

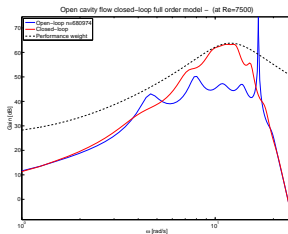
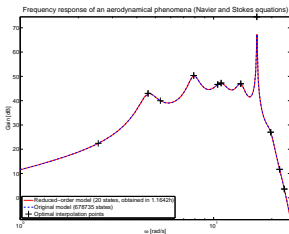
Fluid flow dynamical model approximation and control

... a case-study on an open cavity flow

C. Poussot-Vassal & D. Sipp



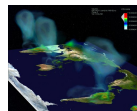
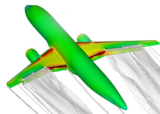
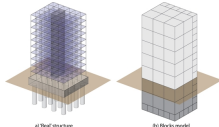
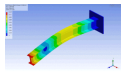
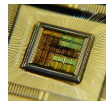
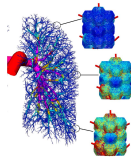
Journée conjointe GT Contrôle de Décollement & GT MOSAR



LARGE-SCALE DYNAMICAL MODELS

... some motivating examples in the simulation & control domains

Large-scale systems are present in many engineering fields: aerospace, computational biology, building structure, VLI circuits, automotive, weather forecasting, fluid flow...



- ▶ difficulties with simulation & memory management (e.g. ODE solvers)
- ▶ difficulties with analysis (e.g. frequency response, μ_{SSV} and \mathcal{H}_∞ computation ...)
- ▶ difficulties with controller design (e.g. robust, optimal, predictive, ...)

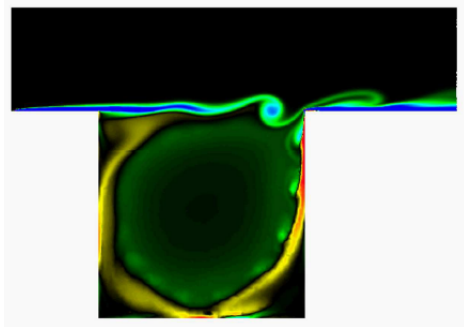
... in fluid flow dynamical problems

Fluid flow dynamical models

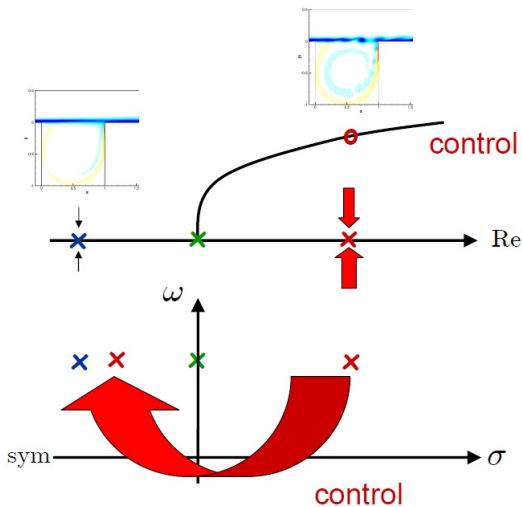
- ▶ Complex phenomena describing the motion of fluid flows,
- ▶ described by [Navier and Stokes equations](#),
- ▶ arising when modeling the weather, ocean currents, water flow in a pipe and [air flow around a wing](#)...

Some challenges arising

- ▶ Modeling and simulation
- ▶ Control turbulences



OPEN-CAVITY FLOW AND HOPF BIFURCATION



Physical model and dynamical modeling

- Navier and Stokes equations and assumptions

- Linearisation and simplifications

- Reduce and control approach

Large-scale dynamical model approximation

Active closed-loop control design

Conclusions

Navier and Stokes equations and assumptions

Navier and Stokes equations

$$\partial_t u + u \cdot \nabla u = -\nabla p + \frac{1}{Re} \Delta u \quad (1)$$

$$\nabla \cdot u = 0 \quad (2)$$

or in a condensed way

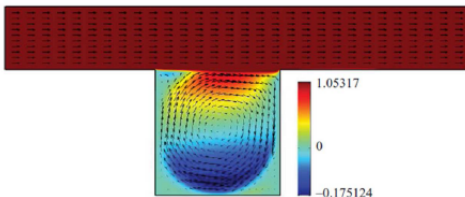
$$\dot{x}(t) = f(x(t), Re) \quad (3)$$

Existence of equilibrium points for a range of Reynolds numbers

Family of base-flows parametrized by the **Reynolds** number: $f(x_0, Re) = 0$

Navier and Stokes equations and assumptions

Navier and Stokes equations



Existence of equilibrium points for a range of Reynolds numbers

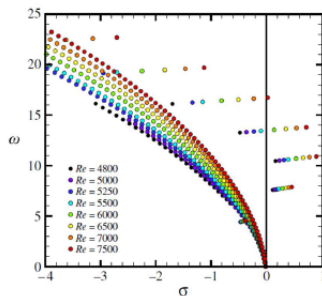
Family of base-flows parametrized by the **Reynolds** number: $f(x_0, Re) = 0$

Linearisation for different Reynolds Numbers

$$x(t) = x_0^{(Re)}(t) + \epsilon x_1^{(Re)}(t) \quad (1)$$

where ϵ is small

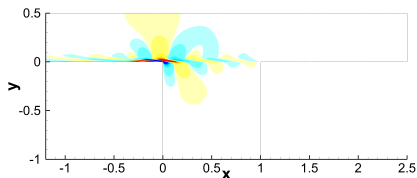
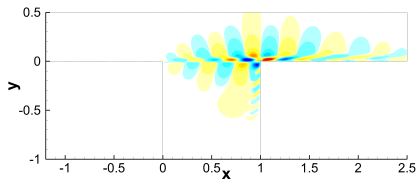
$$\dot{x}_1^{(Re)}(t) = \frac{\partial f}{\partial x} \Big|_{x_0^{(Re)}} x_1(t) = A(Re)x_1(t) \quad (2)$$



PHYSICAL MODEL AND DYNAMICAL MODELING

Linearisation and simplifications - Eigenvectors

Right and left eigenvectors



Localization of sensor and actuator

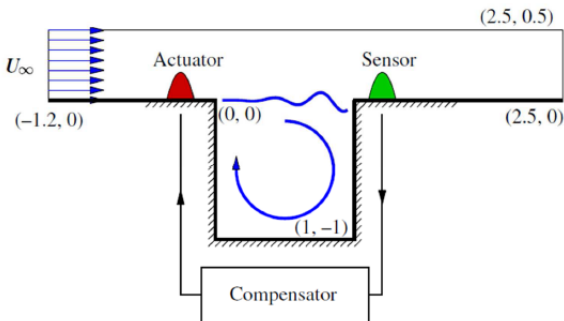
PHYSICAL MODEL AND DYNAMICAL MODELING

Linearisation and simplifications - Dynamical model and control setting

Actuator/sensor

$$\begin{aligned} E\dot{x}(t) &= A(Re)x(t) + Bu(t) \\ y(t) &= Cx(t) \end{aligned} \quad (3)$$

