



**UNIVERSIDADE FEDERAL
DE SANTA CATARINA**

On Robustly Invariant Polyhedral Sets and Bilinear Programming for Designing Constrained Controllers

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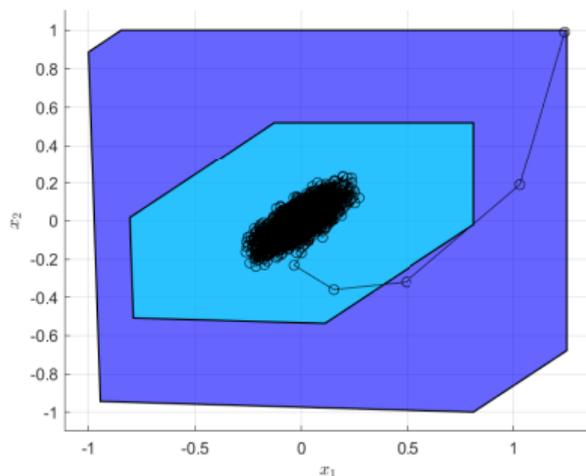
- 1 Introduction
- 2 Set-theoretic output feedback control: A bilinear programming approach
- 3 Output feedback design for LPV systems subject to disturbances and control rate constraints
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- 1 Introduction
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- Considering physical and safety limits that occur in control systems is the primary concern of the *Constrained Control* research field
- *Set- invariance* and *contractivity* are fundamental concepts used to guarantee constraints fulfillment and determine regions of local stability [Tarbouriech et al., 2011, Blanchini and Miani, 2015]
- In practice, state and control constraints and exogenous disturbances are mostly bounded in amplitude and can be represented by *polyhedral sets*. Furthermore, *Output Feedback* (OF) control laws are often required in real-world applications
- **Objective:** To show that **Bilinear Programming** is an effective optimization tool to design **Output Feedback** controllers for constrained LTI and LPV systems using **Polyhedral Set-Invariance**

- *Set-invariance* properties relate convex (C-)sets (ellipsoidal, polyhedral, or composite) to a dynamical system (linear, nonlinear, LPV, or Fuzzy T-S) [Tarbouriech et al., 2011, Blanchini and Miani, 2015]

- For systems subject to persistent disturbances, the property of *Robust Positive Invariance* (RPI) ensures that any trajectory originating from a set within the state space will stay within that set. Additionally, if the set is *contractive*, the trajectory will ultimately be bounded within a subset surrounding the origin



- RPI reduces to the *Positive Invariance* property in the absence of disturbances, and the set contractivity guarantees the convergence of the system's trajectories to the origin [Many authors, 20th Century]
- *Robust Controlled Invariance* (RCI) ensures the existence of a control law that will make a set Robustly Positively Invariant

Lemma (Extended Farkas' Lemma (EFL) [Hennet, 1995])

Consider two polyhedral sets, $\mathcal{P}_i = \{x : P_i x \leq \phi_i\}$, $i = 1, 2$, $P_i \in \mathbb{R}^{l_{P_i} \times n}$, and positive vectors $\phi_i \in \mathbb{R}^{l_{P_i}}$. Then $\mathcal{P}_1 \subseteq \mathcal{P}_2$ iff

$$\begin{array}{l} P_2 x \leq \phi_2 \\ \forall x : P_1 x \leq \phi_1 \end{array} \iff \exists Q \geq 0 ; \begin{array}{l} Q P_1 = P_2 \\ Q \phi_1 \leq \phi_2 \end{array}$$

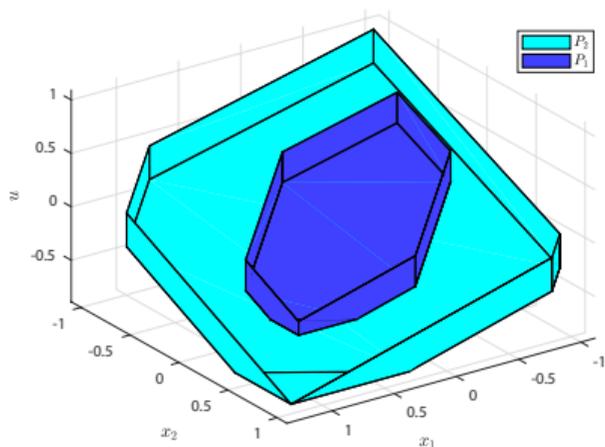


Figure 1: Set Inclusion

Definition

For a given the discrete-time system $x_{k+1} = (A + BKC)x_k$, the polyhedral set

$$\mathcal{L} = \{x_k \in \mathbb{R}^n : Lx_k \leq \mathbf{1}_{l_r}\}, \quad L \in \mathbb{R}^{l_r \times n}, \quad \mathbf{1}_{l_r} = [1 \quad \dots \quad 1]^T$$

is Positively Invariant and λ -contractive, with $\lambda \in [0, 1)$, if and only if

$$Lx_{k+1} = L(A + BKC)x_k \leq \lambda \mathbf{1}_{l_g} \iff \exists H \geq 0; \quad \begin{array}{l} HL = L(A + BKC) \\ H\mathbf{1}_{l_g} \leq \lambda \mathbf{1}_{l_g} \end{array}$$

