Vincent's theorem of 1836:

Overview and Future Research

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Outline of the talk

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• We will present Vincents theorem and review the various real root isolation methods derived from it (2 bisection methods and 1 Continued Fractions method).

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- We will concentrate on the Continued Fractions method (the fastest of them all) and show how it was recently speeded up by 40% over its initial implementation.
- We will indicate new directions for future research.

Descartes' rule of signs — for the open interval $]0,\infty[$

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Consider the polynomial

$$p(x) = a_n x^n + \cdots + a_1 x + a_0,$$

where $p(x) \in \mathbb{R}[x]$ and let var(p) represent the number of sign *changes* or *variations* (positive to negative and vice-versa) in the sequence of coefficients $a_n, a_{n-1}, \ldots, a_0$.

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Theorem

The number $\varrho_+(p)$ of real roots — multiplicities counted — of the polynomial $p(x) \in \mathbb{R}[x]$ in the open interval $]0, \infty[$ is bounded above by var(p); that is, we have $var(p) \ge \varrho_+(p)$.

Special Cases of Descartes' rule of signs

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▶ 0 sign variations \Leftrightarrow 0 positive roots

Special Cases of Descartes' rule of signs

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▶ 1 sign variation \Rightarrow 1 positive root

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Vincent's theorem of 1836 — the Continued Fractions version

If in a polynomial, p(x), of degree *n*, with rational coefficients and simple roots we perform sequentially replacements of the form

$$x \leftarrow \alpha_1 + \frac{1}{x}, x \leftarrow \alpha_2 + \frac{1}{x}, x \leftarrow \alpha_3 + \frac{1}{x}, \dots$$

where $\alpha_1 \ge 0$ is an arbitrary non negative integer and $\alpha_2, \alpha_3, \ldots$ are arbitrary positive integers, $\alpha_i > 0$, i > 1, then the resulting polynomial either has *no* sign variations or it has *one* sign variation. In the first case there are *no* positive roots whereas in the last case the equation has exactly one positive root, represented by the continued fraction

$$\alpha_1 + \frac{1}{\alpha_2 + \frac{1}{\alpha_3 + \frac{1}{\alpha_3}}}$$

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▶ was rediscovered by Akritas in 1976 and formed the subject of his Ph.D. Thesis (1978).

Vincent's theorem of 1836 and real root isolation

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Historical note on Uspensky (1883-1947)

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Historical note on Uspensky (1883-1947)

► Uspensky never read Vincent's original paper of 1836, which included several examples (applications). Instead,

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Historical note on Uspensky (1883-1947)

► Uspensky never read Vincent's original paper of 1836, which included several examples (applications). Instead,

▶ he read only a proof of Vincent's theorem in the Russian translation of Serrets (French) book on Algebra (1866).

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Vincent's Bisection theorem of 2000 — by Alesina and Galuzzi

Let f(z), be a real polynomial of degree n, which has only simple roots. It is possible to determine a positive quantity δ so that for every pair of positive real numbers a, b with $|b - a| < \delta$, every transformed polynomial of the form

$$\phi(z) = (1+z)^n f(\frac{a+bz}{1+z})$$

has exactly 0 or 1 variations. The second case is possible if and only if f(z) has a simple root within]a, b[.

Vincent's theorem of 1836 and real root isolation

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Proof of Vincent's theorem

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Proof of Vincent's theorem

▶ The proof by Alezina and Galuzzi (2000) is the most recent one; it uses Obreschkoff's theorem of 1920-23 which gives the necessary condition for a polynomial with one positive root to have one sign variation!

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► A similar proof was presented earlier by Ostrowski (1950), who rediscovered Obreschkoff's theorem 30 years later.

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Obreschkoff's Cone or Sector Theorem

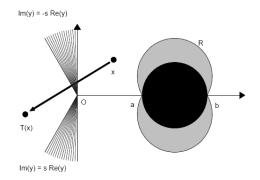
If a real polynomial has one positive simple root x_0 and all the other — possibly multiple — roots lie in the sector

$$S_{\sqrt{3}} = \{x = -\alpha + \imath \beta \mid \alpha > 0 \text{ and } \beta^2 \leq 3\alpha^2\}$$

then the sequence of its coefficients has exactly one sign variation.

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View of Obreschkoff's Cone and Circles



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Real root isolation using Vincent's theorem

To isolate the positive roots of a polynomial p(x), all we have to do is compute — for *each* root — the variables *a*, *b*, *c*, *d* of the corresponding Möbius transformation

$$M(x) = \frac{ax+b}{cx+d}$$

that leads to a transformed polynomial

$$f(x) = (cx+d)^n p(\frac{ax+b}{cx+d})$$

with one sign variation.

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Two different ways to isolate the real roots:

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Two different ways to isolate the real roots:

Crucial observation:

The variables a, b, c, d of a Möbius transformation $M(x) = \frac{ax+b}{cx+d}$ (in Vincent's theorem) leading to a transformed polynomial with one sign variation can be computed:

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▶ either by continued fractions leading to the continued fractions method developed by Vincent, Akritas and Strzeboński, which in the sequel will be called the VAS *continued fractions* method — to distinguish it from other continued fraction methods,

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▶ or, by bisections leading to (among others) the bisection method developed by (Vincent), Collins and Akritas, which in the sequel will be called the VCA *bisection* method — to distinguish it from Sturm's bisection method.

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Historical note on the two methods

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Historical note on the two methods

The VCA-bisection method was developed first — in 1976 by Collins and Akritas. Its fastest implementation was developed in 2004 by Rouillier and Zimmermann.

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Historical note on the two methods

The VCA-bisection method was developed first — in 1976 by Collins and Akritas. Its fastest implementation was developed in 2004 by Rouillier and Zimmermann.

The VAS-continued fractions method was developed later — in 1978 by Akritas and in 1993 by Akritas, Botcharov and Strzeboński. Its fastest implementation was developed in 2008 by Akritas, Strzeboński and Vigklas.

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Why bother with these methods and not use numeric ones?

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Why bother with these methods and not use numeric ones?

• Numeric methods cannot isolate just the positive roots! They isolate all the roots (real and complex).

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Why bother with these methods and not use numeric ones?

- Numeric methods cannot isolate just the positive roots! They isolate all the roots (real and complex).
- They can give wrong answers as the following example demonstrates.

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Using Mma 5 or 6	

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Consider the polynomial

$$f = 10^{999} (x - 1)^{50} - 1$$

which has just 2 positive roots \neq 1:

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► The same numeric method using 1020 digits of accuracy successfully finds the roots but it takes 18000 ms!

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► The same numeric method using 1020 digits of accuracy successfully finds the roots but it takes 18000 ms!

► The VAS Continued Fractions method, discussed below, isolates the two positive roots in 4 ms!

