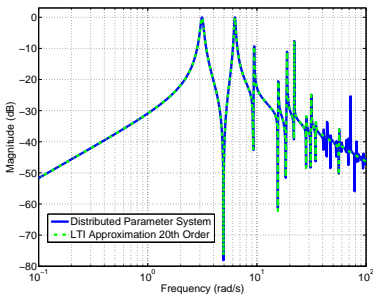


Model reduction of infinite dimensional systems: An application to TDS and linear PDE cases

I. Pontes Duff, P. Vuillemin, C. Pousot-Vassal, C. Seren
& C. Briat



Séminaire MOSAR Novembre 2014



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Optimal model approximation

Projection-based approximation framework

Rational interpolation Loewner framework

Stability regions estimation

Conclusions

Introduction**Context****Considered benchmarks**

Optimal model approximation

Projection-based approximation framework

Rational interpolation Loewner framework

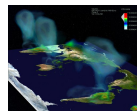
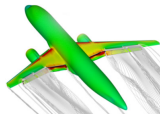
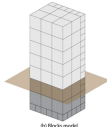
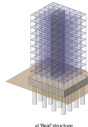
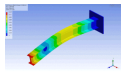
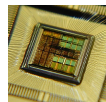
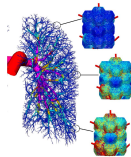
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LARGE-SCALE DYNAMICAL MODELS

... 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, ...)

Topics addressed in this presentation about model approximation:

- ▶ Some projection based methods in the finite dimensional case where a realization is available : IRKA/ITIA¹, IETIA² ...
- ▶ Interpolation method using Loewner framework^{3 4}
- ▶ Approximation of stability regions for large-scale time-delay systems^{5 6}


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

²  C. Poussot-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.

³  A.J. Mayo and A.C. Antoulas, "*A framework for the solution of the generalized realization problem*", Linear Algebra and its Applications 425(2-3), 2007, pp. 634-662.

⁴  C. Beattie and S. Gugercin, "*Realization-independent \mathcal{H}_2 -approximation*", IEEE Conference on Decision and Control, 2012, pp. 4953-4958.

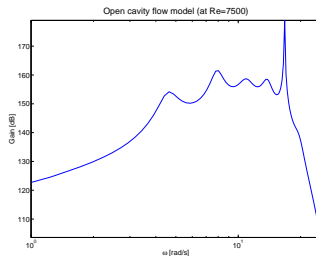
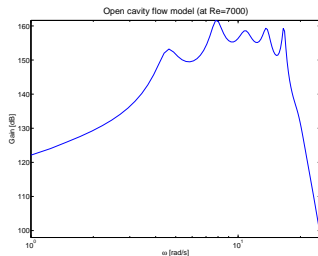
⁵  I. Pontes Duff, P. Vuillemin, C. Poussot-Vassal, C. Seren and C. Briat, "*Approximation of stability regions for large-scale time-delay systems using model reduction techniques*", submitted to ECC 2015.

⁶  I. Pontes Duff, P. Vuillemin, C. Poussot-Vassal, C. Seren and C. Briat, "*Stability and Performance Analysis of a Large-Scale Aircraft Anti-Vibration Control Subject to Delays Using Model Reduction Techniques*", submitted to EuroGNC 2015.

Considered benchmarks

Benchmark NSS : Navier-Stokes equation in a open cavity flow : discretization and linearisation for different Reynolds Numbers⁷

$$\begin{aligned} E\dot{x}(t) &= A(Re)x(t) + Bu(t) \\ y(t) &= Cx(t) \end{aligned} \quad (1)$$



- ▶ Two Reynolds cases ($Re = 7000$ and $Re = 7500$)
- ▶ SISO DAE, 8 unstable modes, order $\approx 700,000$ states

⁷  A. Barbagallo, D. Sipp and P. Schmid, "Closed-loop control of an open cavity flow using reduced order models", Journal of Fluid Mechanics, 641, 2009, pp. 1-50.

