

CNRS - RAS cooperation

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Quadratic Separation for Robustness and Design

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What is LAAS-CNRS?

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French National Center for Scientific Research.

- Public basic-research org. producing knowledge and making it available to society.
- 26,000 employees (11,600 researchers).
- 1,260 units, spread throughout the country, cover all fields of research.



Analysis and Architecture of Systems

- Part of CNRS - STIC department
(Science and Technology for Information and Communication)
- 500 employees (200 researchers)
- 12 research groups
- **Control Theory**, Robotics, Micro and Nano-Systems, Computer science
- In Toulouse, France.



MAC group

<http://www.laas.fr/MAC>

- ❑ Research fields : **Robust control** & Non-linear control
- ❑ Application fields : Aeronautics & **Space industry** & Environment

Research in robust control

- ❑ MIMO LTI systems with parametric uncertainty
- ❑ State-space modeling and Lyapunov theory
- ❑ Stability and performance (H_∞ , H_2 , pole location, impulse to peak)
- ❑ Analysis & Control design (state-feedback, full-order and static output-feedback)
- ❑ LMI based results & Semi-definite programming
- ❑ Robust MULTI-Objective Control toolbox (V1 in September)
<http://www.laas.fr/OLOCEP>

Quadratic separation for LTI systems

Examples of results for robustness and design

- Preliminaries and notations
- Robust analysis and losslessness of quadratic separators
- Quadratic separation and control design

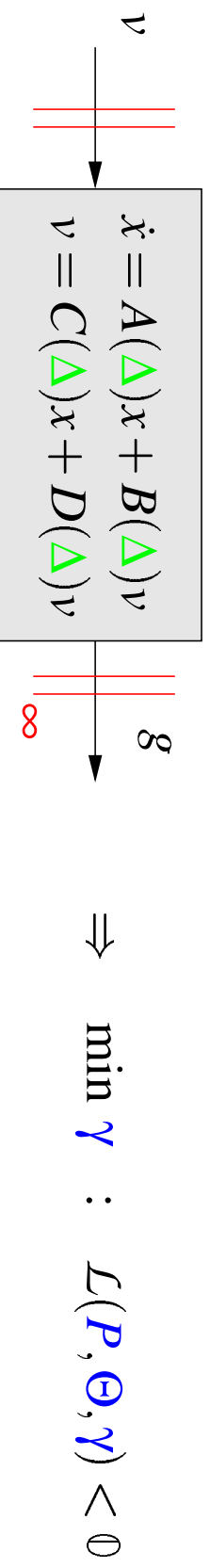
Uncertain model

- Engineering problem modeled as uncertain differential equations with constraints
- State-space LTI systems / parametric uncertainty / pole and induced norm constraints



Optimization problem

- At best: lossless formulation with a global polynomial-time algorithm
- Conveniently: Conservative formulation with a global polynomial-time algorithm
- Usually: Conservative with sub-optimal heuristic algorithm
- LMI formulated results \Rightarrow convex SDP & $O(n^{6.5})$ algorithms
- YALMIP interface in Matlab & Solvers: SeDuMi, SDPT3, CSDP,...

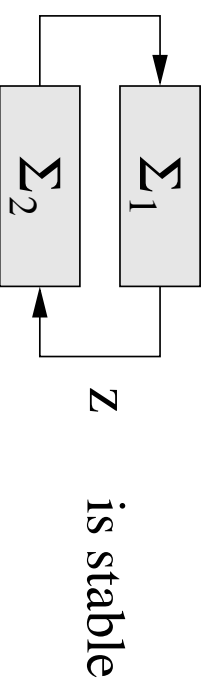


Graph of Σ_1 and inverse graph of Σ_2 :

$$\mathcal{G}(\Sigma_1) = \left\{ x = \begin{pmatrix} z \\ w \end{pmatrix} : \Sigma_1(z, w) = 0 \right\} \quad \mathcal{G}^I(\Sigma_2) = \left\{ x = \begin{pmatrix} z \\ w \end{pmatrix} : \Sigma_2(w, z) = 0 \right\}$$

Stability result [Safonov]:

The interconnected system



□ iff $\mathcal{G}(\Sigma_1) \cap \mathcal{G}^I(\Sigma_2) = \{0\}$

□ iff $\exists d$:
$$\begin{cases} d(x) > 0, \forall x \in \mathcal{G}(\Sigma_1) \\ d(x) \leq 0, \forall x \in \mathcal{G}^I(\Sigma_2) \end{cases}$$

d : topological separator (see also “supply rate” in dissipative theory [Willems])

Quadratic Separation

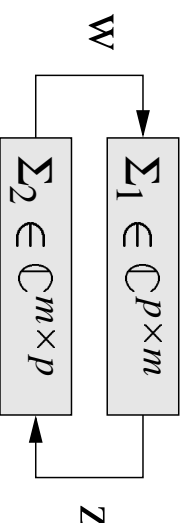
Quadratic function for the topological separator:

$$d(x) = x^* \Theta x \quad \Theta = \Theta^* \in \mathbb{C}^{(m+p) \times (m+p)}$$

Lossless for linear systems

E.g. for matrix gains

The interconnected system

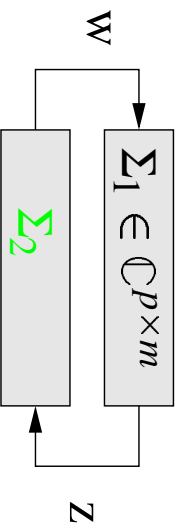


is well-posed

$$\square \text{ iff } \det(\mathbb{1} - \Sigma_1 \Sigma_2) \neq 0$$

$$\square \text{ iff } \exists \Theta : \left\{ \begin{array}{l} w^* \begin{bmatrix} \Sigma_1^* & \mathbb{1} \end{bmatrix} \Theta \begin{bmatrix} \Sigma_1 \\ \mathbb{1} \end{bmatrix} w > 0, \quad \forall w \neq 0 \\ z^* \begin{bmatrix} \mathbb{1} & \Sigma_2^* \end{bmatrix} \Theta \begin{bmatrix} \mathbb{1} \\ \Sigma_2 \end{bmatrix} z \leq 0, \quad \forall z \neq 0 \end{array} \right.$$

Robust analysis



is robustly well-posed for all $\Sigma_2 \in \mathcal{U} \subset \mathbb{C}^{m \times p}$

□ iff $\det(\mathbb{1} - \Sigma_1 \Sigma_2) \neq 0$, $\forall \Sigma_2 \in \mathcal{U}$

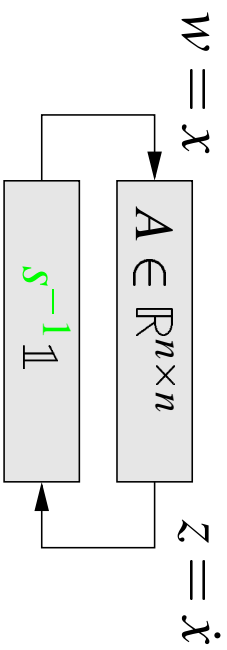
$$\square \text{ iff } \exists \Theta : \begin{cases} \begin{bmatrix} \Sigma_1^* & \mathbb{1} \end{bmatrix} \Theta \begin{bmatrix} \Sigma_1 \\ \mathbb{1} \end{bmatrix} > 0 \\ \begin{bmatrix} \mathbb{1} & \Sigma_2^* \end{bmatrix} \Theta \begin{bmatrix} \mathbb{1} \\ \Sigma_2 \end{bmatrix} \leq 0, \forall \Sigma_2 \in \mathcal{U} \end{cases}$$

→ Quadratic separation results are LMI-based.

→ Is it possible to handle the infinite-dimensional constraint without conservatism?