Vincent's theorem of 1836 and real root isolation

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Using Mma 7 (1/3 frames)

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Using Mma 7 (1/3 frames)

Consider again the same polynomial

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▶ On the other hand, the function RootIntervals, i.e. the VAS continued fractions method, isolates the two positive roots in $5 * 10^{-16}$ seconds ...

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```
ints = RootIntervals[f][[1]] // Timing
{5.60316×10<sup>-16</sup>, {{0, 1}, {1, 2}}}
```

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Using Mma 7 (3/3 frames)

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▶ ... and approximates them to 30 digits of accuracy in practically no time at all!

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Figure: Using the function FindRoot in Mma 7

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Therefore,	

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Therefore,	

• We do need the real root isolation methods derived from Vincents Theorem of 1836.

Therefore, ...

- We do need the real root isolation methods derived from Vincents Theorem of 1836.
- Especially so since the method developed by Vincent in 1836 has exponential computing time!

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Vincent's exponential method of 1836

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Vincent's exponential method of 1836

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Vincent's exponential method of 1836

- ... uses continued fractions and is described in his original paper of 1836!,
- ... was erroneously and unintentionally attributed by Uspensky to himself (1947).

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

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3 The continued fractions method derived from Vincent's theorem

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

The bisection methods derived from Vincents theorem differ in:

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

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• the termination test (criterion) used, and

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VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

The bisection methods derived from Vincents theorem differ in:

- the termination test (criterion) used, and
- the open interval,]a, b[, they bisect.

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

The first bisection method, VCA,

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- It uses Uspensky's termination test, explained below, and

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The first bisection method, VCA,

- ... was developed in 1976, in order to overcome the exponential behavior of Vincent's method.
- It uses Uspensky's termination test, explained below, and
- bisects the open interval]0, 1[.

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

The termination test used in VCA is named after Uspensky because ...

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VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

The termination test used in VCA is named after Uspensky because ...

- ... Uspensky was the one to use it as a test, in the (also exponential) isolation method that he erroneously attributed to himself.
- Uspensky developed the test since he was not aware of Budan's theorem of 1807, which was eclipsed by Fourier's theorem of 1819.

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

Budan's theorem of 1807 — to be found in Vincent's paper of 1836 and Akritas' work

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If in an equation p(x) = 0 we make two transformations, x = x' + p and x = x'' + q, where p and q are real numbers such that p < q, then:

Budan's theorem of 1807 — to be found in Vincent's paper of 1836 and Akritas' work

If in an equation p(x) = 0 we make two transformations, x = x' + p and x = x'' + q, where p and q are real numbers such that p < q, then:

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▶ the number of real roots of the equation p(x) = 0, located between p and q, can never be more than the number of sign variations lost in passing from the transformed equation in x' = x - p to the transformed equation in x'' = x - q;

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▶ when the first number is less than the second, the difference is always an even number.

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

Vincent vs Uspensky (1/2)

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

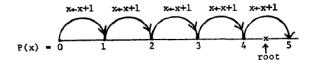
Vincent vs Uspensky (1/2)

▶ Vincent knew Budan's theorem and so he applied the transformation $x \leftarrow x + 1$ repeatedly until he detected a loss of sign variations in the transformed polynomial; in that case he knew there are real roots in the open interval]0, 1[. So he proceeded as indicated in the following figure:

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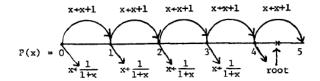
Vincent vs Uspensky (2/2)

▶ On the other hand, Uspensky did not know Budan's theorem and so he had to invent a test to check whether there are real roots in the open interval]0,1[. That test is the transformation $x \leftarrow \frac{1}{1+x}$ and it was performed before Uspensky proceeded with the transformation $x \leftarrow x + 1$. So he proceeded as indicated in the following figure:

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Uspenskys termination test, for the interval]a, b[=]0,1[

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Uspenskys termination test, for the interval]a, b[=]0,1[

The number $\varrho_{01}(p)$ of real roots in the open interval]0,1[— multiplicities counted — of the polynomial $p(x) \in \mathbb{R}[x]$ is bounded above by the number of sign variations $var_{01}(p)$, where

$$\mathit{var}_{01}(p) = \mathit{var}((x+1)^{\mathit{deg}(p)}p(rac{1}{x+1}))$$

and we have $var_{01}(p) \ge \varrho_{01}(p)$.

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VCA, 1976: The *original* version of the 1st bisection method derived from Vincent's theorem

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Input: The square-free polynomial p(ub ⋅ x) ∈ Z[x], p(0) ≠ 0, and the open interval]a, b[=]0, ub[, where ub is an upper bound on the values of the positive roots of p(x).
Output: A list of isolating intervals of the positive roots of p(x)

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Output: A list of isolating intervals of the positive roots of p(x)

- 1 var \leftarrow the number of sign changes of $(x+1)^{deg(p)}p(\frac{1}{x+1})$;
- 2 if var = 0 then RETURN \emptyset ;
- 3 if var = 1 then RETURN {]a, b[};

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$$\begin{array}{lll} & var \longleftarrow \text{ the number of sign changes of } (x+1)^{deg(p)}p(\frac{1}{x+1});\\ 2 & \text{if } var = 0 \text{ then RETURN } \emptyset;\\ 3 & \text{if } var = 1 \text{ then RETURN } \{]a, b[\};\\ 4 & p_{0\frac{1}{2}} \longleftarrow 2^{deg(p)}p(\frac{x}{2}) \ // \text{ Look for real roots in }]0, \frac{1}{2}[;\\ 5 & m \longleftarrow \frac{a+b}{2} \ // \text{ Is } \frac{1}{2} \text{ a root? };\\ 6 & p_{\frac{1}{2}1} \longleftarrow 2^{deg(p)}p(\frac{x+1}{2}) \ // \text{ Look for real roots in }]\frac{1}{2}, 1[;] \end{array}$$

VCA, 1976: The *original* version of the 1st bisection method derived from Vincent's theorem

Input: The square-free polynomial $p(ub \cdot x) \in \mathbb{Z}[x], p(0) \neq 0$, and the open interval]a, b[=]0, ub[, where ub is an upper bound on the values of the positive roots of p(x).

Output: A list of isolating intervals of the positive roots of p(x)