Actuator (volumic forcing in momentum equations) Sensor (shear stress)

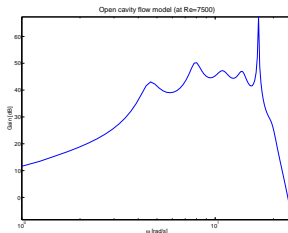
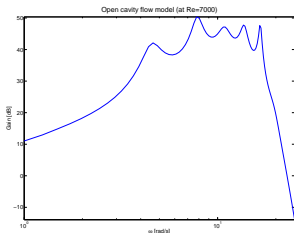


Linearisation and simplifications - Dynamical model and control setting

Actuator/sensor

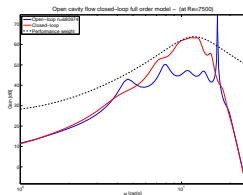
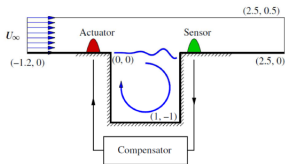
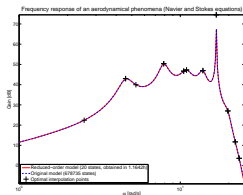
$$\begin{aligned} E\dot{x}(t) &= A(Re)x(t) + Bu(t) \\ y(t) &= Cx(t) \end{aligned} \quad (4)$$

- ▶ Two Reynolds cases ($Re = 7000$ and $Re = 7500$)
- ▶ Single Input Single Output Differential Algebraic Equations (SISO DAE)
- ▶ 8 unstable modes, order $\approx 650,000$ states



Proposed procedure

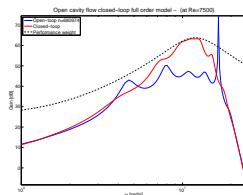
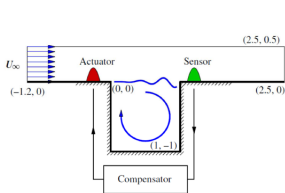
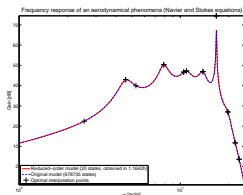
- ▶ Approximate the large-scale dynamical model
- ▶ Design a stabilizing active closed loop control strategy



Challenge of simulating and controlling such high complexity system...

Proposed procedure

- ▶ Approximate the large-scale dynamical model
- ▶ Design a stabilizing active closed loop control strategy



Challenge of simulating and controlling such high complexity system...

Physical model and dynamical modeling

Large-scale dynamical model approximation

Projection-based approximation framework

Approximation in the \mathcal{H}_2 , $\mathcal{H}_{2,\Omega}$ and \mathcal{L}_2 -norm

MIMO IRKA (or ITIA)

IETIA

Balanced Truncation POD

Fluid flow dynamical model approximation

Active closed-loop control design

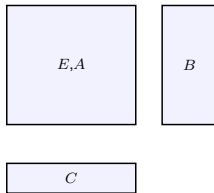
Conclusions

LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Projection-based approximation framework

Let $H : \mathbb{C} \rightarrow \mathbb{C}^{n_y \times n_u}$ be a n_u inputs n_y outputs, full order $\mathcal{H}_2^{n_y \times n_u}$ (or $\mathcal{L}_2^{n_y \times n_u}$) complex-valued function describing a **LTI** dynamical system as a DAE of order n , with realization \mathbf{H} :

$$\mathbf{H} : \begin{cases} E\dot{x}(t) & = Ax(t) + Bu(t) \\ y(t) & = Cx(t) \end{cases} \quad (5)$$



LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Projection-based approximation framework

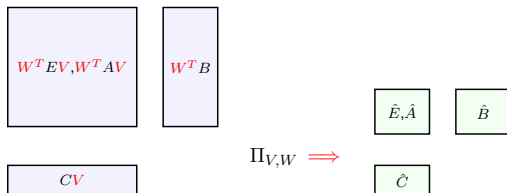
Let $H : \mathbb{C} \rightarrow \mathbb{C}^{n_y \times n_u}$ be a n_u inputs n_y outputs, full order $\mathcal{H}_2^{n_y \times n_u}$ (or $\mathcal{L}_2^{n_y \times n_u}$) complex-valued function describing a **LTI** dynamical system as a DAE of order n , with realization \mathbf{H} :

$$\mathbf{H} : \begin{cases} E\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) \end{cases} \quad (5)$$

the approximation problem consists in finding $V, W \in \mathbb{R}^{n \times r}$ (with $r \ll n$) **spanning** \mathcal{V} **and** \mathcal{W} **subspaces** and forming a **projector** $\Pi_{V,W} = VW^T$, such that

$$\hat{\mathbf{H}} : \begin{cases} W^T E V \hat{x}(t) &= W^T A V \hat{x}(t) + W^T B u(t) \\ \hat{y}(t) &= C V \hat{x}(t) \end{cases} \quad (6)$$

well approximates \mathbf{H} .



LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Projection-based approximation framework

Let $H : \mathbb{C} \rightarrow \mathbb{C}^{n_y \times n_u}$ be a n_u inputs n_y outputs, full order $\mathcal{H}_2^{n_y \times n_u}$ (or $\mathcal{L}_2^{n_y \times n_u}$) complex-valued function describing a **LTI** dynamical system as a DAE of order n , with realization \mathbf{H} :

$$\mathbf{H} : \begin{cases} E\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) \end{cases} \quad (5)$$

the approximation problem consists in finding $V, W \in \mathbb{R}^{n \times r}$ (with $r \ll n$) **spanning** \mathcal{V} **and** \mathcal{W} **subspaces** and forming a **projector** $\Pi_{V,W} = VW^T$, such that

$$\hat{\mathbf{H}} : \begin{cases} W^T E V \hat{x}(t) &= W^T A V \hat{x}(t) + W^T B u(t) \\ \hat{y}(t) &= C V \hat{x}(t) \end{cases} \quad (6)$$

well approximates \mathbf{H} .

