

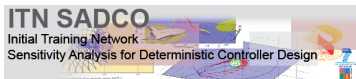
# Computation of Local ISS Lyapunov Function Via Linear Programming

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Joint work with Robert Baier, Lars Grüne,  
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# Outline

- 1 Introduction
- 2 Relationship between ISS and Robust Lyapunov Functions
- 3 Computing Local Robust Lyapunov Functions by Linear Programming
- 4 Computing Local ISS Lyapunov Functions by Linear Programming
- 5 Example
- 6 Conclusion and Future Works

$$\dot{x} = f(x, u), \quad x \in \mathbb{R}^n, u \in \mathcal{U}_R, \quad (1)$$

$f : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n$  is Lipschitz continuous and  $f(0,0) = 0$ .

### Definition (Input to state stability (ISS))

System (1) is **input to state stable (ISS)** if there exist a  $\mathcal{KL}$  function  $\beta : \mathbb{R}_{\geq 0} \times \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}$ ,  $\gamma \in \mathcal{K}$  such that

$$\|x(t, x(0), u)\|_2 \leq \beta(\|x(0)\|_2, t) + \gamma(\|u\|_{L_\infty}), \quad \text{for each } t \geq 0.$$

### Definition (ISS Lyapunov function)

A smooth function  $V : \mathbb{R}^n \rightarrow \mathbb{R}_{\geq 0}$  is an **ISS Lyapunov function** of system (1) if there exist  $\varphi_1, \varphi_2, \alpha, \sigma \in \mathcal{K}_\infty$  such that

$$\begin{aligned} \varphi_1(\|x\|_2) &\leq V(x) \leq \varphi_2(\|x\|_2), \\ \dot{V} = \langle \nabla V(x), f(x, u) \rangle &\leq -\alpha(\|x\|_2) + \sigma(\|u\|_2), \text{dissipative form.} \end{aligned}$$

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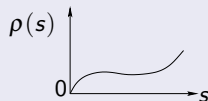
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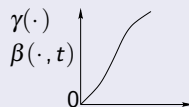
# Comparison Functions

## Positive definite function

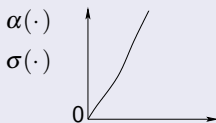


$\rho : \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}; \rho(0) = 0;$   
 $\rho(s) > 0, \text{ for } s > 0.$

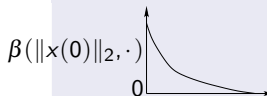
## $\mathcal{K}$ function



## $\mathcal{K}_{\infty}$ function



## $\mathcal{L}$ function



## Definition

System (1) is **locally input to state stable (ISS)** if there exist  $\rho^0 > 0$ ,  $\rho^u > 0$ ,  $\mathcal{KL}$  function  $\beta : \mathbb{R}_{\geq 0} \times \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}$ ,  $\gamma \in \mathcal{K}$  such that

$$\|x(t, x(0), u)\|_2 \leq \beta(\|x(0)\|_2, t) + \gamma(\|u\|_{L_\infty}), \quad t \geq 0,$$

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## Definition

A smooth function  $V : \mathcal{D} \rightarrow \mathbb{R}_{\geq 0}$ , with  $\mathcal{D} \subset \mathbb{R}^n$  open, is a **local ISS Lyapunov function** for system (1) if there exist  $\rho^0 > 0$ ,  $\rho^u > 0$ ,  $\varphi_1$ ,  $\varphi_2$ ,  $\alpha$ ,  $\sigma \in \mathcal{K}_\infty$  such that  $B(0, \rho^0) \subset \mathcal{D}$  and

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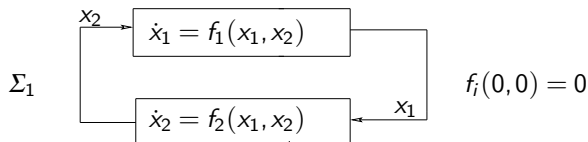
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for all  $\|x\|_2 \leq \rho^0$ ,  $\|u\|_2 \leq \rho^u$ .

