# Real solving polynomial equations with semidefinite programming

Jean Bernard Lasserre - Monique Laurent - Philipp Rostalski

LAAS, Toulouse - CWI, Amsterdam - ETH, Zürich

LAW 2008

Real solving polynomial equations with semidefinite programming - p.1

Given polynomials  $h_1, \ldots, h_m \in \mathbb{R}[\mathbf{x}] = \mathbb{R}[\mathbf{x}_1, \ldots, \mathbf{x}_n]$ 

- Compute all common real roots (assuming finitely many), i.e. compute the real variety  $V_{\mathbb{R}}(I)$  of the ideal  $I := (h_1, \ldots, h_m)$
- Find a basis of the real radical ideal  $I(V_{\mathbb{R}}(I))$

$$V_{\mathbb{R}}(I) := \{ v \in \mathbb{R}^n \mid f(v) = 0 \ \forall f \in I \}$$
$$I(V_{\mathbb{R}}(I)) := \{ f \in \mathbb{R}[\mathbf{x}] \mid f(v) = 0 \ \forall v \in V_{\mathbb{R}}(I) \}$$
$$\underset{\{ f \in \mathbb{R}[\mathbf{x}] \mid \exists m \in \mathbb{N} \ s_i \in \mathbb{R}[\mathbf{x}] \ f^{2m} + \sum_i s_i^2 \in I \}$$
Real Nullstellensatz

- **1.** A semidefinite characterization of  $I(V_{\mathbb{R}}(I))$ [as the kernel of some positive semidefinite *moment matrix*]
- **2.** Assuming  $|V_{\mathbb{R}}(I)| < \infty$ , an algorithm for finding:
- a generating set (**border** or **Gröbner basis**) of  $I(V_{\mathbb{R}}(I))$
- the real variety  $V_{\mathbb{R}}(I)$

## **Remarks about the method:**

- real algebraic in nature: no complex roots computed
- works if  $V_{\mathbb{R}}(I)$  is finite (even if  $V_{\mathbb{C}}(I)$  is not)
- no preliminary Gröbner basis of I is needed
- *numerical*, based on semidefinite programming (SDP)

**1.** The moment-matrix method for  $V_{\mathbb{R}}(I)$ 

- **2.** Adapt the moment-matrix method for  $V_{\mathbb{C}}(I)$  [drop PSD]
- **3.** Relate to the 'prolongation-projection' algorithm of Zhi and Reid for  $V_{\mathbb{C}}(I)$

**4.** Adapt the prolongation-projection algorithm for  $V_{\mathbb{R}}(I)$  [add PSD]

Given an ideal  $I \subseteq \mathbb{R}[\mathbf{x}]$  with  $|V_{\mathbb{C}}(I)| < \infty$ ,

find the (complex) variety  $V_{\mathbb{C}}(I)$  and the radical ideal  $I(V_{\mathbb{C}}(I))$ .

Linear algebra in the finite dimensional space  $\mathbb{R}[\mathbf{x}]/I$ 

 $\rightsquigarrow$  Need a linear basis of  $\mathbb{R}[\mathbf{x}]/I$  and a normal form algorithm

 $V_{\mathbb{C}}(I)$  can be computed e.g. with:

- Linear algebra methods: ~> **Eigenvalue method** [Stetter-Möller, Stickelberger, Rouillier]
- Homotopy methods [Verschelde] . . .

Seidenberg [1974]:  $I(V_{\mathbb{C}}(I)) = (I \cup \{q_1, \dots, q_n\})$ , where  $q_i$  is the square-free part of  $p_i$ , the monic generator of  $I \cap \mathbb{R}[\mathbf{x}_i]$ .

**Stickelberger theorem:** 

Let  $m_f$  be the *'multiplication by f' linear operator* in  $\mathbb{R}[\mathbf{x}]/I$ .

- **1.** The eigenvalues of  $m_f$  are  $\{f(v) \mid v \in V_{\mathbb{C}}(I)\}$ .
- **2.** The eigenvectors of  $m_f^T$  give the points  $v \in V_{\mathbb{C}}(I)$ .

$$M_f^T \zeta_{\mathcal{B},v} = f(v) \zeta_{\mathcal{B},v} \quad \forall v \in V_{\mathbb{C}}(I)$$

where  $M_f$  is the matrix of  $m_f$  in a base  $\mathcal{B}$  of  $\mathbb{R}[\mathbf{x}]/I$  and  $\zeta_{\mathcal{B},v} := (b(v))_{b \in \mathcal{B}}$ 

Moreover, when  $\mathcal{B}$  is a set of monomials and  $1 \in \mathcal{B}$ , a **border basis** of *I* can be read directly from the multiplication matrices  $M_{x_1}, \ldots, M_{x_n}$ .