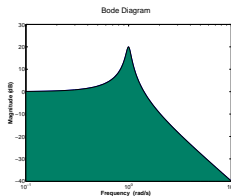
- ▶ **Small approximation error and/or global error bound**
- ▶ **Stability / passivity preservation**
- ▶ **Numerically stable & efficient procedure**

LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Approximation in the \mathcal{H}_2 , \mathcal{H}_2, Ω and \mathcal{L}_2 -norm^{1 2}

\mathcal{H}_2 model approximation

$$\hat{H} := \arg \min_{\substack{G \in \mathcal{H}_2^{n_y \times n_u} \\ \text{rank}(G) = r \ll n}} \|H - G\|_{\mathcal{H}_2} \quad (7)$$



¹  S. Gugercin and A. C. Antoulas and C. A. Beattie, " \mathcal{H}_2 Model Reduction for Large Scale Linear Dynamical Systems", SIAM Journal on Matrix Analysis and Applications, vol. 30(2), June 2008, pp. 609-638.

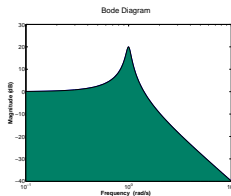
²  K. A. Gallivan, A. Vanderoppe, and P. Van-Dooren, "Model reduction of MIMO systems via tangential interpolation", SIAM Journal of Matrix Analysis and Application, vol. 26(2), February 2004, pp. 328-349.

LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Approximation in the \mathcal{H}_2 , $\mathcal{H}_{2,\Omega}$ and \mathcal{L}_2 -norm^{1 2}

\mathcal{H}_2 model approximation

$$\hat{H} := \arg \min_{\substack{G \in \mathcal{H}_2^{n_y \times n_u} \\ \text{rank}(G) = r \ll n}} \|H - G\|_{\mathcal{H}_2} \quad (7)$$



Energy to an impulse input

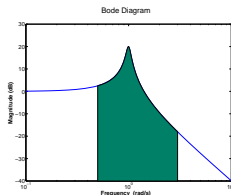
$$\begin{aligned} \|H\|_{\mathcal{H}_2}^2 &:= \text{trace} \left(\frac{1}{2\pi} \int_{-\infty}^{\infty} \overline{H(i\nu)} H(i\nu) d\nu \right) \\ &:= \text{trace} \left(C P C^T \right) = \text{trace} \left(B^T Q B \right) \\ &:= \sum_{i=1}^n \text{trace} \left(\phi_i H(-\lambda_i)^T \right) \end{aligned}$$


¹  S. Gugercin and A. C. Antoulas and C. A. Beattie, " \mathcal{H}_2 Model Reduction for Large Scale Linear Dynamical Systems", SIAM Journal on Matrix Analysis and Applications, vol. 30(2), June 2008, pp. 609-638.


²  K. A. Gallivan, A. Vanderoppe, and P. Van-Dooren, "Model reduction of MIMO systems via tangential interpolation", SIAM Journal of Matrix Analysis and Application, vol. 26(2), February 2004, pp. 328-349.

$\mathcal{H}_{2,\Omega}$ model approximation

$$\hat{H} := \arg \min_{\substack{G \in \mathcal{H}_\infty^{n_y \times n_u} \\ \text{rank}(G) = r \ll n}} \|H - G\|_{\mathcal{H}_{2,\Omega}} \quad (8)$$

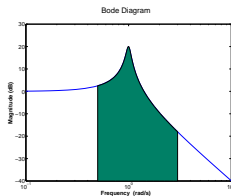


³  P. Vuillemin, C. Pousot-Vassal and D. Alazard, "A Spectral Expression for the Frequency-Limited \mathcal{H}_2 -norm", Available as <http://arxiv.org/abs/1211.1858>, 2012.

⁴  P. Vuillemin, C. Pousot-Vassal and D. Alazard, "Spectral expression for the Frequency-Limited \mathcal{H}_2 -norm of LTI Dynamical Systems with High Order Poles", European Control Conference, 2014, pp. 55-60.


\mathcal{H}_2, Ω model approximation


$$\hat{H} := \arg \min_{\substack{G \in \mathcal{H}_\infty^{n_y \times n_u} \\ \text{rank}(G) = r \ll n}} \|H - G\|_{\mathcal{H}_2, \Omega} \quad (8)$$



Energy (in a finite frequency) to an impulse input

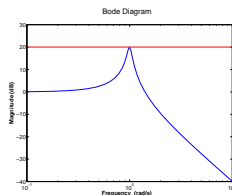
$$\begin{aligned} \|H\|_{\mathcal{H}_2, \Omega}^2 &:= \text{trace} \left(\frac{1}{\pi} \int_{\Omega} (\overline{H(i\nu)} H(i\nu)) d\nu \right) \\ &:= \text{trace} \left(C P_{\Omega} C^T \right) = \text{trace} \left(B^T Q_{\Omega} B \right) \\ &:= \sum_{i=1}^n \text{trace} \left(\phi_i H(-\lambda_i)^T \right) \left[-\frac{2}{\pi} \text{atan} \left(\frac{\omega}{\lambda_i} \right) \right] \end{aligned}$$


³  P. Vuillemin, C. Pousot-Vassal and D. Alazard, "A Spectral Expression for the Frequency-Limited \mathcal{H}_2 -norm", Available as <http://arxiv.org/abs/1211.1858>, 2012.

⁴  P. Vuillemin, C. Pousot-Vassal and D. Alazard, "Spectral expression for the Frequency-Limited \mathcal{H}_2 -norm of LTI Dynamical Systems with High Order Poles", European Control Conference, 2014, pp. 55-60.

\mathcal{H}_∞ model approximation

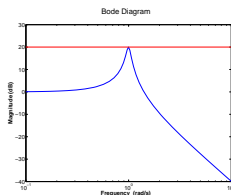
$$\hat{H} := \arg \min_{\substack{G \in \mathcal{H}_\infty^{n_y \times n_u} \\ \text{rank}(G) = r \ll n}} \|H - G\|_{\mathcal{H}_\infty} \quad (9)$$



⁵  P. Vuillemin, C. Pousot-Vassal, D. Alazard, "Two upper bounds on the \mathcal{H}_∞ -norm of LTI dynamical systems", 19th IFAC World Congress, pp. 5562-5567, 2014.


\mathcal{H}_∞ model approximation

$$\hat{H} := \arg \min_{\substack{G \in \mathcal{H}_\infty^{n_y \times n_u} \\ \text{rank}(G) = r \ll n}} \|H - G\|_{\mathcal{H}_\infty} \quad (9)$$



Worst case to an impulse input
(numerically complex to compute)

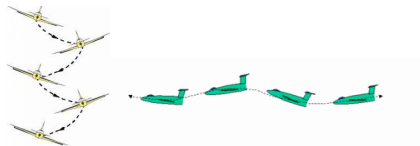
$$\begin{aligned} \|H\|_{\mathcal{H}_\infty} &:= \sup_{\omega \in \mathbb{R}} \bar{\sigma}(H(j\omega)) \\ &:= \max_{w \in L_2} \frac{\|y\|_2}{\|u\|_2} \end{aligned}$$

⁵  P. Vuillemin, C. Pousset-Vassal, D. Alazard, "Two upper bounds on the \mathcal{H}_∞ -norm of LTI dynamical systems", 19th IFAC World Congress, pp. 5562-5567, 2014.