# Basic small gain theorem



## Theorem

*If there exist ISS Lyapunov functions  $v_1, v_2$  for subsystems of system  $\Sigma_1$  satisfying*

$$\dot{v}_1 \leq -\|x_1\|_2 + K_1\|x_2\|_2, \quad K_1 > 0,$$

$$\dot{v}_2 \leq -\|x_2\|_2 + K_2\|x_1\|_2, \quad K_2 > 0,$$

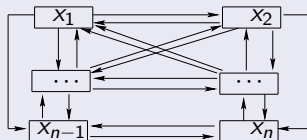
*and  $K_1 K_2 < 1$ , then there exists a vector  $\zeta \in \mathbb{R}_{>0}^2$  such that function  $V = \langle \zeta, v \rangle$  ( $v = (v_1, v_2)$ ) is a Lyapunov function for system  $\Sigma_1$ .*

*Furthermore, system  $\Sigma_1$  is asymptotically stable at the origin.*



# Motivation

Estimate the domain of attraction of large scale systems



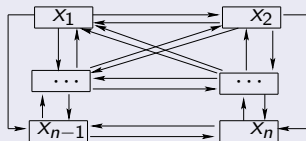
$$\begin{cases} \dot{x}_1 = f_1(x_1, x_2, \dots, x_n), \\ \vdots \\ \dot{x}_n = f_n(x_1, x_2, \dots, x_n), \end{cases} \quad f_i(0, 0, \dots, 0) = 0, x_i \in \mathbb{R}^{n_i}, \sum_{i=1}^n n_i = N.$$

If subsystems are **input to state stable (ISS)**, an effective way to estimate the domain of attraction is **using small gain theorems with subsystems' input to state stability (ISS) Lyapunov functions**.

Main question: How to compute local ISS Lyapunov functions for subsystems?

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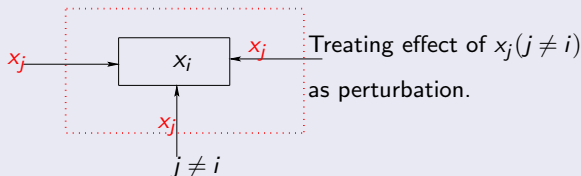


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## Subsystems



$$S_i: \quad \dot{x}_i = f_i(x_1, x_2, \dots, x_{i-1}, x_i, x_{i+1}, \dots, x_n).$$

## System with perturbation

Original system:

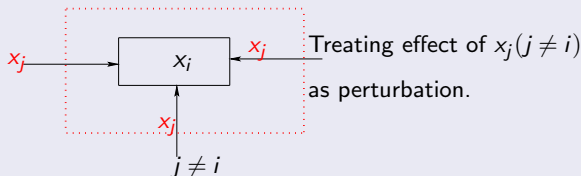
$$\dot{x} = f(x, u), \quad x \in \mathbb{R}^n, \quad u \in \mathcal{U}_R \subset \mathbb{R}^m, \quad f(0, 0) = 0. \quad (2)$$

Assumption:

- System (2) is locally input to state stable(ISS).

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# Robust Lyapunov function

## System

$$\dot{x} = f(x, u), \quad x \in \mathbb{R}^n, u \in \mathbb{R}^m, f(0, u) = 0.$$

## Definition (Robust Lyapunov function)

A smooth function  $V : \mathcal{D} \rightarrow \mathbb{R}_{\geq 0}$ , with  $\mathcal{D} \subset \mathbb{R}^n$  open is said to be a **local robust Lyapunov function** of system if there exist  $\rho^0 > 0$ ,  $\rho^u > 0$ ,  $\varphi_1$ ,  $\varphi_2 \in \mathcal{K}_\infty$ , and a positive definite function  $\alpha$  such that  $B(0, \rho^0) \subset \mathcal{D}$  and

$$\begin{aligned} \varphi_1(\|x\|_2) &\leq V(x) \leq \varphi_2(\|x\|_2), \\ \sup_{u \in U_R} \langle \nabla V, f(x, u) \rangle &\leq -\alpha(\|x\|_2), \end{aligned}$$

for all  $\|x\|_2 \leq \rho^0$ ,  $\|u\|_2 \leq \rho^u$ .

If  $\rho^0 = \rho^u = \infty$  then  $V$  is called a **global robust Lyapunov function**.

