



MAX PLANCK INSTITUTE
FOR DYNAMICS OF COMPLEX
TECHNICAL SYSTEMS
MAGDEBURG



COMPUTATIONAL METHODS IN
SYSTEMS AND CONTROL THEORY

LQG and \mathcal{H}_∞ Balanced Truncation for Active Flow Control

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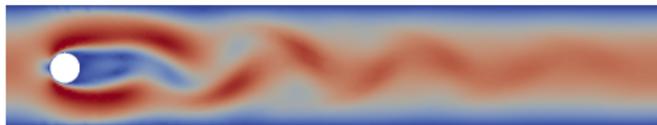
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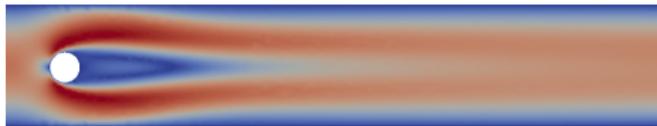
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1. Introduction
2. Robust Control
3. Reduced-Order Controllers
4. Numerical Realization
5. Conclusions



- Consider the cylinder wake at moderate *Reynolds* numbers.
- The steady state is a solution, but unstable \rightsquigarrow transition to turbulent flow due to unavoidable perturbations, if they are not attenuated.
- Goal: Stabilizing feedback controller that works in experiments.
- Thus, the simulation needs to cope with:
 - limited measurements,
 - short evaluation times,
 - system uncertainties.



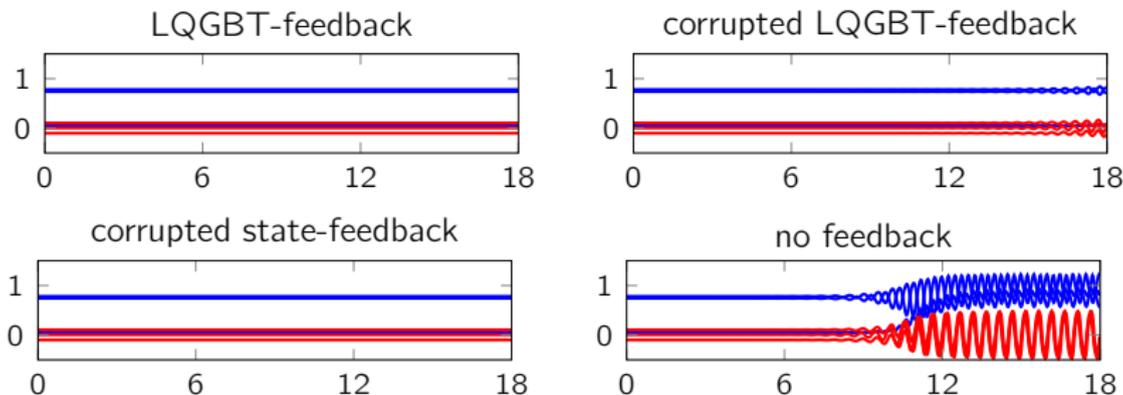
Idea: Linearization based feedback to stabilize a steady-state:

$$\begin{aligned} \dot{v} + (v \cdot \nabla)v - \frac{1}{\text{Re}} \Delta v + \nabla p &= Bu \\ \nabla \cdot v &= 0 \end{aligned} \quad \begin{array}{l} \text{linearization} \\ \text{and} \\ \text{(semi-)} \\ \text{discretization} \end{array} \quad \begin{aligned} \dot{v} + Av + J^T p &= Bu \\ Jv &= 0 \end{aligned}$$

- see [RAYMOND'05, '06] for theory concerning flows;
- see [PB&JH'15] for low-rank output-based feedback;
- see also [BREITEN&KUNISCH'14].

X Fragility of Observer-Based Controllers

LQG controllers have no guaranteed robustness margins and will likely fail in the presence of system uncertainties.



