quotient sequence of f and g is the polynomial sequence $\{Q_i\}_{0 \leq i < k}$, where $Q_i = \mathbf{quo}(R_i, R_{i+1})$ and the quotient boot is $(Q_0, Q_1, \dots, Q_{k-1}, R_k)$.

There is a huge bibliography on signed polynomial remainder sequences (c.f [1,45,47] and references there in). [46] presents a unified approach to subresultants. For the Sturm-Habicht (or Sylvester-Habicht) sequences the reader may refer to [22] (see also [1,29,30]).

In this paper we consider the Sturm-Habicht sequence of f and g , i.e $\mathbf{StHa}(f, g)$, which contains polynomials that are proportional to the polynomials in $\mathbf{SPRS}(f, g)$. Sturm-Habicht sequences achieve better bounds on the bit size of the coefficients

and have good specialization properties, since they are defined through determinants.

Let M_j be the matrix which has as rows the coefficient vectors of the polynomials $fx^{q-1-j}, fx^{q-2-j}, \dots, fx, f, g, gx, \dots, gx^{p-2-j}, gx^{p-1-j}$ with respect to the monomial basis $x^{p+q-1-j}, x^{p+q-2-j}, \dots, x, 1$. The dimension of M_j is $(p+q-1-2j) \times (p+q-1-j)$. For $l = 0, \dots, p+q-1-j$ let M_j^l be the square matrix of dimension $(p+q-2j) \times (p+q-2j)$ obtained by taking the first $p+q-1-2j$ columns and the l -th column of M_j .

Definition 2. *The Sturm-Habicht sequence of f and g , is the sequence*

$$\mathbf{StHa}(f, g) = (H_p = H_p(f, g), \dots, H_0 = H_0(f, g))$$

where $H_p = f, H_{p-1} = g$ and $H_j = \sum_{l=0}^j \det(M_j^l)x^l$. The sequence of principal Sturm-Habicht coefficients $(h_p = h_p(f, g), \dots, h_0(f, g))$ is defined as $h_p = 1$ and h_j is the coefficient of x^j in the polynomial H_j , for $0 \leq j \leq p$. When $h_j = 0$ for some j then the sequence is called defective, otherwise non-defective.

If $\mathbf{StHa}(f, g)$ is non-defective then it coincides up to sign with the classical subresultant sequence introduced by Collins [8] (see also [47]). However, in the defective case, can have better control on the bit size of the coefficients in the sequence (see e.g [29,30]).

Theorem 1. [1,38,30,45] *There is an algorithm that computes $\mathbf{StHa}(f, g)$ in $\mathcal{O}_B(pqM(p\tau))$, or $\tilde{\mathcal{O}}_B(p^2q\tau)$. Moreover, $\mathcal{L}(H_j(f, g)) = \mathcal{O}(p\tau)$.*

Let the quotient boot that corresponds to $\mathbf{StHa}(f, g)$, be $\mathbf{StHaQ}(f, g) = (Q_0, Q_1, \dots, Q_{k-1}, H_k)$. The number of coefficients in $\mathbf{StHaQ}(f, g)$ is $\mathcal{O}(q)$ and their bit size is $\mathcal{O}(p\tau)$ (c.f [1,38]).

Theorem 2. [1,29,38,45] *The quotient boot, the resultant and the gcd of f and g , can be computed in $\mathcal{O}_B(q \log(q)M(p\tau))$ or $\tilde{\mathcal{O}}_B(pq\tau)$.*

Theorem 3. [29,38] *There is an algorithm that computes the evaluation of $\mathbf{StHa}(f, g)$ over a number \mathbf{a} , where $\mathbf{a} \in \mathbb{Q} \cup \{\pm\infty\}$ and $\mathcal{L}(\mathbf{a}) = \sigma$ with complexity $\mathcal{O}_B(q \log q M(\max(p\tau, q\sigma)))$ or $\mathcal{O}_B(qM(\max(p\tau, q\sigma)))$ if $\mathbf{StHaQ}(f, g)$ is already computed. In both cases the complexity is $\tilde{\mathcal{O}}_B(q \max(p\tau, q\sigma))$.*

In many cases, e.g real root isolation, sign evaluation, comparison of algebraic numbers, we need the evaluation of $\mathbf{StHa}(f, f')$ over a rational number of bit size $\mathcal{O}(p\tau)$. If we perform the evaluation by Horner's rule then for every polynomial in sequence, there are $\Omega(p)$, we must perform $\Omega(p)$ multiplications between numbers of bit size $\mathcal{O}(p\tau)$ and $\mathcal{O}(p^2\tau)$, thus the overall complexity is $\mathcal{O}_B(p^3M(p\tau))$.

However, when we compute the complete $\mathbf{StHa}(f, f')$ in $\mathcal{O}_B(p^2M(p\tau))$ (Th. 1), the quotient boot is computed implicitly [38,1]. Thus, we can use the quotient boot in order to perform the evaluation even if we have already computed all the polynomials in the Sturm-Habicht sequence. Notice also that the computation should be started by the last element of the quotient boot so as to avoid the costly computation of two polynomial evaluations using Horner's scheme.

Even though this approach is optimal, it involves big constants in its complexity, thus it is not efficient in practice when the length of the sequence is not sufficiently big or when the sequence is defective (see e.g [12]). Moreover, special techniques should be used for its implementation to avoid costly operations with rational numbers. So, as it is always the case with optimal algebraic algorithms, the implementation is far from a trivial task.

Theorem 4. [1] *The square-free part of f , i.e. f_{red} , can be computed from $\mathbf{StHa}(f, f')$, in $\mathcal{O}_B(p \log p M(p\tau))$ or $\tilde{\mathcal{O}}_B(p^2\tau)$, and $\mathcal{L}(f_{red}) = \mathcal{O}(p + \tau)$.*

Let $W_{(f,g)}(\mathbf{a})$ denote the number of modified sign changes of the evaluation of $\mathbf{StHa}(f, g)$ over \mathbf{a} . Notice that $W_{(f,g)}(\mathbf{a})$ does not refer to the usual counting of sign variations, since special care should be taken for the presence of consecutive zeros [1,22,29].