## Finding a linear basis $\mathcal{B}$ of $\mathbb{R}[\mathbf{x}]/I$ and a basis G of the ideal I

• Typically: *G* is a **Gröbner basis** and *B* is the set of **standard monomials** for a given monomial ordering (e.g. via Buchberger's algorithm)

• More generally: Assume  $\mathcal{B} = \{b_1 = 1, b_2, \dots, b_N\}$  is a set of monomials with **border**  $\partial \mathcal{B} := (\mathbf{x}_1 \mathcal{B} \cup \ldots \cup \mathbf{x}_n \mathcal{B}) \setminus \mathcal{B}$ . Write any border monomial

$$\mathbf{x}_i b_j = \underbrace{\sum_{k=1}^N a_k^{(ij)} b_k}_{\in \mathbf{Span}(\mathcal{B})} + \underbrace{g^{(ij)}}_{\in I}$$

**Then:**  $G := \{g^{(ij)} \mid \mathbf{x}_i b_j \in \partial B\}$  is a (border) basis of I and carries the *same information* as the multiplication matrices  $M_{\mathbf{x}_1}, \ldots, M_{\mathbf{x}_n}$ Real solving polynomial equations with semidefinite programming – p.7 **Counting real roots with the Hermite quadratic form** 

For 
$$f \in \mathbb{R}[\mathbf{x}]$$

Hermite bilinear form:

**Theorem:** For f = 1

 $\operatorname{rank}(H_1) = |V_{\mathbb{C}}(I)|, \operatorname{Sign}(H_1) = |V_{\mathbb{R}}(I)|, \operatorname{Rad}(H_1) = I(V_{\mathbb{C}}(I))$ 

- rank $(H_f) = |\{v \in V_{\mathbb{C}}(I) \mid f(v) \neq 0\}|$
- Sign( $H_f$ ) =  $|\{v \in V_{\mathbb{R}}(I) \mid f(v) > 0\}| - |\{v \in V_{\mathbb{R}}(I) \mid f(v) < 0\}|$

To find  $V_{\mathbb{R}}(I)$  and a basis of the real radical ideal  $I(V_{\mathbb{R}}(I))$  ...

... it suffices to have a linear basis  $\mathcal{B}$  of  $\mathbb{R}[\mathbf{x}]/I(V_{\mathbb{R}}(I))$  and the **multiplication matrices** in  $\mathbb{R}[\mathbf{x}]/I(V_{\mathbb{R}}(I))$  !

**New tool: Moment matrices** 

$$y \in \mathbb{R}^{\mathbb{N}_{2s}^n} \rightsquigarrow M_s(y) := (y_{\alpha+\beta})_{\alpha,\beta \in \mathbb{N}_s^n}$$

$$\mathbb{N}_s^n := \{ \alpha \in \mathbb{N}^n \mid |\alpha| = \sum_i \alpha_i \le s \}$$
  
\$\sim monomials \$\mathbf{x}^\alpha\$ of degree \$\le s\$

Motivation: For  $y = (v^{\alpha})_{\alpha \in \mathbb{N}_{2s}^n} =: \zeta_{2s,v}$  where  $v \in \mathbb{R}^n$ 

$$M_s(y) = \zeta_{s,v} \zeta_{s,v}^T \succeq 0$$
 and  $\operatorname{Ker} M_s(y) \subseteq I(v)$ 

**Real roots of**  $I = (h_1, \ldots, h_m)$  **and PSD moment matrices** 

**Lemma:** For  $v \in V_{\mathbb{R}}(I)$  and  $t \ge D := \max_j \deg(h_j)$ the vector  $y = \zeta_{t,v} = (v^{\alpha})_{|\alpha| \le t}$  satisfies:

• the linear constraints (LC):  $[v \in V_{\mathbb{C}}(I)]$ 

 $y^T(h_j \mathbf{x}^{\alpha}) = 0 \quad \forall j = 1 \dots m \; \forall \alpha \text{ s.t. } |\alpha| + \deg(h_j) \le t$ 