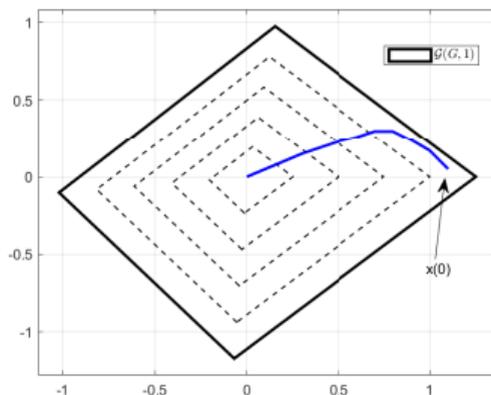


Figure 2: Positive Invariance with λ -contractivity

Glimpse on Positive Invariance and Bilinear Programming

- Brião et al. ***Explicit Computation of Stabilizing Feedback Control Gains Using Polyhedral Lyapunov Functions***. 2018 IEEE ICA-ACCA, Chile:

Min
 $H, K, L, J, \lambda, \gamma$

$$\lambda + \alpha \gamma$$

s.t.

$$HL = L(A + BKC)$$

$$\|H\|_{\infty} \leq \lambda, \quad 0 \leq \lambda < 1$$

$$\|K\|_{\infty} \leq \gamma, \quad 0 \leq \gamma \leq \bar{\gamma}$$

$$LU = I_n$$

$$(\underline{L}, \underline{U}) \leq (L, U) \leq (\bar{L}, \bar{U})$$

Lyapunov function

$$v(x_k) = \|Lx_k\|_{\infty}, \quad L \in \mathbb{R}^{n \times n}$$

with $\text{rank}(L) = n$

- Lower and upper bounds on the unconstrained variables reduce the optimization search space \implies *Bilinear Program can be solved using nonlinear solvers, as e.g. KNITRO, which implements a multistart strategy to find local minima under convergence [Brião et al.(2021)]*

- 1 Briño; Castelan; Ernesto; Camponogara. ***Output feedback design for discrete-time constrained systems subject to persistent disturbances via bilinear programming.*** Journal of the Franklin Institute, 2021.
Asymmetrical constraints and disturbance bounds, Static and Dynamic OF design
- 2 Lucia; Ernesto; Castelan. ***Set-theoretic output feedback control: A bilinear programming approach.*** Automatica, 2023.
- 3 Ernesto, Castelan, Lucia. ***Control-rate Constrained Output Feedback Design for LPV Systems subject to Bounded Disturbance.*** CBA 2024, Brazil.

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Set-theoretic output feedback control: A bilinear programming approach

Lucia, Walter; Ernesto, Jackson G. and Castelan, Eugênio B.
In: **Automatica** 2023

Problem Formulation

Consider the LTI discrete-time system:

$$x_{k+1} = Ax_k + Bu_k + B_p p_k \quad (1a)$$

$$y_k = Cx_k + D_\eta \eta_k, \quad (1b)$$

$$k \in \mathbb{N}, x_k \in \mathbb{R}^n, u_k \in \mathbb{R}^m, y_k \in \mathbb{R}^p, p_k \in \mathbb{R}^s, \eta_k \in \mathbb{R}^q$$

State and control constraints:

$$\mathcal{X} = \{x_k : Xx_k \leq \mathbf{1}_{l_x}\}, \text{ with } X \in \mathbb{R}^{l_x \times n}, \quad (2a)$$

$$\mathcal{U} = \{u_k : Uu_k \leq \mathbf{1}_{l_u}\}, \text{ with } U \in \mathbb{R}^{l_u \times m}, \quad (2b)$$

Bounded persistent disturbances:

$$\mathcal{P} = \{p_k : Pp_k \leq \mathbf{1}_{l_p}\}, \text{ with } P \in \mathbb{R}^{l_p \times s}, \quad (3a)$$

$$\mathcal{N} = \{\eta_k : N\eta_k \leq \mathbf{1}_{l_n}\}, \text{ with } N \in \mathbb{R}^{l_n \times r}. \quad (3b)$$

Definition (Robust Control Invariant (RCI) [Borrelli et al., 2017])

A set $\mathcal{Q} \subseteq \mathcal{X}$ is said RCI for the LTI discrete system under state and control constraints, also subject to the bounded persistent disturbances, if:

$$\begin{aligned} \forall x_k \in \mathcal{Q} \rightarrow \exists u_k \in \mathcal{U} : \\ Ax_k + Bu_k + B_p p_k \in \mathcal{Q}, \quad \forall p_k \in \mathcal{P} \end{aligned} \quad (4)$$

Definition (Robustly One-Step Controllable (ROSC)[Borrelli et al., 2017])

Consider the LTI discrete system under state and control constraints, also subject to the bounded persistent disturbances, and a set $\mathcal{L}_i \subset \mathcal{X}$. The set of states ROSC to \mathcal{L}_i in one-step, namely $\mathcal{L}_{i+1} \subseteq \mathcal{X}$, is defined as:

$$\mathcal{L}_{i+1} := \{ x \in \mathcal{X}, \exists u \in \mathcal{U} : Ax + Bu + B_p p \in \mathcal{L}_i, \forall p \in \mathcal{P} \} \quad (5)$$