Theorem 5. [1,47,39] *Let $f, g \in \mathbb{Z}[x]$ be relatively prime polynomials, where f is square-free and f' is the derivative of f . If $\mathbf{a} < \mathbf{b}$ are both non-roots of f and γ ranges over the roots of f in (\mathbf{a}, \mathbf{b}) , then $W_{(f,g)}(\mathbf{a}) - W_{(f,g)}(\mathbf{b}) = \sum_{\gamma} \text{sign}(f'(\gamma)g(\gamma))$.*

Corollary 1. *If $g = f'$ then $\mathbf{StHa}(f, f')$ is the Sturm sequence and Th. 5 counts the number of real roots of f in (\mathbf{a}, \mathbf{b}) .*

For the Sturm solver by $V(f, [\mathbf{a}, \mathbf{b}])$ will denote $V(f, [\mathbf{a}, \mathbf{b}]) = W_{(f,f')}(\mathbf{a}) - W_{(f,f')}(\mathbf{b})$.

3 Preliminaries for Bernstein Basis Representation

In this section we present the main ingredients needed for the representation of polynomials in the Bernstein basis.

Let $\mathbb{R}[x]_d$ be the set of real polynomials of degree d . For $a < b \in \mathbb{R}$, we denote by $B_d^i(x; a, b) = \binom{d}{i} \frac{(x-a)^i (b-x)^{d-i}}{(b-a)^d}$ ($i = 0, \dots, d$) the Bernstein basis of $\mathbb{R}[x]_d$ on an interval $[a, b]$.

For any polynomial $f \in \mathbb{R}[x]_d = \sum_{i=0}^d b_i B_d^i(x; a, b)$, represented in the Bernstein basis, the coefficients $\mathbf{b} = (b_i)_{i=0, \dots, d}$ are called *control coefficients* of f . We denote by $V(f, [a, b]) \equiv V(\mathbf{b})$, the number of sign changes in this sequence \mathbf{b} (after removing zero coefficients).

The following theorem, which is a direct consequence of Descartes' rule, allows us to bound the number of real roots of f on the interval $[a, b]$

Proposition 1. [1] *The number N of real roots of f on (a, b) is bounded by $V(f, [a, b])$. Moreover $N \equiv V(f, [a, b]) \pmod{2}$.*

Given a polynomial f represented in the Bernstein basis on an interval $[a, b]$, de Castel'jau's algorithm (see e.g [1,34]), allows us to compute its representation on the Bernstein bases of the two subintervals, $I_L = [a, (1-t)a + tb]$ and $I_R = [(1-t)a + tb, b]$, where $0 \leq t \leq 1$. Namely, $\mathbf{b}_L = (b_0^i)_{i=0, \dots, d}$ (resp. $\mathbf{b}_R =$

$(b_i^{d-i})_{i=0,\dots,d}$ are the control coefficients of f on I_L (resp. I_R), where $b_i^0 = b_i, i = 0, \dots, d$, and

$$b_i^r = (1-t)b_i^{r-1} + tb_{i+1}^{r-1}(t), \quad 0 \leq i \leq d-r, \quad 0 \leq r \leq d. \quad (1)$$

In order to analyse the complexity of the de Casteljau algorithm we recall some polynomial transformations related to the Bernstein representation (see [34] for more details). Let $\mathbb{R}[x, y]_{[d]}$ be the set of homogeneous polynomials of degree d in (x, y) . For any $f \in \mathbb{R}[x]_d$, we denote by \bar{f} the homogenisation of f in degree d . For $\lambda \neq 0, \mu \in \mathbb{R}$, consider the following maps, $\mathbb{R}^2 \rightarrow \mathbb{R}^2$:

$$\begin{aligned} - \rho &: (x, y) \mapsto (y, x), \\ - H_\lambda &: (x, y) \mapsto (\lambda x, y), \quad H'_\lambda : (x, y) \mapsto (x, \lambda y), \\ - T_\mu &: (x, y) \mapsto (x - \mu y, y), \quad T'_\mu : (x, y) \mapsto (x, y - \mu x). \end{aligned}$$

The composition of the previous maps with \bar{f} induces invertible transformations on the set of polynomials of degree d , and the corresponding maps for non-homogeneous polynomials, which we denote using the same names, are: $\forall f \in \mathbb{R}[x]_d, \rho(f) = x^d f(1/x), H_\lambda(f) = f(\lambda x), H'_\lambda(f) = f(\lambda^{-1}x), T_\mu(f) = f(x - \mu), T'_\mu(f) = (1 - \mu x)^d f(\frac{x}{1 - \mu x})$.

For any polynomial, $f(x) = \sum_{i=0}^d b_i B_d^i(x; a, b)$, we have

$$\rho \circ T_1 \circ \rho \circ H_{b-a} \circ T_{-a}(f) = \sum_{i=0}^d \binom{d}{i} b_i x^i.$$

Now consider another interval $[c, d]$. The representation of p in the Bernstein basis on $[c, d]$ is $p(x) = \sum_{i=0}^d b'_i B'_d(x; c, d)$. The map which transforms p from its Bernstein representation on $[a, b]$ to its Bernstein representation on $[c, d]$, i.e from $\sum_{i=0}^d \binom{d}{i} b_i x^i$ to $\sum_{i=0}^d \binom{d}{i} b'_i x^i$ is

$$\rho \circ T_1 \circ \rho \circ H_{d-c} \circ T_{-c} \circ T_a \circ H_{\frac{1}{b-a}} \circ \rho \circ T_{-1} \circ \rho = T'_1 \circ H_{d-c} \circ T_{a-c} \circ H_{\frac{1}{b-a}} \circ T'_{-1} \quad (2)$$

If we consider $[a, b] = [0, 1]$ and $[c, d] = [0, \frac{1}{2}]$ then map (2) becomes: $\rho \circ T_{-1} \circ \rho \circ H_{\frac{1}{2}} \circ \rho \circ T_1 \circ \rho$. And after simplifications, we obtain

$$\Delta_L : \bar{f} \mapsto \bar{f}(x + \frac{y}{2}, \frac{y}{2}) = \bar{f} \circ T_{-1} \circ H'_{\frac{1}{2}}. \quad (3)$$

If we consider the symmetric case, i.e $[a, b] = [0, 1]$ and $[c, d] = [\frac{1}{2}, 1]$ then map (2) becomes: $\rho \circ T_{-1} \circ \rho \circ H_{\frac{1}{2}} \circ T_{-\frac{1}{2}} \circ \rho \circ T_1 \circ \rho$. It corresponds to the following map on the homogeneous polynomials:

$$\Delta_R : \bar{f} \mapsto \bar{f}(\frac{x}{2}, \frac{x}{2} + y) = \bar{f} \circ T'_{-1} \circ H_{\frac{1}{2}}.$$

In both cases, multiplication by 2^d yields the map $\bar{\Delta}_R : \bar{p} \mapsto \bar{p}(x, x + 2y)$, (resp. $\bar{\Delta}_L : \bar{p} \mapsto \bar{p}(2x + y, y)$), which operates on polynomials with integer coefficients.