Considered benchmarks

Benchmark TDS-#1: Feedback delay and controller gain⁸

Let us consider

$$\dot{x}(t) = Ax(t) + Bu(t); y(t) = Cx(t), \quad (2)$$

where

$$A = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ -10 & 10 & 0 & 0 \\ 5 & -15 & 0 & -0.25 \end{pmatrix}, B = \begin{pmatrix} 0 \\ 0 \\ 1 \\ 0 \end{pmatrix}, C = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}^T. \quad (3)$$

We add to this model the delayed static output feedback $u(t) = -ky(t) + ky(t - \tau)$

The resulted model \mathbf{H}_{TDS1} is governed by

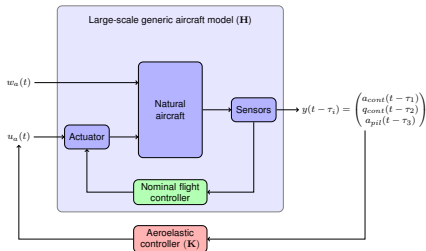
$$\dot{x}(t) = A_0x(t) + A_1x(t - \tau) \quad (4)$$

where $A - BCk$ and $A_1 = BCk$.

Question: Given (k, τ) , what is the stability of (4)?

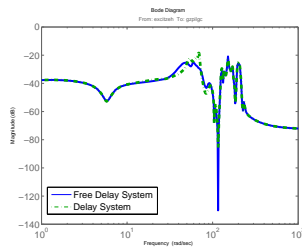
⁸  A. Seuret and F. Gouaisbaut, "Hierarchy of LMI conditions for the stability analysis of time delay systems", Report LAAS 14429.


Benchmark TDS-#2: Multiple delays (in feedback) large-scale system ⁹



Question: Stability function of $\{\tau_1, \tau_2, \tau_3\}$?
 How to measure loss of performance ?

- ▶ Vibration control of aircraft model.
- ▶ $\text{order}(\mathbf{H}) \approx 600$ states.
- ▶ Controller \mathbf{H} designed without taking into account time-delays.
- ▶ Three output delays $\{\tau_1, \tau_2, \tau_3\}$.



⁹  C. Pousot-Vassal and T. Loquen and P. Vuillemin and O. Cantinaud and J-P. Lacoste, "Business Jet Large-Scale Model Approximation and Vibration Control", IFAC ALCOSP, 2013, pp. 199-204.

Considered benchmarks

Benchmark PDE: Example string vibration with dissipation¹⁰

Vibrating string of length $L = 1$ whose ends are fixed with control and observation are both distributed along the string.


$$\frac{\partial^2 z(x, t)}{\partial t^2} + \varepsilon \left\langle \frac{\partial z(x, t)}{\partial t}, \mathbf{1}_{[0, \frac{1}{2}]} \right\rangle \mathbf{1}_{[0, \frac{1}{2}]}(x) = \frac{\partial^2 z(x, t)}{\partial x^2} + \mathbf{1}_{[0, \frac{1}{2}]}(x) u(t), \quad 0 < x < 1, \quad t \geq 0 \quad (5)$$

where, $\mathbf{1}_{[0, \frac{1}{2}]}(x) = \begin{cases} 1 = & 0 \leq x \leq 1/2 \\ 0 = & 1/2 < x \leq 1 \end{cases}$, with

$$z(0, t) = 0, z(1, t) = 0$$

, and

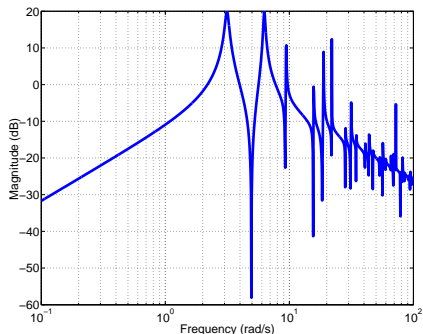
$$y(t) = \int_0^1 \frac{\partial z(x, t)}{\partial t} \mathbf{1}_{[0, \frac{1}{2}]}(x) dx.$$

¹⁰  R. Curtain and K. Morris, "Transfer functions of distributed parameter systems: A tutorial", Automatica, 45(5), 2009, pp. 1101-1116.

