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Constructing Piecewise-Polynomial Lyapunov Functions for Nonlinear Systems Using Handelman's Theorem

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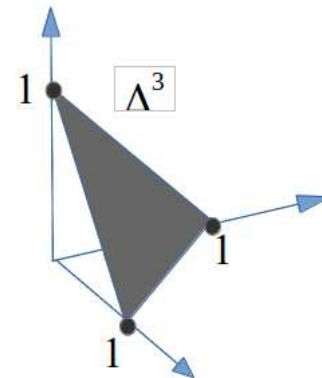
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Solving Large-scale Problems in Control

- We designed a parallel version of Polya's algorithm to verify stability of **linear uncertain** systems with 100+ states.
(TAC 2013)

$$\dot{x}(t) = A(\alpha)x(t), \quad \alpha^n \in \Delta^n$$

$$\Delta^n := \{\alpha \in \mathbb{R}^n : \|\alpha_i\|_1 = 1, \alpha_i \geq 0\}$$

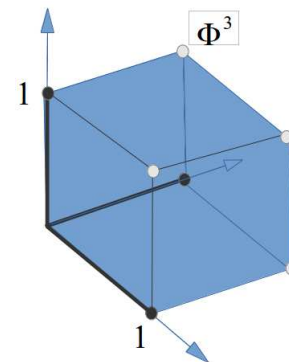


- We designed a parallel algorithm which uses a multi-simplex version of Polya's theorem to verify stability of

(CDC 2012) $\dot{x}(t) = A(\alpha)x(t), \quad \alpha \in \Phi^n$

(CDC 2013) $\dot{x}(t) = f(x(t)), \quad x(t) \in \Phi^n$

$$\Phi^n := \{x \in \mathbb{R}^n : |x_i| \leq r_i\}$$



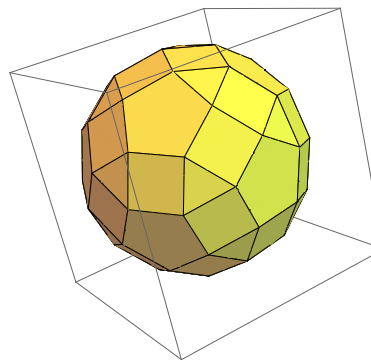
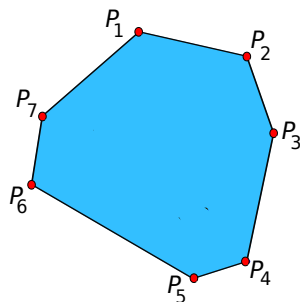
Analysis on More Complicated Geometries

In this talk, we address local stability of **nonlinear** systems defined by polynomial vector fields

$$\dot{x}(t) = f(x) = \sum_{\|\alpha\|_1 \leq d_f} b_\alpha x^\alpha = \sum_{\|\alpha\|_1 \leq d_f} b_\alpha x_1^{\alpha_1} \cdots x_n^{\alpha_n}$$

over **convex polytopes**:

$$\Gamma_p := \{x \in \mathbb{R}^n : x = \sum_{i=1}^K \mu_i p_i, \mu_i \in [0, 1], \sum_{i=1}^K \mu_i = 1\}$$



Alternatively, convex polytopes can be represented as

$$\Gamma_K := \{x \in \mathbb{R}^n : w_i^T x + u_i \geq 0, i = 1, \dots, K\}.$$

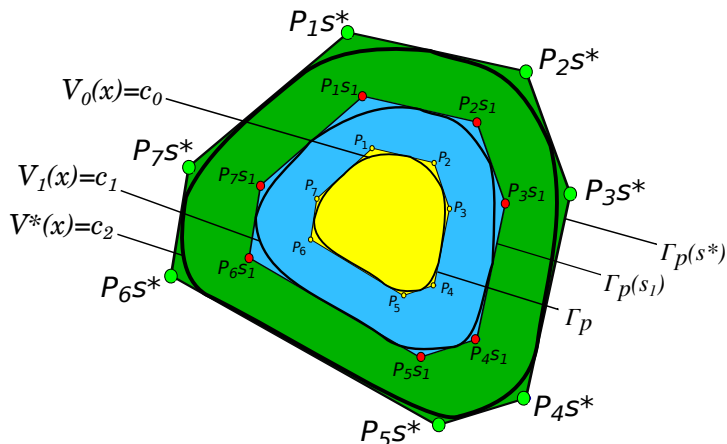
We Search for Lyapunov Polynomials on Polytopes

Given $p_i \in \mathbb{R}^n$, we would like to find a **polynomial** V which solves

$$s^* := \max_{V \in \mathbb{R}[x], s > 0} s$$

$$\begin{aligned} \text{subject to } & V(x) - \epsilon x^T x \geq 0 && \text{for all } x \in \Gamma_p(s) \\ & \nabla V(x)^T f(x) + \epsilon x^T x \leq 0 && \text{for all } x \in \Gamma_p(s). \end{aligned}$$

$$\Gamma_p(s) := \{x \in \mathbb{R}^n : x = \sum_{i=1}^K \mu_i p_i, \mu_i \in [0, s], \sum_{i=1}^K \mu_i = s\}$$



Then $\{x : \{y : V^*(y) \leq V^*(x)\} \subset \Gamma_p(s^*)\}$ is the ROA of the origin.