Proposition 2. *Let $(b_i)_{i=0,\dots,d} \in \mathbb{Z}^{d+1}$ be the coefficients of a polynomial f in the Bernstein basis on the interval $[a, b]$, and let ν be a bound on their size. The complexity of computing the Bernstein coefficients of f for the two subintervals $[a, \frac{a+b}{2}]$, $[\frac{a+b}{2}, b]$ is bounded by $\tilde{\mathcal{O}}_B(d(d+\nu))$ and their size is bounded by $d+\nu$.*

Proof. Using the de Casteljau scheme, Eq. (1) using $t = \frac{1}{2}$, we prove by induction that the coefficients $b_i^r = \frac{(b_i^{r-1} + b_{i+1}^{r-1})}{2}$ are of the form $\frac{\bar{b}_i^r}{2^r}$, where $\bar{b}_i^r \in \mathbb{Z}$ is of size $\leq \nu + r$. Reducing to the same denominator 2^d , we obtain integer coefficients of size $\leq \nu + d$.

We denote by ν' the size of the coefficients $\binom{d}{i} b_i$ where $(b_i)_{i=0,\dots,d}$ are the coefficients of f in the Bernstein basis $(B_d^i(x; a, b))_{i=0,\dots,d}$. Notice that $\nu' \leq \nu + d$.

For computing the coefficients of f on $[a, \frac{a+b}{2}]$ and $[\frac{a+b}{2}, b]$, we apply the same operations as when we compute the coefficients of a polynomial for the Bernstein bases on $[0, \frac{1}{2}]$ and $[\frac{1}{2}, 1]$, when it is given in the Bernstein basis on $[0, 1]$.

According to (3), applying de Casteljau algorithm corresponds first to multiply by the binomial coefficients, then to shift $y \rightarrow x + y$, then to scale one variable of the homogeneous polynomial \bar{f} by $\frac{1}{2}$, and finally to divide by the binomial coefficients³.

Since the bit size of the binomial coefficients is bounded by d (their sum is 2^d), multiplying the b_i by them costs at most $\tilde{\mathcal{O}}_B(d(\nu + d))$. The shift by 1 of a polynomial of degree d with coefficients of size $\leq \nu + d$ requires $\tilde{\mathcal{O}}_B(d(d+\nu))$ bit operations [45, Th. 9.15]. This produces a polynomial which coefficients are of size $\mathcal{O}(\nu + d)$. Thus scaling in this polynomial, the variable by $\frac{1}{2}$ and computing the quotient by the binomial coefficients requires $\tilde{\mathcal{O}}_B(d(\nu + d))$ bit-operations.

Therefore, the complexity of computing the Bernstein coefficients of f on the subinterval $[a, \frac{a+b}{2}]$ is bounded by $\tilde{\mathcal{O}}_B(d(\nu + d))$. By symmetry, inverting the order of the coefficients of f , we obtain the same bound for the coefficients of f on $[\frac{a+b}{2}, b]$, which ends the proof. \square

4 Subdivision Solver

Let $f = \sum_{i=0}^d a_i x^i \in \mathbb{Z}[x]$, such that $\deg(f) = d$ and $\mathcal{L}(f) = \tau$ and let f_{red} be its square-free part. We want to isolate the real roots of f , i.e to compute intervals with rational endpoints that contain one and only one root of f , as well as the multiplicity of every real root. In Fig. 1, we present the general scheme of the (subdivision) solvers that we consider. It uses an external function $V(f, I)$, which bounds the number of roots of f on an interval I . In the case of Sturm solver, $V(f, I)$ is exactly the number of roots (counted without multiplicities) of f on I (see Sec. 2). In the case of Bernstein solver, $V(f, I)$ is equal to the number of roots of f on I (counted with multiplicities) modulo 2 (see Sec. 3).

Separation bounds. An important ingredient for the analysis of our solvers is a good bound on the minimal distance $\text{sep}(f)$ between the roots of a univariate

³ Not needed, if we have to apply repeatedly the shift operation.

ALGORITHM: Real Root Isolation
Input: A polynomial $f \in \mathbb{Z}[x]$, with $\deg(f) = d$ and $\mathcal{L}(f) = \tau$.
Output: A list of intervals with rational endpoints, which contain one and only one root of f and the multiplicity over every real root.
1. Compute the square-free part of f , i.e f_{red}
2. Compute an interval $I_0 = (-B, B)$ with rational endpoints that contains all the real roots. Initialize a queue Q with I_0 .
3. While Q is not empty do
a) Pop an interval I from Q and compute $v := V(f, I)$.
b) If $v = 0$, discard I .
c) If $v = 1$, output I .
d) If $v \geq 2$, split I into I_L and I_R and push them to Q .
4. Determine the multiplicities of the real roots, using the square-free factorization of f .

Fig. 1. Real root isolation subdivision algorithm

polynomial f (also called *separation bound*), or more generally on the product of distances between roots. We recall here classical results, slightly adapted to our context.

For the separation bound it is known [1,31,47] that $\text{sep}(f) \geq d^{-\frac{d+2}{2}}(d+1)^{\frac{1-d}{2}}2^{\tau(1-d)}$, thus $\log(\text{sep}(f)) = \mathcal{O}(d\tau)$. The latter provides a bound on the bit size of the endpoints of the isolating intervals. Recall that Mahler's measure of a polynomial f is $\mathcal{M}(f) = |a_d| \prod_{i=1}^d \max\{1, |\gamma_i|\}$, where a_d is the leading coefficient and γ_i are all the roots of f . We know that $\mathcal{M}(f) < 2^\tau \sqrt{d+1}$ [1,31]. Thus, the following inequality [1,31] holds:

$$\mathcal{M}(f_{red}) \leq \mathcal{M}(f) < 2^\tau \sqrt{d+1} \quad (4)$$

For the minimum distance between two consecutive real roots of a square-free polynomial, Davenport-Mahler bound is known [12] (see also [24,27]). The conditions for this bound to hold were generalized by Du et al [13]. Moreover, a similar bound, with less strict hypotheses, also appeared in [32]. Using (4) we can provide a similar bound to [12,24,27]) for non square-free polynomials.

