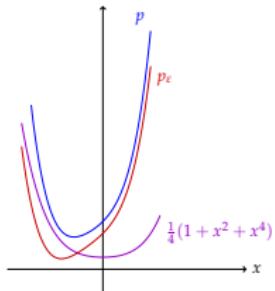


# Exact Polynomial Optimization

**Victor Magron**, LAAS CNRS



TU Chemnitz  
22 October 2019



# Deciding Nonnegativity & Exact Optimization

---

$X = (X_1, \dots, X_n)$

**co-NP hard problem:** check  $f \geq 0$  on  $\mathbf{K}$

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**NP hard problem:**  $\min\{f(x) : x \in \mathbf{K}\}$

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2 Constrained

$$\rightsquigarrow \mathbf{K} = \{x \in \mathbb{R}^n : g_1(x) \geq 0, \dots, g_m(x) \geq 0\} \quad g_j \in \mathbb{Q}[X]$$

$$\deg f, \deg g_j \leq d$$

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[Collins 75]💡 CAD doubly exp. in  $n$  poly. in  $d$

[Grigoriev-Vorobjov 88, Basu-Pollack-Roy 98]

💡 Critical points singly exponential time  $(m+1)\tau d^{\mathcal{O}(n)}$

# Deciding Nonnegativity & Exact Optimization

---

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$$\sigma = h_1^2 + \cdots + h_p^2$$

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 [Artin 27] YES!

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$\simeq \rightarrow =$

The Question of Exact Certification

How to go from **approximate** to **exact** certification?

# Motivation

---

## Positivity certificates

- Stability proofs of critical control systems (Lyapunov)
- Certified function evaluation [Chevillard et. al 11]
- Formal verification of real inequalities [Hales et. al 15]:



HOL-LIGHT

# Decomposing Nonnegative Polynomials

---

1 Polya's representation

positive definite form  $f$

[Reznick 95]

$$f = \frac{\sigma}{(X_1^2 + \dots + X_n^2)^D}$$

2 Hilbert-Artin's representation

$f \geq 0$

[Artin 27]

$$f = \frac{\sigma}{h^2}$$

3 Putinar's representation

$f = \sigma_0 + \sigma_1 g_1 + \dots + \sigma_m g_m f > 0$  on compact  $K$

$\deg \sigma_i \leq 2D$

[Putinar 93]

# Decomposing Nonnegative Polynomials

---

- Deciding **polynomial nonnegativity**

$$f(a, b) = a^2 - 2ab + b^2 \geq 0$$

- $f(a, b) = \begin{pmatrix} a & b \end{pmatrix} \underbrace{\begin{pmatrix} z_1 & z_2 \\ z_2 & z_3 \end{pmatrix}}_{\succcurlyeq 0} \begin{pmatrix} a \\ b \end{pmatrix}$
- $a^2 - 2ab + b^2 = z_1 a^2 + 2z_2 ab + z_3 b^2 \quad (\mathbf{A} \mathbf{z} = \mathbf{d})$
- $\begin{pmatrix} z_1 & z_2 \\ z_2 & z_3 \end{pmatrix} = \underbrace{\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}}_{\mathbf{F}_1} z_1 + \underbrace{\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}}_{\mathbf{F}_2} z_2 + \underbrace{\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}}_{\mathbf{F}_3} z_3 \succcurlyeq \underbrace{\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}}_{\mathbf{F}_0}$

# Decomposing Nonnegative Polynomials

---

- Choose a cost  $\mathbf{c}$  e.g.  $(1, 0, 1)$  and solve **SDP**

$$\begin{aligned} \min_{\mathbf{z}} \quad & \mathbf{c}^\top \mathbf{z} \\ \text{s.t.} \quad & \sum_i \mathbf{F}_i z_i \succcurlyeq \mathbf{F}_0 , \quad \mathbf{A} \mathbf{z} = \mathbf{d} \end{aligned}$$

- Solution  $\begin{pmatrix} z_1 & z_2 \\ z_2 & z_3 \end{pmatrix} = \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix} \succcurlyeq 0$  (eigenvalues 0 and 2)
- $a^2 - 2ab + b^2 = (a - b) \underbrace{\begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix}}_{\succcurlyeq 0} \begin{pmatrix} a \\ b \end{pmatrix} = (a - b)^2$
- Solving **SDP**  $\implies$  Finding **SUMS OF SQUARES** certificates

# Decomposing Nonnegative Polynomials

---

## 4 Circuit polynomial

$$f = b_{\alpha(1)} X^{\alpha(1)} + \cdots + b_{\alpha(r)} X^{\alpha(r)} + b_{\beta} X^{\beta}$$

$$b_{\alpha(j)} > 0 \quad \alpha(j) \in (2\mathbb{N})^n$$

$$\beta = \lambda_1 \alpha(1) + \cdots + \lambda_r \alpha(r) \quad \lambda_j > 0 \text{ and } \lambda_1 + \cdots + \lambda_r = 1$$

# Decomposing Nonnegative Polynomials

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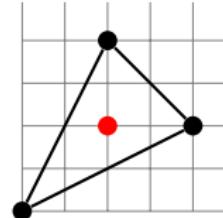
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$$f = 1 + X_1^2 X_2^4 + X_1^4 X_2^2 - 3X_1^2 X_2^2$$



# Decomposing Nonnegative Polynomials

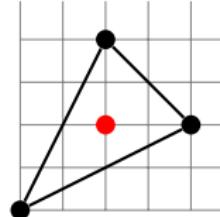
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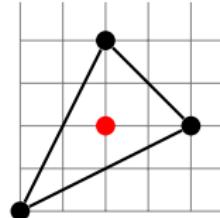
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Theorem (Illman-de Wolff 16)

$$f \geq 0 \Leftrightarrow |b_{\beta}| \leq \Theta_f \text{ or } (b_{\beta} \geq -\Theta_f, \beta \text{ even})$$

💡 SONC (SUMS OF NONNEGATIVE CIRCUITS)

