

# A Macaulay 2 Package for Computing Rational Sum of Squares Decompositions

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# Outline

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1. Introduction: Sum of squares (SOS)
2. Computing rational SOS
3. Macaulay 2 SOS package
4. Outlook

# Polynomial Nonnegativity

Given a polynomial  $p(x) \in \mathbb{R}[x_1, \dots, x_n]$  of degree  $d$ , consider the question

$$p(x) \geq 0 \quad \text{for all } x \in \mathbb{R}^n \quad ?$$

If true,  $p(x)$  is called **positive semidefinite (PSD)**.

- ▶ Problem NP-hard, but decidable
- ▶ Many applications

**A sufficient condition:** there exist polynomials  $g_i(x)$  s.t.

$$p(x) = \sum_i g_i(x)^2,$$

i. e.,  $f(x)$  can be written as a **sum of squares (SOS)**.

# SOS - Background Information

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Hilbert 1888: SOS=PSD only in 3 special cases

- ▶  $d = 1$ , i.e., univariate polynomials
- ▶  $n = 2$ , i.e., quadratic polynomials
- ▶  $d = 4, n = 2$ , i.e., quartic polynomials in two variables

Famous counter examples for  $\text{SOS} \neq \text{PSD}$  found by

- ▶ Motzkin, Robinson, Choi and Lam

The good news: SOS condition is “easily checkable”.

⇒ Semidefinite Programming (SDP)

goes back to Shor (1987) and Choi, Lam, Reznick (1995)

# SDP Formulation

Suppose  $p(x) \in \mathbb{R}[x_1, \dots, x_n]$  with degree  $2d$ .

Express  $p(x)$  as a quadratic form in the monomials:

$$p(x) = z^T Q z$$

$z$ : vector of monomials of degree less than or equal to  $d$

- ▶  $Q$  not unique
- ▶ Matching coefficients yields linear constraints on  $Q$   
(denote this affine space by  $\mathcal{L}$ )

$p(x)$  is SOS if and only if there exists a  $Q$  s.t.:

$$\begin{array}{rcl} Q & \succeq & 0 \\ p(x) & = & z^T Q z \end{array}$$

# SDP Formulation

$$\begin{array}{rcl} Q & \succeq & 0 \\ p(x) & = & z^T Q z \end{array}$$

This is an SDP

- ▶ solvable in **polynomial time** by **interior point methods** (for fixed  $d$  or  $n$ )
- ▶ efficient free solvers available (e.g. SeDuMi, SDPT3, CSDP, ...)
- ▶ problem is **linear in coefficients** of  $p(x)$   
 $\Rightarrow$  optimization over coefficients possible

# Numerical SOS Packages

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Available numerical SOS packages:

- ▶ YALMIP (Löfberg)  
<http://control.ee.ethz.ch/~joloef/yalmip.php>
- ▶ SOSTOOLS (Prajna, Papachristodoulou, Seiler, Parrilo)  
<http://www.cds.caltech.edu/sostools/>

Both packages

- ▶ are free
- ▶ depend on MATLAB
- ▶ give numerical answers

# Rational SOS

- ▶ Interior point solvers attain **only finite accuracy**
- ▶ We want an **exact** solution

## Rational SOS

Given  $p(x) \in \mathbb{Q}[x]$ , if there exist  $g_i(x) \in \mathbb{Q}[x]$  s.t.

$$p(x) = \sum g_i(x)^2,$$

then we call  $p(x)$  a *rational sum of squares*.



# Rational SOS - Some Background

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Is every SOS  $p(x) \in \mathbb{Q}[x]$  a rational SOS?

Some results:

- ▶ Landau 1906: true for  $n = 1$
- ▶ Schweighofer 1999: algorithmic proof,  $n = 1$
- ▶ Hillar 2007: If  $p(x)$  is SOS in  $K[x]$ ,  $K$  being a totally real number field, then  $p(x)$  is a rational SOS.

In general: open problem (to our knowledge)

# Rational SOS - SDP connection

Rational SOS  $\iff$  rational Gram matrix  $Q$

Assume  $Q \in \mathbb{Q}^{k \times k}$ . Consider  $L^T D L$  decomposition:

$$p(x) = z^T Q z = z^T L^T D L z = \sum_{i=1}^k d_i (L_i z)^2 = \sum_{i=1}^k d_i g_i(x)^2,$$

where  $0 < d_i \in \mathbb{Q}$  and  $g_i(x) \in \mathbb{Q}[x]$ .