Benchmark PDE: Example string vibration with dissipation

The transfer function of this model is given by

$$H(s) = \frac{\frac{s}{2} \sinh(s) + 2 \cosh(\frac{s}{2}) - 3 \cosh^2(\frac{s}{2}) + 1}{s(s + \frac{1}{2}) \sinh(s) + 2 \cosh(\frac{s}{2}) - 3 \cosh^2(\frac{s}{2}) + 1} \quad (6)$$



Introduction

Optimal model approximation

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

\mathcal{H}_2 and \mathcal{L}_2 optimality conditions

Projection-based approximation framework

Rational interpolation Loewner framework

Stability regions estimation


Conclusions

Approximation in the \mathcal{H}_2 , \mathcal{H}_2, Ω and \mathcal{L}_2 -norm¹¹ \mathcal{L}_2 model approximation

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

$$\|H\|_{\mathcal{L}_2}^2 := \text{trace} \left(\frac{1}{2\pi} \int_{-\infty}^{\infty} (\overline{H(i\nu)} H(i\nu)) d\nu \right) \quad (8)$$


- ▶ $\mathcal{L}_2(i\mathbb{R})$ the Hilbert space of matrix-valued functions $\mathbb{C} \rightarrow \mathbb{C}^{n_y \times n_u}$ satisfying $\int_{\mathbb{R}} \text{trace}[F(i\omega)F(i\omega)^T] d\omega < \infty$.
- ▶ $\mathcal{L}_2(i\mathbb{R}) = \mathcal{H}_2(\mathbb{C}^-) \oplus \mathcal{H}_2(\mathbb{C}^+)$

¹¹  C. Magruder and C A. Beattie and S. Gugercin, "Rational Krylov methods for optimal \mathcal{L}_2 model reduction", IEEE Conference on Decision and Control, 2010.

\mathcal{H}_2 model approximation

$$\hat{H} := \arg \min_{\substack{G \in \mathcal{H}_2^{n_y \times n_u} \\ \dim(G) = r}} \|H - G\|_{\mathcal{H}_2} \quad (9)$$

$$\begin{aligned} \|H\|_{\mathcal{H}_2}^2 &:= \text{trace} \left(\frac{1}{2\pi} \int_{-\infty}^{\infty} (\overline{H(i\nu)} H(i\nu)) d\nu \right) \\ &:= \text{trace} (CPC^T) = \text{trace} (B^T QB) \\ &:= \sum_{i=1}^n \text{trace} (\phi_i H(-\lambda_i)^T) \end{aligned} \quad (10)$$

¹²  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.

¹³  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.

Assume that \mathbf{H} and $\hat{\mathbf{H}}$ have semi-simple poles and suppose that $\hat{\mathbf{H}}$ is a r^{th} -order finite-dimensional model with transfer function

$$\hat{H}(s) = \sum_{k=1}^r \frac{\hat{c}_k \hat{b}_k^T}{s - \hat{\lambda}_k}. \quad (11)$$


\mathcal{H}_2 -optimality conditions

If $\mathbf{H}, \hat{\mathbf{H}} \in \mathcal{H}_2$ and $\hat{\mathbf{H}}$ is a local minimum of the \mathcal{H}_2 approximation problem, then the following interpolations equations hold

$$H(-\hat{\lambda}_k) \hat{b}_k = \hat{H}(-\hat{\lambda}_k) \hat{b}_k, \quad \hat{c}_k^T H(-\hat{\lambda}_k) = \hat{c}_k^T \hat{H}(-\hat{\lambda}_k) \quad (12)$$

$$\hat{c}_k^T \left. \frac{dH}{ds} \right|_{s=-\hat{\lambda}_k} \hat{b}_k = \hat{c}_k^T \left. \frac{d\hat{H}}{ds} \right|_{s=-\hat{\lambda}_k} \hat{b}_k, \quad (13)$$

for all $k = 1, \dots, r$ where $\hat{\lambda}_k$ are the poles of $\hat{\mathbf{H}}$ and \hat{b}_k and \hat{c}_k are its tangential directions, respectively.