# Decomposing Nonnegative Polynomials

---

## 5 arithmetic-geometric-mean-exponential (AGE)

$$f = c_1 \exp[X \cdot \alpha(1)] + \cdots + c_t \exp[X \cdot \alpha(t)] + \beta \exp[X \cdot \alpha(0)]$$
$$c_j \in \mathbb{Q}_{>0} \quad \beta \in \mathbb{Q} \quad \alpha(j) \in \mathbb{N}^n$$

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**relative entropy**

$$D(\nu, \mathbf{c}) = \sum_j \nu_j \log \frac{\nu_j}{c_j}$$

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**relative entropy**       $D(\nu, \mathbf{c}) = \sum_j \nu_j \log \frac{\nu_j}{c_j}$

Theorem (Chandrasekaran-Shah 16)

$$f \geq 0 \Leftrightarrow \exists \nu \mid D(\nu, \mathbf{c}) \leq \beta \text{ and } \sum_j \alpha(j) \nu_j = (\mathbf{1} \cdot \nu) \alpha(0)$$

💡 SAGE (SUMS OF AGE)

# From Approximate to Exact Solutions

## APPROXIMATE SOLUTIONS

sum of squares of  $a^2 - 2ab + b^2$ ?



$(1.00001a - 0.99998b)^2$ !



$$a^2 - 2ab + b^2 \simeq (1.00001a - 0.99998b)^2$$

$$a^2 - 2ab + b^2 \neq 1.0000200001a^2 - 1.9999799996ab + 0.9999600004b^2$$

$\simeq \rightarrow = ?$

# Rational SOS Decompositions

---

- Let  $f \in \mathbb{R}[X]$  and  $f \geq 0$  on  $\mathbb{R}$  ( $n = 1$ )

## Theorem

There exist  $f_1, f_2 \in \mathbb{R}[X]$  s.t.  $f = f_1^2 + f_2^2$ .

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## Proof.

$$f = h^2(q + ir)(q - ir)$$

□

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## Examples

$$1 + X + X^2 = \left(X + \frac{1}{2}\right)^2 + \left(\frac{\sqrt{3}}{2}\right)^2$$

$$\begin{aligned}1 + X + X^2 + X^3 + X^4 &= \left(X^2 + \frac{1}{2}X + \frac{1 + \sqrt{5}}{4}\right)^2 + \\&\quad \left(\frac{\sqrt{10 + 2\sqrt{5}} + \sqrt{10 - 2\sqrt{5}}}{4}X + \frac{\sqrt{10 - 2\sqrt{5}}}{4}\right)^2\end{aligned}$$

# Rational SOS Decompositions

---

- $f \in \mathbb{Q}[X] \cap \mathring{\Sigma}[X]$  (interior of the SOS cone)

## Existence Question

Does there exist  $f_i \in \mathbb{Q}[X], c_i \in \mathbb{Q}^{>0}$  s.t.  $f = \sum_i c_i f_i^2$ ?

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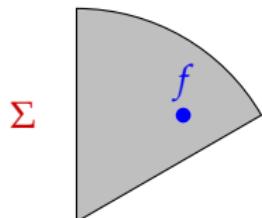
## Examples

$$1 + X + X^2 = \left(X + \frac{1}{2}\right)^2 + \left(\frac{\sqrt{3}}{2}\right)^2 = 1 \left(X + \frac{1}{2}\right)^2 + \frac{3}{4}(1)^2$$

$$\begin{aligned} 1 + X + X^2 + X^3 + X^4 &= \left(X^2 + \frac{1}{2}X + \frac{1+\sqrt{5}}{4}\right)^2 + \\ &\quad \left(\frac{\sqrt{10+2\sqrt{5}} + \sqrt{10-2\sqrt{5}}}{4}X + \frac{\sqrt{10-2\sqrt{5}}}{4}\right)^2 = ??? \end{aligned}$$

# Round & Project Algorithm [Peyrl-Parrilo 08]

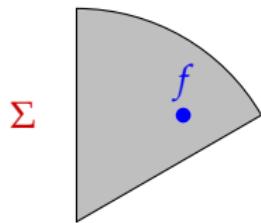
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$f \in \mathring{\Sigma}[X]$  with  $\deg f = 2D$

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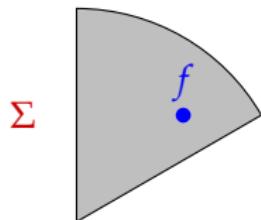
💡 Find  $\tilde{\mathbf{G}}$  with SDP at tolerance  $\tilde{\delta}$  satisfying

$$f(X) \simeq \mathbf{v}_D^T(X) \tilde{\mathbf{G}} \mathbf{v}_D(X) \quad \tilde{\mathbf{G}} \succ 0$$

$\mathbf{v}_D(X)$ : vector of monomials of  $\deg \leq D$

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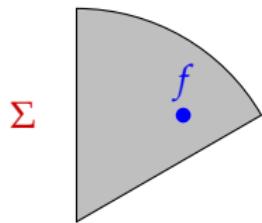
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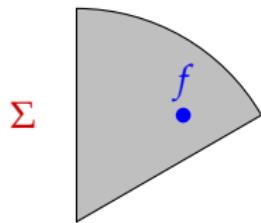
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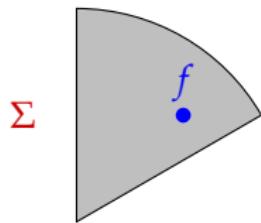
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1 Rounding step  $\hat{G} \leftarrow \text{round}(\tilde{G}, \hat{\delta})$

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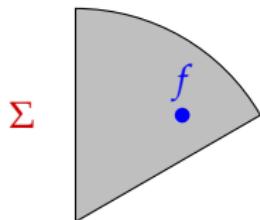
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💡 Small enough  $\tilde{\delta}, \hat{\delta} \implies f(X) = \mathbf{v}_D^T(X) \mathbf{G} \mathbf{v}_D(X)$  and  $\mathbf{G} \succcurlyeq 0$

# One Answer when $\mathbf{K} = \{\mathbf{x} \in \mathbb{R}^n : g_j(\mathbf{x}) \geq 0\}$

---

## 💡 Hybrid SYMBOLIC/NUMERIC methods

📄 Magron-Allamigeon-Gaubert-Werner 14

$$f \simeq \tilde{\sigma}_0 + \tilde{\sigma}_1 g_1 + \cdots + \tilde{\sigma}_m g_m$$

$$u = f - \tilde{\sigma}_0 + \tilde{\sigma}_1 g_1 + \cdots + \tilde{\sigma}_m g_m$$