- corrupted linearization – about the not quite converged steady state
 - visually undistinguishable from the *exact* linearization point
 - relative difference in norm: 5%

Uncertainty A_Δ in the linearization $A...$

$$\dot{x} = Ax + Bu$$

$$y = Cx$$



$$\dot{x} = [A + A_\Delta]x + Bu$$

$$y = Cx$$

or

$$G_0(s) = C(sI - A)^{-1}B$$



$$\begin{aligned} G(s) &= C(sI - A - A_\Delta)^{-1}B \\ &= G_0(s) + G_\Delta(s) \end{aligned}$$

$$\text{with } G_\Delta(s) = CA_\Delta(sI - A)^{-1}(sI - A - A_\Delta)^{-1}B$$

... is an additive uncertainty in the transfer function.

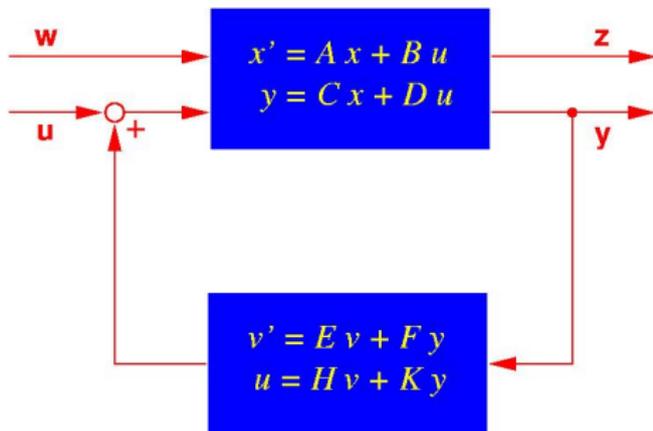
→ Robust \mathcal{H}_∞ controllers can compensate for that.

Linear Time-Invariant Systems (finite or infinite)

$$\Sigma : \begin{cases} \dot{x} = Ax + B_1 w + B_2 u, \\ z = C_1 x + D_{11} w + D_{12} u, \\ y = C_2 x + D_{21} w + D_{22} u, \end{cases}$$

where A , B_j , C_i , and D_{ij} are matrices of suitable sizes, $j, i \in \{1, 2\}$.

- x – states of the system,
- w – exogenous inputs
- u – control inputs,
- z – performance outputs
- y – measured outputs



Laplace transform \implies **transfer function** (in frequency domain)

$$G(s) = \begin{bmatrix} G_{11}(s) & G_{12}(s) \\ G_{21}(s) & G_{22}(s) \end{bmatrix} \equiv \left[\begin{array}{c|cc} A & B_1 & B_2 \\ \hline C_1 & D_{11} & D_{12} \\ C_2 & D_{21} & D_{22} \end{array} \right],$$

where for $x(0) = 0$, G_{ij} are the transfer functions

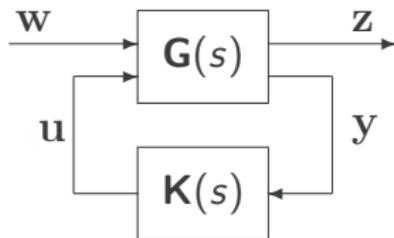
$$G_{ij}(s) = C_i(sI - A)^{-1}B_j + D_{ij}$$

with $i, j \in \{1, 2\}$, describing the transfer from inputs to outputs of Σ via

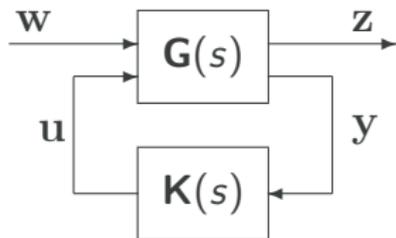
$$z(s) = G_{11}(s)w(s) + G_{12}(s)u(s),$$

$$y(s) = G_{21}(s)w(s) + G_{22}(s)u(s).$$

Consider **closed-loop** system, where $K(s)$ is an **internally stabilizing** controller, i.e., K stabilizes G for $w \equiv 0$.