Output feedback function:

$$u_k = f(y_k) \quad (6)$$

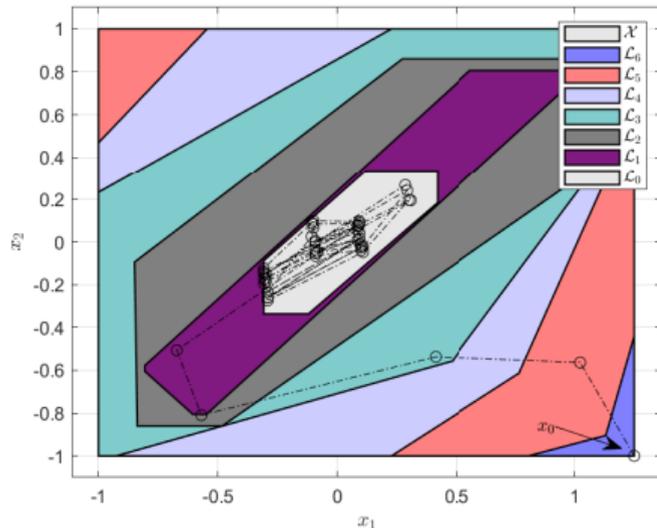
Problem

Find a stabilizing output feedback control function and an associated domain of attraction $\mathcal{L}_D \subseteq \mathcal{X}$, $0_n \in \mathcal{L}_D$ such that $\forall x_0 \in \mathcal{L}_D$ and under the effect of the bounded persistent disturbances, the following properties are met:

- \mathcal{L}_D is a RCI set.
- There exist a small RCI region $\mathcal{L}_0 \subseteq \mathcal{L}_D$, $0_n \in \mathcal{L}_0$ where the state trajectory is ultimately bounded in a finite and a-priori known numbers of steps.
- The state and input constraints are fulfilled.

— Offline computations —

- 1: Build a small terminal RCI region \mathcal{L}_0 and associated SoF gain K_0 ;
- 2: Build a family of \bar{N} ROSC sets $\{\mathcal{L}_i\}$ and associated SoF controller gains $\{K_i\}$, until the set growth saturates;
- 3: Store $\{K_i, \mathcal{L}_i\}_{i=0}^{\bar{N}}$ for online use.



RCI and ROSC sets construction using \mathcal{X}

Defining

$$\mathcal{R}_i = \{x_k \in \mathfrak{R} : R_i x_k \leq \mathbf{1}_{r_i}\}, R_i \in \mathfrak{R}^{r_i \times n} \quad (7)$$

\mathcal{L}_i is described as

$$\mathcal{L}_i = \{x_k \in \mathfrak{R} : L_i x_k \leq \mathbf{1}_{l_{r,i}}\}, L_i \in \mathfrak{R}^{l_{r,i} \times n}, \text{rank}(L_i) = n, \quad (8)$$

where, by construction,

$$L_i = \begin{bmatrix} R_i \\ \delta_i X \end{bmatrix} \quad \text{and} \quad \mathbf{1}_{l_{r,i}} = \begin{bmatrix} \mathbf{1}_{r_i} \\ \mathbf{1}_{l_x} \end{bmatrix}, \quad (9)$$

with set complexity $l_{r,i} = r_i + l_x > n$ and $0 < \delta_i \leq 1, \forall i \implies \mathcal{L}_i \subseteq \mathcal{X}$.

RCI Bilinear Optimization Problem

Min
 $\Gamma_0(\cdot), \delta_0$

δ_0

s.t.

$$H_0 L_0 = L_0 (A + BK_0 C)$$

$\Leftrightarrow \mathcal{L}_0$ is RPI for the system controlled with $u_k = K_0 y_k$:

$$V_0 P = L_0 B_p$$

$$(A + BK_0 C)\mathcal{L}_0 \oplus B_p \mathcal{P} \oplus BK_0 D_\eta \mathcal{N} \subseteq \mathcal{L}_0$$

$$W_0 N = L_0 BK_0 D_\eta$$

From EFL, $[H_0 \ V_0 \ W_0] \geq 0$

$$H_0 \mathbf{1}_{l_{r,0}} + V_0 \mathbf{1}_{l_p} + W_0 \mathbf{1}_{l_\eta} \leq \mathbf{1}_{l_{r,0}}$$

$$M_0 L_0 = UK_0 C$$

\Leftrightarrow Control constraints admissibility

$$Z_0 N = UK_0 D_\eta$$

$$K_0 C \mathcal{L}_0 \oplus K_0 D_\eta \mathcal{N} \subseteq \mathcal{U}$$

$$M_0 \mathbf{1}_{l_{r,0}} + Z_0 \mathbf{1}_{l_\eta} \leq \mathbf{1}_{lu}$$

$$[Z_0 \ M_0] \geq 0$$

$$T_0 S = L_0 \quad , \quad T_0 \mathbf{1}_{l_s} \leq \mathbf{1}_{l_{r,0}}$$

$\Leftrightarrow S \subseteq \mathcal{L}_0$ for good conditioning

$$J_0 L_0 = I_n \quad , \quad 0 < \delta_0 \leq 1$$

$\Leftrightarrow \text{rank}(L) = n$ and $\mathcal{L}_0 \subseteq \delta_0 \mathcal{X}$

$$\underline{\Gamma}_0(\cdot) \leq \Gamma_0(\cdot) \leq \bar{\Gamma}_0(\cdot)$$

\Rightarrow To bound unconstrained variables

ROSC Bilinear Optimization Problem - $i = 1, \dots, \bar{N}$

$$\begin{array}{ll}
 \text{Max} & \mathcal{J}_i = \sum_{t=1}^n \gamma_t \\
 \Gamma_i(\cdot), \delta_i, \gamma_t & \\
 \text{s.t.} & H_i L_i = L_{i-1}(A + BK_i C) \\
 & V_i P = L_{i-1} B_p \\
 & W_i N = L_{i-1} BK_i D_\eta \\
 & H_i \mathbf{1}_{l_r} + V_i \mathbf{1}_{l_p} + W_i \mathbf{1}_{l_\eta} \leq \mathbf{1}_{l_r} \\
 & M_i L_i = UK_i C, \quad Z_i N = UK_i D_\eta \\
 & M_i \mathbf{1}_{l_r} + Z_i \mathbf{1}_{l_\eta} \leq \mathbf{1}_{l_u} \\
 & T_i L_{i-1} = L_i, \quad \delta_{i-1} < \delta_i \leq 1 \\
 & T_i \mathbf{1}_{l_r} \leq \mathbf{1}_{l_r}, \quad T_i \geq 0 \\
 & \gamma_t L_i v_t, \quad t = 1, \dots, t \\
 & J_i L_i = I_n, \quad \underline{\Gamma}_i(\cdot) \leq \Gamma_i(\cdot) \leq \bar{\Gamma}_i(\cdot)
 \end{array}$$