Theorem 6 (Davenport-Mahler bound revisited). *Let $A = \{\alpha_1, \dots, \alpha_k\}$ and $B = \{\beta_1, \dots, \beta_k\}$ be subsets of distinct (complex) roots of f (not necessarily square-free) such that $\beta_i \notin \{\alpha_1, \dots, \alpha_i\}$ and $|\beta_i| \leq |\alpha_i|$, for all $i \in \{1, \dots, k\}$. Then*

$$\prod_{i=1}^k |\alpha_i - \beta_i| \geq \mathcal{M}(f)^{-d+1} d^{-\frac{d}{2}} \left(\frac{\sqrt{3}}{d}\right)^k$$

The bound also holds when $\alpha_1 > \beta_1 = \alpha_2 > \beta_2 = \dots \alpha_k > \beta_k := \alpha_{k+1}$, are distinct real roots of f .

Proof. Use [24] (or [47,27]) and (4). □

5 Complexity Analysis of Real Root Isolation

In this section, we bound the number of bit operations for isolating the real roots of a polynomial using Sturm and Bernstein solver. We consider the tree

associated with a run of the subdivision algorithm on a polynomial f . Each node represents an interval. The root of the tree corresponds to the initial interval $I_0 = [a, b]$. Each interval which is not a leaf of the tree is splitted to two half intervals. The depth of a node of the tree (associated with an interval I) is $\log(|I_0|/|I|)$. This is also the number of subdivisions performed to obtain the subinterval I from I_0 . Notice that the number of steps (subdivisions) that the algorithm performs equals the total number of nodes of the subdivision tree, or in other words equals the number of intervals (subintervals of I_0) that are tested.

In what follows we present in detail the complexity of each step of the subdivision algorithm (see Fig 1).

5.1 Square-Free Factorisation [step 1]

The computation of f_{red} can be done in $\tilde{\mathcal{O}}_B(d^2\tau)$ (Th. 4). Notice that $\mathcal{L}(f_{red}) = \mathcal{O}(d+\tau)$. We assume that $d = \mathcal{O}(\tau)$, thus $\mathcal{L}(f_{red}) = \mathcal{O}(\tau)$. Notice also that after this computation, the Sturm-Habicht sequence $\mathbf{StHa}(f)$ is available. We do not need the complete sequence but only the quotient boot, thus this computation can be done in $\tilde{\mathcal{O}}_B(d^2\tau)$ (Th. 2). However, we may also assume that the complete sequence is computed, with complexity $\tilde{\mathcal{O}}_B(d^3\tau)$ (Th. 1), since this step is not the bottleneck of the algorithm.

5.2 Root Bounds and Initialization [step 2]

The Cauchy bound states that if α is a real root of f then $|\alpha| \leq B = 1 + \max\left(\left|\frac{a_{d-1}}{a_n}\right|, \dots, \left|\frac{a_0}{a_n}\right|\right) \leq 2^\tau$. Various upper bounds are known for the absolute value of the real roots [1,47,45]. However, asymptotically the bit size of all the bounds is the same, i.e $B \leq 2^\tau$. Thus, we can take $I_0 = [a, b]$, with $a \geq -2^\tau$, $b \leq 2^\tau$.

For the Sturm solver, before starting the main loop, we have to compute the Sturm-Habicht sequence of f , which costs $\tilde{\mathcal{O}}_B(d^3\tau)$ (Th. 2).

For the Bernstein solver, we have to represent f_{red} in the Bernstein basis of $[a, b]$. This can be done in $\mathcal{O}(d^2)$ arithmetic operations and it produces coefficients of size $\mathcal{O}(d(d+\tau))$. The cost of this transformation is bounded by $\tilde{\mathcal{O}}_B(d^3(d+\tau))$.

In both methods, the initialisation step can be done in $\tilde{\mathcal{O}}_B(d^3(d+\tau))$.

5.3 Computing $V(f, I)$ and Splitting [steps 3.a-d]

Suppose that the algorithm is at depth h of the subdivision tree. The tested interval, say I , has endpoints of bit size bounded by $\tau+h$, since each subdivision step increases the bit size by one.

Using Sturm solver, we compute $V(f, I)$, Cor. 1, by evaluating $\mathbf{StHa}(f)$ over endpoints of I . The cost of the evaluation is $\tilde{\mathcal{O}}_B(d^2(\tau+h))$ (Th. 3). Then we split I , i.e compute the middle point of it, in $\mathcal{O}_B(\tau+h)$.

Using Bernstein solver, we compute $V(f, I)$ by counting the number of sign changes in the control coefficients of f in I . This can be done in $\mathcal{O}(d)$ operations. We denote by $\tau_0 = \mathcal{O}(d(d + \tau))$ (Sec. 5.2) a bound on the bit size of the coefficients of f in the Bernstein basis on the interval I_0 . By proposition 2, since we performed h subdivisions starting from a polynomial with coefficients of size τ_0 , the coefficients of f on I are of bit size $\tau_0 + dh$ and the complexity of the splitting operation is in $\tilde{\mathcal{O}}_B(d(d + \tau_0 + dh)) = \tilde{\mathcal{O}}_B(d^2(d + \tau + h))$ (Sec. 5.2).

Finally for both solvers, the steps 3.a-d can be performed in $\tilde{\mathcal{O}}_B(d^2(d + \tau + h))$.

5.4 Subdivision Tree Analysis [step 3]

In this section, we analyse the total number of subdivisions. A bound on this number was derived in [27, Th. 5.5, 5.6], where in Rem. 5.7 the authors state: “The theorem (5.6) implies the dominance relations $hk \preceq n \log(nd)$ and $h \preceq n \log(nd)$ which can be used in an asymptotic analysis of the Algorithm 1 when the ring S of the coefficients is \mathbb{Z} ”, where k is the number of internal nodes of depth h in the recursion tree of the subdivision algorithm based on Descartes’ rule, n is the degree and d is the Euclidean norm of the polynomial. In [43, Th. 5], a $\mathcal{O}(d\tau + d \log d)$ bound is derived and, later on, [15] proved optimality under the mild assumption that $\tau = \Omega(\log d)$. Our arguments for this bound are a combination and/or simplification of the arguments in [27,13,43]. Our proof (prop. 4) is simpler than the one in [15,43] since the handling of the subdivision tree stops at an earlier level and we use Th. 6 (as stated in [24] and [27]) without any modifications. We also simplify substantially the proof of [13], for Sturm solver.