One interconnected operator is uncertain

Example: Lyapunov matrix

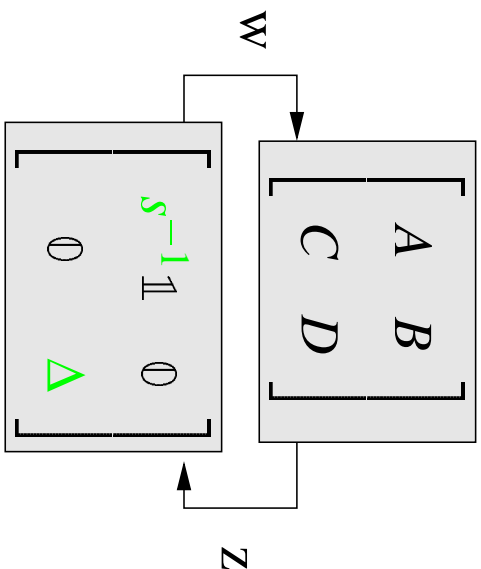


is stable (interconnection well-posed for all $s^{-1} \in \mathbb{C}^+$)

- iff $\exists \Theta : \begin{cases} \begin{bmatrix} A^T & \mathbb{1} \end{bmatrix} \Theta \begin{bmatrix} A \\ \mathbb{1} \end{bmatrix} > 0 \\ \begin{bmatrix} \mathbb{1} & s^{-*} \mathbb{1} \end{bmatrix} \Theta \begin{bmatrix} \mathbb{1} \\ s^{-1} \mathbb{1} \end{bmatrix} \leq 0, \forall s^{-1} \in \mathbb{C}^+ \end{cases}$
 - iff $\exists \Theta = \begin{bmatrix} 0 & -P \\ -P & 0 \end{bmatrix} : P > 0, \begin{bmatrix} A^T & \mathbb{1} \end{bmatrix} \Theta \begin{bmatrix} A \\ \mathbb{1} \end{bmatrix} = -A^T P - PA > 0$
- ↳ Lossless quadratic separator.
 ↳ Other sets than $\mathbb{C}^+ \supseteq$ pole location.

One interconnected operator is uncertain

Example: bounded real lemma



is robustly stable ($s^{-1} \in \mathbb{C}^+$, $\Delta^T \Delta \leq \mathbb{1}$)

□ iff there exists a separator such as:

$$\Theta = \begin{bmatrix} 0 & 0 & -P & 0 \\ 0 & -\tau \mathbb{1} & 0 & 0 \\ -P & 0 & 0 & 0 \\ 0 & 0 & 0 & \tau \mathbb{1} \end{bmatrix}, \quad \begin{matrix} P > 0 \\ \tau > 0 \end{matrix}$$

that satisfies the LMI constraint:

$$\begin{bmatrix} \Sigma_1^* & \mathbb{1} \end{bmatrix} \Theta \begin{bmatrix} \Sigma_1 \\ \mathbb{1} \end{bmatrix} = - \begin{bmatrix} A^T P + P A + \tau C^T C & P B + \tau C^T D \\ B^T P + \tau D^T C & -\tau \mathbb{1} + \tau D^T D \end{bmatrix} > 0$$

↳ Lossless quadratic separator.

Lossless quadratic separators

- Full-block dissipative

(S-procedure)

$$\begin{bmatrix} \mathbb{1} & \Delta_D^* \\ X & Y \\ Y^* & Z \end{bmatrix} \begin{bmatrix} \mathbb{1} \\ \Delta_D \end{bmatrix} \leq 0 \quad \Leftrightarrow \quad \Theta = \tau \begin{bmatrix} X & Y \\ Y^* & Z \end{bmatrix}, \quad \tau > 0$$

- Disk located, repeated, complex valued scalar

$$\Delta = \delta_c \mathbb{1} \quad \Leftrightarrow \quad \Theta = \begin{bmatrix} \alpha P & \beta P \\ \beta^* P & \gamma P \end{bmatrix}, \quad P > 0$$

$$\alpha + \delta_c \beta + \delta_c^* \beta^* + \delta_c \delta_c^* \gamma \leq 0$$