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Compact  $\mathbf{K} \subseteq [0, 1]^n$

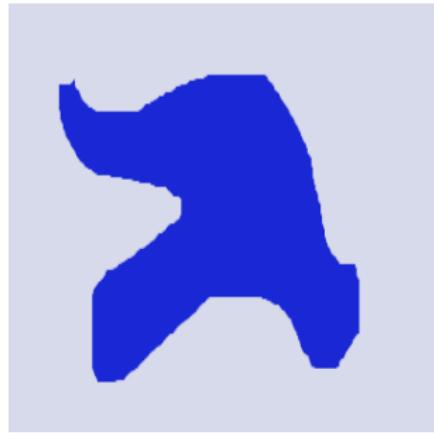
$$u = f - \tilde{\sigma}_0 + \tilde{\sigma}_1 g_1 + \cdots + \tilde{\sigma}_m g_m$$

$$\boxed{\simeq \quad \rightarrow \quad =}$$

💡  $\forall \mathbf{x} \in [0, 1]^n, u(\mathbf{x}) \leq -\varepsilon$

$\min_{\mathbf{K}} f \geq \varepsilon$  when  $\varepsilon \rightarrow 0$

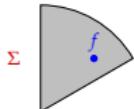
**COMPLEXITY?**



# From Approximate to Exact Solutions

---

Win TWO-PLAYER GAME



sum of squares of  $f$ ?

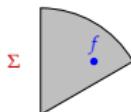


$\simeq$  Output!



# From Approximate to Exact Solutions

Win TWO-PLAYER GAME



**Hybrid** Symbolic/Numeric Algorithms

sum of squares of  $f + \varepsilon$ ?



Error Compensation

$\approx \rightarrow =$

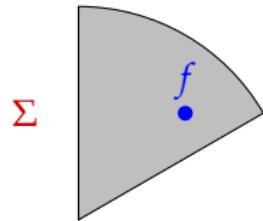
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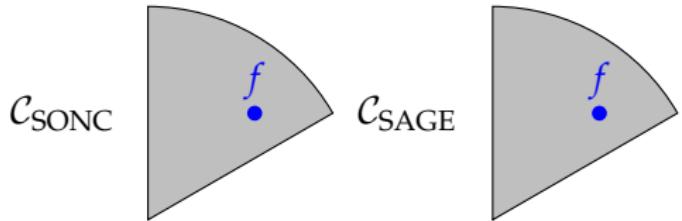
# From Approximate to Exact Solutions

---

Exact SOS



Exact SONC/SAGE



## Software: RealCertify and POEM

---

### Exact optimization via SOS: [RealCertify](#)

Maple & arbitrary precision SDP solver SDPA-GMP  
[Nakata 10]

univsos       $n = 1$

multivsos       $n > 1$

### Exact optimization via SONC/SAGE: [POEM](#)

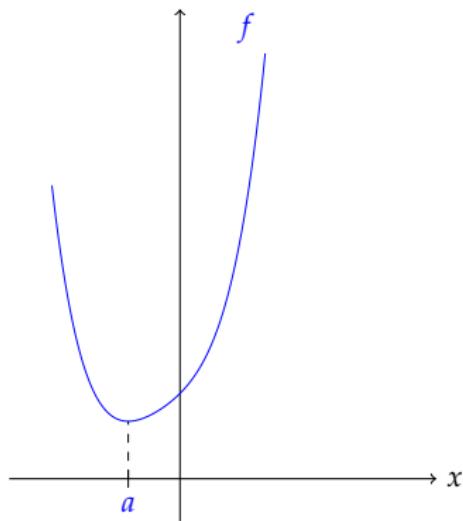
Python (SymPy) & geometric programming/relative entropy ECOS  
[Domahidi-Chu-Boyd 13]

# univsos: Outline [Schweighofer 99]

---

$f \in \mathbb{Q}[X]$  and  $f > 0$

Minimizer  $a$  may not be in  $\mathbb{Q}$ ...



$$f = 1 + X + X^2 + X^3 + X^4$$

$$a = \frac{5}{4(135+60\sqrt{6})^{1/3}} - \frac{4(135+60\sqrt{6})^{1/3}}{12} - \frac{1}{4}$$

# univsos: Outline [Schweighofer 99]

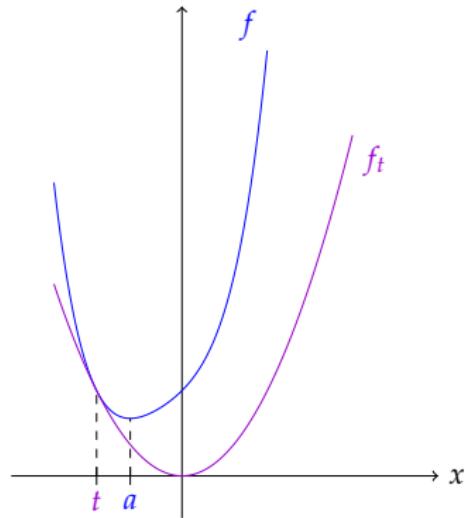
---

$f \in \mathbb{Q}[X]$  and  $f > 0$

Minimizer  $a$  may not be in  $\mathbb{Q}$  ...