• the PSD constraint:

$$M_{\lfloor t/2 \rfloor}(y) \succeq 0$$

$$[v \in \mathbb{R}^n]$$

**Set:** 
$$\mathcal{K}_t := \{ y \in \mathbb{R}^{\mathbb{N}_t^n} \mid (\mathbf{LC}), M_{\lfloor t/2 \rfloor}(y) \succeq 0 \}$$

**Obviously:**  $\mathcal{K}_t \supseteq \operatorname{cone}(\zeta_{t,v} \mid v \in V_{\mathbb{R}}(I))$ 

**Theorem:**  $\exists t \geq s \geq D \quad \pi_s(\mathcal{K}_t) = \operatorname{cone}(\zeta_{s,v} \mid v \in V_{\mathbb{R}}(I))$ 

Semidefinite characterization of  $I(V_{\mathbb{R}}(I))$ 

**Theorem 1:** Let y be a **generic element** of  $\mathcal{K}_t$ , i.e. y lies in the **relative interior** of the cone  $\mathcal{K}_t$ . Then  $(\operatorname{Ker} M_{\lfloor t/2 \rfloor}(y)) \subseteq I(V_{\mathbb{R}}(I))$ with equality for t large enough.

• Geometric property of SDP:  $y \text{ is generic} \iff \operatorname{rank} M_{\lfloor t/2 \rfloor}(y) \text{ is maximum}$  $\iff \operatorname{Ker} M_{\lfloor t/2 \rfloor}(y) \subseteq \operatorname{Ker} M_{\lfloor t/2 \rfloor}(z) \quad \forall z \in \mathcal{K}_t$ 

**Thus:** for  $v \in V_{\mathbb{R}}(I)$ ,  $\operatorname{Ker} M_{\lfloor t/2 \rfloor}(y) \subseteq \operatorname{Ker} M_{\lfloor t/2 \rfloor}(\zeta_{t,v}) \subseteq I(v)$ .

• Let  $\{g_1, \ldots, g_L\}$  be a basis of  $I(V_{\mathbb{R}}(I))$ . Real Nullstellensatz:  $g_l^{2m} + \sum_i s_i^2 = \sum_{j=1}^m u_j h_j$ . This implies:  $g_l \in \operatorname{Ker} M_{\lfloor t/2 \rfloor}(y)$  for t large enough. **Theorem 2:** Let *y* be a **generic** element of  $\mathcal{K}_t$ . Assume one of the following two *flatness conditions* holds:

(F1) rank
$$M_s(y) = \operatorname{rank} M_{s-1}(y)$$
 for some  $D \le s \le \lfloor t/2 \rfloor$ 

(Fd) rank $M_s(y) = \operatorname{rank} M_{s-d}(y)$  for some  $d = \lceil D/2 \rceil \le s \le \lfloor t/2 \rfloor$ .

## Then:

- $I(V_{\mathbb{R}}(I)) = (\operatorname{Ker} M_s(y))$
- Any base  $\mathcal{B}$  of the column space of  $M_{s-1}(y)$ is a base of  $\mathbb{R}[\mathbf{x}]/I(V_{\mathbb{R}}(I))$
- The multiplication matrices can be constructed from  $M_s(y)$ .

## Sketch of proof: Assume $\operatorname{rank} M_s(y) = \operatorname{rank} M_{s-1}(y)$

- Thm [Curto-Fialkow 1996]  $\pi_{2s}(y)$  has a flat extension  $\tilde{y} \in \mathbb{R}^{\mathbb{N}^n}$ , i.e. such that  $\operatorname{rank} M(\tilde{y}) = \operatorname{rank} M_s(y)$ .
- Thm [La 2005] As  $M(\tilde{y}) \succeq 0$ ,  $(\text{Ker}M_s(y))=\text{Ker}M(\tilde{y})$  is a real radical 0-dimensional ideal.

• 
$$I \underbrace{\subseteq}_{(LC)} (\operatorname{Ker} M_s(y)) \underbrace{\subseteq}_{y \text{ generic}} I(V_{\mathbb{R}}(I))$$

**Thus:**  $(\operatorname{Ker} M_s(y)) = I(V_{\mathbb{R}}(I))$ 

•  $\mathcal{B}$  indexes a base of  $M_{s-1}(y) \Longrightarrow \mathcal{B}$  indexes a base of  $M(\tilde{y}) \Longrightarrow \mathcal{B}$  is a base of  $\mathbb{R}[\mathbf{x}]/\mathrm{Ker}M(\tilde{y}) = \mathbb{R}[\mathbf{x}]/I(V_{\mathbb{R}}(I))$ 

 $\rightsquigarrow$  Use linear dependencies in  $M_s(y)$  to construct the multiplication matrices.