- Bounded, repeated, real valued scalar

$$\Delta = \delta_r \mathbb{1} \quad \Leftrightarrow \quad \Theta = \begin{bmatrix} \alpha P & \beta P + Q \\ \beta P + Q^* & \gamma P \end{bmatrix}, \quad P > 0$$

$$\alpha + 2\delta_r \beta + \delta_r^2 \gamma \leq 0 \quad , \quad Q = -Q^*$$

Conservative quadratic separators for block diagonal uncertainty

□ Repeated full-block dissipative

$$\Delta = \mathbb{1}_r \otimes \Delta_D$$

$$\begin{bmatrix} \mathbb{1} & \Delta_D^* \\ Y^* & Z \end{bmatrix} \begin{bmatrix} X & Y \\ Y^* & Z \end{bmatrix} \begin{bmatrix} \mathbb{1} \\ \Delta_D \end{bmatrix} \leq 0$$

□ Block diagonal polytopic

$$\Delta = \text{diag}(\Delta_1, \dots, \Delta_r)$$

$$\Delta = \sum \zeta_i \Delta^{[i]} \quad : \quad \sum \zeta_i = 1, \quad \zeta_i \geq 0$$

□ Any block diagonal structure of previous types (lossless if $2(m_r + m_c) + m_D \leq 3$)

$$\Delta = \text{diag}(\Delta_1, \dots) \Rightarrow$$

$$\Theta = \begin{bmatrix} \text{diag}(\Theta_1^{11}, \dots) & \text{diag}(\Theta_1^{12}, \dots) \\ \text{diag}(\Theta_1^{21}, \dots) & \text{diag}(\Theta_1^{22}, \dots) \end{bmatrix}$$

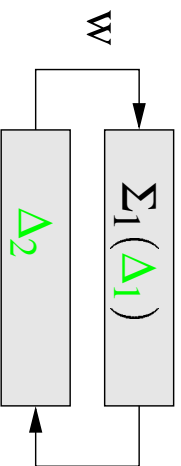
$$\Theta = \begin{bmatrix} T \otimes X & T \otimes Y \\ T \otimes Y^* & T \otimes Z \end{bmatrix}$$

$$T > 0$$

$$\Rightarrow \begin{bmatrix} \mathbb{1} & \Delta^{[i]} \\ \Delta^{[i]} & \mathbb{1} \end{bmatrix} \Theta \begin{bmatrix} \mathbb{1} \\ \Delta^{[i]} \end{bmatrix} \leq 0$$

$$\Theta_{ff}^{22} \geq 0$$

Robust analysis: parameter-dependent separators

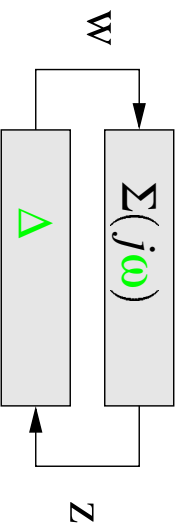


is robustly well-posed for all $\forall \Delta_1 \in \mathcal{U}_1$, $\forall \Delta_2 \in \mathcal{U}_2$

$$\square \text{ iff } \forall \Delta_1 \in \mathcal{U}_1 \exists \Theta(\Delta_1) : \left\{ \begin{array}{l} \left[\Sigma_1^*(\Delta_1) \quad \mathbb{1} \right] \Theta(\Delta_1) \left[\begin{array}{l} \Sigma_1(\Delta_1) \\ \mathbb{1} \end{array} \right] > 0 \\ \left[\mathbb{1} \quad \Delta_2^* \right] \Theta(\Delta_1) \left[\begin{array}{l} \mathbb{1} \\ \Delta_2 \end{array} \right] \leq 0, \forall \Delta_2 \in \mathcal{U}_2 \end{array} \right.$$

→ Infinite number of LMI variables & infinite number of constraints

Example: μ -analysis



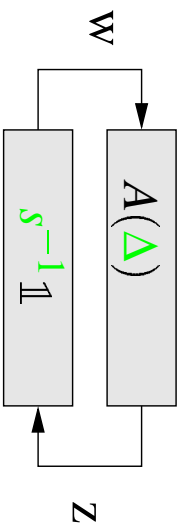
is robustly stable ($\omega \in \mathbb{R}^+$, $\Delta \in \mathcal{U}$)

$$\left\{ \begin{array}{l} \left[\begin{array}{cc} \Sigma^*(j\omega) & \mathbb{1} \end{array} \right] \Theta(j\omega) \left[\begin{array}{c} \Sigma(j\omega) \\ \mathbb{1} \end{array} \right] > 0 \\ \left[\begin{array}{cc} \mathbb{1} & \Delta^* \end{array} \right] \Theta(j\omega) \left[\begin{array}{c} \mathbb{1} \\ \Delta \end{array} \right] \leq 0, \forall \Delta \in \mathcal{U} \end{array} \right.$$