💡 Find  $f_t \in \mathbb{Q}[X]$  s.t. :

- $\deg f_t \leq 2$
- $f_t \geq 0$
- $f \geq f_t$
- $f - f_t$  has a root  $t \in \mathbb{Q}$



$$f = 1 + X + X^2 + X^3 + X^4$$

$$a = \frac{5}{4(135+60\sqrt{6})^{1/3}} - \frac{4(135+60\sqrt{6})^{1/3}}{12} - \frac{1}{4}$$

$$f_t = X^2$$

$$t = -1$$

# univsos: Outline [Schweighofer 99]

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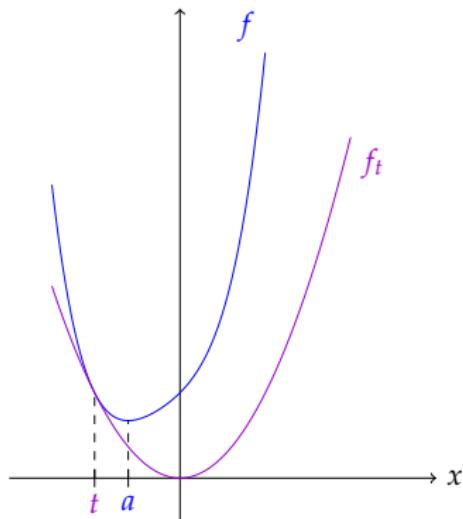
$f \in \mathbb{Q}[X]$  and  $f > 0$

Minimizer  $a$  may not be in  $\mathbb{Q}$ ...

💡 Square-free decomposition:

$$f - f_t = gh^2$$

- $\deg g \leq \deg f - 2$
- $g > 0$
- Do it again on  $g$



$$f = 1 + X + X^2 + X^3 + X^4$$

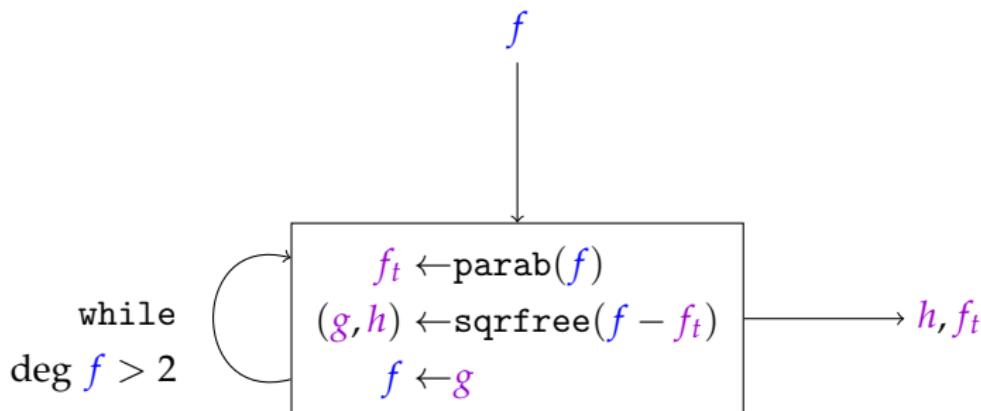
$$f_t = X^2$$

$$f - f_t = (X^2 + 2X + 1)(X + 1)^2$$

# univsos: Algorithm [Schweighofer 99]

---

- **Input:**  $f \geq 0 \in \mathbb{Q}[X]$  of degree  $d \geq 2$
- **Output:** SOS decomposition with coefficients in  $\mathbb{Q}$



# univsos: Output Bitsize

---

## Theorem

Let  $0 < f \in \mathbb{Q}[X]$  with bitsize  $\tau$ ,  $\deg f = d$ .

The output bitsize  $\tau'$  of univsos on  $f$  is  $\mathcal{O}\left((\frac{d}{2})^{\frac{3d}{2}}\tau\right)$ .

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The output bitsize  $\tau'$  of univsos on  $f$  is  $\mathcal{O}\left((\frac{d}{2})^{\frac{3d}{2}}\tau\right)$ .

## Proof.

💡 Worst-case:  $k = d/2$  induction steps

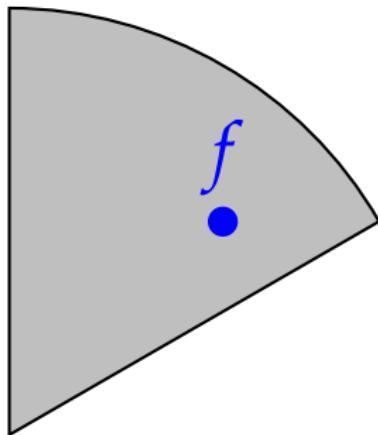
$$\implies \tau' = \mathcal{O}(\tau + k^3\tau + (k-1)^3k^3\tau + \dots + (k!)^3\tau)$$



## intsos with $n \geq 1$ : Perturbation

---

$\Sigma$



### PERTURBATION idea

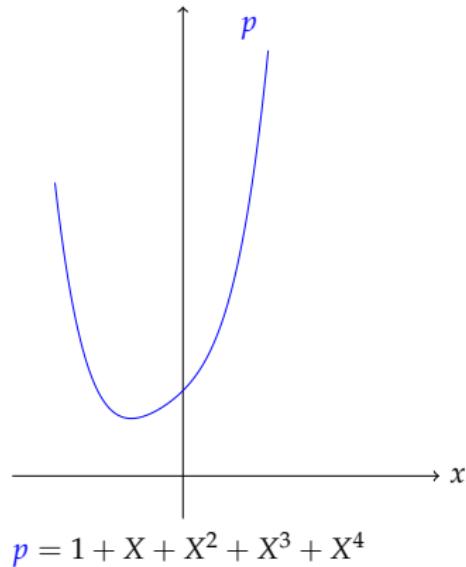
💡 Approximate SOS Decomposition

$$f(X) - \varepsilon \sum_{\alpha \in \mathcal{P}/2} X^{2\alpha} = \tilde{\sigma} + \textcolor{violet}{u}$$

## intsos with $n = 1$ [Chevillard et. al 11]

---

$p \in \mathbb{Q}[X], \deg p = d = 2k, p > 0$



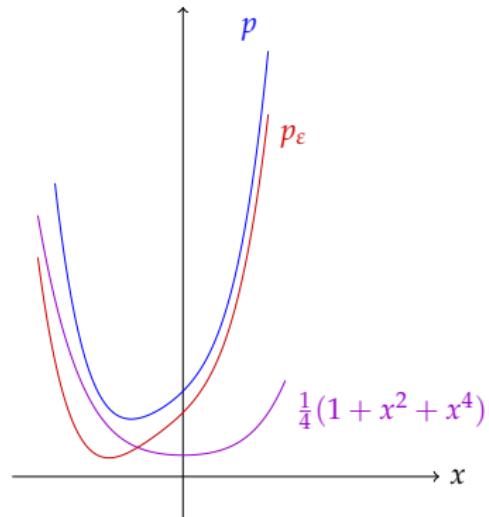
# intsos with $n = 1$ [Chevillard et. al 11]