$\Leftrightarrow \mathcal{L}_i \subseteq \mathcal{X}$ is ROSC to \mathcal{L}_{i-1} for the system controlled with $u_k = K_i y_k$:

$$(A + BK_i C)\mathcal{L}_i \oplus B_p \mathcal{P} \oplus BK_i D_\eta \mathcal{N} \subseteq \mathcal{L}_{i-1}$$

From EFL, $[H_i \ V_i \ W_i] \geq 0$

\Leftrightarrow Control constraints admissibility:

$$\begin{aligned}
 K_i C \mathcal{L}_i \oplus K_i D_\eta \mathcal{N} \mathcal{L}_i &\subseteq \mathcal{U} \\
 [Z_i \ M_i] &\geq 0
 \end{aligned}$$

\Leftrightarrow Recursive sets inclusion: $\mathcal{L}_{i-1} \subseteq \mathcal{L}_i$

\Rightarrow To enlarge \mathcal{L}_i in given directions v_t

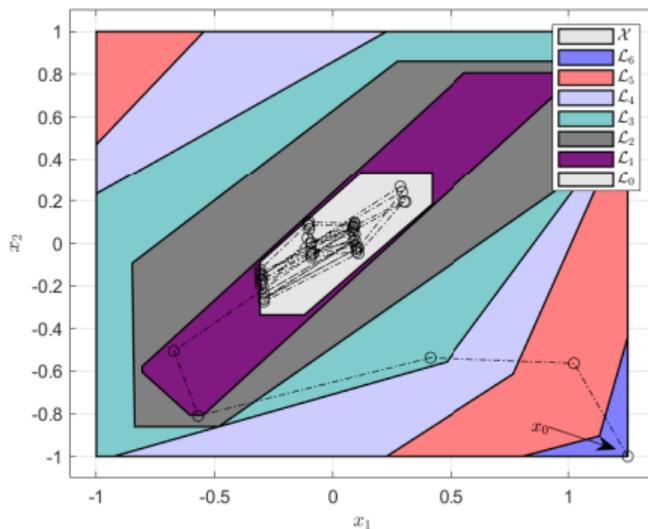
$\Leftrightarrow \text{rank}(L) = n$ and bounded variables

— Online switching rule —
($\forall k, x_0 \in \mathcal{L}_D$)

- 1: Given y_k , compute i_k using Propositions 1 and 2 in [Lucia et al., 2023]:

$$i_k = \begin{cases} \bar{i}_k & \text{if } \text{rank}(C) < n \\ \underline{i}_k & \text{if } \text{rank}(C) = n \end{cases}$$

- 2: Compute and apply $u_k = K_{i_k} y_k$.



Offline numerical complexity

Number of variables, equality and inequality constraints in the RCI and ROSC optimization problems

	RCI set \mathcal{L}_0
# of Variables	$mp + l_0(n^2 + l_0 + l_p + l_n + l_u + l_s) + l_u l_n$
# of Equalities	$l_0(n^2 + s + r) + l_u(n + r) + n^2$
# of Inequalities	$l_0 + l_u + l_s$

	ROSC sets \mathcal{L}_i
# of Variables	$mp + l_{i-1}(n + l_i^2 + l_p + l_n) + l_u(l_i + l_n) + nl_i$
# of Equalities	$l_{i-1}(n + s + r) + l_u(n + r) + l_i n + n^2$
# of Inequalities	$l_{i-1} + l_u + l_i$

Example

Consider the Double Integrator system:

$$x_{k+1} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} x_k + \begin{bmatrix} 2 \\ 1 \end{bmatrix} u_k + \begin{bmatrix} 1 \\ 1 \end{bmatrix} p_k \quad (10a)$$

$$y_k = \begin{bmatrix} 1 & 0 \end{bmatrix} x_k + \eta_k, \quad (10b)$$

subject to

$$-1 \leq x_{k,1} \leq 1.25, \quad |x_{k,2}| \leq 1, \quad -0.8 \leq u_k \leq 1, \quad |p_k| \leq 0.1, \quad |\eta_k| \leq 0.1$$

which implies the matrices

$$X^T = \begin{bmatrix} 0.8 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{bmatrix}, \quad U^T = \begin{bmatrix} 1 & -1.25 \end{bmatrix}$$
$$P^T = \begin{bmatrix} 10 & -10 \end{bmatrix}^T, \quad N^T = \begin{bmatrix} 10 & -10 \end{bmatrix}$$

i	K_i	\mathcal{L}_i Area	\mathcal{J}_i
0	$[-0.7500]$	0.3120	2.3466
1	$[-0.7803]$	0.4054	2.7227
2	$[-0.8686]$	0.7949	4.1941
3	$[-0.8476]$	1.7372	5.9305
4	$[-0.6979]$	2.5962	7.3990
5	$[-0.6111]$	3.0652	8.0587
6	$[-0.6111]$	3.1510	8.2741

Table 1: ST-OF, offline design.

