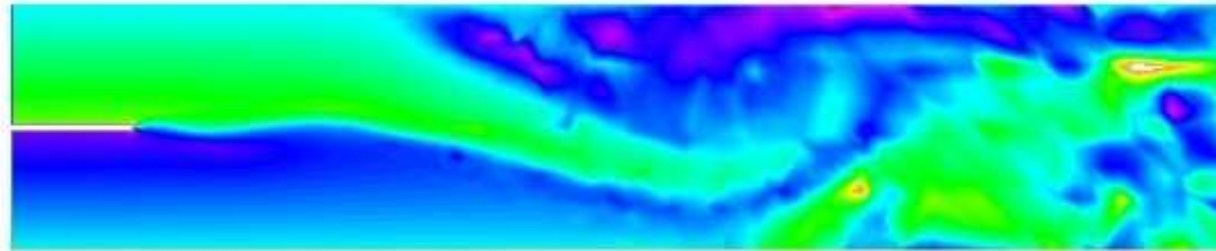


# Robust control theory for flow control via reduced order models

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## 👉 Procedure 1 [Kasnakoglu 2007]

- NSNL  $\xrightarrow{Lin.}$  NSL  $\xrightarrow{Red.}$  Linear synthesis on LROM
- NSNL  $\xrightarrow{Red.}$  NLROM  $\xrightarrow{Lin.}$  Linear synthesis on LROM
- ✓ Proper Orthogonal Decomposition (POD modes + actuation mode)
- ✓ Snapshots
- ✓ Galerkin Projection (GP)

The nonlinear ROM:

$$\begin{aligned}\dot{x}(t) &= C + Lx(t) + x^T Qx + \hat{C}u(t) + \hat{L}x(t)u(t) + \hat{Q}u^2(t) \\ y(t) &= Cx(t) + Du(t)\end{aligned}$$

where  $x(t) \in \mathbb{R}^6$ ,  $y(t) \in \mathbb{R}^6$  and  $u(t) \in \mathbb{R}$

Linearization around an equilibrium state  $x_{eq}$  (or not) to get the **LTI ROM**:

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) + Du(t)\end{aligned} \quad \text{where } x \in \mathbb{R}^6, y \in \mathbb{R}^6 \text{ and } u = f(x, t) \in \mathbb{R}?$$

➔ Closed-form solutions  $\neq$  Numerical solutions (LAAS-CNRS + S. Boyd 1990-1995)

Control pb. = Mathematical programming pb. + efficient numerical method

➔ Weak results  $\neq$  Strong results

Complexity :  $P = NP$  ? and global proof : convexity

① LMI formalism emerging (Willems 1971)

② Developments of theoretical and numerical SDP tools:

Interior points [Nesterov et Nemirovskii 1994] and cutting planes [Kelley 1960]

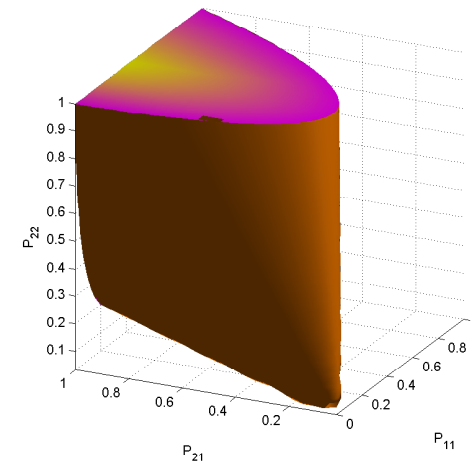
- "Easy" problems: numerical alternative to usual closed-form solutions
- "Tough" problems: convex relaxations