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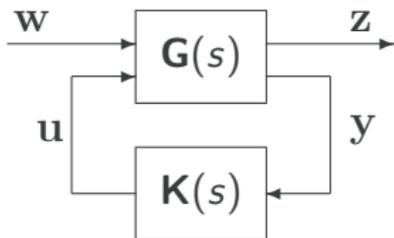
Goal:

Find **robust controller**, i.e., K that minimizes error outputs

$$z = \left(G_{11} + G_{12}K(I - G_{22}K)^{-1}G_{21} \right) w =: \mathcal{F}(G, K)w,$$

where $\mathcal{F}(G, K)$ is the **linear fractional transformation** of G, K .

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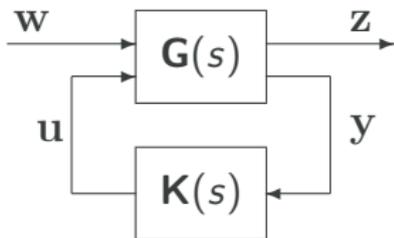
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\mathcal{H}_∞ -optimal Control Problem:

$$\min_{K \text{ stabilizing}} \|\mathcal{F}(G, K)\|_{\mathcal{H}_\infty}$$

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\mathcal{H}_∞ -Suboptimal Control Problem:

For given constant $\gamma > 0$, find all internally stabilizing controllers satisfying

$$\|\mathcal{F}(G, K)\|_{\mathcal{H}_\infty} < \gamma.$$

Simplifying Assumptions

1. $D_{11} = 0$
2. $D_{22} = 0$
3. (A, B_1) stabilizable, (C_1, A) detectable
4. (A, B_2) stabilizable, (C_2, A) detectable ($\implies \Sigma$ internally stabilizable)
5. $D_{12}^T [C_1 \ D_{12}] = [0 \ I]$
6. $\begin{bmatrix} B_1 \\ D_{21} \end{bmatrix} D_{21}^T = \begin{bmatrix} 0 \\ I \end{bmatrix}$

Remark. 1.,2.,5.,6. only for notational convenience, 3. can be relaxed, but derivations get even more complicated.

Theorem [DOYLE/GLOVER/KHARGONEKAR/FRANCIS '89, VAN KEULEN '93]

Given the Assumptions 1.–6., there exists an admissible controller $K(s)$ solving the \mathcal{H}_∞ -suboptimal control problem \iff

(i) There exists a stabilizing solution $X_\infty = X_\infty^T \geq 0$ to the Riccati equation

$$C_1^T C_1 + A^T X + XA + X(\gamma^{-2} B_1 B_1^T - B_2 B_2^T) X = 0.$$

(ii) There exists a stabilizing solution $Y_\infty = Y_\infty^* \geq 0$ to the Riccati equation

$$B_1 B_1^T + AY + YA^T + Y(\gamma^{-2} C_1^T C_1 - C_2^T C_2) Y = 0.$$

(iii) $\gamma^2 > \rho(X_\infty Y_\infty)$.

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\mathcal{H}_∞ -optimal Control

Find minimal γ for which (i)–(iii) are satisfied \rightsquigarrow γ -iteration based on solving the Riccati equations above repeatedly for different γ .

\mathcal{H}_∞ -(sub-)optimal Controller

If (i)–(iii) hold, a suboptimal controller is given by

$$\hat{K}(s) = \left[\begin{array}{c|c} \hat{A} & \hat{B} \\ \hline \hat{C} & 0 \end{array} \right] = \hat{C}(sI - \hat{A})^{-1}\hat{B},$$

where for

$$Z_\infty := (I - \gamma^{-2}Y_\infty X_\infty)^{-1},$$

$$\hat{A} := A + (\gamma^{-2}B_1B_1^T - B_2B_2^T)X_\infty - Z_\infty Y_\infty C_2^T C_2,$$

$$\hat{B} := Z_\infty Y_\infty C_2^T,$$

$$\hat{C} := -B_2^T X_\infty.$$

$\hat{K}(s)$ is the **central** or **minimum entropy** controller.