¹⁴  S. Gugercin and A. C. Antoulas and C. Beattie, " \mathcal{H}_2 Model Reduction for Large Scale Linear Dynamical Systems", SIAM Journal on Matrix Analysis and Applications, vol. 30(2), 2008, pp. 609-638.

In the case where $\mathbf{H} \in \mathcal{L}_2(i\mathbb{R})$ is a SISO LTI system and $\mathbf{H} = \mathbf{H}^+ + \mathbf{H}^-$ where $\mathbf{H}^+ \in \mathcal{H}(\mathbb{C}^+)$ and $\mathbf{H}^- \in \mathcal{H}(\mathbb{C}^-)$, it is possible to state the following result:


\mathcal{L}_2 optimality conditions

Given $\mathbf{H} \in \mathcal{L}_2(i\mathbb{R})$ and its decomposition $\mathbf{H} = \mathbf{H}^+ + \mathbf{H}^-$ where $\mathbf{H}^+ \in \mathcal{H}(\mathbb{C}^+)$ and $\mathbf{H}^- \in \mathcal{H}(\mathbb{C}^-)$. Let $\hat{\mathbf{H}}$ be the local minimizer of order r whose poles are all simple $\{\hat{\lambda}_1, \dots, \hat{\lambda}_k\} \in \mathbb{C}^-$ and $\{\hat{\lambda}_{k+1}, \dots, \hat{\lambda}_r\} \in \mathbb{C}^+$. If $\hat{H}(s)$ is given as (11) and if it is a local minimal of the \mathcal{L}_2 approximation problem, then following hold for $i = 1, \dots, k$

$$H^+(-\hat{\lambda}_i) = \hat{H}^+(-\hat{\lambda}_i), \quad \left. \frac{dH^+}{ds} \right|_{s=-\hat{\lambda}_i} = \left. \frac{d\hat{H}^+}{ds} \right|_{s=-\hat{\lambda}_i} \quad (14)$$

and for $i = k + 1, \dots, r$,


$$H^-(-\hat{\lambda}_i) = \hat{H}^-(\hat{\lambda}_i), \quad \left. \frac{dH^-}{ds} \right|_{s=-\hat{\lambda}_i} = \left. \frac{d\hat{H}^-}{ds} \right|_{s=-\hat{\lambda}_i} \quad (15)$$


¹⁵  C. Magruder and C A. Beattie and S. Gugercin, "Rational Krylov methods for optimal \mathcal{L}_2 model reduction", IEEE Conference on Decision and Control, 2010.

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

$$\hat{H} := \arg \min_{\substack{G \in \mathcal{H}_{\infty}^{n_y \times n_u} \\ \dim(G) = r}} \|H - G\|_{\mathcal{H}_{2,\Omega}} \quad (16)$$

$$\begin{aligned} \|H\|_{\mathcal{H}_{2,\Omega}}^2 &:= \text{trace} \left(\frac{1}{\pi} \int_{\Omega} \overline{H(i\nu)} H(i\nu) d\nu \right) \\ &:= \text{trace} \left(C P_{\Omega} C^T \right) = \text{trace} \left(B^T Q_{\Omega} B \right) \\ &:= \sum_{i=1}^n \text{trace} \left(\phi_i H(-\lambda_i)^T \right) \left[-\frac{2}{\pi} \text{atan} \left(\frac{\omega}{\lambda_i} \right) \right] \end{aligned} \quad (17)$$


¹⁶  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.