The moment-matrix algorithm for  $V_{\mathbb{R}}(I)$ 

Input:  $h_1, \ldots, h_m \in \mathbb{R}[\mathbf{x}]$ **Output:**  $\mathcal{B}$  base of  $\mathbb{R}[\mathbf{x}]/I(V_{\mathbb{R}}(I))$ The multiplication matrices  $M_{\mathbf{x}_i}$  in  $\mathbb{R}[\mathbf{x}]/I(V_{\mathbb{R}}(I))$ 

Algorithm: For t > D

**Step 1:** Compute a generic element  $y \in \mathcal{K}_t$ .

**Step 2:** Check if (F1) or (Fd) holds.

If yes, return a column basis  $\mathcal{B}$  of  $M_{s-1}(y)$  and  $M_{\mathbf{x}_i} = M_{\mathcal{B}}^{-1} P_i$ ,

- $M_{\mathcal{B}}$ := principal submatrix of  $M_{s-1}(y)$  indexed by  $\mathcal{B}$
- $P_i$ := submatrix of  $M_s(y)$  with rows in  $\mathcal{B}$  and columns in  $\mathbf{x}_i \mathcal{B}$ .

If no, go to Step 1 with  $t \rightarrow t + 1$ .

# **Theorem:** The algorithm terminates. Real solving polynomial equations with semidefinite programming – p.14

#### The algorithm terminates: (F1) holds for t large enough.

- For  $t \ge t_0$ ,  $\operatorname{Ker} M_{\lfloor t/2 \rfloor}(y)$  contains a Gröbner base  $\{g_1, \ldots, g_L\}$  of  $I(V_{\mathbb{R}}(I))$  for a total degree ordering.
- $\mathcal{B} := \{b_1, \dots, b_N\}$ : set of standard monomials ~> base of  $\mathbb{R}[\mathbf{x}]/I(V_{\mathbb{R}}(I))$ .

Set:  $s := 1 + \max_{b \in \mathcal{B}} \deg(b)$  and assume  $t \ge t_0$ ,  $\lfloor t/2 \rfloor > s$ . For  $|\alpha| \le s$ , write  $x^{\alpha} = \sum_{\substack{i=1 \ deg \le s-1}}^{N} \lambda_i b_i + \sum_{\substack{l=1 \ deg \le |\alpha| \le s < \lfloor t/2 \rfloor}}^{L} u_l g_l$ Thus:  $x^{\alpha} - \sum_{i=1}^{N} \lambda_i b_i \in \operatorname{Ker} M_{\lfloor t/2 \rfloor}(y)$ . That is:  $\operatorname{rank} M_s(y) = \operatorname{rank} M_{s-1}(y)$ .

## **Two small examples**

**Ex. 1:** 
$$I = (h := \mathbf{x}_1^2 + \mathbf{x}_2^2)$$
  
 $V_{\mathbb{R}}(I) = \{0\}, |V_{\mathbb{C}}(I)| = \infty.$   
 $M_1(y) \succeq 0, \ 0 = y^T \vec{h} = y_{20} + y_{02} \Longrightarrow y_{\alpha} = 0 \ \forall \alpha \neq 0.$   
 $\rightsquigarrow$  Any generic  $y \in K_2$  is  $y = (y_0, 0, \dots, 0)$  with  $y_0 > 0.$   
**Thus:**  $(\operatorname{Ker} M_1(y)) = (\mathbf{x}_1, \mathbf{x}_2) = I(V_{\mathbb{R}}(I)).$   
**Ex. 2:**  $I = (h_i := \mathbf{x}_i(\mathbf{x}_i^2 + 1) | i = 1, \dots, n)$   
 $V_{\mathbb{R}}(I) = \{0\}, |V_{\mathbb{C}}(I)| = 3^n.$   
 $M_2(y) \succeq 0, \ 0 = y^T(\mathbf{x}_i \vec{h}_i) = y_{4e_i} + y_{2e_i} \ \forall i \Longrightarrow y_{\alpha} = 0 \ \forall \alpha \neq 0.$   
 $\rightsquigarrow$  Any generic  $y \in K_4$  is  $y = (y_0, 0, \dots, 0)$  with  $y_0 > 0.$   
**Thus:**  $(\operatorname{Ker} M_1(y)) = (\mathbf{x}_1, \dots, \mathbf{x}_n) = I(V_{\mathbb{R}}(I)).$ 

How to find a generic  $y \in \mathcal{K}_t$ , i.e. with rank $M_t(y)$  max. ?