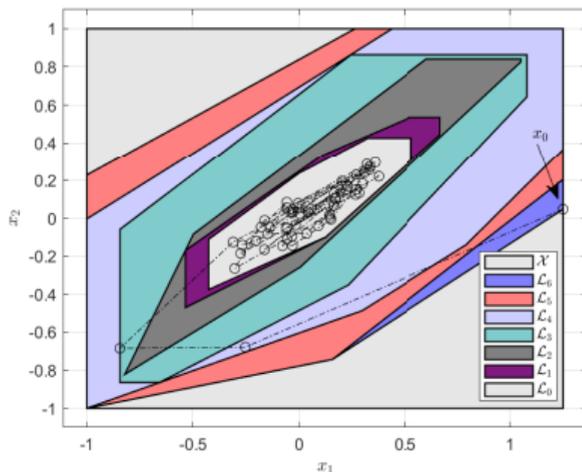


Figure 3: ST-OF: DoA for $(\bar{t} = 8, r = 4,)$, and state trajectory for $x_0 = [1.25, 0.047]^T \in \mathcal{L}_6$.

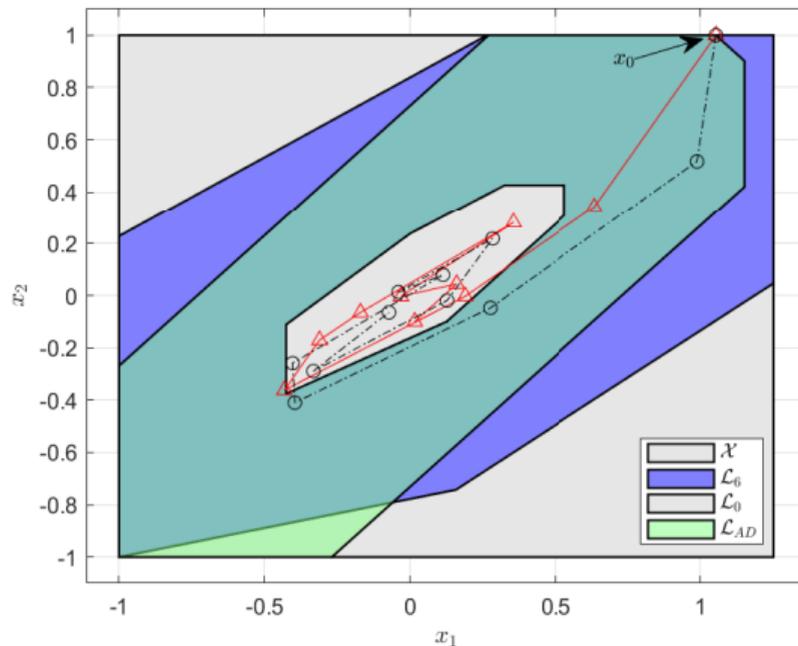


Figure 4: DoA and state trajectory: ST-OF \circ vs \triangle De Almeida and Dorea [2020]
(online optimization to find u_k)

$$Area_{\mathcal{L}_6} = 3.1510 \quad \text{vs} \quad Area_{\mathcal{L}_{AD}} = 2.4837$$

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Control-rate Constrained Output Feedback Design for LPV Systems Subject to Disturbances

Ernesto, Jackson G., Eugênio B. Castelan e Walter Lucia.
In: **CBA 2024**.

Problem Formulation

LPV discrete-time system

$$x_{k+1} = A(\alpha_k)x_k + B(\alpha_k)u_k + B_p(\alpha_k)p_k \quad (11a)$$

$$y_k = Cx_k + D_\eta\eta_k \quad (11b)$$

$$[A(\alpha_k) \quad B(\alpha_k) \quad B_p(\alpha_k)] = \sum_{i=1}^{\nu} \alpha_{k,i} [A_i \quad B_i \quad B_{pi}], \quad \alpha_k \in \mathcal{S}_{\text{simplex}}$$

State, control input and rate variation constraints:

$$\mathcal{X} = \{x_k : Xx_k \leq \mathbf{1}_x\}, \quad X \in \mathbb{R}^{l_x \times n_x} \quad (12a)$$

$$\mathcal{U} = \{u_k : Uu_k \leq \mathbf{1}_u\}, \quad U \in \mathbb{R}^{l_u \times n_u} \quad (12b)$$

$$\mathcal{U}_d = \{\delta u_k : U_d\delta u_k \leq \mathbf{1}_d\}, \quad U_d \in \mathbb{R}^{l_d \times n_u}, \quad \delta u_k = u_{k+1} - u_k \quad (12c)$$

Bounded persistent disturbances:

$$\mathcal{P} = \{p_k : Pp_k \leq \mathbf{1}_p\}, \quad P \in \mathbb{R}^{l_p \times n_p} \quad (13a)$$

$$\mathcal{N} = \{\eta_k : N\eta_k \leq \mathbf{1}_n\}, \quad N \in \mathbb{R}^{l_n \times n_\eta} \quad (13b)$$

Augmented state vector:

$$\xi_k = \begin{bmatrix} x_k^T & u_k^T \end{bmatrix}^T \in \mathbb{R}^{n_\xi}, \quad n_\xi = n_x + n_u \quad (14)$$

Augmented output vector:

$$v_k = \begin{bmatrix} y_k^T & u_k^T & y_{k+1}^T \end{bmatrix}^T \in \mathbb{R}^{n_v}, \quad n_v = 2n_y + n_u \quad (15)$$