¹⁷  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.

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Mismatch objective and eigenvector preservation

$$\hat{H} := \arg \min_{\substack{G \in \mathcal{L}_2^{n_y \times n_u} \\ \dim(G) = r \ll n \\ \lambda_k(G) \subseteq \lambda(H) \quad k = 1, \dots, q_1 < r}} \|H - G\|_{\mathcal{H}_2} \quad (18)$$

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



¹⁸ 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.

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MIMO IRKA (or ITIA)

IETIA

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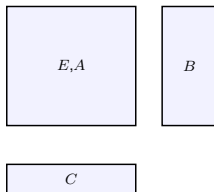
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Projectors

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} :

$$\mathbf{H} : \begin{cases} E\dot{x}(t) & = Ax(t) + Bu(t) \\ y(t) & = Cx(t) \end{cases} \quad (19)$$



PROJECTION-BASED APPROXIMATION FRAMEWORK

Projectors

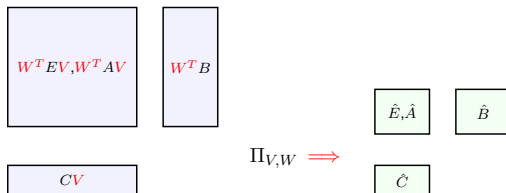
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the approximation problem consists in finding $V, W \in \mathbb{R}^{n \times r}$ (with $r \ll n$) **spanning** \mathcal{V} **and** \mathcal{W} **subspaces** and forming a **projector** $\Pi_{V,W} = VW^T$, such that

$$\hat{\mathbf{H}} : \begin{cases} W^T E V \dot{\hat{x}}(t) &= W^T A V \hat{x}(t) + W^T B u(t) \\ \hat{y}(t) &= C V \hat{x}(t) \end{cases} \quad (20)$$

well approximates \mathbf{H} .



PROJECTION-BASED APPROXIMATION FRAMEWORK

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well approximates \mathbf{H} .

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


PROJECTION-BASED APPROXIMATION FRAMEWORK


MIMO IRKA (or ITIA) - \mathcal{H}_2 optimality conditions (Tangential subspace approach) ^{19 20}

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 $r = q_2$ such that $W^*V = I_r$. If, for $j = 1, \dots, q_2$,

$$\left[(\sigma_j E - A)^{-1} B \hat{b}_j \right] \in \mathbf{span}(V) \text{ and } \left[(\sigma_j E - A^T)^{-1} C^T \hat{c}_j^* \right] \in \mathbf{span}(W) \quad (21)$$

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.

¹⁹  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.

²⁰  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.


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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

$$\begin{aligned} H(-\hat{\sigma}_j) \hat{b}_j &= \hat{H}(-\hat{\sigma}_j) \hat{b}_j \\ \hat{c}_j^* H(-\hat{\sigma}_j) &= \hat{c}_j^* \hat{H}(-\hat{\sigma}_j) \\ \hat{c}_j^* \frac{d}{ds} H(s) \Big|_{s=-\hat{\sigma}_j} \hat{b}_j &= \hat{c}_j^* \frac{d}{ds} \hat{H}(s) \Big|_{s=-\hat{\sigma}_j} \hat{b}_j \end{aligned} \quad (22)$$

¹⁹  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.

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PROJECTION-BASED APPROXIMATION FRAMEWORK

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 $r = q_2 \in \mathbb{N}$

1: Construct,

$$\text{span}(V(\sigma_j^{(0)}, \hat{b}_j)) \text{ and } \text{span}(W(\sigma_j^{(0)}, \hat{c}_j^*)) \quad (23)$$

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

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

9: Construct,

$$\text{span}(V(\sigma_j^{(k)}, \hat{b}_j)) \text{ and } \text{span}(W(\sigma_j^{(k)}, \hat{c}_j^*)) \quad (24)$$