Solve the SDP program:  $\min_{y \in \mathcal{K}_t} 1$  with a SDP solver using the 'extended self-dual embedding property'.

Then the central path converges to a solution in the relative interior of the optimum face, i.e., to a **generic** point  $y \in \mathcal{K}_t$ .

## How to compute ranks of matrices ?

We use SVD decomposition, but this is a sensitive numerical issue ...

## The method may work without (F1) or (Fd):

If rank  $M_{\mathcal{B}}(y) = \operatorname{rank} M_{\mathcal{B} \cup \partial \mathcal{B}}(y)$  and the formal multiplication matrices commute.

**Extension of the moment-matrix algorithm to**  $V_{\mathbb{C}}(I)$ 

Omit the PSD condition and work with the linear space:

 $\mathbf{K}_{t} = \{ y \in \mathbb{R}^{\mathbb{N}_{t}^{n}} \mid y^{T}(h_{j} \mathbf{x}^{\alpha}) = 0 \ \forall j, \alpha \text{ with } |\alpha| + \deg(h_{j}) \leq t \}$ 

The *same* algorithm applies: For  $t \ge D$ 

- Pick generic  $y \in K_t$ , i.e. rank $M_s(y)$  maximum  $\forall s \leq \lfloor t/2 \rfloor$ [choose  $y \in K_t$  randomly]
- Check if the flatness condition (F1) or (Fd) holds.
- If yes, find a basis of  $\mathbb{R}[\mathbf{x}]/J$  where  $J := (\text{Ker}M_s(y))$ satisfies  $I \subseteq J \subseteq I(V_{\mathbb{C}}(I))$  and thus  $V_{\mathbb{C}}(J) = V_{\mathbb{C}}(I)$ .
- If not, iterate with t + 1.

## Find the ideal $(Ker M_s(y)) = I$ in the Gorenstein case

The inclusion  $I \subseteq (\text{Ker}M_s(y)) \subseteq I(V_{\mathbb{C}}(I))$  may be strict for any generic y.

**Example:** For  $I = (\mathbf{x}_1^2, \mathbf{x}_2^2, \mathbf{x}_1\mathbf{x}_2), V_{\mathbb{C}}(I) = \{0\},$  $I(V_{\mathbb{C}}(I)) = (\mathbf{x}_1, \mathbf{x}_2), \dim \mathbb{R}[\mathbf{x}]/I = 3, \dim \mathbb{R}[\mathbf{x}]/I(V_{\mathbb{C}}(I)) = 1,$ while  $\dim \mathbb{R}[\mathbf{x}]/(\operatorname{Ker} M_s(y)) = 2$  for any generic y !

**Recall:** The algebra  $\mathcal{A} := \mathbb{R}[\mathbf{x}]/I$  is *Gorenstein* if there exists a non-degenerate bilinear form on  $\mathcal{A}$  satisfying (f, gh) = (fg, h) $\forall f, g, h \in \mathcal{A}$ , i.e. if there exists  $y \in K_{\infty}$  with I = KerM(y)

**Hence:**  $\exists y \in K_t$  s.t. rank $M_s(y) = \text{rank}M_{s-1}(y)$  and  $I = (\text{Ker}M_s(y))$  IFF  $\mathcal{A}$  is Gorenstein.