---

$p \in \mathbb{Q}[X]$ ,  $\deg p = d = 2k$ ,  $p > 0$

💡 PERTURB: find  $\varepsilon \in \mathbb{Q}$  s.t.

$$p_\varepsilon := p - \varepsilon \sum_{i=0}^k X^{2i} > 0$$



$$p = 1 + X + X^2 + X^3 + X^4$$

$$\varepsilon = \frac{1}{4}$$

$$p > \frac{1}{4}(1 + X^2 + X^4)$$

## intsos with $n = 1$ [Chevillard et. al 11]

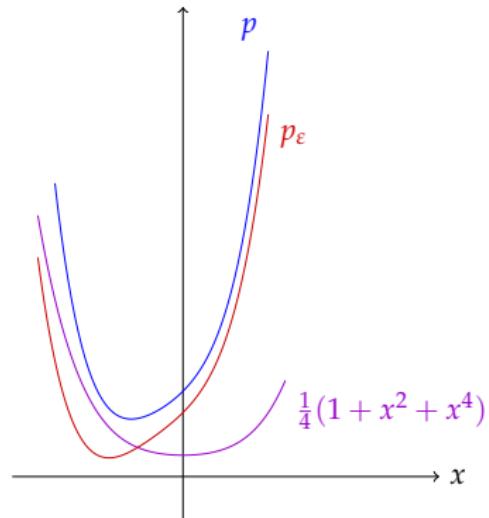
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💡 SDP Approximation:

$$p - \varepsilon \sum_{i=0}^k X^{2i} = \tilde{\sigma} + u$$



$$p = 1 + X + X^2 + X^3 + X^4$$

💡 ABSORB: small enough  $u_i$

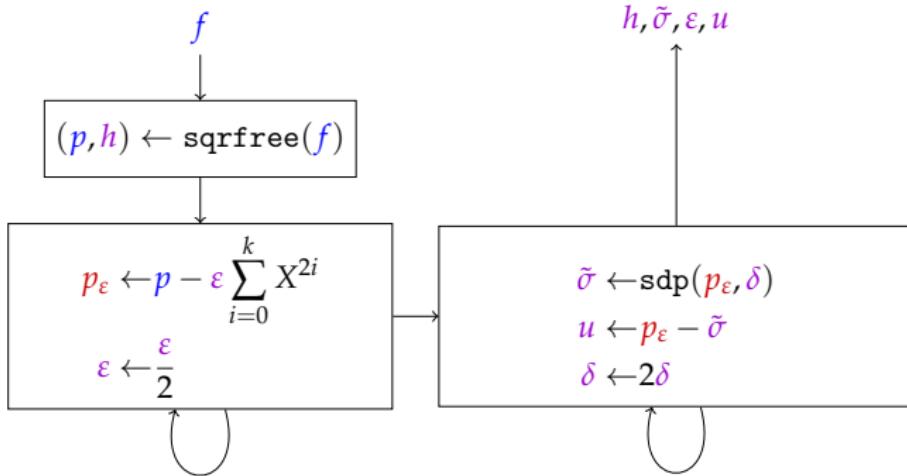
$$\implies \varepsilon \sum_{i=0}^k X^{2i} + u \text{ SOS}$$

$$\varepsilon = \frac{1}{4}$$

$$p > \frac{1}{4}(1 + X^2 + X^4)$$

# intsos with $n = 1$ and SDP Approximation

- **Input**  $f \geq 0 \in \mathbb{Q}[X]$  of degree  $d \geq 2$ ,  $\varepsilon \in \mathbb{Q}^{>0}$ ,  $\delta \in \mathbb{N}^{>0}$
- **Output**: SOS decomposition with coefficients in  $\mathbb{Q}$



while

$$p_\varepsilon \leq 0$$

while

$$\varepsilon < \frac{|u_{2i+1}| + |u_{2i-1}|}{2} - u_{2i}$$

## intsos with $n = 1$ : Absorbtion

---

💡  $X = \frac{1}{2}[(X + 1)^2 - 1 - X^2]$

💡  $-X = \frac{1}{2}[(X - 1)^2 - 1 - X^2]$

## intsos with $n = 1$ : Absorbtion

---

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$$u_{2i+1} X^{2i+1} = \frac{|u_{2i+1}|}{2} [(X^{i+1} + \operatorname{sgn}(u_{2i+1}) X^i)^2 - X^{2i} - X^{2i+2}]$$

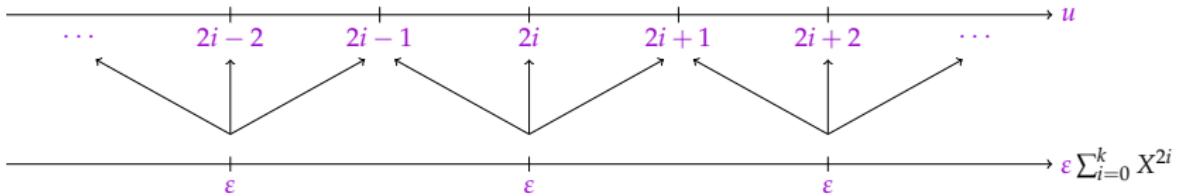
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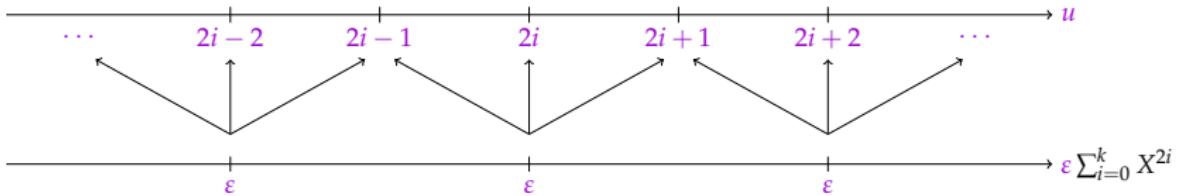
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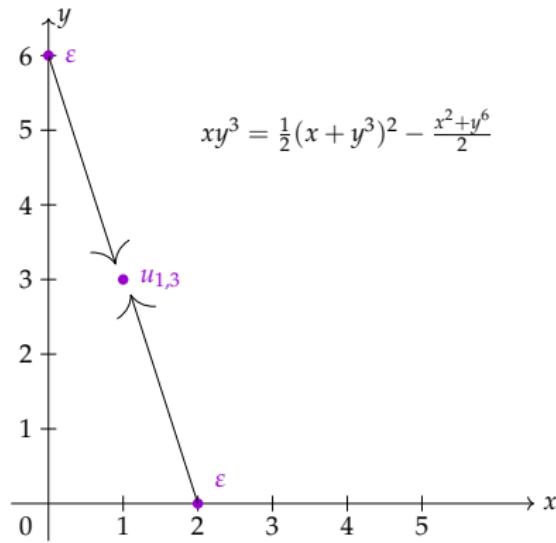
$$\varepsilon \geq \frac{|u_{2i+1}| + |u_{2i-1}|}{2} - u_{2i} \implies \varepsilon \sum_{i=0}^k X^{2i} + u \quad \text{SOS}$$