We denote by \mathcal{I} the set of intervals which are the parent of two leaves in the subdivision tree in Sturm (resp. Bernstein) method. By construction, for $I \in \mathcal{I}$, $V(f, I) \geq 2$ but for the two subintervals I_L, I_R of I , $V(f, I_L)$ and $V(f, I_R)$ are in $\{0, 1\}$ (because these intervals are leaves of the subdivision tree). Moreover, for the Sturm solver, we have $V(f, I) = 2$ and $V(f, I_L) = V(f, I_R) = 1$.

Notice that $|\mathcal{I}|$ is less than $V(f, I_0)$, since at each subdivision the sum of the variations of f on all the intervals cannot increase, for both methods (see [36,34] for the Bernstein solver). In particular, we have $|\mathcal{I}| \leq d$.

Proposition 3. *Let $I \in \mathcal{I}$. Then, there exist two distinct (complex) roots $\alpha_I \neq \beta_I$ of f such that $|\alpha_I - \beta_I| < 2|I|$.*

Proof. Consider an interval $I \in \mathcal{I}$ which contains two leaves I_L, I_R of the subdivision tree. We have the following possibilities for the sign variation of f on the two subintervals I_L, I_R :

- (1, 1): for both methods, there are two distinct real roots $\alpha \in I_L, \beta \in I_R$ in I and $|\alpha - \beta| \leq |I|$. This is the only case, which can happen in Sturm method.
- (0, 0): this may happen only in the Bernstein method. Since the sign variation of $V(f, I) \geq 2$, by the first circle theorem [34,1,27], there exist two complex conjugate roots $\beta, \bar{\beta}$ in the disc $D(m(I), \frac{|I|}{2})$. Therefore $|\beta - \bar{\beta}| \leq |I|$.

- (1, 0) or (0, 1): this may also happen only in the Bernstein method. Then, there is a real root α in I . Since $V(f, I) \geq 2$, by the second circle theorem [34,1,27], there exists two complex conjugate roots $\beta, \bar{\beta}$ in the union of the discs $D(m(I) \pm \frac{1}{2\sqrt{3}}\mathbf{i}|I|, \frac{1}{\sqrt{3}}|I|)$, which is contained in a disc of diameter $2|I|$. Therefore $|\beta - \alpha| < 2|I|$.

Thus the proposition holds. \square

In addition, we can prove the following result.

Lemma 1. *Let $\{\alpha_I, \beta_I\} \cap \{\alpha_{I'}, \beta_{I'}\} \neq \emptyset$, then $I \cap I' \neq \emptyset$.*

Proof. For the Sturm method, this property is clear since $\alpha_I, \beta_I \in I$.

Let us consider the Bernstein subdivision method. Without loss of generality, we can assume in the proof that $I \neq I'$, $|I'| \leq |I|$, and that $\forall x \in I, \forall y \in I', x \leq y$.

We suppose that $I \cap I' = \emptyset$. Then since the intervals are obtained by binary subdivision, we can assume that the distance between I and I' is at least $|I'|$. Then by scaling and translation, we can assume that the right endpoint of I is 0, that $I' = [1 + u, 2 + u]$, ($u \geq 0$). Then, the tangents to the larger circles containing I and the roots α_I, β_I at $(0, 0)$ are $\frac{\sqrt{3}}{2}x \pm \frac{y}{2} = 0$. We denote by R_I the union of the corresponding discs, so that $\alpha_I, \beta_I \in R_I$.

The center of the discs whose union $R_{I'}$ contains the roots $\alpha_{I'}, \beta_{I'}$ are $(\frac{3}{2} + u, \pm \frac{\sqrt{3}}{6})$ and their radius $\frac{\sqrt{3}}{3}$. A direct computation of the distance between these centers and the two tangent lines shows that $R_I \cap R_{I'} = \emptyset$. Consequently, if $I \cap I' = \emptyset$, then we have $\{\alpha_I, \beta_I\} \cap \{\alpha_{I'}, \beta_{I'}\} = \emptyset$. \square

Let us number the intervals of \mathcal{I} by increasing order and denote by \mathcal{I}' the subset with an even index and by \mathcal{I}'' the subset with an odd index. By lemma 1, the pairs $\{\alpha_I, \beta_I\}$ for $I \in \mathcal{I}'$ (resp. \mathcal{I}'') are disjoint. Thus, by Th. 6 (exchanging the role of α_I and β_I if necessary), we have

$$\prod_{I \in \mathcal{J}} |\alpha_I - \beta_I| \geq \mathcal{M}(f)^{-d+1} d^{-\frac{d}{2} - |\mathcal{J}|} \sqrt{3}^{|\mathcal{J}|}, \quad (5)$$

for $\mathcal{J} = \mathcal{I}'$ or $\mathcal{J} = \mathcal{I}''$. This is the key ingredient of the following result:

Proposition 4. *The number of subdivisions in both methods is in $\mathcal{O}(d\tau + d \log(d))$.*

Proof. The number N of subdivisions equals the number of internal nodes in the subdivision tree. It is less than the sum of the depth of I , for $I \in \mathcal{I}$:

$$\begin{aligned} N &\leq \sum_{I \in \mathcal{I}} \log \frac{|\mathbf{b} - \mathbf{a}|}{|I|} \\ &\leq |\mathcal{I}| \log |\mathbf{b} - \mathbf{a}| - \sum_{I \in \mathcal{I}} \log |I| \\ &\leq |\mathcal{I}| \log |\mathbf{b} - \mathbf{a}| + |\mathcal{I}| - \sum_{I \in \mathcal{I}} \log |\alpha_I - \beta_I| \text{ (Prop. 3)} \end{aligned}$$

By (5), we have $-\sum_{I \in \mathcal{I}'} \log |\alpha_I - \beta_I| \leq (d-1) \log(\mathcal{M}(f)) + (\frac{d}{2} + |\mathcal{I}'|) \log d - |\mathcal{I}'| \log \sqrt{3}$. A similar bound applies for \mathcal{I}'' .