# Relationship between ISS and Robust Lyapunov Functions

Introduce two Lipschitz continuous functions  $\eta_1 : \mathbb{R}^n \rightarrow \mathbb{R}^n$  and  $\eta_2 : \mathbb{R}^m \rightarrow \mathbb{R}_{\geq 0}$ , then consider the **auxiliary system**

$$\begin{aligned}\dot{x} &= f_\eta(x, u) := f(x, u) - \eta_1(x)\eta_2(u), \quad \|u\| \leq R. \\ x(0) &= x^0, \quad f_\eta(0, u) = 0.\end{aligned}$$

## Theorem

*If there exists a local robust Lyapunov function  $V(x)$  for the auxiliary system and*

- $\exists \alpha(\cdot) \in \mathcal{K}_\infty$  such that  $\langle \nabla V(x), f_\eta(x, u) \rangle \leq -\alpha(\|x\|_2)$ , for all  $u \in U_R$ ,
- $\exists K > 0, \beta(\cdot) \in \mathcal{K}_\infty$  such that  $\langle \nabla V(x), \eta_1(x) \rangle \leq K (K > 0)$ ,  $|\eta_2(u)| \leq \beta(\|u\|_2)$ ,

*then  $V(x)$  is a local ISS Lyapunov function for the original system.*

**Main question:** How to compute such a robust Lyapunov function  $V(x)$  with  $\eta_1(x)$  and  $\eta_2(u)$  ?

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## Grids

### Aim:

Compute such a robust Lyapunov function  $V(x)$  which is continuous piecewise affine on each simplex.

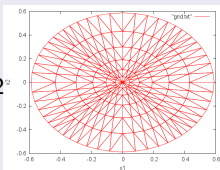
We divide a compact set  $\Omega \subset \mathbb{R}^n$  into  $N$   $n$ -simplices

$$\mathcal{T} = \{\Gamma_v \mid v = 1, \dots, N\}, \Gamma_v := \text{co}\{x_0, x_1, \dots, x_n\}, h_x := \text{diam}(\Gamma_v), \\ \text{diam}(\Gamma_v) := \max_{x, y \in \Gamma_v} \|x - y\|_2.$$

We divide a compact set  $\Omega_u \subset U_R$  into  $N_u$   $m$ -simplices

$$\mathcal{T}_u = \{\Gamma_v^u \mid v = 1, \dots, N_u\}, \Gamma_v^u := \text{co}\{u_0, u_1, \dots, u_m\}, h_u := \text{diam}(\Gamma_v^u),$$

$$n \text{ or } m = 2^q$$

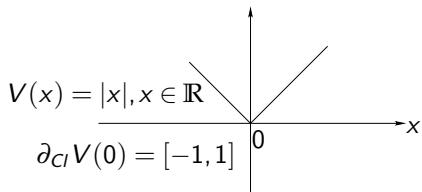


# Clarke's subdifferential for Lipschitz continuous functions

## Proposition (Clarke, 1998)

*For a Lipschitz continuous function  $V : \mathbb{R}^n \rightarrow \mathbb{R}$  Clarke's subdifferential satisfies*

$$\partial_{Cl} V(x) = \text{co}\left\{ \lim_{i \rightarrow \infty} \nabla V(x_i) \mid x_i \rightarrow x, \nabla V(x_i) \text{ and } \lim_{i \rightarrow \infty} \nabla V(x_i) \text{ exist} \right\}.$$



# Nonsmooth robust Lyapunov function

## System

$$\dot{x} = f(x, u), \quad x \in \mathbb{R}^n, u \in \mathbb{R}^m, f(0, u) = 0.$$

## Definition (Nonsmooth robust Lyapunov function)

A Lipschitz continuous function  $V : \mathcal{D} \rightarrow \mathbb{R}_{\geq 0}$ , with  $\mathcal{D} \subset \mathbb{R}^n$  open is said to be a **local nonsmooth robust Lyapunov function** for system if there exist  $\rho^0 > 0$ ,  $\rho^u > 0$ ,  $\mathcal{K}_\infty$  functions  $\phi_1$  and  $\phi_2$ , and a positive definite function  $\alpha$  such that  $B(0, \rho^0) \subset \mathcal{D}$  and

$$\begin{aligned} \phi_1(\|x\|_2) &\leq V(x) \leq \phi_2(\|x\|_2), \\ \sup_{u \in U_R} \langle \xi, f(x, u) \rangle &\leq -\alpha(\|x\|_2), \quad \forall \xi \in \partial_{cl} V(x), \end{aligned}$$

for all  $\|x\|_2 \leq \rho^0$ ,  $\|u\|_2 \leq \rho^u$ .