Mismatch objective and eigenvector preservation

$$\hat{H} := \arg \min_{\substack{G \in \mathcal{L}_2^{n_y \times n_u} \\ \text{rank}(G) = r \ll n \\ \lambda_k(G) \subseteq \lambda(H) \quad k = 1, \dots, q_1 < r}} \|H - G\|_{\mathcal{H}_2} \quad (10)$$

- ▶ More than a \mathcal{H}_2 (sub-optimal) criteria
- ▶ **Keep some user defined eigenvalues... e.g. the unstable/well known ones**



⁶  C. Pousot-Vassal and P. Vuillemin, "An Iterative Eigenvector Tangential Interpolation Algorithm for Large-Scale LTI and a Class of LPV Model Approximation", European Control Conference, 2013, pp. 4490-4495.

MIMO Iterative Krylov Interpolation Algorithm (or ITIA)

- ▶ \mathcal{H}_2 -optimal, but still do not theoretically preserves stability
- ▶ Numerically very efficient (e.g. with sparse methods, $Ax = b$)

Iterative Eigenvector Tangential Interpolation Algorithm (IETIA)

- ▶ \mathcal{H}_2 sub-optimal, but still do not theoretically preserves stability
- ▶ Numerically very efficient (e.g. with sparse methods, $Ax = b$ and $AV = EV\lambda$)
- ▶ Applicable to \mathcal{L}_2 dynamical systems

Balanced Truncation Proper Orthogonal Decomposition (BT POD)

- ▶ Provides a \mathcal{H}_∞ -norm mismatch error (not tight), preserves stability
- ▶ Costly to compute, but a Matrix free version alleviate this problem by replacing by simulation (direct and adjoint)


LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

MIMO IRKA (or ITIA) - \mathcal{H}_2 optimality conditions (Tangential subspace approach)^{7 8}

Given $H(s)$, let $V \in \mathbb{C}^{n \times r}$ and $W \in \mathbb{C}^{n \times r}$ be matrices of full column rank r such that $W^*V = I_r$. If, for $j = 1, \dots, r$,

$$\left[(\sigma_j E - A)^{-1} B \hat{b}_j \right] \in \mathbf{span}(V) \text{ and } \left[(\sigma_j E - A^T)^{-1} C^T \hat{c}_j^* \right] \in \mathbf{span}(W) \quad (11)$$

where $\sigma_j \in \mathbb{C}$, $\hat{b}_j \in \mathbb{C}^{n_u}$ and $\hat{c}_j \in \mathbb{C}^{n_y}$, be given sets of interpolation points and left and right tangential directions, respectively.

⁷  P. Van-Dooren, K. A. Gallivan, and P. A. Absil, " \mathcal{H}_2 -optimal model reduction of MIMO systems", Applied Mathematics Letters, vol. 21(12), December 2008, pp. 53-62.

⁸  S. Gugercin and A. C. Antoulas and C. A. Beattie, " \mathcal{H}_2 Model Reduction for Large Scale Linear Dynamical Systems", SIAM Journal on Matrix Analysis and Applications, vol. 30(2), June 2008, pp. 609-638.


MIMO IRKA (or ITIA) - \mathcal{H}_2 optimality conditions (Tangential subspace approach)^{7 8}


Given $H(s)$, let $V \in \mathbb{C}^{n \times r}$ and $W \in \mathbb{C}^{n \times r}$ be matrices of full column rank r such that $W^*V = I_r$. If, for $j = 1, \dots, r$,

$$\left[(\sigma_j E - A)^{-1} B \hat{b}_j \right] \in \mathbf{span}(V) \text{ and } \left[(\sigma_j E - A^T)^{-1} C^T \hat{c}_j^* \right] \in \mathbf{span}(W) \quad (11)$$

where $\sigma_j \in \mathbb{C}$, $\hat{b}_j \in \mathbb{C}^{n_u}$ and $\hat{c}_j \in \mathbb{C}^{n_y}$, be given sets of interpolation points and left and right tangential directions, respectively. Then, the reduced order system $\hat{H}(s)$ satisfies the tangential interpolation conditions

$$\begin{aligned} H(-\hat{\sigma}_j) \hat{b}_j &= \hat{H}(-\hat{\sigma}_j) \hat{b}_j \\ \hat{c}_j^* H(-\hat{\sigma}_j) &= \hat{c}_j^* \hat{H}(-\hat{\sigma}_j) \\ \hat{c}_j^* \frac{d}{ds} H(s) \Big|_{s=-\hat{\sigma}_j} \hat{b}_j &= \hat{c}_j^* \frac{d}{ds} \hat{H}(s) \Big|_{s=-\hat{\sigma}_j} \hat{b}_j \end{aligned} \quad (12)$$

⁷  P. Van-Dooren, K. A. Gallivan, and P. A. Absil, " \mathcal{H}_2 -optimal model reduction of MIMO systems", Applied Mathematics Letters, vol. 21(12), December 2008, pp. 53-62.

⁸  S. Gugercin and A. C. Antoulas and C. A. Beattie, " \mathcal{H}_2 Model Reduction for Large Scale Linear Dynamical Systems", SIAM Journal on Matrix Analysis and Applications, vol. 30(2), June 2008, pp. 609-638.

LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Require: $\mathbf{H} = (E, A, B, C), \{\sigma_1^{(0)}, \dots, \sigma_{q_2}^{(0)}\} \in \mathbb{C}^{q_2}, \{\hat{b}_1, \dots, \hat{b}_{q_2}\} \in \mathbb{C}^{n_u \times q_2},$
 $\{\hat{c}_1, \dots, \hat{c}_{q_2}\} \in \mathbb{C}^{n_y \times q_2}$ and $r \in \mathbb{N}$

1: Construct,

$$\text{span}(V(\sigma_j^{(0)}, \hat{b}_j)) \text{ and } \text{span}(W(\sigma_j^{(0)}, \hat{c}_j^*)) \quad (13)$$

2: Compute $W \leftarrow W(V^T W)^{-1}$

3: **while** Stopping criteria **do**

4: $k \leftarrow k + 1$

5: $\hat{E} = W^T E V, \hat{A} = W^T A V, \hat{B} = W^T B, \hat{C} = C V$

6: Compute $\hat{A}R = \Lambda(\hat{A}, \hat{E})R$ and $L\hat{A} = \Lambda(\hat{A}, \hat{E})L$

7: Compute $\{\hat{b}_1, \dots, \hat{b}_r\} = \hat{B}^T L$ and $\{\hat{c}_1^*, \dots, \hat{c}_r^*\} = \hat{C}R$