As we can take $\mathbf{a} = -2^\tau, \mathbf{b} = 2^\tau$ (by Cauchy bound) and $\log \mathcal{M}(f) \leq \tau + \frac{1}{2} \log(d+1)$ (Eq. 4) and $|\mathcal{I}'| + |\mathcal{I}''| = |\mathcal{I}| \leq d$, the number of internal nodes N in the subdivision tree is bounded by

$$\begin{aligned} N &< |\mathcal{I}| + |\mathcal{I}| \log |\mathbf{b} - \mathbf{a}| - \sum_{I \in \mathcal{I}} \log |\alpha_I - \beta_I| \\ &\leq d + d(\tau + 1) + (d-1)(2\tau + \log(d+1)) + 2d \log d \\ &= \mathcal{O}(d\tau + d \log d). \end{aligned}$$

□

Remark 1. The constant in this bound on the number of subdivisions can be divided by 2, in Sturm method, by applying directly Th. 6 to α_I, β_I for $I \in \mathcal{I}$.

5.5 Multiplicities [step 4]

In order to compute the multiplicities of the roots, we compute the square-free factorization, i.e a sequence of square-free coprime polynomials (g_1, g_2, \dots, g_m) with $f = g_1 g_2^2 \cdots g_m^m$ and $g_m \neq 1$. The algorithm of Yun [45] computes the square-free factorization in $\tilde{\mathcal{O}}_B(d^2\tau)$. To be more specific the cost is twice the cost of the computation of $\mathbf{StHa}(f, f')$ [21]. Moreover $\deg(g_i) = \delta_i \leq d$ and $\mathcal{L}(g_i) = \mathcal{O}(d\tau)$ by Mignotte's bound [31], where $1 \leq i \leq m$.

At every isolating interval, one and only one g_i must have opposite signs at its endpoints, since g_i are square-free and pairwise coprime. If g_i changes sign at an interval then the multiplicity of the real root that the interval contains is i . Each g_i can be evaluated over an isolating point in $\tilde{\mathcal{O}}_B(\delta_i^2 d\tau)$, using Horner's rule. We can evaluate it over all the isolating points (there are at most $d+1$), in $\tilde{\mathcal{O}}_B(\delta_i d^2\tau)$ [45,47]. Since $\sum_{i=1}^m \delta_i \leq d$ the overall cost is $\tilde{\mathcal{O}}_B(d^3\tau)$.

5.6 Complexity of Real Root Isolation

In this section, we prove that the two subdivision solvers has a bit complexity $\tilde{\mathcal{O}}_B(d^4\tau^2)$:

Theorem 7. *Let $f \in \mathbb{Z}[x]$, with $\deg(f) = d$ and $\mathcal{L}(f) = \tau$, not necessarily square-free. We can isolate the real roots of f and determine their multiplicities using Sturm or Bernstein methods in $\tilde{\mathcal{O}}_B(d^4\tau^2)$. Moreover, the endpoints of the isolating intervals have bit size bounded by $\mathcal{O}(d\tau)$.*

Proof. In order to isolate the real roots of f , we first compute its square-free part (step 1). This can be done in $\tilde{\mathcal{O}}_B(d^2\tau)$ arithmetic operations and yields a polynomial f_{red} , which coefficients are of size bounded by $\mathcal{O}(d+\tau)$ (see section 5.1). This step is not necessary in Sturm method.

The initialisation step costs $\tilde{\mathcal{O}}_B(d^3(d+\tau))$ (Sec. 5.2).

Then we run the main loop of the subdivision algorithm. The cost of a subdivision step at level h is $\tilde{\mathcal{O}}_B(d^2(d+\tau+h))$ (Sec. 5.3).

By Prop. 4, the number of subdivisions and the depth h of any node of the subdivision tree is $\tilde{\mathcal{O}}(d\tau)$. Therefore, the overall complexity of both subdivision solvers is $\tilde{\mathcal{O}}_B(d^2(d+\tau+d\tau)d\tau) = \tilde{\mathcal{O}}_B(d^4\tau^2)$. □

6 Real Algebraic Numbers

The real algebraic numbers, i.e. those real numbers that satisfy a polynomial equation with integer coefficients, form a real closed field denoted by $\mathbb{R}_{alg} = \overline{\mathbb{Q}}$. From all integer polynomials that have an algebraic number α as root, the primitive one (the gcd of the coefficients is 1) with the minimum is called *minimal*. The minimal polynomial is unique (up to a sign), primitive and irreducible [47]. Since we use Sturm-Habicht sequences, it suffices to deal with algebraic numbers, as roots of any square-free polynomial and not as roots of their minimal ones. In order to represent a real algebraic number we choose the *isolating interval representation*.

Definition 3. *The isolating-interval representation of real algebraic number $\alpha \in \mathbb{R}_{alg}$ is $\alpha \cong (f(x), I)$, where $f(x) \in \mathbb{Z}[x]$ is square-free and $f(\alpha) = 0$, $I = [a, b]$, $a, b \in \mathbb{Q}$ and f has no other root in I .*

Using the results of Sec. 2 and 3 we can compute the isolating interval representation of all the real roots a polynomial f , with $\deg(f) = d$ and $\mathcal{L}(f) = \tau$, in $\tilde{\mathcal{O}}_B(d^4\tau^2)$ and the endpoints of the isolating intervals have bit size $\mathcal{O}(d\tau)$.

Comparison and sign evaluation. We can use Sturm-Habicht sequences in order to find the sign of a univariate polynomial, evaluated over a real algebraic number and to compare two algebraic numbers. We improve existing bounds by one factor.