Balancing Related Methods

1. Solve the primal and dual matrix equations defining the characteristic matrices P and Q .
2. Balance the system with respect to P and Q .
3. Truncate states corresponding to small characteristic values of PQ .

■ LQG:

$$A^T P_{LQG} + P_{LQG} A - P_{LQG} B_2 B_2^T P_{LQG} + C_1^T C_1 = 0,$$

$$A Q_{LQG} + Q_{LQG} A^T - Q_{LQG} C_2^T C_2 Q_{LQG} + B_1 B_1^T = 0$$

■ \mathcal{H}_∞ :

$$A^T P_{\mathcal{H}_\infty} + P_{\mathcal{H}_\infty} A + P_{\mathcal{H}_\infty} (\gamma^{-2} B_1 B_1^T - B_2 B_2^T) P_{\mathcal{H}_\infty} + C_1^T C_1 = 0,$$

$$A Q_{\mathcal{H}_\infty} + Q_{\mathcal{H}_\infty} A^T + Q_{\mathcal{H}_\infty} (\gamma^{-2} C_1^T C_1 - C_2^T C_2) Q_{\mathcal{H}_\infty} + B_1 B_1^T = 0$$

The \mathcal{H}_∞ controller design for the stabilization of incompressible flows comes with two immediate numerical challenges:

1. High-dimensional model equations

- ← direct approach not feasible because of memory constraints
- low-rank Riccati iteration

2. Differential-algebraic structure

- ← due to the incompressibility constraint
- implicit realization of the discrete *Leray projector*

Low-Rank Riccati Iteration

[LANZON/FENG/ANDERSON '07, B. '08/'12]

1. Solve the ARE

$$C_1^T C_1 + A^T Z_0 + Z_0 A - Z_0 B_2 B_2^T Z_0 = 0$$

using Newton-ADI / RADI, yielding Y_0 with $Z_0 \approx Y_0 Y_0^T$.

2. Set $R_1 := Y_0$.

{% $R_1 R_1^T \approx X_1$ }

3. FOR $k = 1, 2, \dots$

(i) Set $A_k = A + U_k V_k^T := A + \gamma^{-2} B_1 (B_1^T R_k) R_k^T - B_2 (B_2^T R_k) R_k^T$.

- (ii) Solve the ARE

$$\gamma^{-2} Z_{k-1} B_1 B_1^T Z_{k-1} + A_k^T Z_k + Z_k A_k - Z_k B_2 B_2^T Z_k = 0$$

using Newton-ADI / RADI, yielding Y_k with $Z_k \approx Y_k Y_k^T$.

(iii) Set $R_{k+1} := \text{rrqr}([R_k, Y_k], \tau)$.

{% $R_{k+1} R_{k+1}^T \approx X_{k+1}$ }

(iv) IF $\|(B_1^T Y_k) Y_k^T\|_2 < \text{tol}$ THEN **Stop**.

Under standard assumptions, the semi-discrete incompressible (Navier-)Stokes equations can be realized as an ODE:

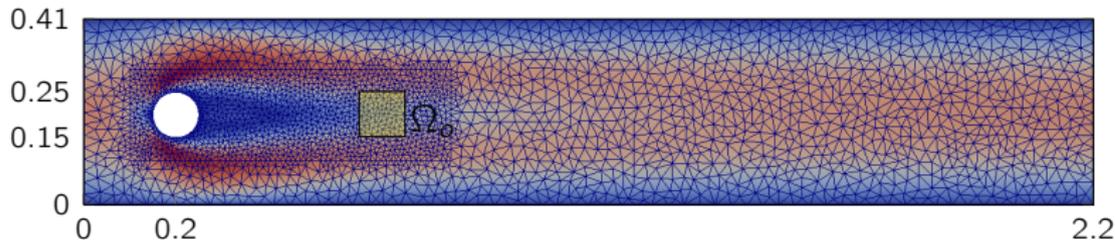
$$\begin{aligned} \dot{v} &= Av + J^T p + Bu \\ 0 &= Jv \end{aligned} \quad \rightarrow \quad \begin{aligned} \dot{v} &= \Pi Av + \Pi Bu \end{aligned}$$