8: Set $\sigma^{(i)} = -\Lambda(\hat{A}, \hat{E})$

9: Construct,

$$\text{span}(V(\sigma_j^{(k)}, \hat{b}_j)) \text{ and } \text{span}(W(\sigma_j^{(k)}, \hat{c}_j^*)) \quad (14)$$

10: Compute $W \leftarrow W(V^T W)^{-1}$

11: **end while**

12: Construct $\hat{\mathbf{H}} := (W^T E V, W^T A V, W^T B, C V)$

Ensure: $V, W \in \mathbb{R}^{n \times r}, W^T V = I_r$

LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Require: $\mathbf{H} = (E, A, B, C), \{\sigma_1^{(0)}, \dots, \sigma_{q_2}^{(0)}\} \in \mathbb{C}^{q_2}, \{\hat{b}_1, \dots, \hat{b}_{q_2}\} \in \mathbb{C}^{n_u \times q_2},$
 $\{\hat{c}_1, \dots, \hat{c}_{q_2}\} \in \mathbb{C}^{n_y \times q_2}$ and $r \in \mathbb{N}$

1: Construct,

$$\text{span}(V(\sigma_j^{(0)}, \hat{b}_j)) \text{ and } \text{span}(W(\sigma_j^{(0)}, \hat{c}_j^*)) \quad (13)$$

2: Compute $W \leftarrow W(V^T W)^{-1}$

3: **while** Stopping criteria **do**

4: $k \leftarrow k + 1$

5: $\hat{E} = W^T E V, \hat{A} = W^T A V, \hat{B} = W^T B, \hat{C} = C V$

6: Compute $\hat{A}R = \Lambda(\hat{A}, \hat{E})R$ and $L\hat{A} = \Lambda(\hat{A}, \hat{E})L$

7: Compute $\{\hat{b}_1, \dots, \hat{b}_r\} = \hat{B}^T L$ and $\{\hat{c}_1^*, \dots, \hat{c}_r^*\} = \hat{C}R$

8: Set $\sigma^{(i)} = -\Lambda(\hat{A}, \hat{E})$

9: Construct,

$$\text{span}(V(\sigma_j^{(k)}, \hat{b}_j)) \text{ and } \text{span}(W(\sigma_j^{(k)}, \hat{c}_j^*)) \quad (14)$$

10: Compute $W \leftarrow W(V^T W)^{-1}$

11: **end while**

12: Construct $\hat{\mathbf{H}} := (W^T E V, W^T A V, W^T B, C V)$

Ensure: $V, W \in \mathbb{R}^{n \times r}, W^T V = I_r$

LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

ITIA - \mathcal{H}_2 & spectral optimality conditions (Tangential subspace approach)⁹

Given $H(s)$, let $V \in \mathbb{C}^n \times r$ and $W \in \mathbb{C}^n \times r$ be matrices of full column rank $r = q_1 + q_2$ such that $W^*V = I_r$. If, for $i = 1, \dots, q_1$ and $j = 1, \dots, q_2$,

$$\left[r_i^* (\sigma_j E - A)^{-1} B \hat{b}_j \right] \in \mathbf{span}(V) \text{ and } \left[l_i^* (\sigma_j E - A^T)^{-1} C^T \hat{c}_j^* \right] \in \mathbf{span}(W) \quad (15)$$

$l_i^* \in \mathbb{C}^n$ and $r_i^* \in \mathbb{C}^n$ are left and right eigenvectors associated to $\lambda_i^* \in \mathbb{C}$ eigenvalues associated to A, E and $\sigma_j \in \mathbb{C}$, $\hat{b}_j \in \mathbb{C}^{n_u}$ and $\hat{c}_j \in \mathbb{C}^{n_y}$, be given sets of interpolation points and left and right tangential directions, respectively.

LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

ITIA - \mathcal{H}_2 & spectral optimality conditions (Tangential subspace approach)⁹

Given $H(s)$, let $V \in \mathbb{C}^{n \times r}$ and $W \in \mathbb{C}^{n \times r}$ be matrices of full column rank $r = q_1 + q_2$ such that $W^*V = I_r$. If, for $i = 1, \dots, q_1$ and $j = 1, \dots, q_2$,

$$\left[r_i^* (\sigma_j E - A)^{-1} B \hat{b}_j \right] \in \mathbf{span}(V) \text{ and } \left[l_i^* (\sigma_j E - A^T)^{-1} C^T \hat{c}_j^* \right] \in \mathbf{span}(W) \quad (15)$$

$l_i^* \in \mathbb{C}^n$ and $r_i^* \in \mathbb{C}^n$ are left and right eigenvectors associated to $\lambda_i^* \in \mathbb{C}$ eigenvalues associated to A, E and $\sigma_j \in \mathbb{C}$, $\hat{b}_j \in \mathbb{C}^{n_u}$ and $\hat{c}_j \in \mathbb{C}^{n_y}$, be given sets of interpolation points and left and right tangential directions, respectively. Then, the reduced order system $\hat{H}(s)$ satisfies the eigenvalue conditions,

$$\{\lambda_1^*, \dots, \lambda_{q_1}^*\} \subset \Lambda(\hat{A}, \hat{E}) \quad (16)$$

and the tangential interpolation conditions

$$\begin{aligned} H(-\hat{\sigma}_j) \hat{b}_j &= \hat{H}(-\hat{\sigma}_j) \hat{b}_j \\ \hat{c}_j^* H(-\hat{\sigma}_j) &= \hat{c}_j^* \hat{H}(-\hat{\sigma}_j) \\ \hat{c}_j^* \frac{d}{ds} H(s) \Big|_{s=-\hat{\sigma}_j} \hat{b}_j &= \hat{c}_j^* \frac{d}{ds} \hat{H}(s) \Big|_{s=-\hat{\sigma}_j} \hat{b}_j \end{aligned} \quad (17)$$

⁹  C. Poussoot-Vassal and P. Vuillemin, "An Iterative Eigenvector Tangential Interpolation Algorithm for Large-Scale LTI and a Class of LPV Model Approximation", European Control Conference, 2013, pp. 4490-4495.

LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Require: $\mathbf{H} = (E, A, B, C)$, $\{\lambda_1^*, \dots, \lambda_{q_1}^*\} \in \mathbb{C}^{q_1}$, $\{\sigma_1^{(0)}, \dots, \sigma_{q_2}^{(0)}\} \in \mathbb{C}^{q_2}$,

$\{\hat{b}_1, \dots, \hat{b}_{q_2}\} \in \mathbb{C}^{n_u \times q_2}$, $\{\hat{c}_1, \dots, \hat{c}_{q_2}\} \in \mathbb{C}^{n_y \times q_2}$ and $r = q_1 + q_2 \in \mathbb{N}$

1: Compute $\{l_1^*, \dots, l_{q_1}^*\}$ and $\{r_1^*, \dots, r_{q_1}^*\}$, eigenvectors of $\{\lambda_1^*, \dots, \lambda_{q_1}^*\}$

2: Construct,

$$\text{span}(V(l_i^*, \sigma_j^{(0)}, \hat{b}_j)) \text{ and } \text{span}(W(r_i^*, \sigma_j^{(0)}, \hat{c}_j^*)) \quad (18)$$

3: Compute $W \leftarrow W(V^T W)^{-1}$

4: **while** Stopping criteria **do**

5: $k \leftarrow k + 1$

6: $\hat{E} = W^T E V$, $\hat{A} = W^T A V$, $\hat{B} = W^T B$, $\hat{C} = C V$

7: Compute $\hat{A} R = \hat{E} \Lambda(\hat{A}, \hat{E}) R$ and $L \hat{A} = \Lambda(\hat{A}) L$

8: Compute $\{\hat{b}_1, \dots, \hat{b}_{q_2}\} = \hat{B}^T L$ and $\{\hat{c}_1^*, \dots, \hat{c}_{q_2}^*\} = \hat{C} R$

9: Set $\sigma^{(i)} = -\Lambda(\hat{A}, \hat{E})$

10: Construct,

$$\text{span}(V(l_i^*, \sigma_j^{(k)}, \hat{b}_j)) \text{ and } \text{span}(W(r_i^*, \sigma_j^{(k)}, \hat{c}_j^*)) \quad (19)$$

11: Compute $W \leftarrow W(V^T W)^{-1}$

12: **end while**

13: Construct $\hat{\mathbf{H}} := (W^T E V, W^T A V, W^T B, C V)$

Ensure: $V, W \in \mathbb{R}^{n \times r}$, $W^T V = I_r$, $\{\lambda_1^*, \dots, \lambda_{q_1}^*\} \subset \Lambda(\hat{A}, \hat{E})$

LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Require: $\mathbf{H} = (E, A, B, C)$, $\{\lambda_1^*, \dots, \lambda_{q_1}^*\} \in \mathbb{C}^{q_1}$, $\{\sigma_1^{(0)}, \dots, \sigma_{q_2}^{(0)}\} \in \mathbb{C}^{q_2}$,
 $\{\hat{b}_1, \dots, \hat{b}_{q_2}\} \in \mathbb{C}^{n_u \times q_2}$, $\{\hat{c}_1, \dots, \hat{c}_{q_2}\} \in \mathbb{C}^{n_y \times q_2}$ and $r = q_1 + q_2 \in \mathbb{N}$
 1: Compute $\{l_1^*, \dots, l_{q_1}^*\}$ and $\{r_1^*, \dots, r_{q_1}^*\}$, eigenvectors of $\{\lambda_1^*, \dots, \lambda_{q_1}^*\}$
 2: Construct,

$$\text{span}(V(l_i^*, \sigma_j^{(0)}, \hat{b}_j)) \text{ and } \text{span}(W(r_i^*, \sigma_j^{(0)}, \hat{c}_j^*)) \quad (18)$$

3: Compute $W \leftarrow W(V^T W)^{-1}$
 4: **while** Stopping criteria **do**
 5: $k \leftarrow k + 1$
 6: $\hat{E} = W^T E V$, $\hat{A} = W^T A V$, $\hat{B} = W^T B$, $\hat{C} = C V$
 7: Compute $\hat{A} R = \hat{E} \Lambda(\hat{A}, \hat{E}) R$ and $L \hat{A} = \Lambda(\hat{A}) L$
 8: Compute $\{\hat{b}_1, \dots, \hat{b}_{q_2}\} = \hat{B}^T L$ and $\{\hat{c}_1^*, \dots, \hat{c}_{q_2}^*\} = \hat{C} R$
 9: Set $\sigma^{(i)} = -\Lambda(\hat{A}, \hat{E})$
 10: Construct,

$$\text{span}(V(l_i^*, \sigma_j^{(k)}, \hat{b}_j)) \text{ and } \text{span}(W(r_i^*, \sigma_j^{(k)}, \hat{c}_j^*)) \quad (19)$$

11: Compute $W \leftarrow W(V^T W)^{-1}$
 12: **end while**
 13: Construct $\hat{\mathbf{H}} := (W^T E V, W^T A V, W^T B, C V)$

Ensure: $V, W \in \mathbb{R}^{n \times r}$, $W^T V = I_r$, $\{\lambda_1^*, \dots, \lambda_{q_1}^*\} \subset \Lambda(\hat{A}, \hat{E})$

LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Balanced Truncation POD - Idea

Assume a **stable** system, the impulse response, $t \geq 0$ such that $h(t) = Ce^{At}B$

- ▶ Input-to-state map $x_c(t) = e^{At}B$
- ▶ State-to-output map $x_o(t) = Ce^{At} = (e^{A^*t}C^*)^*$

Corresponding to Gramian:

$$\begin{aligned}
 \mathcal{P} &= \sum_t x_c(t)x_c^*(t) = \int_0^\infty e^{At}BB^*e^{A^*t}dt \\
 \mathcal{Q} &= \sum_t x_o^*(t)x_o(t) = \int_0^\infty e^{A^*t}C^*Ce^{At}dt
 \end{aligned} \tag{20}$$

solution of the Lyapunov equations,

$$\begin{cases}
 A\mathcal{P} + \mathcal{P}A^* + BB^* = 0 \\
 A^*\mathcal{Q} + \mathcal{Q}A + C^*C = 0
 \end{cases} \tag{21}$$

LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Balanced Truncation POD - in its Balanced basis $T = [T_1, \dots, T_n]$

Meaning of Gramians:

- ▶ $x_f^* \mathcal{P}^{-1} x_f$, is the minimal energy required to steer the state from 0 to x_f as $t \rightarrow \infty$.
- ▶ $x_0^* \mathcal{Q} x_0$ is the maximal energy produced by observing the output of the system corresponding to an initial state x_0 when no input is applied.

Balanced basis $T = [T_1, \dots, T_n]$:

$$\mathcal{P} = \mathcal{Q} = \mathcal{S} = \text{diag}(\sigma_1, \dots, \sigma_n) \text{ with } \sigma_1 > \sigma_2 > \dots > \sigma_n \quad (22)$$

- ▶ $T_1^* \mathcal{P}^{-1} T_1 = \frac{1}{\sigma_1}$ (easily controllable) and $T_1^* \mathcal{Q} T_1 = \sigma_1$ (easily observable)
- ▶ $T_n^* \mathcal{P}^{-1} T_n = \frac{1}{\sigma_n}$ (weakly controllable) and $T_n^* \mathcal{Q} T_n = \sigma_n$ (weakly observable)

moreover,

- ▶ stability is preserved
- ▶ error between original and reduced systems is upper-bounded by,

$$\sigma_r \leq \|H - \hat{H}\|_{\mathcal{H}_\infty} \leq 2(\sigma_{r+1} + \dots + \sigma_n) \quad (23)$$

where $\sigma_i, i = 1, \dots, n$ are the Hankel singular values.

LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Balanced Truncation POD

Require: $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times n_u}$, $C \in \mathbb{R}^{n_y \times n}$, $r \in \mathbb{N}^*$

- 1: Solve $AP + PA^T + BB^T = 0$
- 2: Solve $A^T Q + QA + C^T C = 0$
- 3: $P = UU^T$ and $Q = LL^T$
- 4: SVD decomposition: $[Z, S, Y] = \mathbf{SVD}(U^T L)$
- 5: Set $V = UZS^{-1/2}$
- 6: Set $W = LYS^{-1/2}$
- 7: Apply projectors V and W and obtain

$$\mathbf{H}_\perp = \left[\begin{array}{cc|c} A_{11} & A_{12} & B_1 \\ A_{21} & A_{22} & B_2 \\ \hline C_1 & C_2 & D \end{array} \right] \quad (24)$$

- 8: Approximation is obtained by $\hat{\mathbf{H}} = \left[\begin{array}{c|c} A_{11} & B_1 \\ \hline C_1 & D \end{array} \right]$

Ensure: Small approximation error

LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Balanced Truncation POD - Factorization of Gramians with snapshot method

Numerical approximation

Solving Lyapunov equations is memory/time consuming.

$$\mathcal{P} = UU^* \quad (25)$$

usually done with Cholesky Factorization, but noticing that (i finite)

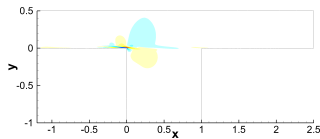
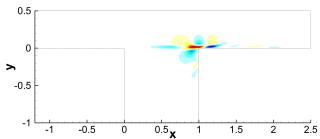
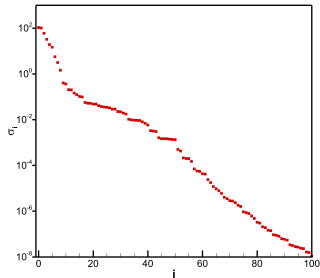
$$\begin{aligned}
 \mathcal{P} &= \int_0^{\infty} e^{At} B B^* e^{A^* t} \\
 &\approx \sum_i x_c(t_i) x_c^*(t_i) \Delta t \\
 &\approx \begin{bmatrix} x_c(t_1) \sqrt{\Delta t} & x_c(t_2) \sqrt{\Delta t} & \dots \end{bmatrix} \begin{bmatrix} x_c^*(t_1) \sqrt{\Delta t} \\ x_c^*(t_2) \sqrt{\Delta t} \\ \vdots \end{bmatrix}
 \end{aligned} \quad (26)$$

where $x_c(t) = e^{At} B$ (here use of the linearized model, $\dot{x}(t) = Ax(t)$, $x(0) = B$).

Same process for \mathcal{Q} with adjoint simulation $\dot{x}(t) = A^* x(t)$, $x(0) = C^*$.

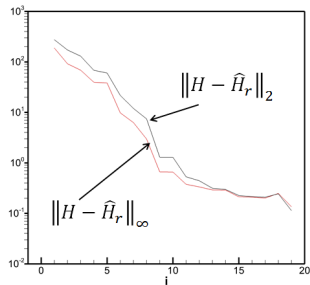
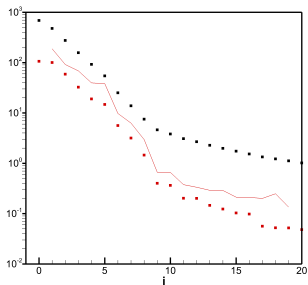
LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Balanced Truncation POD - Hankel singular values and bpod structures)



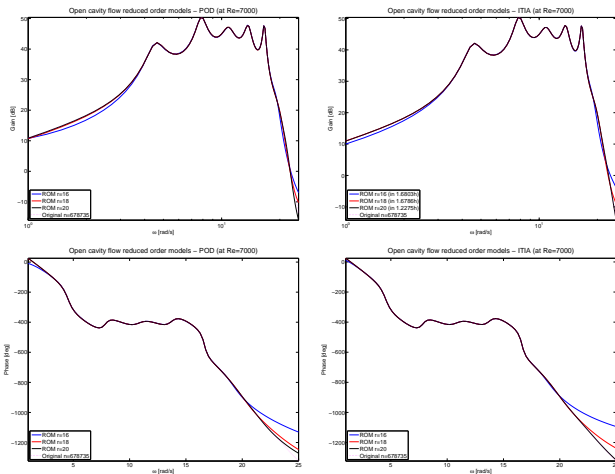
LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Balanced Truncation POD - Properties (lower/upper bounds & mismatch errors)



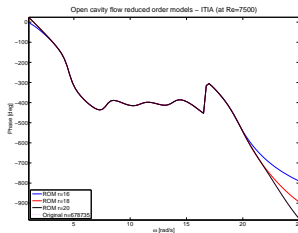
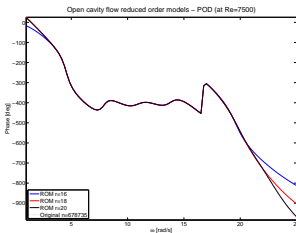
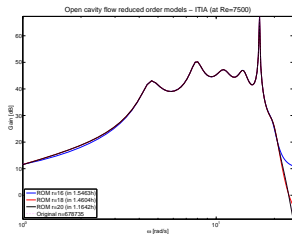
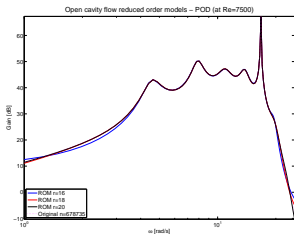
LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Fluid flow dynamical model approximation - Re=7000



LARGE-SCALE DYNAMICAL MODEL APPROXIMATION

Fluid flow dynamical model approximation - Re=7500



Physical model and dynamical modeling

Large-scale dynamical model approximation

Active closed-loop control design

Objectives

Results

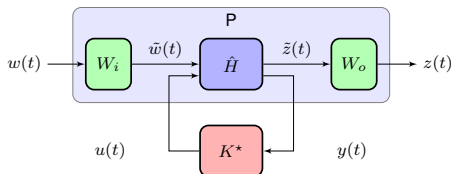
Conclusions

Objectives and \mathcal{H}_∞ control approach

- ▶ Stabilize the system
- ▶ Damp modes
- ▶ Potentially attenuate the \mathcal{H}_∞ -norm
- ▶ Engineering appealing / structured in view of RT implementation