Optimization of Polynomials is NP-hard

The problem

$$\begin{aligned} s^* &:= \max_{V \in \mathbb{R}[x], s > 0} s \\ \text{subject to} \quad & V(x) - \epsilon x^T x \geq 0 && \text{for all } x \in \Gamma_p(s) \\ & \nabla V(x)^T f(x) + \epsilon x^T x \leq 0 && \text{for all } x \in \Gamma_p(s) \end{aligned}$$

is an instance of the more general problem of **Optimization Of Polynomials (OOP)**:

$$\begin{aligned} \gamma^* &:= \max_{x, F_0} c^T x \\ \text{subject to} \quad & F(x, y) := F_0(y) + \sum_{i=1}^n x_i F_i(y) \geq 0 \text{ for all } y \in \mathbb{R}^n \end{aligned}$$

- The OOP is a **convex** optimization problem
- The OOP is an **NP-hard** problem. Difficult part is to verify $F \geq 0$
 - ↳ Indeed the question: “**Is $p(x) \geq 0$ for all $x \in \mathbb{R}^n$?**” is NP-hard

Tests for Non-negativity of Polynomials

- **Quantifier Elimination** (QE) algorithms yield **exact** solutions to OOPs
 - ↳ Tarski-Seidenberg (1954): Computational complexity grows **double-exponentially** with the number of variables
 - ↳ Basu-Pollack (1996): Computational complexity $\sim d^{O(n)}$
- **Sum-of-Squares programming** and Positivstellensatz results
 - ↳ If $f(x) > 0$ for all $x \in \{x \in \mathbb{R}^n : g_i(x) \geq 0\}$, then
$$f = s_0 + \sum_i s_i g_i \quad \text{for some} \quad s_i = \sum_j h_j(x)^2$$
 - ↳ The search for s_i is a Semi-Definite Program with complexity $\sim n^{O(d)}$

Schweighofer's Parameterization of Positive Polynomials

Suppose the semi-algebraic set

$$S := \{x \in \mathbb{R}^n : g_i(x) \geq 0, g_i \in \mathbb{R}[x], i = 1, \dots, K\}$$

is bounded and define the Cone

$$\Theta_d := \left\{ \sum_{\lambda \in \mathbb{N}^K : \lambda_1 + \dots + \lambda_K \leq d} s_\lambda g_1^{\lambda_1} \cdots g_K^{\lambda_K} : s_\lambda \text{ are SOS} \right\}.$$

Theorem(*Schweighofer's parameterization*): If polynomial f satisfies

1. $f \geq 0$ on S
2. $f = q_1 p_1 + q_1 p_2 + \dots$ for some $q_i \in \Theta_d$ and $p_i > 0$ on $S \cap \{x \in \mathbb{R}^n : f(x) = 0\}$

then, $f \in \Theta_d$.

We are interested in a **special case** of this parameterization:

$$g_i \text{ are affine, } p_i = 1 \text{ and } s_\lambda = \text{Constant} \geq 0$$

Handelman's Theorem: A Parameterization Using Polytopes

Handelman's Theorem:

Given $w_i \in \mathbb{R}^n$ and $u_i \in \mathbb{R}$, suppose

$$\Gamma := \{x \in \mathbb{R}^n : w_i^T x + u_i \geq 0, i = 1, \dots, K\}$$

is bounded. If polynomial $f(x) > 0$ on Γ , then there exist $b_\alpha \geq 0$ with $\alpha \in \mathbb{N}^K$ such that for some $d \in \mathbb{N}$,

$$f(x) = \sum_{\substack{\alpha \in \mathbb{N}^K \\ \alpha_1 + \dots + \alpha_K \leq d}} b_\alpha (w_1^T x + u_1)^{\alpha_1} \dots (w_K^T x + u_K)^{\alpha_K}.$$

- Theorem uses **products of affine functions** as a basis to parameterize polynomials that are positive on convex polytopes
- Converse of theorem yields a test for nonnegativity on polytopes

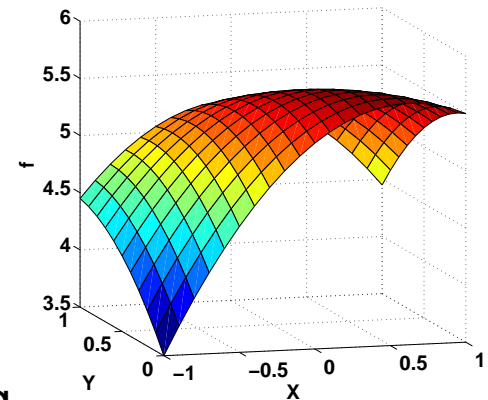
Handelman's Theorem: A Test for Positivity

Example 1:

Is $f(x, y) = -\frac{4}{5}x^2 - xy + x - \frac{3}{4}y^2 + 0.7y + 5.3 \geq 0$
over $\Gamma := [-1, 1] \times [0, 1]$?