```
 \begin{array}{l|ll} & var \longleftarrow \text{ the number of sign changes of } (x+1)^{deg(p)} p(\frac{1}{x+1}); \\ 2 & \text{ if } var = 0 \text{ then RETURN } \emptyset; \\ 3 & \text{ if } var = 1 \text{ then RETURN } \{]a, b[\}; \\ 4 & \rho_{0\frac{1}{2}} \longleftarrow 2^{deg(p)} p(\frac{x}{2}) // \text{ Look for real roots in } ]0, \frac{1}{2}[; \\ 5 & m \longleftarrow \frac{a+b}{2} // \text{ Is } \frac{1}{2} \text{ a root? }; \\ 6 & \rho_{\frac{1}{2}1} \longleftarrow 2^{deg(p)} p(\frac{x+1}{2}) // \text{ Look for real roots in } ]\frac{1}{2}, 1[; \\ 7 & \text{ if } p(\frac{1}{2}) \neq 0 \text{ then} \\ 8 & | & \text{RETURN VCA}(p_{0\frac{1}{2}}, ]a, m[) \cup \text{VCA}(p_{\frac{1}{2}1}, ]m, b[) \\ 9 & \text{ else} \\ 10 & | & \text{RETURN VCA}(p_{0\frac{1}{2}}, ]a, m[) \cup \{[m, m]\} \cup \text{VCA}(p_{\frac{1}{2}1}, ]m, b[) \\ 11 & \text{ end} \\ \end{array}
```

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

VCA has been implemented in maple — version 11 shown below

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

VCA has been implemented in maple — version 11 shown below

— and it takes 170 seconds to isolate the roots of Mignotte's polynomial of degree 300!

VCA has been implemented in maple — version 11 shown below

— and it takes 170 seconds to isolate the roots of Mignotte's polynomial of degree 300!

> with(RootFinding) :
>
$$f := x^{300} - 2(5x-1)^2$$
;
 $f := x^{300} - 2(5x-1)^2$
> $st := time() : Isolate(f, digits = 250) : time() - st;$
170.431

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem



VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

VCA's other names

• For the period 1976-1986 VCA was called *Uspensky's* modified method.

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

- For the period 1976-1986 VCA was called *Uspensky's* modified method.
- In 1986 Akritas wrote the paper There is no Uspensky's method...

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

- For the period 1976-1986 VCA was called *Uspensky's modified method*.
- In 1986 Akritas wrote the paper *There is no Uspensky's method*...
- ... after which the method was called *Descartes*' or *Collins-Akritas*.

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

- For the period 1976-1986 VCA was called *Uspensky's modified method*.
- In 1986 Akritas wrote the paper There is no Uspensky's method...
- ... after which the method was called *Descartes*' or *Collins-Akritas*.
- In 2007 Akritas wrote the paper *There is no Descartes' method*. By that time the correct name VCA had already been coined by Francois Boulier at the University of Lille, France.

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

What's in a name?

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

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▶ It makes clear the existing relation with Vincent's theorem.

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

What's in a name?

▶ It makes clear the existing relation with Vincent's theorem.

▶ We can use the results by Alesina & Galuzzi (2000) to prove the termination and estimate the computing time of all methods derived from Vincents theorem.

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

The computing time of the VCA bisection method

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

The computing time of the VCA bisection method

. . . is

 $O(n^4\tau^2)$.

where *n* is the degree of the polynomial, and τ bounds the coefficient bitsize.

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

The computing time of the VCA bisection method

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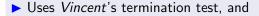
► The fastest implementation of the VCA method, REL, was developed by Rouillier & Zimmermann in 2004!

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

B, 2000: The second bisection method derived from Vincent's theorem

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

B, 2000: The second bisection method derived from Vincent's theorem



VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

B, 2000: The second bisection method derived from Vincent's theorem

Uses Vincent's termination test, and

▶ bisects the interval]a, b[=]0, ub[, where ub is an upper bound on the values of the positive roots.

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

Vincent's termination test for the interval]a, b[=]0, ub[

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

Vincent's termination test for the interval]a, b[=]0, ub[

If $a \ge 0$ and b > a then the number $\varrho_{ab}(p)$ of real roots in the open interval]a, b[— multiplicities counted — of the polynomial $p(x) \in \mathbb{R}[x]$ is bounded above by the number of sign variations $var_{ab}(p)$, where

$$var_{ab}(p) = var((1+x)^{deg(p)}p(rac{a+bx}{1+x}))$$

and we have $var_{ab}(p) = var_{ba}(p) \ge \varrho_{ab}(p)$.

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

B, 2000: The second bisection method derived from Vincent's theorem

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

B, 2000: The second bisection method derived from Vincent's theorem

Input: The square-free polynomial p(x) ∈ Z[x], p(0) ≠ 0, and the open interval]a, b[=]0, ub[, where ub is an upper bound on the values of the positive roots of p(x).
Output: A list of isolating intervals of the positive roots of p(x)

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

B, 2000: The second bisection method derived from Vincent's theorem

Input: The square-free polynomial $p(x) \in \mathbb{Z}[x], p(0) \neq 0$, and the open interval]a, b[=]0, ub[, where ub is an upper bound on the values of the positive roots of p(x).

Output: A list of isolating intervals of the positive roots of p(x)

var ← the number of sign changes of (1 + x)^{deg(p)}p(^{a+bx}/_{1+x});
 if var = 0 then RETURN Ø;
 if var = 1 then RETURN {]a, b[};

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

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1 var \leftarrow the number of sign changes of $(1 + x)^{deg(p)} p(\frac{a+bx}{1+x})$; 2 if var = 0 then RETURN \emptyset ; 3 if var = 1 then RETURN {]a, b[}; 4 $m \leftarrow \frac{a+b}{2}$ // Subdivide the interval]a, b[in two equal parts ;

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

B, 2000: The second bisection method derived from Vincent's theorem

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```
1 var \leftarrow the number of sign changes of (1 + x)^{deg(p)}p(\frac{a+bx}{1+x});

2 if var = 0 then RETURN \emptyset;

3 if var = 1 then RETURN \{]a, b[\};

4 m \leftarrow \frac{a+b}{2} // Subdivide the interval ]a, b[ in two equal parts ;

5 if p(m) \neq 0 then

6 | RETURN B(p, ]a, m[) \cup B(p, ]m, b[)

7 else

8 | RETURN B(p, ]a, m[) \cup \{[m, m]\} \cup B(p, ]m, b[)

9 end
```

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

Comparison of the two bisection methods

VCA, the first bisection method derived from Vincent's theorem B, the second bisection method derived from Vincent's theorem

Comparison of the two bisection methods

VCA, the method using the simpler termination test, i.e. Uspensky's test, is faster than B, which is using Vincent's more complex termination test!

Improving the performance of VAS Speeding up VAS with linear complexity bounds Speeding up VAS with quadratic complexity bounds

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Improving the performance of VAS Speeding up VAS with linear complexity bounds Speeding up VAS with quadratic complexity bounds

The continued fractions method derived from Vincent's theorem ...

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The continued fractions method derived from Vincent's theorem . . .

... uses Descartes rule of signs as the termination test, and

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The continued fractions method derived from Vincent's theorem ...

... uses Descartes rule of signs as the termination test, and

▶ ... relies, heavily, on the repeated estimation of lower bounds on the values of the positive roots of polynomials.