If  $\rho^0 = \rho^u = \infty$  then  $V$  is called a **global nonsmooth robust Lyapunov function**.

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# Computing local robust Lyapunov functions by linear programming

## System

$$\dot{x} = f(x, u), \quad x \in \mathbb{R}^n, u \in \mathbb{R}^m, f(0, u) = 0.$$

## Assumption:

- System is locally asymptotically stable at the origin uniformly in  $u$ .

## Problem

How to compute a local robust Lyapunov function  $V(x)$  for system?

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# Linear programming based algorithm for computing continuous piecewise affine Lyapunov functions

## Previous results of Linear programming based algorithm for computing continuous piecewise affine Lyapunov functions

- ① Linear programming based algorithm for computing piecewise affine Lyapunov function was first presented in [Marinósson,2002] for ordinary differential equations.
  - ② Further developed in [Hafstein,2007] for systems with switching time.
  - ③ Extended to nonlinear differential inclusions in [Baier, Grüne, Hafstein,2012].
- \* Gives a true Lyapunov function, i.e., not an approximation of a Lyapunov function.

# Computing local robust Lyapunov functions by linear programming

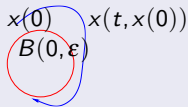
**Notation:**  $PL(\Omega)$ : the space of continuous functions  $V : \Omega \rightarrow \mathbb{R}$  which are linear affine on each simplex, i.e.  $\nabla V_v := \nabla V|_{\text{int}\Gamma_v} \equiv \text{const}$ , for all  $\Gamma_v \in \Omega$ .

**System:**

$$\dot{x} = f(x, u), \quad f(0, u) = 0.$$

**Aim:** compute a local robust Lyapunov function  $V(x) \in PL(\Omega \setminus B(0, \varepsilon))$  ( $\varepsilon > 0$  small enough) satisfying by linear programming.

- $\langle \nabla V(x), f(x, u) \rangle \leq -\|x\|_2.$



Reason for excluding  $B(0, \varepsilon)$

$$\langle \nabla V(x), f(x, u) \rangle \leq -\|x\|_2.$$



# Computing robust Lyapunov functions by linear programming

$$\mathcal{T}^\varepsilon := \{\Gamma_V | \Gamma_V \cap B(0, \varepsilon) = \emptyset\} \subset \mathcal{T}.$$

## Linear programming based algorithm 1

1. For all vertices  $x_i$  of  $\Gamma_V$ , introduce  $V(x_i)$  as the variables and demand  $V(x_i) \geq \|x_i\|_2 \implies V(x) \geq \|x\|_2, x \in \Gamma_V \in \mathcal{T}^\varepsilon$ .
2. For every  $\Gamma_V \in \mathcal{T}^\varepsilon$ , introduce the variables  $C_{V,i}$  ( $i = 1, 2, \dots, n$ ),  $G$  and require

$$|\nabla V_{V,i}| \leq C_{V,i} \leq G,$$

$\nabla V_{V,i}$  is the  $i$ -th component of the vector  $\nabla V_V$ .

3. For every  $\Gamma_V \in \mathcal{T}^\varepsilon$ , and every  $\Gamma_V^u$ , demand

$$\langle \nabla V_V, f(x_i, u_j) \rangle + \sum_{k=1}^n C_{V,k} \underbrace{(A_x(u, h_x) + A_u(x, h_u))}_{\text{compensate for interpolation errors for the points } (x, u) \text{ with } x \neq x_i, u \neq u_j, i = 0, 1, \dots, n, j = 0, 1, \dots, m.} \leq -\|x_i\|_2,$$

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## Linear programming based algorithm 1

1. For all vertices  $x_i$  of  $\Gamma_V$ , introduce  $V(x_i)$  as the variables and demand  $V(x_i) \geq \|x_i\|_2 \implies V(x) \geq \|x\|_2, x \in \Gamma_V \in \mathcal{T}^\varepsilon$ .
2. For every  $\Gamma_V \in \mathcal{T}^\varepsilon$ , introduce the variables  $C_{V,i}$  ( $i = 1, 2, \dots, n$ ),  $G$  and require

$$|\nabla V_{V,i}| \leq C_{V,i} \leq G,$$

$\nabla V_{V,i}$  is the  $i$ -th component of the vector  $\nabla V_V$ .