# intsos with $n \geq 1$ : Absorbtion

---

$$f(X) - \varepsilon \sum_{\alpha \in \mathcal{P}/2} X^{2\alpha} = \tilde{\sigma} + u$$

Choice of  $\mathcal{P}$ ?

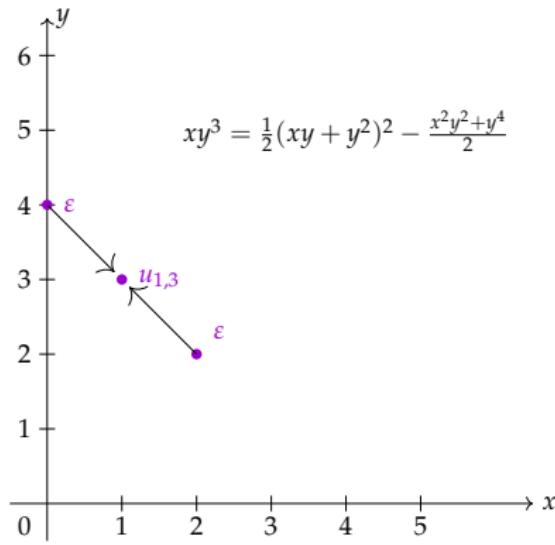


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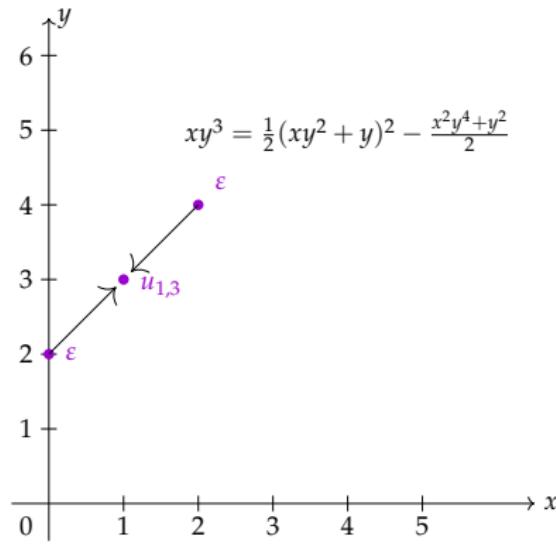


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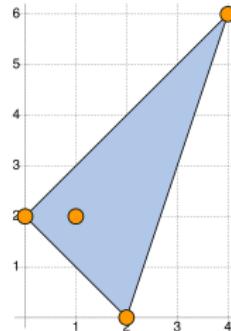
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$$\textcolor{blue}{f}(X) - \varepsilon \sum_{\alpha \in \mathcal{P}/2} X^{2\alpha} = \tilde{\sigma} + u$$

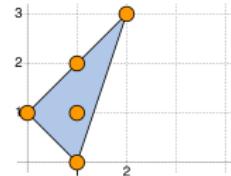
Choice of  $\mathcal{P}$ ?

$$\textcolor{blue}{f} = 4x^4y^6 + x^2 - xy^2 + y^2$$
$$\text{spt}(\textcolor{blue}{f}) = \{(4, 6), (2, 0), (1, 2), (0, 2)\}$$

Newton Polytope  $\mathcal{P} = \text{conv}(\text{spt}(\textcolor{blue}{f}))$

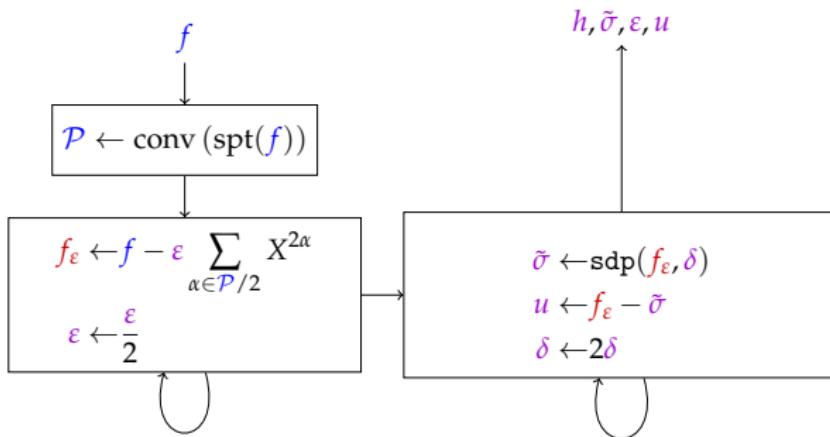


Squares in SOS decomposition  $\subseteq \frac{\mathcal{P}}{2} \cap \mathbb{N}^n$   
[Reznick 78]



# Algorithm int sos

- **Input**  $f \in \mathbb{Q}[X] \cap \dot{\Sigma}[X]$  of degree  $d$ ,  $\varepsilon \in \mathbb{Q}^{>0}$ ,  $\delta \in \mathbb{N}^{>0}$
- **Output**: SOS decomposition with coefficients in  $\mathbb{Q}$



while

$f_\varepsilon \leq 0$

while

$u + \varepsilon \sum_{\alpha \in P/2} X^{2\alpha} \notin \Sigma$

## Algorithm intsos

---

Theorem (Exact Certification Cost in  $\mathring{\Sigma}$ )

$f \in \mathbb{Q}[X] \cap \mathring{\Sigma}[X]$  with  $\deg f = d = 2k$  and bit size  $\tau$

$\implies$  intsos terminates with SOS output of bit size  $\boxed{\tau d^{\mathcal{O}(\textcolor{blue}{n})}}$

# Algorithm intsos

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Proof.