Improving the performance of VAS Speeding up VAS with linear complexity bounds Speeding up VAS with quadratic complexity bounds

VAS, 1978:

Alkiviadis G. Akritas PCA 2009, St. Petersburg, Russia

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VAS, 1978:

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VAS, 1978:

- 1 var \leftarrow the number of sign changes of p(x);
- 2 if var = 0 then RETURN \emptyset ;
- 3 if var = 1 then RETURN {]a, b[} // a = min(M(0), M(∞)), b = max(M(0), M(∞));

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- 1 var \leftarrow the number of sign changes of p(x);
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- 5 if $\ell b > 1$ then $\{p \leftarrow p(x + \ell b), M \leftarrow M(x + \ell b)\};$

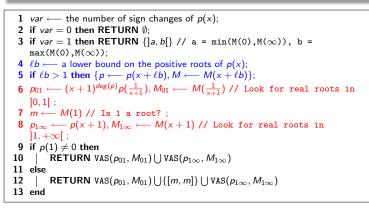
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- $\begin{array}{l} \mathbf{6} \quad p_{01} \longleftarrow (x+1)^{deg(p)} p(\frac{1}{x+1}), M_{01} \longleftarrow M(\frac{1}{x+1}) // \text{ Look for real roots in} \\]0,1]; \end{array}$
- 7 $m \leftarrow M(1) // \text{ Is 1 a root? ;}$

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VAS, 1978:



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Comments on the VAS-continued fractions real root isolation method:

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Strzebonski's contribution is omitted for simplicity.

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Comments on the VAS-continued fractions real root isolation method:

Strzebonski's contribution is omitted for simplicity.

▶ Without steps 4 and 5 it is simply Vincent's original exponential method.

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VAS has been implemented in *Mathematica* — version 7 shown below

Improving the performance of VAS Speeding up VAS with linear complexity bounds Speeding up VAS with quadratic complexity bounds

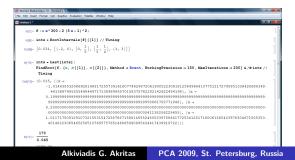
VAS has been implemented in *Mathematica* — version 7 shown below

► — and it takes 0.046 seconds to isolate and approximate the roots of Mignotte's polynomial of degree 300!

Improving the performance of VAS Speeding up VAS with linear complexity bounds Speeding up VAS with quadratic complexity bounds

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Improving the performance of VAS Speeding up VAS with linear complexity bounds Speeding up VAS with quadratic complexity bounds

Over the past 30 years ...

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► VAS has been using Cauchy's bound on the values of the positive roots,

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Over the past 30 years ...

► VAS has been using Cauchy's bound on the values of the positive roots,

▶ For random polys, VAS has been several thousand times faster than the VCA bisection method — even up to 50000 times faster than VCA, for Mignotte polys.

Improving the performance of VAS Speeding up VAS with linear complexity bounds Speeding up VAS with quadratic complexity bounds

Over the past 30 years ...

► VAS has been using Cauchy's bound on the values of the positive roots,

► For random polys, VAS has been several thousand times faster than the VCA bisection method — even up to 50000 times faster than VCA, for Mignotte polys.

▶ Only in the case of very many, (> 50), very large roots, (10^{100}) , had VAS been up to 4 times slower than VCA.

Improving the performance of VAS Speeding up VAS with linear complexity bounds Speeding up VAS with quadratic complexity bounds

Computing time of the VAS continued fractions method

Computing time of the VAS continued fractions method

► Using a plausible hypothesis and the fast translation algorithm by von zu Gathen, the computing time of VAS is

 $O(n^4\tau^2),$

where *n* is the degree of the polynomial, and τ bounds the coefficient bitsize. (Akritas 1978, Tsigaridas-Emiris 2005)

Computing time of the VAS continued fractions method

► Using a plausible hypothesis and the fast translation algorithm by von zu Gathen, the computing time of VAS is

 $O(n^4\tau^2),$

where *n* is the degree of the polynomial, and τ bounds the coefficient bitsize. (Akritas 1978, Tsigaridas-Emiris 2005)

Without any hypotheses the computing time of VAS is

 $O(n^8 \tau^3)$

(Sharma 2007). However, this bound does not match the performance of VAS.

Improving the performance of VAS Speeding up VAS with linear complexity bounds Speeding up VAS with quadratic complexity bounds

To improve the performance of the VAS method even further ...

Improving the performance of VAS Speeding up VAS with linear complexity bounds Speeding up VAS with quadratic complexity bounds

To improve the performance of the VAS method even further ...

▶ ... new bounds on the values of the positive roots of polynomials were needed.

Improving the performance of VAS Speeding up VAS with linear complexity bounds Speeding up VAS with quadratic complexity bounds

To improve the performance of the VAS method even further ...

... new bounds on the values of the positive roots of polynomials were needed.

► To understand the nature of these bounds we used Doru \$tefănescu's inspirational work!

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Stefănescu's theorem (2005): matching a positive-coefficient term with a negative-coefficient one — when the number of sign variations is even

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Ștefănescu's theorem (2005): matching a positive-coefficient term with a negative-coefficient one — when the number of sign variations is even

Let $p(x) \in \mathbb{R}[x]$ be such that the number of sign variations in the sequence of its coefficients is even. If

$$p(x) = c_1 x^{d_1} - b_1 x^{m_1} + c_2 x^{d_2} - b_2 x^{m_2} + \ldots + c_k x^{d_k} - b_k x^{m_k} + g(x),$$

with $g(x) \in \mathbb{R}_+[x], c_i > 0, b_i > 0, d_i > m_i > d_{i+1}$ for all i, then the number

$$ub(p) = \max\left\{\left(\frac{b_1}{c_1}\right)^{1/(d_1-m_1)}, \ldots, \left(\frac{b_k}{c_k}\right)^{1/(d_k-m_k)}\right\}$$

is an upper bound on the values of the positive roots of the polynomial p for any choice of c_1, \ldots, c_k .

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Remarks on Ștefănescu's theorem:

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Remarks on Ștefănescu's theorem:

▶ It does not work if the number of sign variations is not even.

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Remarks on Ștefănescu's theorem:

It does not work if the number of sign variations is not even.

► It also fails to work if a positive-coefficient term is followed by two negative-coefficient terms.

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Remarks on Ștefănescu's theorem:

It does not work if the number of sign variations is not even.

► It also fails to work if a positive-coefficient term is followed by two negative-coefficient terms.

► The following theorem by Akritas, Strzeboński and Vigklas generalizes Ștefănescu's theorem and works in all cases. This is achieved by breaking up a positive-coefficient term into several parts to be matched with the corresponding negative-coefficient terms.