3. For every  $\Gamma_V \in \mathcal{T}^\varepsilon$ , and every  $\Gamma_V^u$ , demand

$$\langle \nabla V_V, f(x_i, u_j) \rangle + \sum_{k=1}^n C_{V,k} \underbrace{(A_x(u, h_x) + A_u(x, h_u))}_{\text{compensate for interpolation errors for the points } (x, u) \text{ with } x \neq x_i, u \neq u_j, i = 0, 1, \dots, n, j = 0, 1, \dots, m.} \leq -\|x_i\|_2,$$

compensate for interpolation errors for the points  $(x, u)$  with  $x \neq x_i, u \neq u_j, i = 0, 1, \dots, n, j = 0, 1, \dots, m$ .

### Theorem (Result of linear programming based algorithm 1)

*If  $f$  satisfies regular conditions, and the linear programming problem with the constraints in algorithm 1 has a feasible solution, then the values  $V(x_i)$  at all the vertices  $x_i$  of all the simplices  $\Gamma_V \in \mathcal{T}^\varepsilon$  and the condition  $V \in PL(\mathcal{T}^\varepsilon)$  uniquely define the function*

$$V: \bigcup_{\Gamma_V \in \mathcal{T}^\varepsilon} \Gamma_V \rightarrow \mathbb{R},$$

*satisfying*

$$\langle \nabla V_V, f(x, u) \rangle \leq -\|x\|_2$$

*for  $x \in \Gamma_V \in \mathcal{T}^\varepsilon$  and  $u \in \Gamma_V^u$ . Furthermore  $V$  is a robust Lyapunov function for system.*

# Recall Main Problem

Original system:

$$\dot{x} = f(x, u), \quad x \in \mathbb{R}^n, u \in \mathbb{R}^m, f(0, 0) = 0.$$

Auxiliary system:

$$\dot{x} = f_{\eta}(x, u) := f(x, u) - \eta_1(x)\eta_2(u), \quad f_{\eta}(0, u) = 0$$

## Problem

How to compute a local ISS Lyapunov function  $V(x)$  with  $\eta_1(x)$  and  $\eta_2(u)$  for the original system by linear programming?

# Nonsmooth ISS Lyapunov function

## System

$$\dot{x} = f(x, u), \quad x \in \mathbb{R}^n, u \in \mathbb{R}^m, f(0, 0) = 0$$

## Definition (Nonsmooth ISS Lyapunov function)

A Lipschitz continuous function  $V : \mathcal{D} \rightarrow \mathbb{R}_+$ , with  $\mathcal{D} \subset \mathbb{R}^n$  open is said to be a **local nonsmooth ISS-Lyapunov function** of system if there exist  $\rho^0 > 0$ ,  $\rho^u > 0$ ,  $\varphi_1, \varphi_2, \alpha, \beta \in \mathcal{K}_\infty$  such that  $B(0, \rho^0) \subset \mathcal{D}$  and

$$\begin{aligned} \varphi_1(\|x\|_2) &\leq V(x) \leq \varphi_2(\|x\|_2), \\ \langle \xi, f(x, u) \rangle &\leq -\alpha(\|x\|_2) + \beta(\|u\|_2), \quad \forall \xi \in \partial_d V(x), \end{aligned}$$

for all  $\|x\|_2 \leq \rho^0$ ,  $\|u\|_2 \leq \rho^u$ .

If  $\rho^0 = \rho^u = \infty$  then  $V$  is called a **global nonsmooth ISS Lyapunov function**.

# Outline

- 1 Introduction
- 2 Relationship between ISS and Robust Lyapunov Functions
- 3 Computing Local Robust Lyapunov Functions by Linear Programming
- 4 Computing Local ISS Lyapunov Functions by Linear Programming**
- 5 Example
- 6 Conclusion and Future Works



# Computing Local ISS Lyapunov Functions by Linear Programming

Introduce two Lipschitz continuous functions  $\eta_1 : \mathbb{R}^n \rightarrow \mathbb{R}^n$  and  $\eta_2 : \mathbb{R}^m \rightarrow \mathbb{R}_{\geq 0}$  in **specified formulations**,

$$\eta_1(x) = \sum_{i=0}^n \lambda_i \eta_1(x_i), \text{ for } x = \sum_{i=0}^n \lambda_i x_i \in \Gamma_v, \sum_{i=0}^n \lambda_i = 1, x_i \text{ vertex of } \Gamma_v,$$

$$\eta_2(u) = r \sum_{j=0}^m \mu_j \|u_j\|_2, \text{ for } u = \sum_{j=0}^m \mu_j u_j \in \Gamma_v^u, \sum_{j=0}^m \mu_j = 1, u_j \text{ vertex of } \Gamma_v^u,$$

$r \geq 0$ .

