Fast high-resolution drawing of algebraic curves and surfaces

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Overview



1 Implicit curve drawing



2 Previous work



3 Our approach



4 Fast multipoint evaluation





6 Experiments

Implicit curve drawing

Scientific visualization

Some scientific visualization applications:

- modeling
- medical imaging
- mechanism design

Goal: build an intuition and get an understanding of the data



3D CT reconstruction of distal tibia fracture



Industrial robots from KUKA by Mixabest (CC BY-SA 3.0)

Implicit curve drawing problem

General problem

Discrete representation of an implicit curve on a fixed grid

- Input:
 - ► function *F*
 - resolution N
 - visualization window

Implicit curve defined as the solution set

$$\{(x,y)\in\mathbb{R}^2\mid F(x,y)=0\}$$

• **Output**: drawing (set of pixels)



Implicit curve drawing problem

Our focus

Discrete representation of an algebraic curve on a fixed grid

- Input:
 - **bivariate polynomial** *P* of **partial degree** *d*
 - resolution N
 - window $[-1,1] \times [-1,1]$

Algebraic curve defined as the solution set

$$\{(x,y)\in\mathbb{R}^2\mid P(x,y)=0\}$$

• **Output**: drawing (set of pixels)

Goal: fast high-resolution drawing of high degree algebraic curves

- $d pprox 100 \longrightarrow d^2 pprox 10,000$ monomials
- $N \approx 1,000$



Why high degree algebraic curves?

Goal of visualization: build an intuition and get an understanding of the data

In robotics, the configuration space could be of high dimension

$$\mathbb{R}^N o \mathbb{R}^M$$

Operations on algebraic varieties:

- cut
- projection

Industrial robots from KUKA by Mixabest (CC BY-SA 3.0)

Correctness of the drawing

For numerical reasons, there may be some:

• False negative pixels



Correctness of the drawing

For numerical reasons, there may be some:

- False negative pixels
- False positive pixels



Previous work

The idea

2D variant of the widely used marching cubes algorithm [Lorensen & Cline, 1987] Implicit curve defined by P(X, Y) = 0



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Complexity

Complexity (number of elementary operations) Naive evaluation

 $\theta(d^2N^2)$

d partial degree N resolution of the grid

Arithmetic complexity of the marching squares

With partial evaluation of P(x, y), assuming d < N

 $\theta(dN^2)$

Slow for high resolutions... Can we have an algorithm in O(dN)?

Local refinements of the grid



Local refinements of the grid



Local refinements of the grid



Local refinements of the grid



Local refinements of the grid



Methods providing topological correctness

Adaptive 2D subdivision with interval arithmetic

- [Snyder, 1992]
- [Plantinga & Vegter, 2004]
- [Burr et al., 2008]
- [Lin & Yap, 2011]
- . . .

Cylindrical algebraic decomposition (CAD)

- [Gonzalez-Vega & Necula, 2002]
- [Eigenwillig et al., 2007]
- [Alberti et al., 2008]
- [Cheng et al., 2009]
- [Kobel & Sagraloff, 2015]
- [Diatta et al., 2018]
- . . .







https://isotop.gamble.loria.fr/

Our approach

Inclusion property

$$P(X) = 2X^3 - X^2 - 1.5X + 0.75$$

How to compute P(I) for I = [-1, 1]?





P(I) = [-0.75, 1.06...]

Inclusion property

$$P(X) = 2X^3 - X^2 - 1.5X + 0.75$$

How to compute P(I) for I = [-1, 1]?



$$\Box P(I) = 2[-1,1]^3 - [-1,1]^2 - 1.5[-1,1] + 0.75$$
$$= [-5.25, 5.25]$$

P(I) = [-0.75, 1.06...]

Inclusion property

$$P(X) = 2X^3 - X^2 - 1.5X + 0.75$$

How to compute P(I) for I = [-1, 1]?



 $\Box P(I) = 2[-1,1]^3 - [-1,1]^2 - 1.5[-1,1] + 0.75$ = [-5.25, 5.25]

With Horner's scheme:

$$\Box P(I) = ((2[-1,1]-1)[-1,1]-1.5)[-1,1] + 0.75$$
$$= [-3.75, 5.25]$$
$$P(I) \subseteq \Box P(I)$$

P(I) = [-0.75, 1.06...]