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Generalization of Ștefănescu's theorem by Akritas, Strzeboński and Vigklas, 2006 (1/2)

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Generalization of Ștefănescu's theorem by Akritas, Strzeboński and Vigklas, 2006 (1/2)

Assumptions

Let p(x)

$$p(x) = \alpha_n x^n + \alpha_{n-1} x^{n-1} + \ldots + \alpha_0, \quad (\alpha_n > 0)$$

be a polynomial with real coefficients and let d(p) and t(p) denote the degree and the number of its terms, respectively. Moreover, assume that p(x) can be written as

$$p(x) = q_1(x) - q_2(x) + q_3(x) - q_4(x) + \ldots + q_{2m-1}(x) - q_{2m}(x) + g(x),$$

where all the polynomials $q_i(x)$, i = 1, 2, ..., 2m and g(x) have only positive coefficients. In addition, assume that for i = 1, 2, ..., m we have

$$q_{2i-1}(x) = c_{2i-1,1}x^{e_{2i-1,1}} + \ldots + c_{2i-1,t(q_{2i-1})}x^{e_{2i-1,t(q_{2i-1})}}$$

and

$$q_{2i}(x) = b_{2i,1}x^{e_{2i,1}} + \ldots + b_{2i,t(q_{2i})}x^{e_{2i,t(q_{2i})}},$$

where $e_{2i-1,1} = d(q_{2i-1})$ and $e_{2i,1} = d(q_{2i})$ and the exponent of each term in $q_{2i-1}(x)$ is greater than the exponent of each term in $q_{2i}(x)$.

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Generalization of Ștefănescu's theorem by Akritas, Strzeboński and Vigklas, 2006 (2/2)

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Generalization of Ștefănescu's theorem by Akritas, Strzeboński and Vigklas, 2006 (2/2)

Theorem

If for all indices $i = 1, 2, \ldots, m$, we have

$$t(q_{2i-1})\geq t(q_{2i}),$$

then an upper bound of the values of the positive roots of p(x) is given by

$$ub = \max_{\{i=1,2,\dots,m\}} \left\{ \left(\frac{b_{2i,1}}{C_{2i-1,1}} \right)^{\frac{1}{e_{2i-1,1}-e_{2i,1}}}, \dots, \left(\frac{b_{2i,t(q_{2i})}}{C_{2i-1,t(q_{2i})}} \right)^{\frac{1}{e_{2i-1,t(q_{2i})}-e_{2i,t(q_{2i})}}} \right\},$$

for any permutation of the positive coefficients $c_{2i-1,j}$, $j = 1, 2, ..., t(q_{2i-1})$. Otherwise, for each of the indices *i* for which we have

$$t(q_{2i-1}) < t(q_{2i}),$$

we break up one of the coefficients of $q_{2i-1}(x)$ into $t(q_{2i}) - t(q_{2i-1}) + 1$ parts, so that now $t(q_{2i}) = t(q_{2i-1})$ and apply the same formula given above.

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Remarks on the theorem by Akritas-Strzeboński and Vigklas, 2006:

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Remarks on the theorem by Akritas-Strzeboński and Vigklas, 2006:

► It is a general theorem from which almost all methods for computing positive bounds on the values of positive roots are derived!

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Remarks on the theorem by Akritas-Strzeboński and Vigklas, 2006:

► It is a general theorem from which almost all methods for computing positive bounds on the values of positive roots are derived!

► This generality is achieved by breaking up and pairing up — in various ways — unmatched positive-coefficient terms with negative-coefficient ones of lower order!

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Remarks on the theorem by Akritas-Strzeboński and Vigklas, 2006:

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► This generality is achieved by breaking up and pairing up — in various ways — unmatched positive-coefficient terms with negative-coefficient ones of lower order!

On terminology

► For simplicity, in the sequel we will simply refer to positive coefficients being matched with negative ones of lower order terms!

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On upper bounds on the values of the positive roots of polynomials:

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In general, bounds in the literature are of linear complexity!

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On upper bounds on the values of the positive roots of polynomials:

In general, bounds in the literature are of linear complexity!

► That is, each negative coefficient of the polynomial is paired up with only one of the preceding (unmatched) positive coefficients and the maximum of all the computed radicals is taken as the estimate of the bound.

Improving the performance of VAS Speeding up VAS with linear complexity bounds Speeding up VAS with quadratic complexity bounds

On upper bounds on the values of the positive roots of polynomials:

In general, bounds in the literature are of linear complexity!

► That is, each negative coefficient of the polynomial is paired up with only one of the preceding (unmatched) positive coefficients and the maximum of all the computed radicals is taken as the estimate of the bound.

▶ We present four linear complexity bounds. Of those, the last two were developed by Akritas, Strzeboński and Vigklas in 2007.

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A: Cauchy's "leading coefficient" bound, (C)

Improving the performance of VAS Speeding up VAS with linear complexity bounds Speeding up VAS with quadratic complexity bounds

A: Cauchy's "leading coefficient" bound, (C)

For the polynomial $p(x) \in \mathbb{R}[x]$

$$p(x) = \alpha_n x^n + \alpha_{n-1} x^{n-1} + \ldots + \alpha_0, \quad (\alpha_n > 0),$$

with λ negative coefficients, Cauchy's method first breaks up its leading coefficient, α_n , into λ equal parts and then pairs up each part with the first unmatched negative coefficient.

That is, we have:

$$ub_{\mathcal{C}} = \max_{\{1 \le k \le n: \alpha_{n-k} < 0\}} \sqrt[k]{-\frac{\alpha_{n-k}}{\frac{\alpha_n}{\lambda}}}.$$

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B: Kioustelidis' "leading coefficient" bound, (K)

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For the polynomial $p(x) \in \mathbb{R}[x]$

$$p(x) = \alpha_n x^n + \alpha_{n-1} x^{n-1} + \ldots + \alpha_0, \quad (\alpha_n > 0),$$

Kioustelidis' method matches the coefficient $-\alpha_{n-k}$ of the term $-\alpha_{n-k}x^{n-k}$ in p(x) with $\frac{\alpha_n}{2^k}$, the leading coefficient divided by 2^k .

That is, we have

$$ub_{\mathcal{K}} = \max_{\{1 \le k \le n: \alpha_{n-k} < 0\}} \sqrt[k]{-\frac{\alpha_{n-k}}{\frac{\alpha_n}{2^k}}}.$$

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That is, we have

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► Kioustelidis' method differs from that by Cauchy only in that the leading coefficient is now broken up in unequal parts — by dividing it with different powers of 2.

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C: "First– λ " bound, (FL)

Alkiviadis G. Akritas PCA 2009, St. Petersburg, Russia

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C: "First– λ " bound, (FL)

▶ For the polynomial $p(x) \in \mathbb{R}[x]$

$$p(x) = q_1(x) - q_2(x) + q_3(x) - q_4(x) + \ldots + q_{2m-1}(x) - q_{2m}(x) + g(x),$$

with λ negative coefficients we:

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C: "First– λ " bound, (FL)

For the polynomial
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with λ negative coefficients we:

▶ first take care of all cases for which $t(q_{2i}) > t(q_{2i-1})$, by breaking up the last coefficient $c_{2i-1,t(q_{2i})}$, of $q_{2i-1}(x)$, into $t(q_{2i}) - t(q_{2i-1}) + 1$ equal parts, and

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with λ negative coefficients we:

▶ first take care of all cases for which $t(q_{2i}) > t(q_{2i-1})$, by breaking up the last coefficient $c_{2i-1,t(q_{2i})}$, of $q_{2i-1}(x)$, into $t(q_{2i}) - t(q_{2i-1}) + 1$ equal parts, and

▶ then pair each of the first λ positive coefficients of p(x), encountered as we move in non-increasing order of exponents, with the first unmatched negative coefficient.