**Example: the moment-matrix algorithm for real/complex roots** 

$$I = (\mathbf{x}_1^2 - 2\mathbf{x}_1\mathbf{x}_3 + 5, \ \mathbf{x}_1\mathbf{x}_2^2 + \mathbf{x}_2\mathbf{x}_3 + 1, \ 3\mathbf{x}_2^2 - 8\mathbf{x}_1\mathbf{x}_3), \ D = 3, \ d = 2$$

	t=2	2 3	4	5	6	7	8	9	
s = 0	) 1	1	1	1	1	1	1	1	_
s = 1	. 4	4	4	4	4	4	4	4	no PCD 8 complex roots
s = 2	2		8	8	8	8	8	8	$10 \text{ FSD} \sim 3 \text{ complex roots}$
s = 3	3				11	10	9	8	
s = 4	Ł						12	10	
	•								
		t =	2	3	4 :	56			
-	e = 0	1		1	1	1 1			

**Ranks of**  $M_s(y)$  **for generic**  $y \in K_t$ ,  $\mathcal{K}_t$ :

	t = 2	3	4	5	6	
s = 0	1	1	1	1	1	
s = 1	4	4	4	2	2	with PSD $\rightsquigarrow$ extract 2 real roots
s = 2			8	8	2	
s = 3					10	

8 complex / 2 real roots:

$$v_{1} = \begin{bmatrix} -1.101, -2.878, -2.821 \end{bmatrix}$$

$$v_{2} = \begin{bmatrix} 0.07665 + 2.243i, 0.461 + 0.497i, 0.0764 + 0.00834i \end{bmatrix}$$

$$v_{3} = \begin{bmatrix} 0.07665 - 2.243i, 0.461 - 0.497i, 0.0764 - 0.00834i \end{bmatrix}$$

$$v_{4} = \begin{bmatrix} -0.081502 - 0.93107i, 2.350 + 0.0431i, -0.274 + 2.199i \end{bmatrix}$$

$$v_{5} = \begin{bmatrix} -0.081502 + 0.93107i, 2.350 - 0.0431i, -0.274 - 2.199i \end{bmatrix}$$

$$v_{6} = \begin{bmatrix} 0.0725 + 2.237i, -0.466 - 0.464i, 0.0724 + 0.00210i \end{bmatrix}$$

$$v_{7} = \begin{bmatrix} 0.0725 - 2.237i, -0.466 + 0.464i, 0.0724 - 0.00210i \end{bmatrix}$$

$$v_{8} = \begin{bmatrix} 0.966, -2.813, 3.072 \end{bmatrix}$$

#### **Extracting real roots without (F1) or (Fd)**

 $I = (5\mathbf{x}_1^9 - 6\mathbf{x}_1^5\mathbf{x}_2 + \mathbf{x}_1\mathbf{x}_2^4 + 2\mathbf{x}_1\mathbf{x}_3, -2\mathbf{x}_1^6\mathbf{x}_2 + 2\mathbf{x}_1^2\mathbf{x}_2^3 + 2\mathbf{x}_2\mathbf{x}_3, \mathbf{x}_1^2 + \mathbf{x}_2^2 - 0.265625)$  $D = 9, d = 5, |V_{\mathbb{R}}(I)| = 8, |V_{\mathbb{C}}(I)| = 20$ 

order	rank sequence of	extract. order s	accuracy	comm. error
t	$M_s(y) \ (1 \le s \le \lfloor t/2 \rfloor)$	MON/SVD	MON/SVD	MON/SVD
10	1 4 8 16 25 34			—
12	1 3 9 15 22 26 32			
14	1 3 8 10 12 16 20 24	3(3)/—(—)	0.12786/—	0.00019754/—
16	1 4 <mark>8 8 8</mark> 12 16 20 24	4(3)/3(3)	4.6789e-5/0.00013406	4.7073e-5/0.00075005

Quotient basis:  $\mathcal{B} = \{1, \mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3, \mathbf{x}_1^2, \mathbf{x}_1\mathbf{x}_2, \mathbf{x}_1\mathbf{x}_3, \mathbf{x}_2\mathbf{x}_3\} \rightsquigarrow \text{border basis } G \text{ of size } 10$ 

Real solutions: 
$$\begin{cases} x_1 = (-0.515, -0.000153, -0.0124) & x_2 = (-0.502, 0.119, 0.0124) \\ x_3 = (0.502, 0.119, 0.0124) & x_4 = (0.515, -0.000185, -0.0125) \\ x_5 = (0.262, 0.444, -0.0132) & x_6 = (-2.07e-5, 0.515, -1.27e-6) \\ x_7 = (-0.262, 0.444, -0.0132) & x_8 = (-1.05e-5, -0.515, -7.56e-7) \end{cases}$$