**Lagrange's Theorem:**  $d_i$  is a sum of at most 4 squares.

$\implies p(x)$  is a sum of at most  $4k$  squares in  $\mathbb{Q}[x]$ .

**Question:** How do we obtain a rational  $Q$ ?

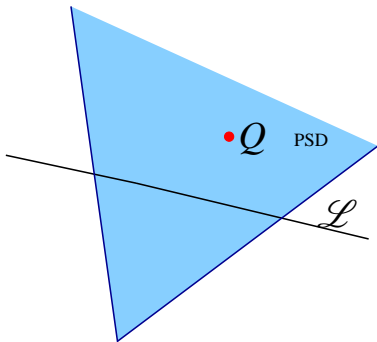
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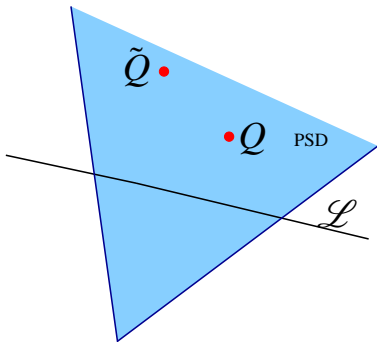
# Computing Rational SOS

1. Compute numerical approximate solution  $Q$



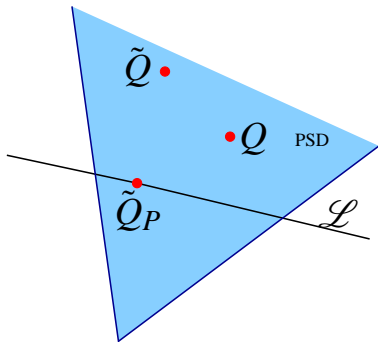
# Computing Rational SOS

1. Compute numerical approximate solution  $Q$
2. Round to rational matrix  $\tilde{Q}$



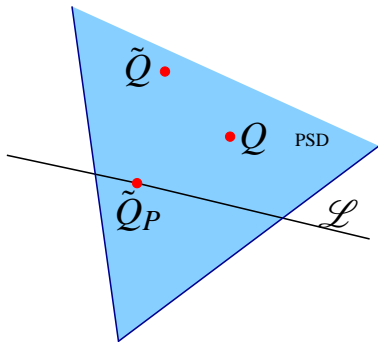
# Computing Rational SOS

1. Compute numerical approximate solution  $Q$
2. Round to rational matrix  $\tilde{Q}$
3. Project back onto  $\mathcal{L}$



# Computing Rational SOS

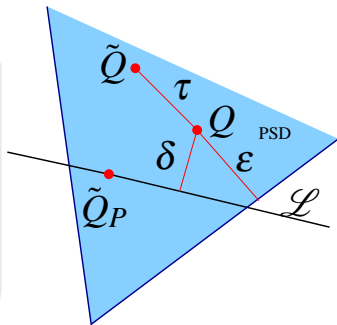
1. Compute numerical approximate solution  $Q$
2. Round to rational matrix  $\tilde{Q}$
3. Project back onto  $\mathcal{L}$
4. Check whether  $Q \succeq 0$ 
  - ▶ yes: done
  - ▶ no: increase precision back to step 2.



# Computing Rational SOS - Remarks

## Proposition

Assume  $\delta < \varepsilon$ ,  $\tau \leq \sqrt{\varepsilon^2 - \delta^2}$ .  
Then the projected matrix  $\tilde{Q}_P$  describes a valid a rational SOS decomposition of  $p(x)$ .



**Hence:** if SDP is strictly feasible (and  $\delta < \varepsilon$ ),  
rational SOS (in principle) always possible by  
using sufficiently many digits.



# Computing Rational SOS - Remarks

Catch: strict feasibility crucial

Partial remedy: eliminate unnecessary monomials in  $z$

- ▶ Newton polytope
- ▶ Symmetry reduction

Caveat: not strictly feasible problems usually fail

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# SOS.m2 - A M2 package

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- ▶ Algorithm implemented in MACAULAY 2  
<http://www.math.uiuc.edu/Macaulay2/>
- ▶ Code may be downloaded for free from  
<http://www.control.ee.ethz.ch/~hpeyrl>

Main package: **SOS.m2**

exports the functions:

- getSOS** computes weighted SOS decomposition
- findSOS** same as getSOS, returns Gram matrix
- sumSOS** just for checks

# Usage – getSOS, sumSOS

Usage:  $(g, d) = \text{getSOS}(f)$

## Example

Is  $p(x, y) = 4x^4 + 4x^3y - 7x^2y^2 - 2xy^3 + 10y^4$  a SOS?