💡  $\{ \varepsilon \in \mathbb{R} : \forall \mathbf{x} \in \mathbb{R}^n, f(\mathbf{x}) - \varepsilon \sum_{\alpha \in \mathcal{P}/2} \mathbf{x}^{2\alpha} \geq 0 \} \neq \emptyset$

Quantifier Elimination [Basu et. al 06]  $\implies \tau(\varepsilon) = \tau d^{\mathcal{O}(n)}$

💡 # Coefficients in SOS output =  $\text{size}(\mathcal{P}/2) = \binom{n+k}{n} \leq d^n$

💡 Ellipsoid algorithm for SDP [Grötschel et. al 93] □

# Algorithm Polyasos

---

$f$  positive definite form has **Polya's** representation:

$$f = \frac{\sigma}{(X_1 + \cdots + X_n)^{2D}} \quad \text{with } \sigma \in \Sigma[X]$$

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## Theorem

$$f (X_1 + \cdots + X_n)^{2D} \in \Sigma[X] \implies f (X_1 + \cdots + X_n)^{2D+2} \in \mathring{\Sigma}[X]$$

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## Theorem (Exact Certification Cost of Polya's representations)

$f \in \mathbb{Q}[X]$  positive definite form with  $\deg f = d$  and bit size  $\tau$

$$\implies D \leq 2^{\tau d^{\mathcal{O}(n)}} \quad \text{OUTPUT BIT SIZE} = \boxed{\tau D^{\mathcal{O}(n)}}$$

## Algorithm Putinarsos

---

**Assumption:**  $\exists i$  s.t.  $g_i = 1 - \|X\|_2^2$   
 $f > 0$  on  $\mathbf{K} := \{\mathbf{x} : g_j(\mathbf{x}) \geq 0\}$  has **Putinar's representation**:

$$f = \sigma_0 + \sum_j \sigma_j g_j \quad \text{with } \sigma_j \in \Sigma[X], \deg \sigma_j \leq 2D$$

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OUTPUT BIT SIZE =  $\boxed{\tau D^{\mathcal{O}(n)}}$

## Algorithm optsonc: numerical steps

---

### SONC (SUMS OF NONNEGATIVE CIRCUITS)

- **Input**  $f = \sum_{\alpha} b_{\alpha} X^{\alpha}$  of degree  $d$ ,  $\hat{\delta} \in \mathbb{Q}^{>0}$ ,  $\tilde{\delta} \in \mathbb{Q}^{>0}$   
Monomial squares = MoSq ( $f$ )    Complement = NoSq ( $f$ )

# Algorithm optsonc: numerical steps

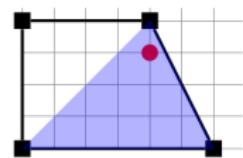
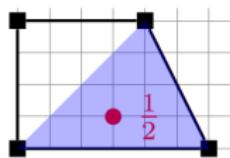
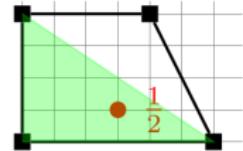
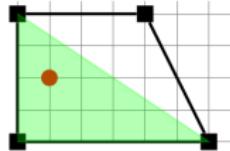
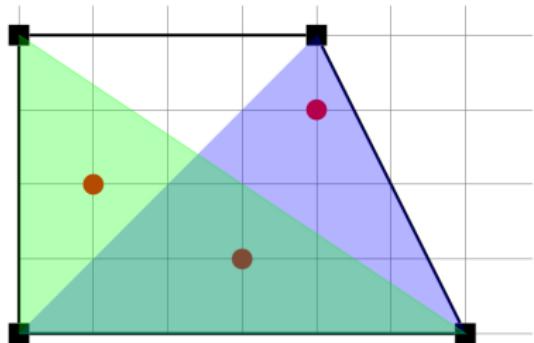
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Monomial squares = MoSq ( $f$ )    Complement = NoSq ( $f$ )

- 1 **Cover** each  $\beta \in \text{NoSq}(f)$  to get nonnegative circuit  $f_{\beta}$

$$\implies \lambda^{\beta} \geqslant 0 \text{ with } \sum_{\alpha \in \text{MoSq}(f)} \lambda_{\alpha}^{\beta} \cdot \alpha = \beta$$



# Algorithm optsonc: numerical steps

---

- **Input**  $f = \sum_{\alpha} b_{\alpha} X^{\alpha}, \hat{\delta}, \tilde{\delta}$

## 2 Numerical resolution of GEOMETRIC PROGRAM

$$f_{\text{SONC}} = \min_{G > 0} \sum_{\beta \in \text{NoSq}(f)} G_{\beta,0}$$

$$\text{s.t. } \sum_{\beta \in \text{NoSq}(f)} G_{\beta,\alpha} \leq b_{\alpha}, \quad \alpha \in \text{MoSq}(f), \alpha \neq 0$$

$$\prod_{\alpha \in \text{Cov}^{\beta}} \left( \frac{G_{\beta,\alpha}}{\lambda_{\alpha}^{\beta}} \right)^{\lambda_{\alpha}^{\beta}} = -b_{\beta}, \quad \beta \in \text{NoSq}(f)$$

# Algorithm optsonc: symbolic steps

---

■ **Input**  $f = \sum_{\alpha} b_{\alpha} X^{\alpha}, \hat{\delta}, \tilde{\delta}$