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D: "Local-Max" bound, (LM)

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D: "Local-Max" bound, (LM)

For the polynomial
$$p(x) \in \mathbb{R}[x]$$

$$\rho(x) = \alpha_n x^n + \alpha_{n-1} x^{n-1} + \ldots + \alpha_0, \quad (\alpha_n > 0),$$

the coefficient $-\alpha_k$ of the term $-\alpha_k x^k$ in p(x) is paired with the coefficient $\frac{\alpha_m}{2^t}$, of the term $\alpha_m x^m$, where α_m is the largest positive coefficient with $n \ge m > k$ and t indicates the number of times the coefficient α_m has been used.

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Example

Consider the polynomial

$$x^3 + 10^{100}x^2 - 10^{100}x - 1,$$

which has one sign variation and, hence, one positive root equal to 1

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Example

Consider the polynomial

$$x^3 + 10^{100}x^2 - 10^{100}x - 1,$$

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With C, Cauchy's bound, we pair the terms:

•
$$\left\{\frac{x^3}{2}, -10^{100}x\right\}$$
 and $\left\{\frac{x^3}{2}, -1\right\}$,

and taking the maximum of the radicals we obtain a bound estimate of $1.41421 \ast 10^{50}.$

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Example

Consider the polynomial

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With K, Kioustelidis' bound, we pair the terms:

•
$$\left\{\frac{x^3}{2^2}, -10^{100}x\right\}$$
 and $\left\{\frac{x^3}{2^3}, -1\right\}$,

and taking the maximum of the radicals we obtain a bound estimate of $2 * 10^{50}$.

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Example

Consider the polynomial

$$x^3 + 10^{100}x^2 - 10^{100}x - 1,$$

which has one sign variation and, hence, one positive root equal to 1

With **FL**, the "First- λ " bound, we pair the terms:

•
$$\{x^3, -10^{100}x\}$$
 and $\{10^{100}x^2, -1\}$,

and taking the maximum of the radicals we obtain a bound estimate of $10^{50}. \label{eq:stability}$

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Example

Consider the polynomial

$$x^3 + 10^{100}x^2 - 10^{100}x - 1,$$

which has one sign variation and, hence, one positive root equal to 1

With LM, the "Local Max" bound, we pair the terms:

•
$$\left\{\frac{10^{100}x^2}{2}, -10^{100}x\right\}$$
 and $\left\{\frac{10^{100}x^2}{2^2}, -1\right\}$,

and taking the maximum of the radicals we obtain a bound estimate of $\ensuremath{2!}$

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Comparison of the 4 linear complexity bounds

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Empirical results have indicated the following:

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Comparison of the 4 linear complexity bounds

Empirical results have indicated the following:

► Kioustelidis' bound is, in general, better (or much better) than Cauchy's; this happens because the former breaks up the leading coefficient in unequal parts, whereas the latter breaks it up in equal parts.

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Comparison of the 4 linear complexity bounds

Empirical results have indicated the following:

► Kioustelidis' bound is, in general, better (or much better) than Cauchy's; this happens because the former breaks up the leading coefficient in unequal parts, whereas the latter breaks it up in equal parts.

• Our *First-* λ bound, as the name indicates, uses additional coefficients and, therefore, it is not surprising that it is, in general, better (or much better) than both previous bounds. In the few cases where Kioustelidis' bound is better than *first-* λ , our *Local-Max* bound takes again the lead.

Conclusions

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► Of the four linear complexity bounds there does not exist one that always computes best estimate values.

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Conclusions

► Of the four linear complexity bounds there does not exist one that always computes best estimate values.

► Therefore, to improve the performance of VAS we used two methods — *First-* λ or *FL* and *Local-Max* or *LM* — and took their minimum as the estimated value of the bound. That is, we used min(FL, LM).

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Conclusions

► Of the four linear complexity bounds there does not exist one that always computes best estimate values.

► Therefore, to improve the performance of VAS we used two methods — *First-* λ or *FL* and *Local-Max* or *LM* — and took their minimum as the estimated value of the bound. That is, we used min(FL, LM).

▶ Using *min*(*FL*, *LM*), instead of using Cauchy's bound, the VAS continued fractions method was speeded up, on average, by 15% and became always faster than the VCA bisections method.

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The only case where VAS was slower than VCA: Products of terms x - r, with random integer r.

The only case where VAS was slower than VCA: Products of terms x - r, with random integer r.

Table: We compare the timings in seconds (s) for: (a) VAS_old, i.e. VAS using Cauchy's rule, (b) VAS_new, i.e. VAS using the new rule min(FL + LM), and (c) VCA_rel. The tests were run on a laptop computer with 1.8 Ghz Pentium M processor, running a Linux virtual machine with 1.78 GB of RAM.

Roots	Deg	VAS_old Time(s)	VAS_new Time(s)	VCA_rel
(bit length)		Average (Min/Max)	Average (Min/Max)	Average (Min/Max)
10	100	0.314 (0.248/0.392)	0.253 (0.228/0.280)	0.346 (0.308/0.384)
10	200	1.74 (1.42/2.33)	1.51 (1.34/1.66)	3.90 (3.72/4.05)
10	500	17.6 (16.9/18/7)	17.4 (16.3/18.1)	129 (122/140)
1000	20	0.066 (0.040/0.084)	0.031 (0.024/0.040)	0.038 (0.028/0.044)
1000	50	1.96 (1.45/2.44)	0.633 (0.512/0.840)	1.03 (0.916/1.27)
1000	100	52.3 (36.7/81.3)	12.7 (11.3/14.6)	17.2 (16.1/18.7)

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Quadratic complexity bounds (1/2)

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► To further improve the performance of the VAS continued fractions method we decided to use quadratic complexity bounds.

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Quadratic complexity bounds (1/2)

► To further improve the performance of the VAS continued fractions method we decided to use quadratic complexity bounds.

Justification:

► Their improved estimates should compensate for the extra time needed to compute these bounds.

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Quadratic complexity bounds (2/2)

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Quadratic complexity bounds (2/2)

Main idea:

► Each negative coefficient of the polynomial is paired with all the preceding positive coefficients and the minimum of the computed values is associated with this coefficient. The maximum of all those minimums is taken as the estimate of the bound.

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Quadratic complexity bounds (2/2)

Main idea:

► Each negative coefficient of the polynomial is paired with all the preceding positive coefficients and the minimum of the computed values is associated with this coefficient. The maximum of all those minimums is taken as the estimate of the bound.