The polytope Γ is defined by the inequalities:

$$x + 1 \geq 0, \quad 1 - x \geq 0, \quad y \geq 0, \quad 1 - y \geq 0$$



Choosing $d = 2$, polytope Γ yields the following
Handelman basis:

$$\{1, x + 1, 1 - x, y, 1 - y, (x + 1)^2, (x + 1)(1 - x), (x + 1)y, y^2, (x + 1)(1 - y), (1 - x)^2, (1 - x)y, (1 - x)(1 - y), y(1 - y), (1 - y)^2\}$$

Example 1: Continued ...

Write polynomial f as

$$f(x, y) = -\frac{4}{5}x^2 - xy + x - \frac{3}{4}y^2 + 0.7y + 5.3 = \begin{bmatrix} 5.3 & 1 & 0.7 & -\frac{4}{5} & -1 & -\frac{3}{4} \end{bmatrix}$$

Convert the Handelman basis to the monomial basis:

$$\begin{bmatrix} 1 \\ x+1 \\ 1-x \\ y \\ 1-y \\ (x+1)^2 \\ (x+1)(1-x) \\ (x+1)y \\ y^2 \\ (x+1)(1-y) \\ (1-x)^2 \\ (1-x)y \\ (1-x)(1-y) \\ y(1-y) \\ (1-y)^2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & -1 & 0 & 0 & 0 \\ 1 & 2 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & -1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 1 & 1 & -1 & 0 & -1 & 0 \\ 1 & -2 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & -1 & 0 \\ 1 & -1 & -1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & -1 \\ 1 & 0 & -2 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ x \\ y \\ x^2 \\ xy \\ y^2 \end{bmatrix}$$

Example 1: Continued ...

From Handelman's Theorem, we search for $b_\alpha \geq 0$ such that

$$f(x, y) = \sum_{\alpha \in \mathbb{N}^K: \alpha_1 + \dots + \alpha_K \leq 2} b_\alpha (x+1)^{\alpha_1} (1-x)^{\alpha_2} y^{\alpha_3} (1-y)^{\alpha_4}.$$

Substitute for the RHS:

$$f(x, y) = \begin{bmatrix} 5.3 \\ 1 \\ 0.7 \\ -\frac{4}{5} \\ -1 \\ -\frac{3}{4} \end{bmatrix}^T \begin{bmatrix} 1 \\ x \\ y \\ x^2 \\ xy \\ y^2 \end{bmatrix} = \begin{bmatrix} b_{[0,0,0,0]} \\ b_{[1,0,0,0]} \\ b_{[0,1,0,0]} \\ b_{[0,0,1,0]} \\ b_{[0,0,0,1]} \\ b_{[2,0,0,0]} \\ b_{[1,1,0,0]} \\ b_{[1,0,1,0]} \\ b_{[1,0,0,1]} \\ b_{[0,2,0,0]} \\ b_{[0,1,1,0]} \\ b_{[0,1,0,1]} \\ b_{[0,0,2,0]} \\ b_{[0,0,1,1]} \\ b_{[0,0,0,2]} \end{bmatrix}^T \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & -1 & 0 & 0 & 0 \\ 1 & 2 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & -1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 1 & 1 & -1 & 0 & -1 & 0 \\ 1 & -2 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & -1 & 0 \\ 1 & -1 & -1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & -1 \\ 1 & 0 & -2 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ x \\ y \\ x^2 \\ xy \\ y^2 \end{bmatrix}$$

Example 1: Continued ...

Finally, positivity of f on $\Gamma := [-1, 1] \times [0, 1]$ can be expressed as:

Find $b_\alpha \geq 0$ subject to

$$\begin{bmatrix}
 1 & 0 & 0 & 0 & 0 & 0 \\
 1 & 1 & 0 & 0 & 0 & 0 \\
 1 & -1 & 0 & 0 & 0 & 0 \\
 0 & 0 & 1 & 0 & 0 & 0 \\
 1 & 0 & -1 & 0 & 0 & 0 \\
 1 & 2 & 0 & 1 & 0 & 0 \\
 1 & 0 & 0 & -1 & 0 & 0 \\
 0 & 0 & 1 & 0 & 1 & 0 \\
 0 & 0 & 0 & 0 & 0 & 1 \\
 1 & 1 & -1 & 0 & -1 & 0 \\
 1 & -2 & 0 & 1 & 0 & 0 \\
 0 & 0 & 1 & 0 & -1 & 0 \\
 1 & -1 & -1 & 0 & 1 & 0 \\
 0 & 0 & 1 & 0 & 0 & -1 \\
 1 & 0 & -2 & 0 & 0 & 1
 \end{bmatrix}^T \begin{bmatrix}
 b_{[0,0,0,0]} \\
 b_{[1,0,0,0]} \\
 b_{[0,1,0,0]} \\
 b_{[0,0,1,0]} \\
 b_{[0,0,0,1]} \\
 b_{[2,0,0,0]} \\
 b_{[1,1,0,0]} \\
 b_{[1,0,1,0]} \\
 b_{[1,0,0,1]} \\
 b_{[0,2,0,0]} \\
 b_{[0,1,1,0]} \\
 b_{[0,1,0,1]} \\
 b_{[0,0,2,0]} \\
 b_{[0,0,1,1]} \\
 b_{[0,0,0,2]}
 \end{bmatrix} = \begin{bmatrix}
 5.3 \\
 1 \\
 0.7 \\
 \frac{-4}{5} \\
 -1 \\
 \frac{-3}{4}
 \end{bmatrix}$$