Corollary 2. *Let $g(x) \in \mathbb{Z}[x]$, where $\deg(g) = d$ and $\mathcal{L}(g) = \tau$, and a real algebraic number $\alpha \cong (f, [a, b])$. We can compute $\text{sign}(g(\alpha))$ in $\tilde{\mathcal{O}}_B(d^3\tau)$.*

Proof. By Th. 5, $\text{sign}(g(\alpha)) = \text{sign}(W_{f,g}[a, b] \cdot f'(\alpha))$. Thus we need to perform two evaluations of $\mathbf{StHa}(f, g)$ over the endpoints of the isolating interval of α . The complexity of each is $\tilde{\mathcal{O}}_B(d^3\tau)$ (Th. 3), which is also the complexity of the operation. \square

Corollary 3. *We can compare two real algebraic numbers in isolating interval representation in $\tilde{\mathcal{O}}_B(d^3\tau)$.*

Proof. Let two algebraic numbers $\gamma_1 \cong (f_1(x), I_1)$ and $\gamma_2 \cong (f_2(x), I_2)$ where $I_1 = [a_1, b_1]$, $I_2 = [a_2, b_2]$. Let $J = I_1 \cap I_2$. When $J = \emptyset$, or only one of γ_1 and γ_2 belong to J , we can easily order the 2 algebraic numbers. If $\gamma_1, \gamma_2 \in J$, then $\gamma_1 \geq \gamma_2 \Leftrightarrow f_2(\gamma_1) \cdot f_2'(\gamma_2) \geq 0$. We obtain the sign of $f_2'(\gamma_2)$, using Cor. 2, thus the complexity of comparison is $\tilde{\mathcal{O}}_B(d^3\tau)$. \square

Simultaneous inequalities. Let $f, A_1, \dots, A_{n_1}, B_1, \dots, B_{n_2}, C_1, \dots, C_{n_3} \in \mathbb{Z}[x]$, with degree bounded by d and coefficient bit size bounded by τ . We wish to compute the number of and the real roots, γ , of f such that $A_i(\gamma) > 0$, $B_j(\gamma) < 0$ and $C_k(\gamma) = 0$ and $1 \leq i \leq n_1, 1 \leq j \leq n_2, 1 \leq k \leq n_3$. Let $n = n_1 + n_2 + n_3$.

Corollary 4. *There is an algorithm that solves the problem of simultaneous inequalities (SI) in $\tilde{\mathcal{O}}_B(d^4\tau \max\{n, \tau\})$.*

Proof. First we compute the isolating interval representation of all the real roots of f in $\tilde{\mathcal{O}}_B(d^4\tau^2)$ (Th. 7). There are at most d . For every real root γ of f , for every polynomial A_i, B_j, C_k we compute the $\text{sign}(A_i(\gamma))$, $\text{sign}(B_j(\gamma))$ and $\text{sign}(C_k(\gamma))$. Sign determination costs $\tilde{\mathcal{O}}_B(d^3\tau)$ (Cor. 2) and in the worst case we must compute n of them. Thus the overall cost is $\tilde{\mathcal{O}}_B(\max\{nd^4\tau, d^4\tau^2\})$.

This improves the known bounds by one or two factors in the bit complexity model.

7 Implementation and Experimentations

In this section, we describe the package for algebraic numbers available in the library SYNAPS [33]. The purpose of this package is to provide a set of tools, for the manipulation of algebraic numbers, needed in applications such as Geometric modeling, and non linear computational geometry. In the problems encountered in these domains, the degree of the involved polynomials is not necessarily very high (< 50), but geometric operations require an intensive use of algebraic solvers. Namely, algebraic numbers are involved as soon as one wants to compute intersections points of curves or surfaces. Predicates such as the comparison of coordinates of points have to be evaluated at such algebraic numbers.

For this reason, in this section we focus on univariate equations of *small degree* in opposition with the first sections, but the input bit size is beyond machine precision. We analyse the behavior of our solvers, in this range of problems which appear in our geometric applications and for which the asymptotic bounds may not be pertinent indicators. We do not consider large degree problems, where memory management issues might influence the solving strategy.

In SYNAPS, there are several solver classes, their interface is as follows

```
template < class T > struct SOLVER {
    typedef NumberTraits<T>::RT   RT;
    typedef NumberTraits<T>::FT   FT;
    typedef NumberTraits<T>::FIT  FIT;

    typedef UPolDse<T>           Poly;
    typedef root_of<T, Poly>     RO_t;
    ... };
```

where RT is the ring number type (typically \mathbb{Z}), FT is the field number type (typically \mathbb{Q}), FIT is the interval type, Poly is the univariate polynomial, RO_t is the type for real algebraic numbers, etc.

Algebraic numbers are of the form:

```
template <class T, class UPOL=UPolDse<T> >
struct root_of {
    NumberTraits<T>::Interval_t interval_;
    UPOL polynomial_;
    ... };
```

parameterized by the type of coefficients and univariate polynomials. This allows flexibility and an easy parameterisation of the code.

In order to construct a real algebraic number the user may select from several different univariate solvers, that we are going to describe hereafter. The other functionalities that we provide are the comparison, `bool compare(const RO_t& a, const RO_t& b)` and the valuation of signs `int sign_at(const Poly& P, const RO_t& a)`, based on interval evaluation and if necessary on the computation of Sturm-Habicht sequences. This involves on several additional functions for computing subresultant sequences with various methods (Euclidean, Subresultants, Sturm-Habicht, etc), for computing the GCD, the square-free part, etc. We also provide the four operations, i.e. $\{+, -, *, /\}$, of `RO_t` with `RT`'s (integer type) and `FT`'s (rational type).

Bivariate problems are also treated in this package, but not reported here (see [18]). Perhaps the most important operation is the construction of real algebraic numbers, i.e real root isolation of univariate polynomials. Several subdivision solvers have been tested for the construction of these algebraic numbers. We report here on the following solvers:

```
(S1) solve(f, IslSturm<ZZ>());
(S2) solve(f, IslBzInteger<QQ>());
(S3) solve(f, IslBzBdgSturm<QQ>());
```

These solvers take as input, polynomials with integer or rational coefficients and output intervals with rational endpoints. All use the same initial interval.

S_1 (`IslSturmQQ` in the plots) is based on the construction of the Sturm-Habicht sequence and subdivisions, using rational numbers or large integers provided by the library GMP.

S_2 (`IslBzIntegerZZ` in the plots) is an implementation of the Bernstein subdivision solver, using integer coefficients. The polynomial is converted to the Bernstein representation on the initial interval, using rational arithmetic. Then, the coefficients are reduced to the same denominator, and the numerators are taken. Finally, the integer version of de Castel'jau algorithm $\overline{\Delta}_{\pm}$ is applied at each subdivision step.