► We will present four quadratic complexity bounds derived from the corresponding four linear complexity bounds discussed before. Of those four, one was developed by Hong in 1998, whereas the other three — including the best and fastest — were developed by Akritas, Argyris, Strzeboński and Vigklas in 2008.

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A: Cauchy's quadratic complexity bound, (CQ)

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A: Cauchy's quadratic complexity bound, (CQ)

For the polynomial $p(x) \in \mathbb{R}[x]$

$$p(x) = \alpha_n x^n + \alpha_{n-1} x^{n-1} + \ldots + \alpha_0, \quad (\alpha_n > 0),$$

each negative coefficient $a_i < 0$ is "paired" with each one of the preceding positive coefficients a_j divided by λ_i — that is, each positive coefficient a_j is "broken up" into equal parts, as is done with *just* the leading coefficient in Cauchy's bound; λ_i is the number of negative coefficients to the right of, and including, a_i — and the minimum is taken over all j; subsequently, the maximum is taken over all i.

That is, we have:

$$ub_{CQ} = \max_{\substack{\{a_j < 0\} \ \{a_j > 0: j > i\}}} \min_{j \to i} \sqrt[j-\frac{a_j}{\frac{a_j}{\lambda_j}}.$$

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B: Kioustelidis' quadratic complexity bound, (KQ),

Hong 1998

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$$p(x) = \alpha_n x^n + \alpha_{n-1} x^{n-1} + \ldots + \alpha_0, \quad (\alpha_n > 0),$$

each negative coefficient $a_i < 0$ is "paired" with each one of the preceding positive coefficients a_j divided by 2^{j-i} — that is, each positive coefficient a_j is "broken up" into unequal parts, as is done with *just* the leading coefficient in Kioustelidis' linear bound — and the minimum is taken over all j; subsequently, the maximum is taken over all i.

That is, we have:

$$ub_{KQ} = \max_{\{a_i < 0\}} \min_{\{a_j > 0: j > i\}} \sqrt{-\frac{a_i}{\frac{a_j}{2^{j-i}}}}$$

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C: "First- λ " quadratic complexity bound, (FLQ)

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with λ negative coefficients we do the following:

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▶ For the polynomial $p(x) \in \mathbb{R}[x]$

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with λ negative coefficients we do the following:

▶ First we take care of all cases for which $t(q_{2\ell}) > t(q_{2\ell-1})$, by breaking up the last coefficient $c_{2\ell-1,t(q_{2\ell})}$, of $q_{2\ell-1}(x)$, into $d_{2\ell-1,t(q_{2\ell})} = t(q_{2\ell}) - t(q_{2\ell-1}) + 1$ equal parts.

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For the polynomial
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▶ Then each negative coefficient $a_i < 0$ is "paired" with each one of the preceding $min(i, \lambda)$ positive coefficients a_j divided by d_j and the minimum is taken over all j; subsequently, the maximum is taken over all i.

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Improving the performance of VAS Speeding up VAS with linear complexity bounds Speeding up VAS with quadratic complexity bounds

. . . .

That is, we have:

$$ub_{FLQ} = \max_{\substack{\{a_i < 0\} \ \{a_j > 0: j > min(i,\lambda): u(j) \neq 0\}}} \min_{\substack{j = i \\ d_j}} \sqrt{-\frac{a_i}{\frac{a_j}{d_j}}},$$
where . . .

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That is, we have:

$$ub_{FLQ} = \max_{\{a_i < 0\}} \min_{\{a_j > 0: j > \min(i,\lambda): u(j) \neq 0\}} \sqrt[j-i]{-\frac{a_i}{a_j}},$$
where ...

► d_j indicates the number of equal parts into which each of the preceding $min(i, \lambda)$ positive coefficients a_j is "broken up". The value of d_j is initially set to 1, for each j, and it changes only if the positive coefficient a_j is broken up into equal parts.

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That is, we have:

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where ...

► d_j indicates the number of equal parts into which each of the preceding $min(i, \lambda)$ positive coefficients a_j is "broken up". The value of d_j is initially set to 1, for each j, and it changes only if the positive coefficient a_j is broken up into equal parts.

▶ u(j) indicates the number of times a_j can be used to calculate the minimum. The value of u(j) is originally set equal to d_j and it decreases each time a_j is used in the computation of the minimum.

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D: "Local Max" quadratic complexity bound, (LMQ)

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D: "Local Max" quadratic complexity bound, (LMQ)

For the polynomial $p(x) \in \mathbb{R}[x]$

$$p(x) = \alpha_n x^n + \alpha_{n-1} x^{n-1} + \ldots + \alpha_0, \quad (\alpha_n > 0),$$

each negative coefficient $a_i < 0$ is "paired" with each one of the preceding positive coefficients a_j divided by 2^{t_j} — where t_j is initially set to 1 and is incremented each time the positive coefficient a_j is used — and the minimum is taken over all j; subsequently, the maximum is taken over all i.

That is, we have:

$$ub_{LMQ} = \max_{\{a_i < 0\}} \min_{\{a_j > 0: j > i\}} \sqrt[j-i]{-\frac{a_i}{\frac{a_j}{2^{t_j}}}},$$

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D: "Local Max" quadratic complexity bound, (LMQ)

For the polynomial $p(x) \in \mathbb{R}[x]$

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each negative coefficient $a_i < 0$ is "paired" with each one of the preceding positive coefficients a_j divided by 2^{t_j} — where t_j is initially set to 1 and is incremented each time the positive coefficient a_j is used — and the minimum is taken over all j; subsequently, the maximum is taken over all i.

That is, we have:

$$ub_{LMQ} = \max_{\{a_i < 0\}} \min_{\{a_j > 0: j > i\}} \sqrt[j-i]{-\frac{a_i}{\frac{a_j}{2^{t_j}}}},$$

Each positive coefficient a_j is "broken up" into unequal parts, as is done with just the locally maximum coefficient in the linear local max bound.

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Example

Consider again the polynomial

$$x^3 + 10^{100}x^2 - 10^{100}x - 1,$$

which has one sign variation and, hence, one positive root equal to 1

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Example

Consider again the polynomial

$$x^3 + 10^{100}x^2 - 10^{100}x - 1,$$

which has one sign variation and, hence, one positive root equal to 1

With CQ, Cauchy's quadratic complexity bound, we compute:

▶ the minimum of the two radicals obtained from the pairs of terms {x³/2, -10¹⁰⁰x} and {10¹⁰⁰x²/2, -10¹⁰⁰x} which is 2, and
▶ the minimum of the two radicals obtained from the pairs of terms {x³, -1} and {10¹⁰⁰x², -1} which is 1/10⁵⁰.
▶ Therefore, the obtained estimate of the bound is max{2, 1/10⁵⁰} = 2.