Link with the elimination method of Zhi and Reid

**Theorem:** If (F1) holds, i.e. for some  $D \le s \le \lfloor t/2 \rfloor$ 

 $\operatorname{rank} M_s(y) = \operatorname{rank} M_{s-1}(y)$  for generic  $y \in K_t$ ,

then 
$$\dim \pi_{2s}(K_t) = \dim \pi_{2s-1}(K_t) = \dim \pi_{2s}(K_{t+1})$$

**Theorem (based on [Zhi-Reid 2004]):** If for some  $D \le s \le t$ 

(ZR) 
$$\dim \pi_s(K_t) = \dim \pi_{s-1}(K_t) = \dim \pi_s(K_{t+1})$$

then one can construct a base of  $\mathbb{R}[\mathbf{x}]/I$  and the multiplication matrices in  $\mathbb{R}[\mathbf{x}]/I$  [and thus extract  $V_{\mathbb{C}}(I)$ ].

**Hence:** The Zhi-Reid criterion (ZR) may be satisfied earlier than the flatness criterion (F1).

	<i>t</i> =	= 2	3	4	5	6	7	8	9
s =	0	1	1	1	1	1	1	1	1
s =	1 4	1	4	4	4	4	4	4	4 $\operatorname{rank} M_3(y) = \operatorname{rank} M_2(y)$
s =	2			8	8	8	8	8	8 for $y \in K_9$
s =	3					11	10	9	8
s =	4							12	10
	t = 3	4		5	6	7	8	9	
s = 1	4	4	Z	1	4	4	4	4	
s = 2	8	8	8	3	8	8	8	8	
s = 3	11	10	ç	)	8	8	8	8	$\dim -(K_{-})$
s = 4		12	1	0	9	8	8	8	$\dim \pi_3(\mathcal{K}_6)$
s = 5			1	2	10	9	8	8	$= \dim \pi_2(\mathcal{K}_6)$ $= \dim \pi_2(\mathcal{K}_6)$
s = 6					12	10	9	8	$= \dim \pi_3(\mathcal{K}_7)$
s = 7						12	10	9	
s = 8							12	10	
s = 9								12	Real solving polynomial equations with semidefinite programming – p.24

**Extending the Zhi-Reid criterion to the real case** 

• In the **complex** case,  $K_t = \mathcal{H}_t^{\perp}$  where

$$\mathcal{H}_t := \{ h_j \mathbf{x}^{\alpha} \quad \forall j, \alpha \text{ with } \deg(h_j \mathbf{x}^{\alpha}) \leq t \}.$$

• In the real case,  $\mathcal{K}_t$  is a cone, contained in the linear space  $\mathcal{P}_t^{\perp}$ , with the same dimensions: dim  $\mathcal{K}_t = \dim \mathcal{P}_t^{\perp}$ , where

$$\mathcal{P}_t := \mathcal{H}_t \cup \{ f \mathbf{x}^\alpha \mid f \in \mathrm{Ker} M_{\lfloor t/2 \rfloor}(y), \ \mathrm{deg}(\mathbf{x}^\alpha) \le \lfloor t/2 \rfloor \}$$

**Theorem:** If for some  $D \le s \le t$ 

(ZR+) dim  $\pi_s(\mathcal{P}_t^{\perp}) = \dim \pi_{s-1}(\mathcal{P}_t^{\perp}) = \dim \pi_s((\mathcal{P}_t \cup \partial \mathcal{P}_t)^{\perp})$ 

then one can construct a base of J with  $I \subseteq J \subseteq I(V_{\mathbb{R}}(I))$  and thus extract  $V_{\mathbb{R}}(I) = V_{\mathbb{C}}(J) \cap \mathbb{R}^n$ . Link with the flatness criterion

**Theorem:** In the PSD case, the flatness criterion (F1):

 $\operatorname{rank} M_s(y) = \operatorname{rank} M_{s-1}(y)$  for generic  $y \in \mathcal{K}_t$ 

is equivalent to the stronger version of the (ZR) criterion:

(ZR++) dim 
$$\pi_{s-1}(\mathcal{P}_t^{\perp}) = \dim \pi_{2s}(\mathcal{P}_t^{\perp}) = \dim \pi_{2s}((\mathcal{P}_t \cup \partial \mathcal{P}_t)^{\perp})$$

in which case we find the real radical ideal  $J = I(V_{\mathbb{R}}(I))$ .

**Hence:** the algorithm based on (ZR) may stop earlier than the moment-matrix algorithm, based on (F1).

**Future work:** Adapt other known efficient algorithms for complex roots to *real* roots by incorporating SDP conditions.