Example 1: Finished!

The problem is in the dual form of **Linear Program**:

$$\begin{aligned} \min_{b_\alpha \geq 0} \quad & c^T b_\alpha \\ \text{subject to} \quad & A b_\alpha = g \end{aligned}$$

By solving the LP we get

$$b_\alpha = [1.5 \ 0 \ 0 \ 2.2 \ 0 \ 0 \ 0.8 \ 0 \ 1 \ 0 \ 0 \ 0 \ 0.75 \ 3.5 \ 2]$$

By plugging b_α in

$$f(x, y) = \sum_{\alpha \in \mathbb{N}^K: \alpha_1 + \dots + \alpha_K \leq 2} b_\alpha (x+1)^{\alpha_1} (1-x)^{\alpha_2} y^{\alpha_3} (1-y)^{\alpha_4}.$$

we get

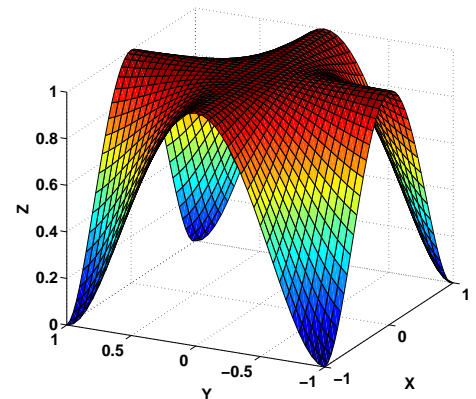
$$\begin{aligned} f(x, y) &= -\frac{4}{5}x^2 - xy + x - \frac{3}{4}y^2 + 0.7y + 5.3 \\ &= 1.5 + 2.2y + 0.8(x+1)(1-x) + y^2 \\ &\quad + 0.75(1-x)(1-y) + 3.5y(1-y) + 2(1-y)^2 \end{aligned}$$

Handelman's Theorem: A Test for Positivity

Example 2: (Motzkin Polynomial)

Is $f(x, y) = x^4y^2 + x^2y^4 - 3x^2y^2 + 1 \geq 0$ over $\Gamma := [-1, 1] \times [-1, 1]$?

- It is well-known that Motzkin polynomial is not SOS of polynomials.
- However $f(x, y)$ it can be represented in Handelman basis with all positive coefficients:



$$\begin{aligned} f(x, y) = & 0.125(\lambda_1^3\lambda_3^2\lambda_4^2 + \lambda_1^2\lambda_2^2\lambda_3^3 + \lambda_1^2\lambda_2^2\lambda_4^3 + \lambda_2^3\lambda_3^2\lambda_4^2 + \lambda_1^3\lambda_2\lambda_3\lambda_4) \\ & + 0.0625(\lambda_1^2\lambda_2\lambda_3^3\lambda_4 + \lambda_1\lambda_2^3\lambda_3^2\lambda_4 + \lambda_1\lambda_2^3\lambda_3\lambda_4^2 + \lambda_1\lambda_2^2\lambda_3^3\lambda_4 \\ & + \lambda_1\lambda_2\lambda_3^2\lambda_4^3 + \lambda_1\lambda_2\lambda_3\lambda_4^4) \end{aligned}$$

$$\lambda_1(x) := 1 - x, \quad \lambda_2(x) := 1 + x, \quad \lambda_3(y) := 1 - y, \quad \lambda_4(y) := 1 + y$$

Handelman's theorem precludes interior zeros

Recall the problem of stability analysis:

$$\begin{aligned} s^* &:= \max_{V \in \mathbb{R}[x], s > 0} s \\ \text{subject to} \quad & V(x) - \epsilon x^T x \geq 0 && \text{for all } x \in \Gamma_p(s) \\ & \nabla V(x)^T f(x) + \epsilon x^T x \leq 0 && \text{for all } x \in \Gamma_p(s). \end{aligned}$$

- We need to search over polynomials $V(x) \geq 0$ with $V(0) = 0$
- Handelman's theorem allows for zeros on the vertices of Γ_p
 - ↳ Motzkin example
- Handelman's theorem does **NOT** parameterize positive polynomials with zeros in the interior of Γ_p (why?)