□ iff $\forall \omega \in \mathbb{R}^+ \exists \Theta(j\omega) :$

- ↪ An optimistic bound on μ can then be obtained by gridding $\{\omega_1, \dots, \omega_N\} \subset \mathbb{R}^+$.
- ↪ For each ω_i , build finite dimensional LMIs.

Example: Parameter-dependent Lyapunov Function



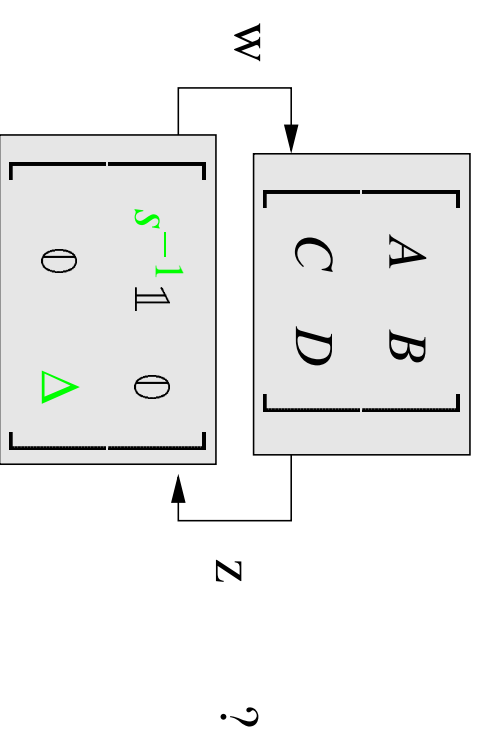
is robustly stable ($s^{-1} \in \mathbb{C}^+$, $\Delta \in \mathcal{U}$)

$$\square \text{ iff } \forall \Delta \in \mathcal{U} \exists P(\Delta) : \begin{cases} A^T(\Delta)P(\Delta) + P(\Delta)A(\Delta) < 0 \\ P(\Delta) > 0 \end{cases}$$

Question:

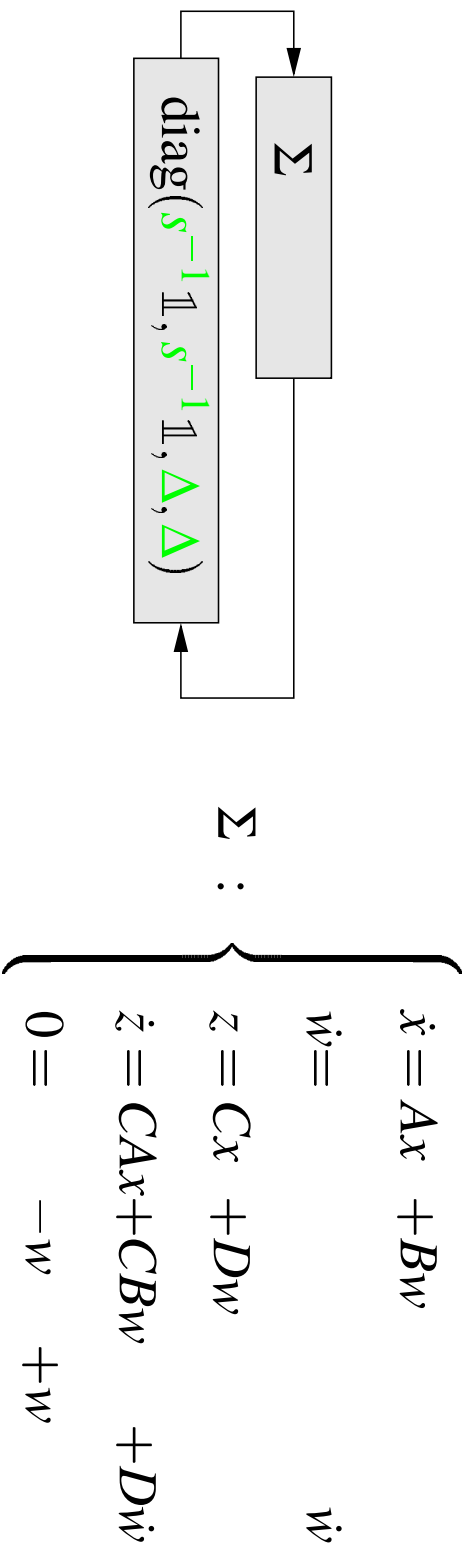
When $A(\Delta) = A + B\Delta(\mathbb{1} - D\Delta)^{-1}C$