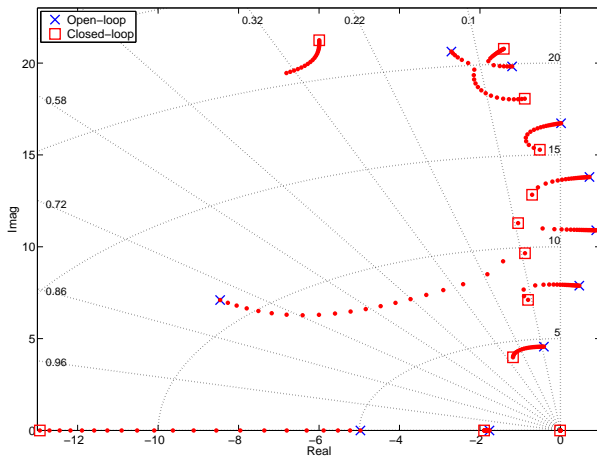
A standard control approach:

$$K^* = \arg \min_{K \subseteq \mathcal{K}} \|\mathcal{F}_l(\hat{H}, K)\|_{\mathcal{H}_\infty} \quad (27)$$



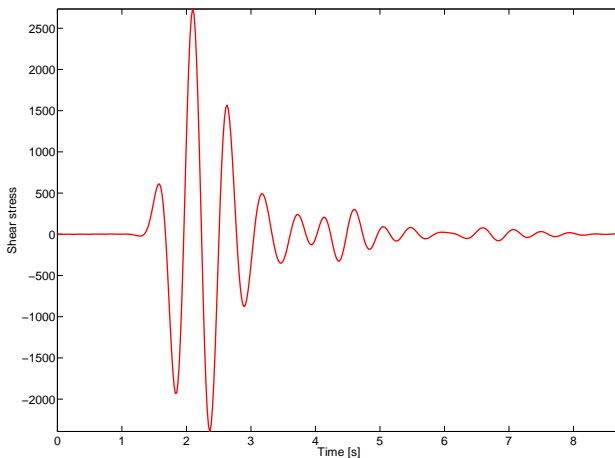
ACTIVE CLOSED-LOOP CONTROL DESIGN

Results - Spectral (ROM)



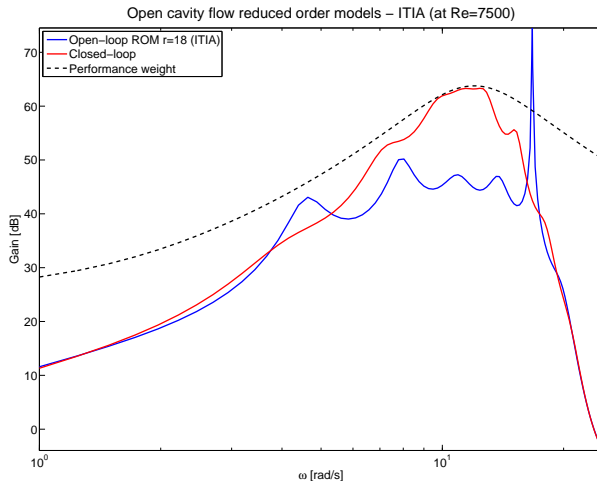
ACTIVE CLOSED-LOOP CONTROL DESIGN

Results - Impulse (un-normalized ROM)



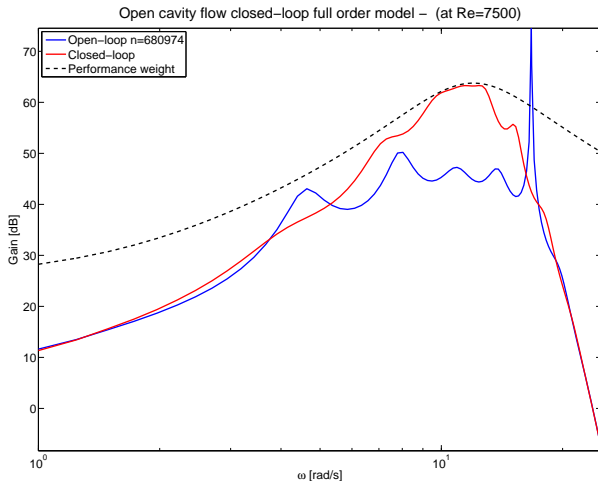
ACTIVE CLOSED-LOOP CONTROL DESIGN

Results - Frequency-domain (un-normalized ROM)



ACTIVE CLOSED-LOOP CONTROL DESIGN

Results - Frequency-domain (original LSS)



Physical model and dynamical modeling

Large-scale dynamical model approximation

Active closed-loop control design

Conclusions

About fluid flow control

About model approximation

About fluid flow control

About today's presentation

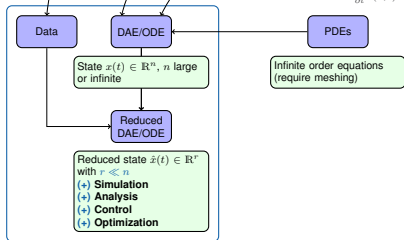
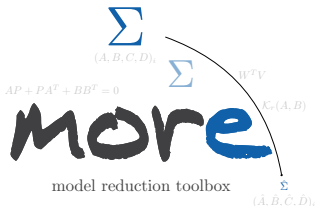
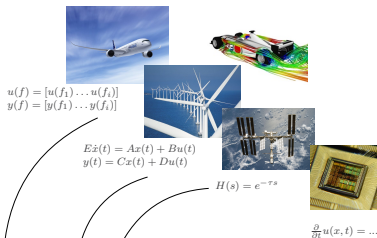
- ▶ Good performance of two model approximation techniques
- ▶ First attempt of \mathcal{H}_∞ control synthesis (controller of order 6)
- ▶ Application on Navier and Stokes equations for an open cavity flow

Some perspectives

- ▶ Extension to robust analysis / parameter dependent control
- ▶ Apply the realization-less approaches (e.g. handle delays)
- ▶ Include learning policy for (on-line) model accuracy enhancement?

About model approximation - MORE toolbox ¹⁰

- ▶ Successful application of advanced model approximation techniques
- ▶ both full and sparse
- ▶ on a complex unstable aerodynamical set of equations



¹⁰ C. Pousot-Vassal and P. Vuillemin, "Introduction to MORE: a Model REduction Toolbox", IEEE Multi Systems Conference, pp. 776-781, 2012.

Fluid flow dynamical model approximation and control

... a case-study on an open cavity flow

C. Poussot-Vassal & D. Sipp

ONERA

THE FRENCH AEROSPACE LAB

Journée conjointe GT Contrôle de Décollement & GT MOSAR