Convergence property

Convergence at a point With $x \in [a, b]$

 $\lim_{[a,b]\longrightarrow[x,x]=\{x\}}\Box P([a,b])=P(x)$

Our approach: guaranteed intersection with the grid



Adaptive subdivision



New approach: evaluation along fibers



 \Rightarrow Make it fast and provide some guarantees

An algorithm

Pixel drawing

- evaluation in X Chebyshev nodes multipoint evaluation with IDCT Taylor approximation
- *subdivision in Y* naive root finding method

Guarantees False positive pixels *only*

 $P([x_k, x_{k+1}], Y) = \sum a_j Y^j$



 $P([x_k, x_{k+1}], Y) = \sum a_j Y^j$



 $P([x_7, x_8], Y)$

 $P([x_k, x_{k+1}], Y) = \sum a_j Y^j$



 $P([x_k, x_{k+1}], Y) = \sum a_j Y^j$



 $P([x_k, x_{k+1}], Y) = \sum a_j Y^j$


Pixel drawing Pixel lighting

- Detect a crossing in pixel of the grid
- Light that pixel



Pixel drawing False positive and false negative pixels

Some incorrect pixels:

- False negative when a connected component lies inside of a pixel
- False positive when the evaluation on an edge of a pixel is close to zero That occurs for a segment *S* when

$$0\in \Box P(S)+[-E,E]$$



Pixel drawing False positive and false negative pixels

Some incorrect pixels:

- False negative when a connected component lies inside of a pixel
- False positive when the evaluation on an edge of a pixel is close to zero That occurs for a segment *S* when

$$0\in \Box P(S)+[-E,E]$$

Certification of segments that are not crossed:

$$0 \notin \Box P(S) + [-E, E]$$

$$\downarrow \\ 0 \notin P(S)$$



Fast multipoint evaluation

A prerequisite to fast multipoint evaluation

Chebyshev polynomials

Definition

The Chebyshev polynomials (T_k) verify $\forall k \in \mathbb{N}, T_k(\cos \theta) = \cos(k\theta)$

The first three Chebyshev polynomials

$$\begin{aligned} \cos(0 \cdot \theta) &= 1 & T_0 = 1 \\ \cos(1 \cdot \theta) &= \cos(\theta) & T_1 = X \\ \cos(2 \cdot \theta) &= 2\cos(\theta)^2 - 1 & T_2 = 2X^2 - 1 \end{aligned}$$

A prerequisite to fast multipoint evaluation

Chebyshev polynomials

Definition

The Chebyshev polynomials (T_k) verify $\forall k \in \mathbb{N}, T_k(\cos \theta) = \cos(k\theta)$

Lemma

An arbitrary polynomial p of degree d can be written in terms of the Chebyshev polynomials:

$$p(X) = \sum_{k=0}^{d} \alpha_k T_k(X)$$

A prerequisite to fast multipoint evaluation

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Lemma

For $N \in \mathbb{N}$, a polynomial p of degree d can be evaluated on the Chebyshev nodes $(c_n)_{0 \le n \le N-1}$ using the IDCT:

$$(p(c_n))_{0 \le n \le N-1} = \frac{1}{2}(\alpha_0, \ldots, \alpha_0) + \mathsf{IDCT}((\alpha_k)_{0 \le k \le N-1})$$

A prerequisite to fast multipoint evaluation Chebyshev nodes

Definition

For $N \in \mathbb{N}$, the Chebyshev nodes are

$$c_n = \cos\left(\frac{2n+1}{2N}\pi\right), \ n=0,\ldots,N-1$$

They are the roots of T_N



Inverse Discrete Cosine Transform (IDCT): $\alpha_k \rightarrow x_n$

$$x_{n} = \frac{1}{2}\alpha_{0} + \sum_{k=1}^{N-1} \alpha_{k} \cos\left[\frac{\pi k(2n+1)}{2N}\right]$$

$$IDCT$$

$$\lim_{linear transformation} \qquad FFT \qquad \lim_{linear transformation} \bigvee_{\mathbf{v}_{k}}$$