Parameter-varying control increment input vector:

$$\delta u_k = \begin{bmatrix} K(\alpha_k) & \bar{K}(\alpha_k) & \hat{K} \end{bmatrix} \begin{bmatrix} y_k \\ u_k \\ y_{k+1} \end{bmatrix} \quad (16)$$

$$\begin{bmatrix} K(\alpha_k) & \bar{K}(\alpha_k) & \hat{K} \end{bmatrix} = \sum_{i=1}^{\nu} \alpha_{k,i} \begin{bmatrix} K_i & \bar{K}_i & \hat{K} \end{bmatrix}$$

$$K_i \in \mathbb{R}^{n_u \times n_y}, \quad \bar{K}_i \in \mathbb{R}^{n_u \times n_u}, \quad \forall i = 1, \dots, \nu, \quad \text{and} \quad \hat{K} \in \mathbb{R}^{n_u \times n_y}$$

Closed-loop augmented system:

$$\xi_{k+1} = \mathbb{A}^{cl}(\alpha_k)\xi_k + \mathbb{B}_d^{cl}(\alpha_k)d_k \quad (17)$$

$$\left[\mathbb{A}^{cl}(\alpha_k) \quad \mathbb{B}_d^{cl}(\alpha_k) \right] = \sum_{i=1}^{\nu} \alpha_{k,i} \left[\mathbb{A}_i^{cl} \quad \mathbb{B}_{d,i}^{cl} \right]$$

$$\mathbb{A}_i^{cl} = \begin{bmatrix} A_i & & \\ (K_i C + \hat{K} C A_i) & (\bar{K}_i + \hat{K} C B_i) + I & \end{bmatrix} = \begin{bmatrix} E_i & \\ \mathbb{E}_i + [0 \ I] & \end{bmatrix}$$

$$\mathbb{B}_{d,i}^{cl} = \begin{bmatrix} B_{p,i} & 0 & 0 \\ \hat{K} C B_{p,i} & K_i D_\eta & \hat{K} D_\eta \end{bmatrix} = \begin{bmatrix} F_i \\ \mathbb{F}_i \end{bmatrix}, \quad d_k = \begin{bmatrix} p_k \\ \eta_k \\ \eta_{k+1} \end{bmatrix} \in \mathfrak{R}^{n_d}$$

Augmented state constraints:

$$\Xi = \{\xi_k : \mathbb{X}\xi_k \leq \mathbf{1}_{l_\xi}\}, \quad \Xi = \begin{bmatrix} X & 0 \\ 0 & U \end{bmatrix} \in \mathbb{R}^{l_\xi \times n_\xi}, \quad (18)$$

Augmented bounded disturbance:

$$\Delta = \{d_k : \mathbb{D}d_k \leq \mathbf{1}_{l_\Delta}\}, \quad \mathbb{D} = \begin{bmatrix} P & 0 & 0 \\ 0 & N & 0 \\ 0 & 0 & N \end{bmatrix} \in \mathbb{R}^{l_\Delta \times n_d} \quad (19)$$

Definition

$\mathcal{L} \in \mathbb{R}^{n_\xi}$ is a contractive robust positive invariant (RPI-)set of the LPV system, with ultimately bounded (UB-)set $\mathcal{L}^0 \subseteq \mathcal{L}$, if for any $\xi_0 = [x_0^T \quad u_0^T]^T \in \mathcal{L}$ and $d_k = [p_k^T \quad \eta_k^T \quad \eta_{k+1}^T]^T \in \Delta$, the corresponding state trajectory remains inside \mathcal{L} , converge to \mathcal{L}^0 in a finite number of steps, and remains ultimately bounded within \mathcal{L}^0 , for all $\alpha_k \in \mathcal{S}$.

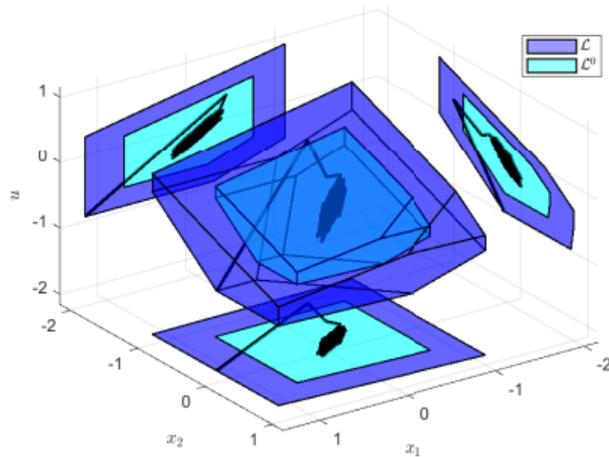
Consider the polyhedral sets:

$$\mathcal{L} = \{\xi_k : \mathbb{L}\xi_k \leq \mathbf{1}_{l_r}\},$$

$$\mathcal{L}^0 = \{\xi_k : \mathbb{L}\xi_k \leq \rho \mathbf{1}_{l_r}\}$$

$$\mathbb{L} \in \mathbb{R}^{l_r \times n_\xi}, \quad \text{rank}(\mathbb{L}) = n_\xi,$$

set complexity $l_r > n$, and $0 < \rho \leq 1$



Problem

Given l_r , find control gains $(K_i, \bar{K}_i, \hat{K})$ and a triplet $(\mathbb{L}, \lambda, \rho)$, which defines a large contractive RPI set $\mathcal{L} \subseteq \Xi$ and a small UB-set $\mathcal{L}^0 \subseteq \mathcal{L}$, such that, for any initial condition $\xi_0 \in \mathcal{L}$, $d_k \in \Delta$, and for all $\alpha_k \in \mathcal{S}$, the state, control, and control-rate variation constraints, $\mathcal{U}_d = \{U_d \delta u_k \leq \mathbf{1}_{l_d}\}$, are fulfilled.