Handelman's theorem precludes interior zeros

Proof:

Suppose $f(a) = 0$ for some $a \in \text{int}(\Gamma)$, where

$$\Gamma := \{x \in \mathbb{R}^n : w_i^T x + u_i \geq 0, i = 1, \dots, K\}.$$

Suppose there exist $b_\alpha \geq 0, \alpha \in \mathbb{N}^K$ such that for some $d \in \mathbb{N}$,

$$f(x) = \sum_{\alpha \in \mathbb{N}^K : \|\alpha_i\|_1 \leq d} b_\alpha (w_1^T x + u_1)^{\alpha_1} \cdots (w_K^T x + u_K)^{\alpha_K}.$$

Then,

$$f(a) = \sum_{\alpha \in \mathbb{N}^K : \|\alpha_i\|_1 \leq d} b_\alpha (w_1^T a + u_1)^{\alpha_1} \cdots (w_K^T a + u_K)^{\alpha_K} = 0.$$

Since $a \in \text{int}(\Gamma)$, $w_i^T a + u_i > 0$ for $i = 1, \dots, K$.

Hence there exist some $\alpha \in \{\alpha \in \mathbb{N}^K : \|\alpha\|_1 \leq d\}$ such that

$$b_\alpha < 0.$$

This contradicts with the assumption that all $b_\alpha \geq 0$.

Applying Handelman's Theorem to Stability Analysis

Step 1: Decomposition

Decompose the polytope

$$\Gamma := \{x \in \mathbb{R}^n : w_i^T x + u_i \geq 0, i = 1, \dots, K\}$$

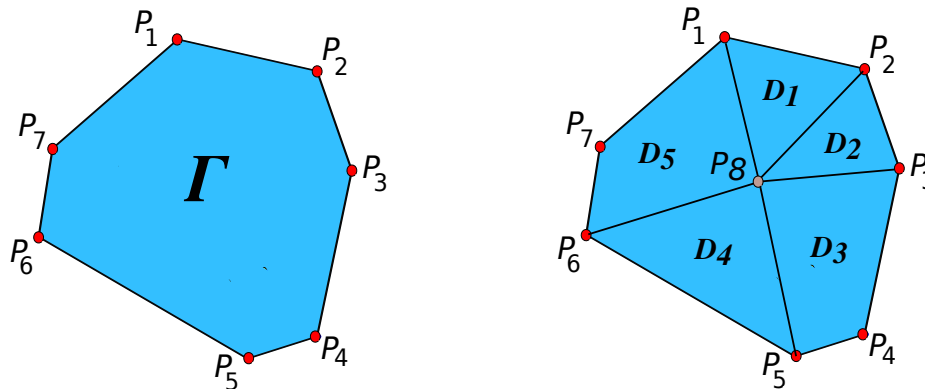
into L subpolytopes

$$D_i := \{x \in \mathbb{R}^n : h_{i,j}^T x + g_{i,j} \geq 0, j = 1, \dots, m_i\}$$

such that

$$\cup_{i=1}^L D_i = \Gamma, \quad \cap_{i=1}^L D_i = \{0\}, \quad \text{int}(D_i) \cap \text{int}(D_j) = \emptyset$$

Example:



Applying Handelman's Theorem to Stability Analysis

Step 2: Enforcing $V(0) = 0$

- For each subpolytope D_i with m edges, let

$$V_i(x) = \sum_{\alpha \in \mathbb{N}^m: \|\alpha\|_1 \leq d} b_{i,\alpha} \lambda_1(x)^{\alpha_1} \cdots \lambda_m(x)^{\alpha_m}$$

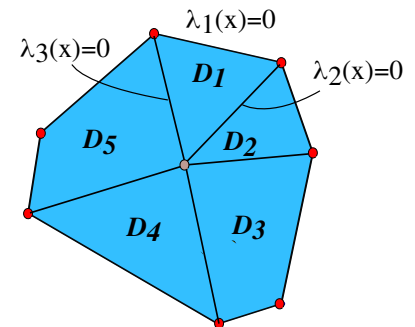
where $\lambda_j(x) := h_{i,j}^T x + g_{i,j}$ define the edges of D_i .