$$\text{with } \Pi := I - J^T(JJ^T)^{-1}J.$$

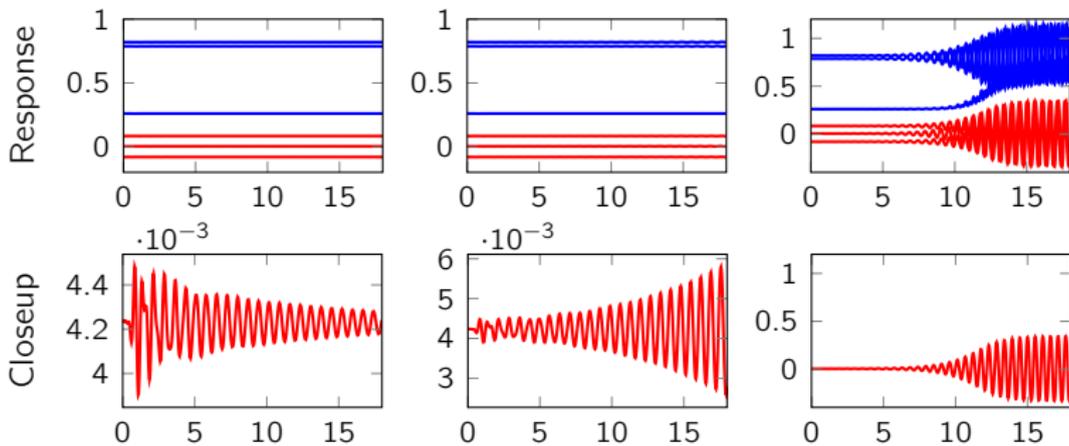
In turn, the feedback defined via the Π -based Riccati equations can be realized via constrained Riccati equations [PH&JH'17]:

$$\begin{aligned} A^T \Pi X + X \Pi A \\ -X \Pi B B^T \Pi X &= -C^T C \end{aligned} \quad \leftarrow \quad \begin{aligned} A^T X + X A - X B B^T X \\ + J^T Y + Y^T J &= -C^T C, \\ JX = 0, \quad XJ &= 0, \end{aligned}$$

avoiding Π altogether.



- 2D cylinder wake
- Navier-Stokes equations
- $Re = 90$
- *Taylor-Hood* finite elements
- 19500 velocity nodes
- Boundary control at 2 outlets at the cylinder periphery
- Distributed observation:
 - 3 *sensors* in the wake
 - measuring both v -components each
- \mathcal{H}_∞ -BT reduced controller
- Target: stabilization of the steady-state solution



- Controller of dimension 14 (left) and 8 (middle)
- Based on an inexact linearization
 - only 3 *Picard* iteration on the *Stokes* steady-state
 - relative difference to the *exact* linearization point: 8%
- Random perturbation of the initial value to trigger instabilities

Summary

- \mathcal{H}_∞ -BT reduced controllers are
 - output based and of low dimensions,
 - robust against system uncertainties – as opposed to LQG.
- The application to flow stabilization becomes feasible with
 - low-rank Riccati iterations,
 - implicit realization of the incompressibility constraint.

Code Availability:

- The Riccati iteration will be available in the M-M.E.S.S. library version 2.0.
- \mathcal{H}_∞ -BT and LQGBT implementations can be found in the MORLAB toolbox.

$$\begin{bmatrix} M & E \\ S & S \end{bmatrix}$$

M	R
O	LAB



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