Maximizing the size of \mathcal{L}

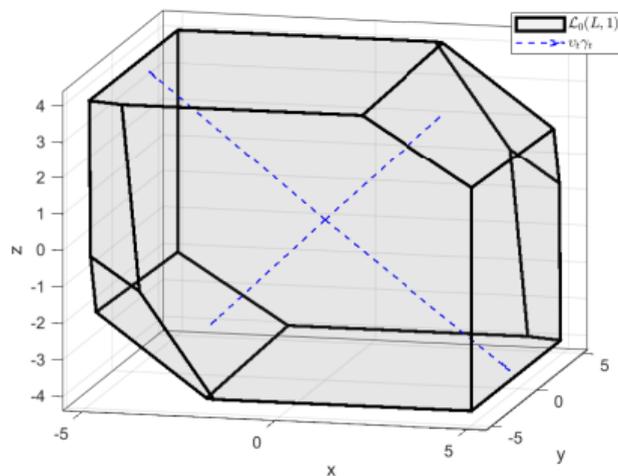
Auxiliary inequalities:

$$\gamma_t \mathbb{L} \psi_t \leq \mathbf{1}_l, t = 1, \dots, \bar{t} \quad (20)$$

where $\gamma_t \in \mathfrak{R}$ are positive scaling factors associated to the a pre-defined set of \bar{t} directions

$$\Psi = \{\gamma_t \psi_t, t = 1, \dots, \bar{t}\} \quad (21)$$

with $\psi_t = [\psi_{x,t}^T \quad \psi_{u,t}^T]^T$, $\psi_{x,t} \in \mathfrak{R}^{n_x}$ and $\psi_{u,t} \in \mathfrak{R}^{n_u}$, which can be set as a variable $\psi_{u,t}$



Bilinear Optimization Problem

$$\begin{array}{ll}
 \text{Max}_{\Gamma(\cdot)} & \sum_{t=1}^{\bar{t}} \gamma_t - \alpha \rho \\
 \text{s.t.} & H_i \mathbb{L} = \mathbb{L} \mathbb{A}_i^{cl} \quad , \quad H_i \geq 0 \\
 & V_i \mathbb{D} = \mathbb{L} \mathbb{B}_{d,i}^{cl} \quad , \quad V_i \geq 0 \\
 & H_i \mathbf{1}_{l_r} + V_i \mathbf{1}_{l_\Delta} \leq \lambda \mathbf{1}_{l_r} \\
 & H_i \rho \mathbf{1}_{l_r} + V_i \mathbf{1}_{l_\Delta} \leq (1 - \epsilon) \rho \mathbf{1}_{l_r} \\
 & \mathbb{G} \mathbb{L} = \mathbb{X} \quad , \quad \mathbb{G} \geq 0 \\
 & \mathbb{G} \mathbf{1}_{l_r} \leq \mathbf{1}_{l_\xi} \\
 & Q_i \mathbb{L} = U_d \mathbb{E}_i \quad , \quad Q_i \geq 0 \\
 & T_i \mathbb{D} = U_d \mathbb{F}_i \quad , \quad T_i \geq 0 \\
 & Q_i \mathbf{1}_{l_r} + T_i \mathbf{1}_{l_\Delta} \leq \mathbf{1}_{l_d} \\
 & \mathbb{J} \mathbb{L} = I_{n_\xi} \quad , \quad \gamma_t \mathbb{L} \psi_t \leq \mathbf{1}_{l_r} \\
 & \underline{\Gamma}(\cdot) \leq \Gamma(\cdot) \leq \bar{\Gamma}(\cdot)
 \end{array}$$

\Leftrightarrow RPI of \mathcal{L} , with λ -contractivity

\Leftrightarrow RPI of the UB-set $\mathcal{L}^0 \subseteq \mathcal{L}$

$\Leftrightarrow \mathcal{L} \subseteq \Xi$: state and control constraints fulfilment

\Leftrightarrow Control increment admissibility

$\Leftrightarrow \text{rank}(\mathbb{L}) = n_\xi$ and set enlargement in given directions

Example

LPV discrete-time system

$$x_{k+1} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} x_k + \begin{bmatrix} [2, 2.25] \\ 1 \end{bmatrix} u_k + \begin{bmatrix} 1 \\ 1 \end{bmatrix} p_k \quad (22a)$$

$$y_k = \begin{bmatrix} 1 & 0 \end{bmatrix} x_k + \eta_k, \quad (22b)$$

subject to

$$-1 \leq x_{k,1} \leq 1.25, \quad |x_{k,2}| \leq 1, \quad -0.8 \leq u_k \leq 1, \quad |p_k| \leq 0.1, \quad |\eta_k| \leq 0.1$$