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Example

Consider again the polynomial

$$x^3 + 10^{100}x^2 - 10^{100}x - 1,$$

which has one sign variation and, hence, one positive root equal to 1

With KQ, Kioustelidis' quadratic complexity bound, we compute:

▶ the minimum of the two radicals obtained from the pairs of terms {^{x³}/_{2²}, -10¹⁰⁰x} and {^{10¹⁰⁰x²}/₂, -10¹⁰⁰x} which is 2, and
▶ the minimum of the two radicals obtained from the pairs of terms {^{x³}/_{2³}, -1} and {^{10¹⁰⁰x²}/_{2²}, -1} which is ²/_{10⁵⁰}.
▶ Therefore, the obtained estimate of the bound is max{2, ²/_{10⁵⁰}} = 2.

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Example

Consider again the polynomial

$$x^3 + 10^{100}x^2 - 10^{100}x - 1,$$

which has one sign variation and, hence, one positive root equal to 1

With **FLQ**, the "First- λ " quadratic bound, we compute:

the minimum of the two radicals obtained from the pairs of terms {x³, -10¹⁰⁰x} and {10¹⁰⁰x², -10¹⁰⁰x} which is 1, and
the minimum of the two radicals obtained from the pairs of terms {x³, -1} and {10¹⁰⁰x², -1} which is 1.
Therefore, the obtained estimate of the bound is max{1,1} = 1.
Note: Once a term with a positive coefficient has been used in obtaining the minimum, it cannot be used again!

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Example

Consider again the polynomial

$$x^3 + 10^{100}x^2 - 10^{100}x - 1,$$

which has one sign variation and, hence, one positive root equal to 1

With LMQ, the "Local Max" quadratic bound, we compute:

▶ the minimum of the two radicals obtained from the pairs of terms {x³/2, -10¹⁰⁰x} and {10¹⁰⁰x²/2, -10¹⁰⁰x} which is 2, and
▶ the minimum of the two radicals obtained from the pairs of terms {x³/2², -1} and {10¹⁰⁰x²/2², -1} which is 2/10⁵⁰.
▶ Therefore, the obtained estimate of the bound is max{2, 2/10⁵⁰} = 2.

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Linear vs quadratic complexity bounds

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Linear vs quadratic complexity bounds

► From the example we see that the estimates of all quadratic complexity bounds are much better than those of their linear complexity counterparts.

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Linear vs quadratic complexity bounds

► From the example we see that the estimates of all quadratic complexity bounds are much better than those of their linear complexity counterparts.

► In general, the quadratic complexity bounds cannot perform worse than the linear complexity ones; most of the times they perform a lot better!

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Comparison of the 4 quadratic complexity bounds

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► FLQ is faster — or quite faster — than all of them because it tests just the first $min(\iota, \lambda)$ positive coefficients. By comparison, all the other quadratic complexity bounds test every preceding positive coefficient.

▶ The estimates computed by *LMQ* are sharper by the factor $2^{\frac{j-i-t_j}{j-i}}$ than those computed by Kioustelidis' *KQ* — because $2^{t_j} \leq 2^{j-i}$, where *i* and *j* are the indices realizing the *max* of *min*.

Equality holds when there are no missing terms in the polynomial.

Comparison of the 4 quadratic complexity bounds

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▶ The estimates computed by LMQ are sharper by the factor $2^{\frac{j-i-t_j}{j-i}}$ than those computed by Kioustelidis' KQ — because $2^{t_j} \leq 2^{j-i}$, where *i* and *j* are the indices realizing the *max* of *min*.

Equality holds when there are no missing terms in the polynomial.

► Experimental results indicated that *FLQ*, *LMQ* and *min*(*FLQ*, *LMQ*) behave equally well! Therefore, we picked pick *LMQ* to improve the performance of *VAS*.

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Conclusions on the quadratic complexity bounds

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Conclusions on the quadratic complexity bounds

▶ Using LMQ, the performance of the VAS real root isolation method was speeded up by an average overall factor of 40%.

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Overall time spent for computing bounds

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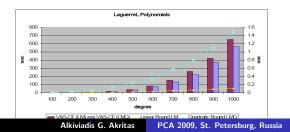
Overall time spent for computing bounds

▶ In the following graph the left scale shows the times in seconds (bars) needed by VAS to isolate the roots of a certain class of polynomials using both LM, the Local Max bound, and LMQ, its quadratic version. The right scale is associated with the two curves which show the total time spent by VAS in computing the bounds.

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Overall time spent for computing bounds

▶ In the following graph the left scale shows the times in seconds (bars) needed by VAS to isolate the roots of a certain class of polynomials using both LM, the Local Max bound, and LMQ, its quadratic version. The right scale is associated with the two curves which show the total time spent by VAS in computing the bounds.



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Products of terms x - r, with random integer r — Revisited

Products of terms x - r, with random integer r — Revisited

Table: We compare the timings in seconds (s) for: (a) VAS(cauchy), i.e. VAS using Cauchy's rule, (b) VAS(fl+lm), i.e. VAS using the linear complexity bound min(FL + LM), and (c) VAS(lmq), i.e. VAS using the Locam Max quadratic complexity bound. The average speed-up for this table is about 35%.

Bit-length	Degree	VAS(cauchy)	VAS(fl+lm)	VAS(lmq)
		t(s)	t(s)	t(s)
of roots		Avg(Min/Max)	Avg(Min/Max)	Avg(Min/Max)
10	100	0.46 (0.28/0.94)	0.24 (0.18/0.28)	0.34 (0.30/0.41)
10	200	1.46 (1.24/1.85)	1.40 (1.28/1.69)	1.40 (1.20/1.69)
10	500	18.1 (16.5/18.9)	18.1 (16.6/18.8)	22.1 (18.7/24.2)
1000	20	0.07 (0.04/0.14)	0.02 (0.02/0.03)	0.03 (0.02/0.04)
1000	50	3.69 (2.38/6.26)	0.81 (0.60/1.28)	0.81 (0.52/1.11)
1000	100	47.8 (37.6/56.9)	13.8 (10.3/19.2)	15.8 (11.3/21.3)

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Future Research

Future Research

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Computing time bound

Sharma's bound on the computing time of the VAS continued fractions method is greatly overestimated.

Hence, theoretical research is needed to see if we can bring it down.

Future Research

Coefficients

The VAS continued fractions method works for integer or rational coefficients.

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Hence, we need to discover new ways to deal with coefficients that are algebraic numbers or approximate reals.

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Sparse polynomials of great degree

The VAS continued fractions method is the fastest real root isolation method when the polynomials are not sparse and their degree is less than a few thousand.

However, *Mathematica* runs out of memory when we try to isolate the roots of a sparse polynomial of degree 100000.

Hence, we need to discover new ways to deal with sparse polynomials of extremely high degrees.

Future Research

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Parallel implementation

Last, but not least, we need to investigate the performance of the $V\!AS$ continued fractions method in a multiprocessor environment.

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Alkiviadis G. Akritas PCA 2009, St. Petersburg, Russia

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