how is this result related to the robust stability of



Answer:

Consider the implicit augmented system

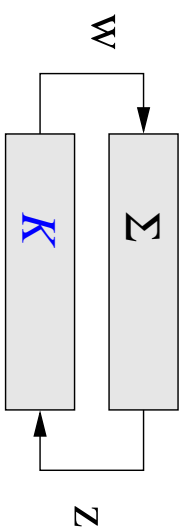


Quadratic separation \Rightarrow parameter-dependent Lyapunov matrix

$$P(\Delta) = \begin{bmatrix} \mathbb{1} & C^T (\mathbb{1} - D\Delta)^{-T} \Delta^T \\ P & \begin{bmatrix} \mathbb{1} \\ \Delta(\mathbb{1} - D\Delta)^{-1} C \end{bmatrix} \end{bmatrix}$$

Prospective work:

- In relation with [Bliman], build asymptotically lossless P-D Lyapunov functions
- Take into account information on the derivatives of Δ .



There exists a matrix K such that Σ is stable

$$\square \text{ iff } \exists K, \exists \Theta : \begin{bmatrix} \Sigma^* & \mathbb{1} \end{bmatrix} \Theta \begin{bmatrix} \Sigma \\ \mathbb{1} \end{bmatrix} > 0, \quad \begin{bmatrix} \mathbb{1} & K^* \end{bmatrix} \Theta \begin{bmatrix} \mathbb{1} \\ K \end{bmatrix} \leq 0$$

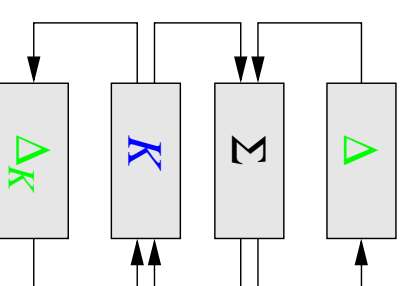
$$\square \text{ iff } \exists \Theta = \begin{bmatrix} X & Y \\ Y^* & Z \end{bmatrix} : \begin{cases} \begin{bmatrix} \Sigma^* & \mathbb{1} \end{bmatrix} \Theta \begin{bmatrix} \Sigma \\ \mathbb{1} \end{bmatrix} > 0 \\ X \leq YZ^{-1}Y^* \quad Z > 0 \end{cases}$$

\Rightarrow All matrices K such that $\begin{bmatrix} \mathbb{1} & K^* \end{bmatrix} \Theta \begin{bmatrix} \mathbb{1} \\ K \end{bmatrix} \leq 0$, stabilize the interconnection.

\hookrightarrow The non-linear constraint $X \leq YZ^{-1}Y^*$ is not convex

Quadratic separation for design, features:

- ❑ Design of sets of controllers: handles fragility
- ❑ LMI formulations for state-feedback and full-order output-feedback
- ❑ All robust multi-objective problems can be recast as LMIs + $X \leq YZ^{-1}Y^*$



Challenge:

- ❑ Algorithms to handle the non-linear matrix inequality.
- ❑ Successful results of a gradient-type algorithm - to be improved.

- ❑ Towards lossless robust analysis - PDLF and beyond
- ❑ Topological separation for control design
- ❑ Algorithms for NLMIs
- ❑ Free academic software to test and compare results.
<http://www.laas.fr/OLOCEP>
- ❑ Control theory and optimization - Duality, separation...