 \Rightarrow Fast thanks to the Fast Fourier Transform (FFT) algorithm in $O(N \log_2 N)$

[Makhoul, 1980]

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$$p(c_n) = \sum_{k=0}^{N-1} \alpha_k T_k \left(\cos \left(\frac{2n+1}{2N} \pi \right) \right)$$

Inverse Discrete Cosine Transform (IDCT): $\alpha_k \rightarrow x_n$

$$x_{n} = \frac{1}{2}\alpha_{0} + \sum_{k=1}^{N-1} \alpha_{k} \cos\left[\frac{\pi k(2n+1)}{2N}\right]$$

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$$p(c_n) = \sum_{k=0}^{N-1} \alpha_k T_k \left(\cos\left(\frac{2n+1}{2N}\pi\right) \right) = \sum_{k=0}^{N-1} \alpha_k \cos\left[\frac{\pi k(2n+1)}{2N}\right]$$

Inverse Discrete Cosine Transform (IDCT): $\alpha_k \rightarrow x_n$

$$x_{n} = \frac{1}{2}\alpha_{0} + \sum_{k=1}^{N-1} \alpha_{k} \cos\left[\frac{\pi k(2n+1)}{2N}\right]$$

$$IDCT$$

$$(\alpha_{k}) - \cdots \rightarrow (V_{k}) \xrightarrow{\mathsf{FFT}} (v_{k}) \xrightarrow{\mathsf{linear transformation}} (x_{k})$$

 \Rightarrow Fast thanks to the Fast Fourier Transform (FFT) algorithm in $O(N \log_2 N)$ [Makhoul, 1980]

$$p(c_n) = \frac{1}{2}\alpha_0 + \frac{1}{2}\alpha_0 + \sum_{k=1}^{N-1} \alpha_k \cos\left[\frac{\pi k(2n+1)}{2N}\right]$$
$$(p(c_n))_{0 \le n \le N-1} = \frac{1}{2}(\alpha_0, \dots, \alpha_0) + \mathsf{IDCT}((\alpha_k)_{0 \le k \le N-1})$$

Error of the IDCT

[Makhoul, 1980] and [Brisebarre et al., 2020, Theorem 3.4] yield

Theorem (H., Moroz, Pouget, 2022)

Assume radix-2, precision-p arithmetic, with rounding unit $u = 2^{-p}$. Let \hat{x} be the computed 2^n -point IDCT of $\alpha \in \mathbb{C}^{2^n}$, and let x be the exact value. Then

 $\|\widehat{x}-x\|_{\infty}=n\|\alpha\|_{\infty}O(u).$

Table: IDCT error bounds for p = 53 (double precision)

$N = 2^n$	1,024	2,048	4,096	8,192	16,384	32,768
$\ \widehat{x} - x\ _{\infty} / \ \alpha\ _{\infty}$	7.97e-15	8.84e-15	9.72e-15	1.06e-14	1.15e-14	1.23e-14

 $26 \, / \, 41$

Algorithms

Illustration

$$P(X, Y) = \sum \left(\sum a_{i,j} X^i\right) Y^j = \sum p_j(X) Y^j$$
$$p_j(X) = \sum a_{i,j} X^i = \sum \alpha_{i,j} T_i(X)$$
$$(p_j(c_n))_{0 \le n \le N-1} = \frac{1}{2} (\alpha_{0,j}, \dots, \alpha_{0,j}) + \mathsf{IDCT}((\alpha_{k,j})_{0 \le k \le N-1})$$

 $\frac{\text{Illustration}}{P(c_n, Y)} = \sum p_j(c_n) Y^j$



 $\frac{\text{Illustration}}{P(c_3, Y)} = \sum p_j(c_3) Y^j$



 $\begin{array}{l} \text{Illustration} \\ P(c_3,Y) = \sum p_j(c_3)Y^j \end{array}$



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An edge enclosing algorithm



IDCT multipoint evaluation in Xat $c_0, c_1 \dots$

subdivision in Y

IDCT multipoint evaluation of the partial polynomials of $P(X, Y) = \sum p_j(X)Y^j$