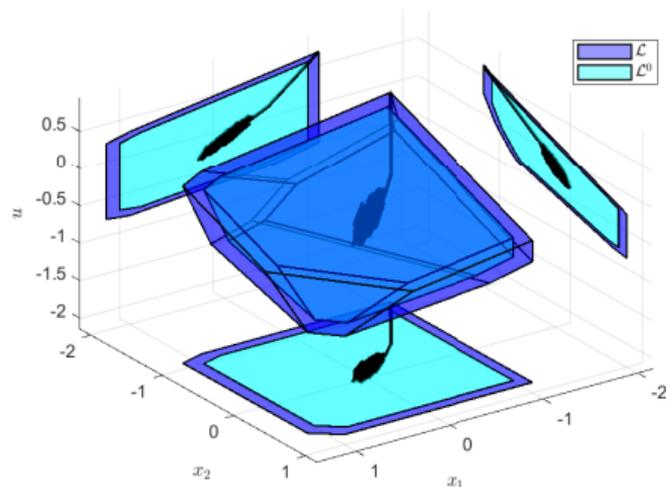
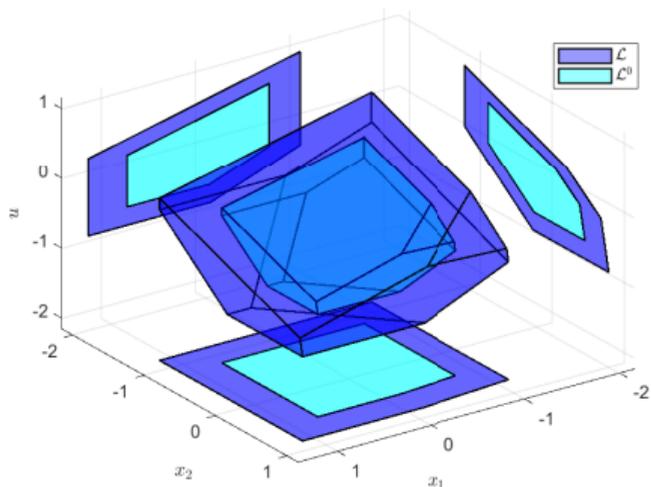
which implies the matrices

$$X^T = \begin{bmatrix} 0.8 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{bmatrix}, \quad U^T = \begin{bmatrix} 1 & -1.25 \end{bmatrix}$$

$$P^T = \begin{bmatrix} 10 & -10 \end{bmatrix}^T, \quad N^T = \begin{bmatrix} 10 & -10 \end{bmatrix}$$

Table 2: Design with $l_r = 9$, $\bar{t} = 16$, $\alpha = 10$, and $\psi_{u,t}$ as a variable

δu_k bounds	\mathcal{L} Vol.	Pr.Area	ρ	$[K_i \quad \bar{K}_i \quad \hat{K}]$
without	2.4217	4.4656	0.6710	$\begin{bmatrix} 0.4431 & -0.5420 & -0.7376 \\ 0.4661 & -0.4377 & -0.7376 \end{bmatrix}$
$[-0.9, 0.6]$	1.3584	4.4679	0.8864	$\begin{bmatrix} 0.4033 & -0.5366 & -0.6016 \\ 0.4033 & -0.4317 & -0.6016 \end{bmatrix}$
$[-0.7, 0.5]$	1.1997	4.4028	0.9980	$\begin{bmatrix} 0.3743 & -0.5575 & -0.5551 \\ 0.3734 & -0.4533 & -0.5551 \end{bmatrix}$



Improved $\delta u_k = \mathcal{K}(\alpha_k)y_k + \bar{\mathcal{K}}(\alpha_k)u_k + \hat{\mathcal{K}}(\alpha_{k+1})y_{k+1}$

- Improved $\delta u_k = u_{k+1} - u_k$ implies

$$\mathbb{A}^{cl}(\alpha_k, \alpha_{k+1}) \text{ and } \mathbb{B}^{cl}(\alpha_k, \alpha_{k+1})$$

- Modified $\mathcal{L}_0 = \{\xi_k ; \mathbb{L}\xi_k \leq \rho\}$,
with

$$\rho = [\rho_1 \ \dots \ \rho_{l_r}]^T$$

- Modified objective function

$$\text{Max} \sum_{t=1}^{\bar{t}} \frac{\gamma_t}{\bar{t}} - \alpha \sum_{i=1}^{l_r} \frac{\rho_i}{l_r}$$

- Numerical complexity increases but
less conservative results are obtained

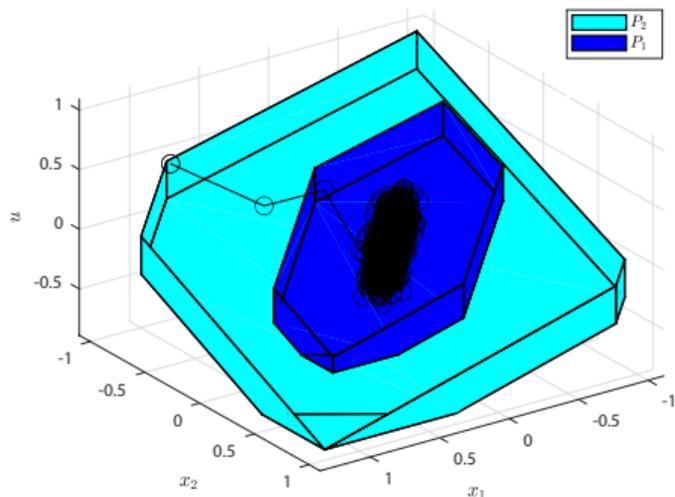


Figure 7

- 1 Introduction
- 2 Set-Theoretic Control
- 3 Constrained LPV Systems under control rate limits
- 4 Concluding Remarks**

Conclusion

- Bilinear programming is an effective optimization tool to design output feedback controllers for constrained LTI and LPV systems through polyhedral set-invariance
- Bilinear optimization problems were solved using the KNITRO solver - Artelys. Free access from <https://neos-server.org/neos/>
- Explicit computation of the control feedback matrices allows for specific consideration of control gain structures and the design of reduced-order dynamical controllers and decentralized control laws
- Ongoing collaborations: time-delay and second-order systems, *PID-like* control design for reference tracking and disturbance rejection
- For dealing with the numerical complexity issue in higher-dimensional and *Complex Systems*, one can explore the system and controller structures

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