$$\min_{x \in \mathbb{R}^m} \sum_{i=1}^m c_i x_i = c'x$$

$$\text{sous } F_0 + \sum_{i=1}^m F_i x_i \succ 0$$

$$A'P + PA \prec 0$$

$$P \succ 0$$



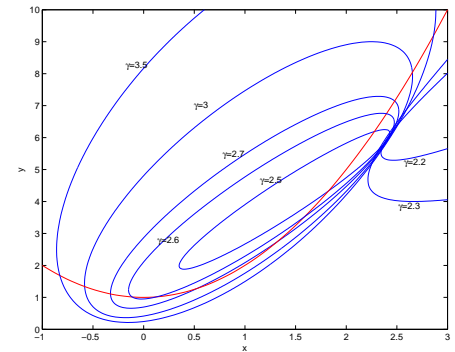
Nonconvexity and/or nonpolynomial complexity obstruction

$$\begin{aligned} \min_{x \in \mathbb{R}^m} \quad & \sum_{i=1}^m c_i x_i = c'x \\ \text{s.t.} \quad & F_0 + \sum_{i=1}^m F_i x_i + \sum_{i=1}^m \sum_{j=i}^m G_{ij} x_i x_j \succ \mathbf{0} \end{aligned}$$

$$M'PM < \gamma^2 P$$

$$\text{rank}(UPV) = 1$$

$$P = \begin{bmatrix} 1 & x & 0 \\ x & y & 0 \\ 0 & 0 & 1 \end{bmatrix}$$



- Convex relaxations (Lagrangian duality)
- Algorithmic Approach (nonsmooth optimization)
- ✓ Take advantage of the theoretical and practical structure of the system in the algorithm
- ✓ Large scale and reliability (guaranteed computations) are still open problems for SDP (sensitivity theory)

Given the model:

$$\begin{aligned} \dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) \end{aligned} \quad \text{where } x \in \mathbb{R}^n, y \in \mathbb{R}^r \text{ and } u \in \mathbb{R}^m$$

➔ **Problem 1 : SOF**

Find  $K \in \mathbb{R}^{r \times m}$  s.t.  $A + BKC$  asymptotically stable (closed-loop spectrum in  $\mathbb{C}^-$ )

Given the uncertain model :

$$\begin{aligned} \dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) \end{aligned} \quad M = \begin{bmatrix} A & B & C \end{bmatrix} \in \Omega \\ &= \text{co} \left\{ \begin{bmatrix} A^1 & B^1 & C^1 \end{bmatrix}, \dots, \begin{bmatrix} A^N & B^N & C^N \end{bmatrix} \right\}$$

$$M(\xi) = \begin{bmatrix} A(\xi) & B(\xi) & C(\xi) \end{bmatrix} = \sum_{i=1}^N \xi_i \begin{bmatrix} A^i & B^i & C^i \end{bmatrix}, \quad \xi \in \Xi = \left\{ \xi = \sum_{i=1}^N \xi_i = 1 \quad \xi_i \geq 0 \right\}$$

➔ **Problem 2 : Robust SOF**

Find a robustly stabilizing **SOF**  $u(t) = Ky(t)$  i.e. find a single matrix  $K \in \mathbb{R}^{r \times m}$  s.t.

$$\Omega_{bf} = \text{co} \left\{ A^1 + B^1 KC, \dots, A^N + B^N KC \right\}$$

is asymptotically stable  $\forall \xi \in \Xi$

- Existence problem is a **decidable** (algorithmic) problem [Anderson 1975]
- **Different approaches based on Lyapunov theory**
  - ✓ Quantifier elimination procedure of Tarski-Seidenberg [Anderson 1975]
  - ✓ Nonlinear Programming methods [Levine-Athans 1970], [Overton 2000]
  - ✓ *BMI* approaches

$$P > \mathbf{0} \quad (A + BKC)'P + P(A + BKC) < \mathbf{0}$$

- ✓ *LMI* optimization with rank constraint:

$$B^\perp (AX + XA') B^{\perp'} < \mathbf{0} \quad X = Y^{-1} \text{ or rank } \begin{bmatrix} X & \mathbf{1} \\ \mathbf{1} & Y \end{bmatrix} = n$$