- To enforce $V_i(0) = 0$, set

$$b_{i,\alpha} = 0 \text{ for all } \alpha \in \{\alpha : \alpha_j = 0 \text{ for all } j : \lambda_j(0) = 0\}$$

Example:

$$\begin{aligned} V_1(x) = & \cancel{b_{1,[0,0,0]}} + \cancel{b_{1,[1,0,0]}} \lambda_1 + b_{1,[0,1,0]} \lambda_2 + b_{1,[0,0,1]} \lambda_3 \\ & + \cancel{b_{1,[2,0,0]}} \lambda_1^2 + b_{1,[1,1,0]} \lambda_1 \lambda_2 + b_{1,[1,0,1]} \lambda_1 \lambda_3 + b_{1,[0,1,1]} \lambda_2 \lambda_3 \\ & + b_{1,[0,2,0]} \lambda_2^2 + b_{1,[0,1,1]} \lambda_2 \lambda_3 + b_{1,[0,0,2]} \lambda_3^2 \end{aligned}$$

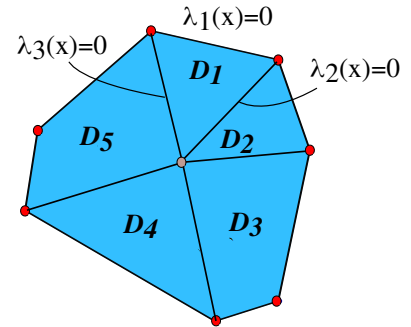


Applying Handelman's Theorem to Stability Analysis

Step 3: Ensuring continuity of $V(x)$

- For any two adjacent subpolytopes D_i and D_j with $D_i \cap D_j = \{x : \lambda(x) = 0\}$, set

$$V_i(x) \big|_{\lambda} = V_j(x) \big|_{\lambda}$$



Example:

- Define V_1 on D_1 as

$$V_1(x) = b_{1,[0,1,0]} \lambda_2 + b_{1,[0,0,1]} \lambda_3 + b_{1,[1,1,0]} \lambda_1 \lambda_2 + b_{1,[1,0,1]} \lambda_1 \lambda_3 + b_{1,[0,1,1]} \lambda_2 \lambda_3 \\ + b_{1,[0,2,0]} \lambda_2^2 + b_{1,[0,1,1]} \lambda_2 \lambda_3 + b_{1,[0,0,2]} \lambda_3^2$$

- Restriction of V_1 to $\lambda_2(x) = 0$:

$$V_1(x) \big|_{\lambda_2} = b_{1,[0,0,1]} \lambda_3 + b_{1,[1,0,1]} \lambda_1 \lambda_3 + b_{1,[0,0,2]} \lambda_3^2 \\ = [l_1(b_1) \ l_2(b_1) \ l_3(b_1) \ l_4(b_1) \ l_5(b_1)] [1 \ x \ y \ xy \ x^2 \ y^2]^T$$

- Similarly, On D_2 we have

$$V_2(x) \big|_{\lambda_2} = [g_1(b_2) \ g_2(b_2) \ g_3(b_2) \ g_4(b_2) \ g_5(b_2)] [1 \ x \ y \ xy \ x^2 \ y^2]^T$$

- To enforce continuity on $\{x : \lambda_2(x) = 0\}$, set

$$l(b_1) = g(b_2)$$

Applying Handelman's Theorem to Stability Analysis

Step 4: Enforcing $\dot{V} < 0$ on Γ

- For each subpolytope D_i , put

$$\begin{aligned}\dot{V}_i(x) &= \langle \nabla V_i, f(x) \rangle = \nabla \left(\sum_{\|\alpha\|_1 \leq d} b_{i,\alpha} \lambda_1(x)^{\alpha_1} \cdots \lambda_m(x)^{\alpha_m} \right)^T f(x) \\ &= [h_1(b_i) \ h_2(b_i) \cdots] [1 \ x_1 \ x_2 \ \cdots]^T \\ &= \sum_{\|\beta\|_1 \leq d+d_f-1} c_{i,\beta} \lambda_1(x)^{\beta_1} \cdots \lambda_m(x)^{\beta_m} \\ &= [q_1(c_i) \ q_2(c_i) \cdots] [1 \ x_1 \ x_2 \ \cdots]^T\end{aligned}$$

- To enforce $\dot{V} < 0$ on D_i we need to solve:

$$\begin{array}{ll}\text{Find} & b_{i,\alpha} \in \mathbb{R} \text{ and } c_{i,\beta} \leq 0 \\ \text{such that} & q(c_i) = h(b_i)\end{array}$$

Applying Handelman's Theorem to Stability Analysis

Step 1: Decomposition of original polytope Γ into D_i

$$\cup_{i=1}^L D_i = \Gamma, \quad \cap_{i=1}^L D_i = \{0\}, \quad \text{int}(D_i) \cap \text{int}(D_j) = \emptyset$$

Step 2: Enforcing $V(0) = 0$

$$V_i(x) = \sum_{\alpha \in \mathbb{N}^m: \|\alpha\|_1 \leq d} b_{i,\alpha} \lambda_1(x)^{\alpha_1} \cdots \lambda_m(x)^{\alpha_m}$$

$$b_{i,\alpha} = 0 \quad \text{for all } \alpha \in \{\alpha : \alpha_j = 0 \text{ for all } j : \lambda_j(0) = 0\}$$