GEOMETRIC PROGRAM provides “IN THEORY”

$$f_{\beta} = \sum_{\alpha \in \text{Cov}^{\beta}} G_{\beta,\alpha} \cdot X^{\alpha} + b_{\beta} X^{\beta}, f + \sum_{\beta} G_{\beta,0} - b_0 = \sum_{\beta} f_{\beta} \geq 0$$

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3 **Rounding step**  $\hat{G} \leftarrow \text{round}(\tilde{G}, \hat{\delta})$

4 **Projection step**

$$G_{\beta,\alpha} \leftarrow b_{\alpha} \cdot \hat{G}_{\beta,\alpha} / \sum_{\beta' \in \text{NoSq}(f)} \hat{G}_{\beta',\alpha}$$

$$\tilde{G}_{\beta,0} \leftarrow \lambda_0^{\beta} \left( -b_{\beta} \cdot \prod_{\alpha \in \text{Cov}^{\beta}} \left( \frac{\lambda_{\alpha}^{\beta}}{G_{\beta,\alpha}} \right)^{\lambda_{\alpha}^{\beta}} \right)^{\frac{1}{\lambda_0^{\beta}}}$$

$$\hat{G}_{\beta,0} \leftarrow \text{round} \uparrow (\tilde{G}_{\beta,0}, \hat{\delta})$$

# Algorithm optsonc: symbolic steps

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💡 
$$f \geq b_0 - \sum_{\beta} \hat{G}_{\beta,0}$$

# SOS Benchmarks

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- rounding-projection (SOS) [Peyrl-Parrilo]
- RAGLib (critical points) [Safey El Din]
- SamplePoints (CAD) [Moreno Maza-Alvandi et al.]

Id	$n$	$d$	RealCertify $\tau_1$ (bits)	RoundProject $\tau_2$ (bits)	RAGLib $t_3$ (s)	CAD $t_4$ (s)
			$t_1$ (s)	$t_2$ (s)		
$f_{20}$	2	20	745 419	110.	78 949 497	141.
$M$	3	8	17 232	0.35	18 831	0.29
$f_2$	2	4	1 866	0.03	1 031	0.04
$f_6$	6	4	56 890	0.34	475 359	0.54
$f_1$	10	4	344 347	2.45	8 374 082	4.59

# SONC vs SAGE

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terms	bit size		time	
	optsonc	optsage	optsonc	optsage
6	432	1005	0.06	0.26
9	806	2696	0.19	0.66
12	1261	5568	0.37	1.29
20	2592	19203	0.64	4.00
24	3826	32543	0.97	6.66
30	5029	53160	1.34	10.58
50	10622	167971	3.95	32.78

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💡 “IN PRACTICE” optsonc faster and more concise than optsage

# SONC vs SAGE

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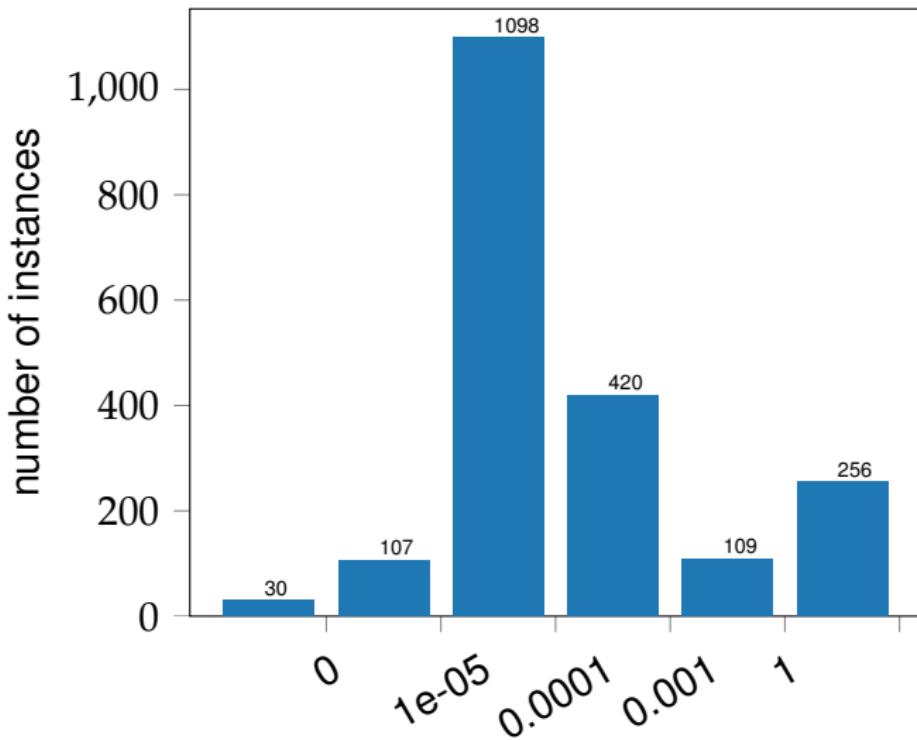
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💡 “IN PRACTICE” optsonc faster and more concise than optsage

💡 “IN THEORY” optsonc less accurate than optsage

# SONC: Gap between Numeric & Symbolic

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# Conclusion and Perspectives

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Input  $f$  on  $\mathbf{K}$  with  $\deg f = d$  and bit size  $\tau$

Algo	Input	$\mathbf{K}$	OUTPUT	BIT SIZE
intsos	$\mathring{\Sigma}$	$\mathbb{R}^n$		
int sage	$\mathring{\mathcal{C}}_{\text{SAGE}}$	$\mathbb{R}^n$		$\tau d^{\mathcal{O}(n)}$

# Conclusion and Perspectives

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Input  $f$  on  $\mathbf{K}$  with  $\deg f = d$  and bit size  $\tau$

Algo	Input	$\mathbf{K}$	OUTPUT	BIT SIZE
intsos	$\mathring{\Sigma}$	$\mathbb{R}^n$		
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- 💡 How to handle degenerate situations?
- 💡 Arbitrary precision SDP/GP/REP solvers
- 💡 Extension to other relaxations

**Crucial need for polynomial systems certification  
Available PhD/Postdoc Positions**



# Thank you for your attention!

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RealCertify      POEM

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