An edge enclosing algorithm





subdivision in Y

IDCT multipoint evaluation of the partial polynomials of $P(X, Y) = \sum p_j(X)Y^j$

An edge enclosing algorithm



IDCT multipoint evaluation of the partial polynomials of $P(X, Y) = \sum p_j(X)Y^j$

General idea: pixel enclosure

 $\frac{\text{Illustration}}{P(I,Y)} = \sum p_j(I)Y^j$



General idea: pixel enclosure

 $\frac{\text{Illustration}}{P(I,Y)} = \sum p_j(I)Y^j$



General idea: pixel enclosure Illustration $P(I, Y) = \sum p_j(I)Y^j$

P(I, Y)

 \checkmark

×





IDCT multipoint evaluation in Xaround $c_0, c_1 \dots$

subdivision in Y

A pixel enclosing algorithm



IDCT multipoint evaluation + Taylor approximation in X

subdivision in Y

Taylor expansion of the partial polynomials of $P(X, Y) = \sum p_j(X)Y^j$

$$\left| p(c_n + r) - \left(p(c_n) + rp'(c_n) + \dots + \frac{r^m}{m!} p^{(m)}(c_n) \right) \right| \le \max_{l_{c_n}} \left| p^{(m+1)} \right| \frac{|r|^{(m+1)}}{(m+1)!}$$

A pixel enclosing algorithm



IDCT multipoint evaluation + Taylor approximation in X

subdivision in Y

Taylor expansion of the partial polynomials of $P(X, Y) = \sum p_j(X)Y^j$

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ight)
ight| \leq \max_{l_{c_n}} \left| p^{(m+1)} \left| rac{|r|^{(m+1)}}{(m+1)!}
ight|$$

Complexities

multipoint evaluation and subdivision $O(d^3 + dN \log_2(N) + dN)$	
multipoint Taylor approximation and subdivision $O(md^3 + mdN \log_2(N) + d)$	dNT) dNT)

d partial degree

N resolution

 ${\cal T}$ maximum number of nodes of the subdivision trees over all vertical fibers / stripes

With a constant number of branches in the window, we expect $T = O(\log_2(N))$

Experiments

Pixel classification

- crossed: blue
- not crossed: white
- undecided: yellow



Drawing for two families of polynomials

Experiments on smooth curves \longrightarrow random polynomials $\xi_{i,j}$: random coefficients in [-100, 100]

Kac polynomial



Kostlan-Shub-Smale (KSS) polynomial





Figure: Kac polynomial of degree d = 110 at a resolution N = 1,024, $\frac{b}{b+y} = 24\%$


Drawing for two families of polynomials

Figure: KSS polynomial of degree d = 40 at a resolution N = 1,024, $\frac{b}{b+y} = 19\%$

Comparison to state-of-the-art software

Our methods

• edge drawing \rightarrow curve enclosing edges false positive and false negative • pixel drawing \rightarrow curve enclosing pixels

Some similar methods

- scikit / NumPy \rightarrow marching squares
- MATLAB \rightarrow could not find the method used
- ImplicitEquations \rightarrow 2D adaptive subdivision

A topologically correct method

• Isotop \rightarrow cylindrical algebraic decomposition

false positive

false negative false negative? false positive





Computation times for a Kac polynomial of degree 40 (in seconds)





Computation times for a Kac polynomial of degree 40 (in seconds)

scikit: $O(dN^2)$

no guarantee slow when d and N are large

Our methods: O(dNT)as expected $T = O(\log_2(N))$

 $\frac{\text{guarantees}}{\text{fast when } d \text{ and } N \text{ are large}}$

Output for a singular curve

Curve: $dfold_{8,1}$ from Challenge 14 of Oliver Labs[13][37] (d = 18)





Conclusion

Contributions

- Two algorithms
 - enclosure of the edges
 - enclosure of the pixels
- Fast implicit curve and surface algorithms for high resolutions: faster than marching squares and marching cubes
- Better guarantees on the drawing than marching squares
- Ability to handle high degrees (d > 20) and high resolutions (N > 1,000)