Step 3: Continuity of V on λ

$$V_i(x) |_{\lambda} = V_j(x) |_{\lambda} \Leftrightarrow l(b_i) = g(b_j)$$

where l and g are affine in b_i and b_j

Step 4: Enforcing $\dot{V} < 0$ on Γ

$$q(c_i) = h(b_i), \quad c_i \leq 0$$

where q and h are affine in c_i and b_i

The Four Steps Define a Linear Program

Given a polytope, the stability analysis problem can be expressed as the following **Linear Program**:

$$\begin{array}{ll} \min_{\substack{b_{i,\alpha} \geq 0 \\ c_{i,\alpha} \leq 0}} & c^T [b_{1,\alpha}, \dots, b_{L,\alpha}] \\ \text{subject to} & b_{i,\alpha} = 0 \quad \text{for all } \alpha \in \{\alpha : \alpha_j = 0, \forall j : \lambda_j(0) = 0\} \\ & l(b_i) = g(b_j) \quad \text{for all } i, j \in \{1, \dots, L\} : D_i \cap D_j \neq \{0\} \\ & q(c_i) = h(b_i) \quad \text{for all } i \in \{1, \dots, L\} \end{array}$$

A solution to the Linear Program yields the Lyapunov function

$$V(x) = V_i(x) = \sum_{\|\alpha\|_1 \leq d} b_{i,\alpha} \lambda_1(x)^{\alpha_1} \dots \lambda_m(x)^{\alpha_m}$$

for $x \in D_i$, $i = 1, \dots, L$

Numerical Example: Van-Der-Pol Oscillator

For the Van-Der-Pol oscillator in reverse-time

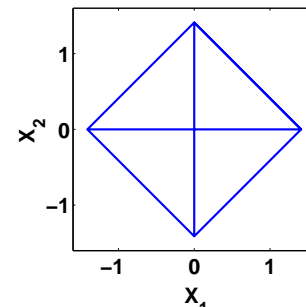
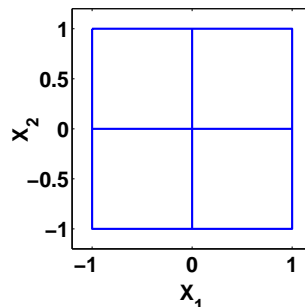
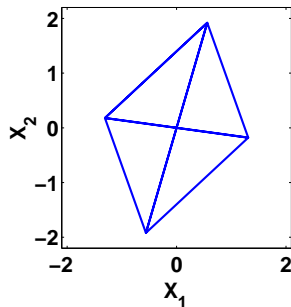
$$\dot{x}_1(t) = -x_2(t), \quad \dot{x}_2(t) = x_1(t) + x_2(t) (x_1^2(t) - 1)$$

We solved

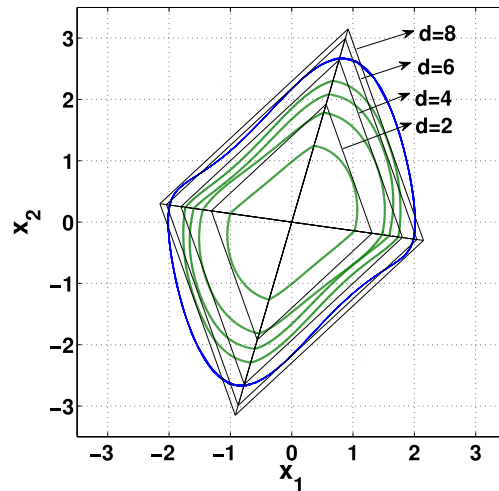
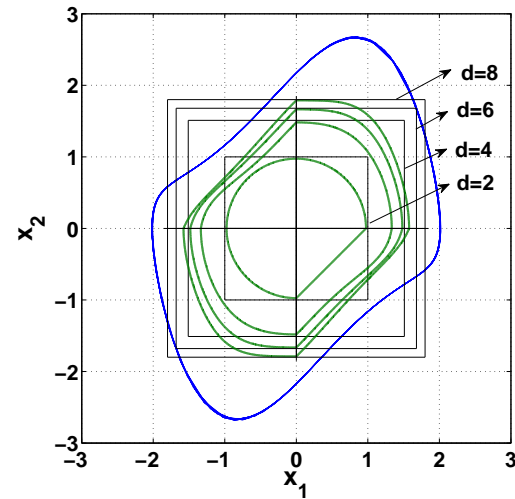
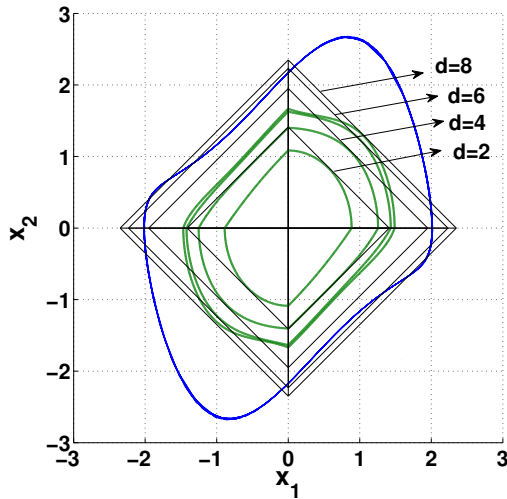
$$s^* := \max_{V \in \mathbb{R}[x], s > 0} s$$