# Usage – getSOS, sumSOS

Usage:  $(g,d) = \text{getSOS}(f)$

## Example

Is  $p(x,y) = 4x^4 + 4x^3y - 7x^2y^2 - 2xy^3 + 10y^4$  a SOS?

```
i1 : loadPackage"SOS";  
i2 : R = QQ[x,y];  
i3 : p = 4*x^4+4*x^3*y-7*x^2*y^2  
      -2*x*y^3+10*y^4;  
i4 : (g,d) = getSOS (p)
```

... omitted output ...

# Usage – getSOS, sumSOS

Usage:  $(g, d) = \text{getSOS}(f)$

## Example

Is  $p(x, y) = 4x^4 + 4x^3y - 7x^2y^2 - 2xy^3 + 10y^4$  a SOS?

$$\text{oS4} = \left( \left( -\frac{11}{25}x^2 - \frac{1}{10}x^2y + y^2, x^2 + \frac{65}{86}x^2y, \right. \right. \\ \left. \left. x^2y \right), \left( 10, \frac{258}{125}, \frac{112}{215} \right) \right)$$

# Usage – getSOS, sumSOS

Usage:  $(g, d) = \text{getSOS}(f)$

## Example

Is  $p(x, y) = 4x^4 + 4x^3y - 7x^2y^2 - 2xy^3 + 10y^4$  a SOS?

Hence:

$$p = 10\left(-\frac{11}{25}x^2 - \frac{1}{10}xy + y^2\right)^2 + \frac{258}{125}\left(x^2 + \frac{65}{86}xy\right)^2 + \frac{112}{215}(xy)^2$$

Let's check ...

# Usage – getSOS, sumSOS

Usage:  $p = \text{sumSOS}(g, d)$

## Example

Is  $p(x, y) = 4x^4 + 4x^3y - 7x^2y^2 - 2xy^3 + 10y^4$  a SOS?

```
i5 : sumSOS (g,d)
```

```
o5 = 4x4 + 4x3y - 7x2y2 - 2x*y3 + 10y4
```

```
o5 : QQ [x, y]
```



# Usage – More Possibilities

- ▶ Linearly parameterized SOS

Usage:  $(g, d, tQ) = \text{getSOS}(p, t)$

$t$  a list of parameters

$tQ$  a list in  $\mathbb{Q}$  s.t.  $p$  is SOS

- ▶ Minimizing a linear functional

Usage:  $(g, d, tQ) = \text{getSOS}(p, t, ob, rndTol=>x)$

$t$  a list of parameters

$ob$  a linear function in  $t$

$rndTol$  optional minimal rounding precision in  $x$   
binary digits

$tQ$  a list in  $\mathbb{Q}$  s.t.  $p$  is SOS

# Usage – More Possibilities

## Example

Compute a lower bound on  $p(x) = x^4 - 2x$

```
i5 : R = QQ[x,t];  
i6 : f = x^4 - 2*x - t;  
i7 : (g,d,tval) = getSOS (f,{t},-t,rndTol=>10)  
  
... omitted output ...
```

# Usage – More Possibilities

## Example

Compute a lower bound on  $p(x) = x^4 - 2x$

$$\begin{aligned}
 \text{lo4} = & \left( \left\{ x - \frac{20480}{25803}, -\frac{665794809}{419422495}x^2 + 1, x^2 \right\}, \right. \\
 & \left. \left\{ \frac{25803}{20480}, \frac{83884499}{211378176}, \frac{41938573}{17179545395200} \right\}, \left\{ -\frac{9753}{8192} \right\} \right)
 \end{aligned}$$

## Planned features:

- ▶ Support multiple SOS constraints  
(for e.g., Lyapunov stability problems)
- ▶ Interface 3rd party SDP solvers (e.g., CSDP)
- ▶ Exploit symmetries to reduce SDP size
- ▶ Your feature request ???

# The last slide

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Thanks for your attention!