**Auxiliary system:**

$$\dot{x} = f_\eta(x, u) := f(x, u) - \eta_1(x)\eta_2(u).$$

# Computing Local ISS Lyapunov Functions by Linear Programming

Auxiliary system:

$$\dot{x} = f_{\eta}(x, u) := f(x, u) - \eta_1(x)\eta_2(u).$$

Aim: compute a local robust Lyapunov function  $V(x) \in PL(\Omega \setminus B(0, \varepsilon))$  for  $x \in \Gamma_v \in \Omega \setminus B(0, \varepsilon)$  satisfying by linear programming

- $K \geq \langle \nabla V(x), \eta_1(x) \rangle \geq 1$  ( $K > 0$ ).
- $\langle \nabla V(x), f_{\eta}(x, u) \rangle \leq -\|x\|_2$ .

Reasons for excluding  $B(0, \varepsilon)$

- (i)  $0 \in \partial_{cl} V(0)$  may hold which is contradictory with  $\langle \nabla V(x), \eta_1(x) \rangle \geq 1$ .
- (ii)  $\langle \nabla V(x), f_{\eta}(x, u) \rangle \leq -\|x\|_2$ .

# Computing Local ISS Lyapunov Functions by Linear Programming

## Linear programming based algorithm 2

### Linear programming problem 1:

1.  $V(x)$  is positive definite. The constraints are the same as 1. in linear programming based algorithm 1.
2.  $|\nabla V_{v,i}| \leq C_{v,i} \leq G$ . The requirements are the same as 2. in linear programming based algorithm 1.
3. For every  $\Gamma_v \in \mathcal{T}^e$ , and every  $\Gamma_v^u$ , introduce a nonnegative variable  $r$  and demand

$$\langle \nabla V_v, f(x_i, u_j) \rangle - r \|u_j\|_2 + \sum_{k=1}^n C_{vk} (A_x(u, h_x) + A_u(x, h_u)) \leq -\|x_i\|_2.$$

Objective function:  $\min\{r\}$ .

# Computing Local ISS Lyapunov Functions by Linear Programming

## Linear programming based algorithm 2

Linear programming problem 1:

1.  $V(x)$  is positive definite. The constraints are the same as 1. in linear programming based algorithm 1.
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# Computing Local ISS Lyapunov Functions by Linear Programming

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Linear programming problem 1:

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**Objective function:**  $\min\{r\}$ .

# Computing Local ISS Lyapunov Functions by Linear Programming

## Linear programming based algorithm 2

Linear programming problem 2:

4. Introduce a nonnegative variable  $K$  and variables  $\eta_{1,k}(x_i)$  ( $k = 1, 2, \dots, n$ ). For every  $\Gamma_v \in \mathcal{T}^\varepsilon$ , demand

$$1 \leq \langle \nabla V_v, \eta_1(x_i) \rangle, i = 1, 2, \dots, n,$$

$$\langle \nabla V_v, \eta_1(x_i) \rangle \leq K, i = 1, 2, \dots, n,$$

$$\eta_1(x_i) = (\eta_{1,1}(x_i), \eta_{1,2}(x_i), \dots, \eta_{1,n}(x_i)).$$

**Objective function:**  $\min\{K\}$

# Result of linear programming based algorithm 2

Original system:

$$\dot{x} = f(x, u), \quad x \in \mathbb{R}^n, u \in \mathcal{U}_R, f(0, 0) = 0. \quad (3)$$