$$C'^\perp (A'Y + YA) C'^{\perp'} < \mathbf{0}$$

- ✓ Lagrangian relaxations + coordinate descent-type algorithm

The triplet  $(A, B, C)$  is stabilizable via SOF iff

$$\exists X \in \mathbb{S}_n^+, K_{sf} \in \mathbb{R}^{m \times n}, Z \in \mathbb{R}^{m \times r}, F \in \mathbb{R}^{m \times m} \text{ s.t.}:$$

$$\mathcal{L}(A, B, C) = \begin{bmatrix} A & B \\ \mathbf{1} & \mathbf{0} \end{bmatrix}' \begin{bmatrix} \mathbf{0} & X \\ X & \mathbf{0} \end{bmatrix} \begin{bmatrix} A & B \\ \mathbf{1} & \mathbf{0} \end{bmatrix} + \text{sym} \left\{ \begin{bmatrix} K_{sf}' \\ -\mathbf{1} \end{bmatrix} \begin{bmatrix} ZC & F \end{bmatrix} \right\} < \mathbf{0}$$

$$K = -F^{-1}Z$$

- First part of  $\mathcal{L}(A, B, C)$  reminds us of **KYP lemma**
- Additional variables  $F, Z$ , allow a decoupling between the computation of the Lyapunov certificate  $X$  and of the SOF  $K$
- The additional variable  $K_{sf}$  is necessarily a stabilizing SF for  $(A, B)$  (application of elimination lemma)

- Testing stability of a polytope of matrices is an  $\mathcal{NP}$ -hard problem [Coxson and DeMarco 1999])
- **Quadratic stabilizability** framework:  $\exists X \in \mathbb{S}_n^+$ , and  $\exists K \in \mathbb{R}^{r \times n}$

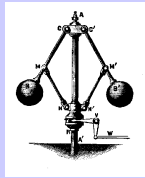
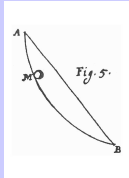
$$\begin{bmatrix} \mathbf{1} \\ KC \end{bmatrix}' \begin{bmatrix} A(\xi) & B(\xi) \\ \mathbf{1} & \mathbf{0} \end{bmatrix}' \begin{bmatrix} \mathbf{0} & X \\ X & \mathbf{0} \end{bmatrix} \begin{bmatrix} A(\xi) & B(\xi) \\ \mathbf{1} & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbf{1} \\ KC \end{bmatrix} < \mathbf{0} \quad \forall \xi \in \Xi$$

- **Robust stabilizability**: For each  $M = \begin{bmatrix} A & B \end{bmatrix} \in \Omega$ ,  $\exists X \in \mathbb{S}_n^+$  and  $\exists K \in \mathbb{R}^{r \times m}$  s.t.:

$$\begin{bmatrix} \mathbf{1} \\ KC \end{bmatrix}' \begin{bmatrix} A(\xi) & B(\xi) \\ \mathbf{1} & \mathbf{0} \end{bmatrix}' \begin{bmatrix} \mathbf{0} & X(\xi) \\ X(\xi) & \mathbf{0} \end{bmatrix} \begin{bmatrix} A(\xi) & B(\xi) \\ \mathbf{1} & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbf{1} \\ KC \end{bmatrix} < \mathbf{0}$$

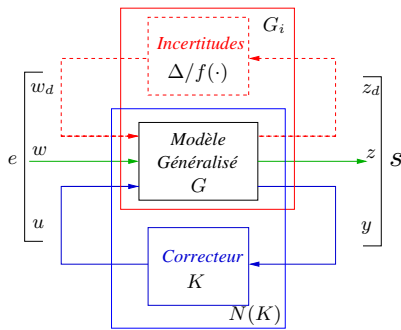
Quadratic stabilizability  $\Rightarrow$  robust stabilizability





- Internal stability (Lyapunov)
- Heterogeneous performances (time and frequency responses)

## ✓ Modern linear framework



- Standard model = central artefact [Doyle 1983]
- Robust stability and performance
- Optimality

## CONSTRUCTIVE approach for C/S



- Lyapunov theory
- Optimization (Duality)