$$\begin{aligned} \text{subject to} \quad & V(x) - \epsilon x^T x \geq 0 && \text{for all } x \in \Gamma_p(s) \\ & \nabla V(x)^T f(x) + \epsilon x^T x \leq 0 && \text{for all } x \in \Gamma_p(s) \end{aligned}$$

using the following polytopes as Γ_p



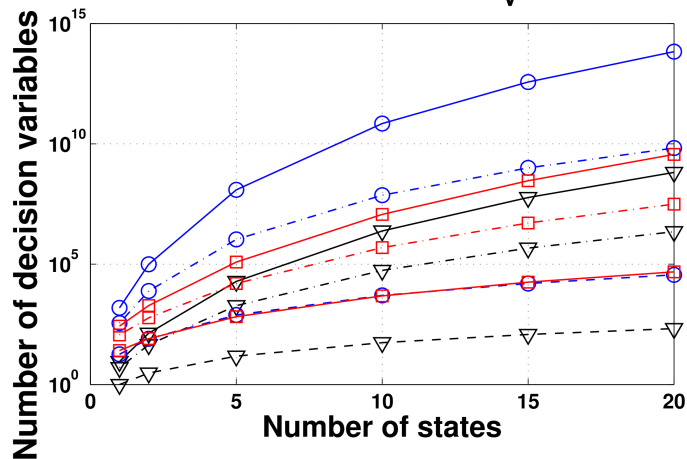
For Fixed d , Quadrilateral Results in Better Estimation



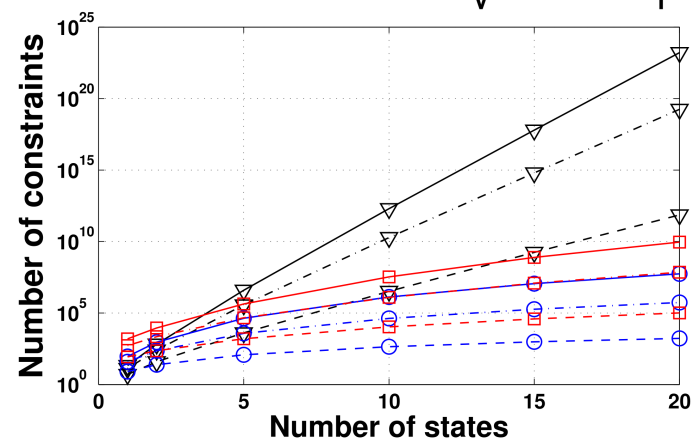
Complexity Scales Polynomially in State Space Dimension

—○— SOS+Putinar's Psatz —▽— Polya —□— Current method

Number of decision variables, $d_V=[2,6,10]$, $d_f = 2$



Number of constraints, $d_V=[2,6,10]$, $d_f = 2$



	SOS	Polya	Current method
Complexity of LP/SDP	$\sim n^{3.5(d_V+d_f)-3}$	$\sim (d_V + d_f + e - 2)^{3n}$	$\sim n^{3(d_V+d_f)}$

d_V : degree of $V(x)$

d_f : degree of $f(x)$

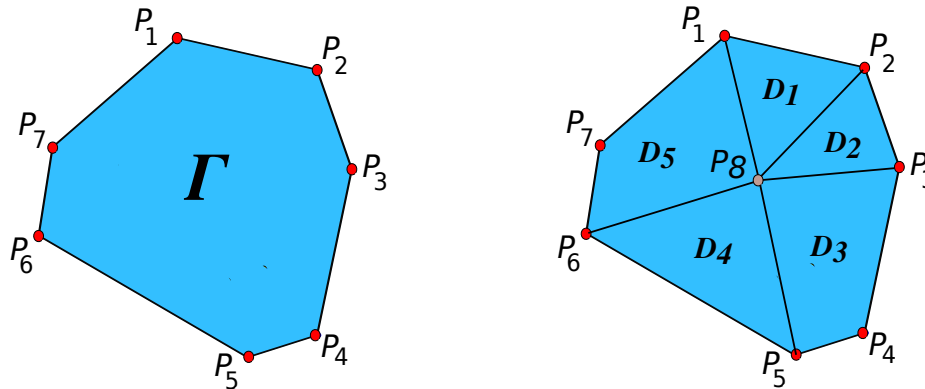
n : No. of states

Conclusion

- We proposed a methodology based on Handelman's theorem to perform stability analysis on nonlinear ODEs

$$\dot{x}(t) = f(x(t)) \quad x(t) \in \Gamma \text{ (Convex polytopes)}$$

using a decomposition of the polytope.



- The method can be readily applied to stability analysis of

$$\dot{x}(t) = f(x(t), \alpha) \quad x(t), \alpha \in \Gamma \text{ (Convex polytopes)}$$