## Theorem

*If  $f$  satisfies regular conditions and the linear programming problems constructed by the algorithm 2 have feasible solutions, then the values  $V(x_i)$  at all the vertices  $x_i$  of all the simplices  $\Gamma_V \in \mathcal{T}^\varepsilon$  and the condition  $V \in PL(\mathcal{T}^\varepsilon)$  uniquely define the function  $V: \bigcup_{\Gamma_V \in \mathcal{T}^\varepsilon} \Gamma_V \rightarrow \mathbb{R}$ . Furthermore  $V(x)$  is a local ISS Lyapunov function and approximately satisfies*

$$\langle \nabla V_V, f(x, u) \rangle \leq -\|x\|_2 + rK\|u\|_2.$$



# Existence of Solutions of Linear Programming Problems

Original system:

$$\dot{x} = f(x, u), \quad x \in \mathbb{R}^n, u \in \mathcal{U}_R, f(0, 0) = 0. \quad (4)$$

## Theorem

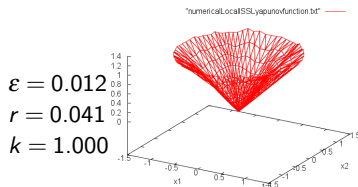
*If  $f$  satisfies regular conditions and system (4) has a local  $C^2$  ISS Lyapunov function  $V^* : \Omega \rightarrow \mathbb{R}$  and let  $\varepsilon > 0$ , then there exist triangulations  $\mathcal{T}^\varepsilon$  and  $\mathcal{T}_u$  such that the linear programming problems constructed by the algorithm 2 have feasible solutions and deliver a local ISS Lyapunov function  $V \in PL(\mathcal{T}^\varepsilon)$ .*

# Example

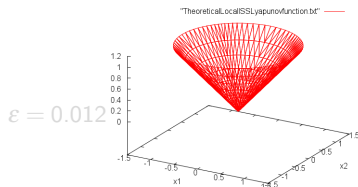
Consider system adapted from [Michel, Sarabudla, Miller, 1982] is described by

$$S_1 : \begin{cases} \dot{x}_1 = -x_1[4 - (x_1^2 + x_2^2)] + 0.1u_1, \\ \dot{x}_2 = -x_2[4 - (x_1^2 + x_2^2)] - 0.1u_2, \end{cases}$$

on  $x = (x_1, x_2)^\top \in [-1.5, 1.5]^2$ ,  $u = (u_1, u_2)^\top \in [-0.6, 0.6]^2$ .



$V(x)$  obtained by the algorithm



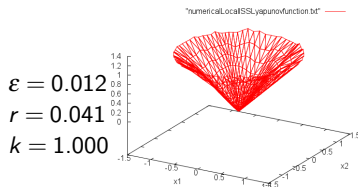
Theoretical  $V_1(x)$  based on  $V^* = \|x\|_2$

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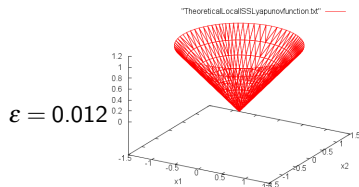
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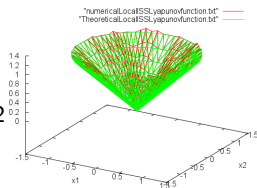
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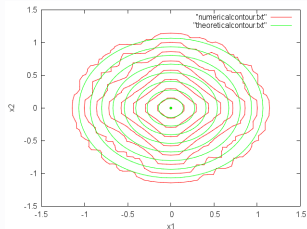
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## Difference between $V(x)$ and $V_1(x)$

$$\varepsilon = 0.012$$

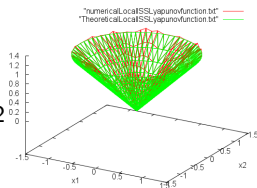


## Difference between contours of $V(x)$ and $V_1(x)$

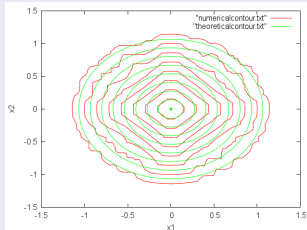


## Difference between $V(x)$ and $V_1(x)$

$$\varepsilon = 0.012$$



## Difference between contours of $V(x)$ and $V_1(x)$



# Conclusion and Future Works

## Conclusion

- A new way of computing local ISS Lyapunov functions is given.

## Future works

- Consider two linear programming problems of algorithm 2 as one quadratic problem.
- Consider the problem of computing local ISS Lyapunov functions without introducing auxiliary systems.
- Estimate the domain of attraction of interconnected systems by small gain theorems with subsystems' local ISS Lyapunov functions obtained by linear programming.

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**Thanks for your attention!**