S_3 (`IslBzBdgSturmQQ` in the plots) is a combination of two solvers. In a first part, the polynomial is converted to the Bernstein representation on the initial interval, using rational arithmetic and its coefficients are rounded to `double` intervals. The Bernstein subdivision solver is applied on this interval representation and stops when it certifies the isolation of a root or when it is not possible to decide the existence and uniqueness of a root from the "sign" ($-$, $+$, $?$) of the interval coefficients. In this case, the solver S_2 is used on the intervals which are suspect (caching the Sturm-Habicht computation), in order to complete the isolation process.

We also compare with the time needed for computing the Sturm-Habicht sequence (`SturmSeq` in the plots). We test against `CORE` [25] (`CORE` in the plots) and `mpsolve` a numerical solver based on Aberth's method [4] and implemented by G. Fiorentino and D. Bini [5] (`SlvAberthQQ` in the plots), that are open source tools with real solving capabilities. Other libraries such as [23], or EXACUS with

Leda [3], or RS [40], have not been tested, due to accessibility obstacles. Namely LEDA is a commercial software and RS can be used only through its MAPLE (v. 9.5) interface which do not have at the time of the experiments.

For experiments against these libraries and the package of Rioboo [39] in AXIOM, for degree ≤ 4 , the reader may refer to [17].

Our data⁴ are polynomials of degree $d \in \{3, \dots, 40\}$ and coefficient bit size $\tau \in \{10, 20, 30, 40, 50\}$ with various attributes. Namely D_1^{τ} denotes random polynomials with few real roots and D_2^{τ} random polynomials with multiple real roots. D_3^{τ} denotes polynomials with d (multiple) integer real roots and D_4^{τ} polynomials with d (multiple) rational real roots. D_5^{τ} denotes Mignotte polynomials, i.e. $x^d - 2(Kx - 1)^2$, D_6^{τ} polynomials that are the product of two Mignotte polynomials and D_7^{τ} Mignotte polynomials with multiple roots.

For reasons of space in the Appendix we present the average times over a run of 100 different polynomials only for D_1^{30} , D_1^{50} , D_2^{30} , D_2^{50} , D_5^{30} , D_5^{50} , D_7^{30} and D_7^{50} . The experiments performed on an Pentium (2.6 GHz), using g++ 3.4.4 (Suse 10). We have to emphasize that we do not consider experimentation as a competition, but rather as a starting point for improving existing implementations.

For polynomials with few, distinct and well separated real roots, this is the case for D_1 and D_2 , S_1 is clearly the worst choice, since the huge time for the computation of the sequence dominates the time for its evaluation. In such data sets, Bernstein or even approximate solvers are the solvers of choice. However when there are multiple roots, or when there are roots that are very close (D_5 , D_7) then the computation time of the Sturm-Habicht sequence is negligible (for the experiments that we performed). In such cases a combined solver is the solver of choice, since it isolates the well separated roots and also provides good initial intervals for the S_2 , if needed. Notice that neither **CORE**, nor **SlvAberthQQ** compute the multiplicities of the roots. For the latter special care should be taken so as to get the correct, if possible, results.

In conclusion, the most interesting solver is S_3 , which is a combination of solvers: it is fast on random instances and comparable to S_2 on all the other instances.

In some geometric problems, it is more important to have controlled approximation of the roots that to isolated them. This is the case in the following example where we want to draw a curve defined by an implicit equation. In this specific problem, the polynomial $f(x, y)$ is of degree 43 in each variable with coefficients of bit size 50 (see [7]). In order to get a picture of the implicit curve in the box $[a, b] \times [c, d]$, we solve the univariate polynomials $f(a + k \frac{(b-a)}{N}, y)_{k=0, \dots, N-1}$ ($N = 200$) and then exchange the role of x and y . The subdivision is stopped, when the precision of 10^{-4} is reached, without checking the existence and uniqueness of the roots in the computed intervals.

Two types of solvers have been tested:

- The first one **SlvBzStd<double>** is a direct implementation of the Bernstein solver with **double** arithmetic. It produces the left part of Fig. 7. We see

⁴ <http://www-sop.inria.fr/galaad/data/upol/>

that in some regions, the solver is more sensible to numerical errors, and behaves almost “like a random generator of points”.

- The second solver (`S1vBzBdg<QQ>`), similar to S_3 , uses exact (rational) arithmetic to convert the input polynomial to its Bernstein representation. Then it normalises the coefficients and rounds up and down the rational numbers to the closest `double` numbers⁵. Then the main subdivision loop is performed on double interval arithmetic, extending the sign count to this context. If all the interval coefficients contain 0, we recompute the representation of the initial polynomial (using exact rational arithmetic) and run again the rounding and subdivision steps with double arithmetic, until we get the required precision. This produces the right part of Fig. 7.

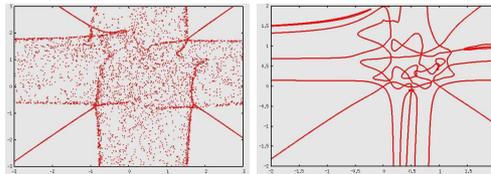


Fig. 2. Left: Approximation with doubles. Right: Approximation with Bernstein solver and intervals.

We see that the Bernstein solver based on interval arithmetic and using this symbolic-numeric strategy can be applied efficiently (even for input polynomials with large coefficient size) to geometric problems where (controlled) approximate results are sufficient. It exploits the performance of machine precision arithmetic for the main loop of the algorithm and the approximation properties of the Bernstein representation. Notice that the size of the problem is prohibitive for exact subdivision based solvers.

8 Current and Future Work

These experimentations show that combining symbolic and numeric techniques leads to very interesting performances. Along these lines, we plan to improve the existing implementation of solvers, which approximate with guarantees the roots of a polynomial with exact coefficients. The applications of Bernstein methods on polynomials with approximate coefficients is also under investigation. We are also extending our package in SYNAPS so that it can handle computations in an extension field.

Here are open questions: Is there any exact subdivision based solver with complexity $\tilde{O}_B(d^3\tau)$, similar to the numerical solvers?

⁵ For that purpose, one can use for instance the function `get_double` of MPFR (<http://www.mpfr.org/>) with correct rounding mode.

Is there any class of polynomials, with few real roots, such that the Bernstein solver performs $\mathcal{O}(d\tau)$ subdivisions steps but the Sturm solver performs only a constant number?

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Experimental results