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

PROJECTION-BASED APPROXIMATION FRAMEWORK

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 $r = q_2 \in \mathbb{N}$

1: Construct,

$$\text{span}(V(\sigma_j^{(0)}, \hat{b}_j)) \text{ and } \text{span}(W(\sigma_j^{(0)}, \hat{c}_j^*)) \quad (23)$$

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 $L\hat{A} = \Lambda(\hat{A}, \hat{E})L$

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11: **end while**

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

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

PROJECTION-BASED APPROXIMATION FRAMEWORK

NETIA - \mathcal{H}_2 & spectral optimality conditions (Tangential subspace approach) ²¹

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 $j = 1, \dots, q_2$,

$$\left[r_i^* (\sigma_j E - A)^{-1} B \hat{b}_j \right] \in \mathbf{span}(V) \text{ and } \left[l_i^* (\sigma_j E - A^T)^{-1} C^T \hat{c}_j^* \right] \in \mathbf{span}(W) \quad (25)$$

$l_i^* \in \mathbb{C}^n$ 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.

²¹ 

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.

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$$\{\lambda_1^*, \dots, \lambda_{q_1}^*\} \subset \Lambda(\hat{A}, \hat{E}) \quad (26)$$

and the tangential interpolation conditions

$$\begin{aligned} H(-\hat{\sigma}_j) \hat{b}_j &= \hat{H}(-\hat{\sigma}_j) \hat{b}_j \\ \hat{c}_j^* H(-\hat{\sigma}_j) &= \hat{c}_j^* \hat{H}(-\hat{\sigma}_j) \\ \hat{c}_j^* \frac{d}{ds} H(s) \Big|_{s=-\hat{\sigma}_j} \hat{b}_j &= \hat{c}_j^* \frac{d}{ds} \hat{H}(s) \Big|_{s=-\hat{\sigma}_j} \hat{b}_j \end{aligned} \quad (27)$$

²¹  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.

PROJECTION-BASED APPROXIMATION FRAMEWORK

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,

$$\text{span}(V(l_i^*, \sigma_j^{(0)}, \hat{b}_j)) \text{ and } \text{span}(W(r_i^*, \sigma_j^{(0)}, \hat{c}_j^*)) \quad (28)$$

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 $L \hat{A} = \Lambda(\hat{A}) L$

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$$\text{span}(V(l_i^*, \sigma_j^{(k)}, \hat{b}_j)) \text{ and } \text{span}(W(r_i^*, \sigma_j^{(k)}, \hat{c}_j^*)) \quad (29)$$

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

PROJECTION-BASED APPROXIMATION FRAMEWORK

- 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 $r = q_1 + q_2 \in \mathbb{N}$
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 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,

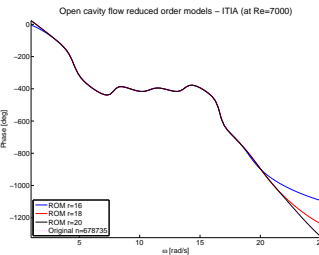
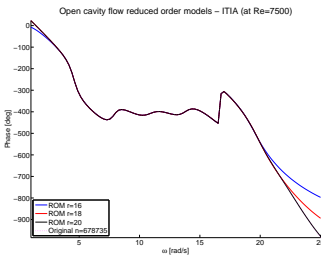
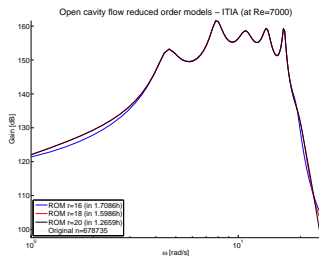
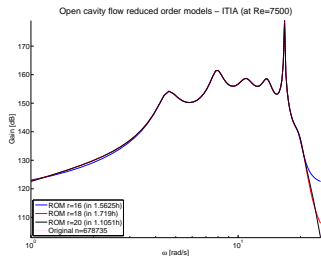
$$\text{span}(V(l_i^*, \sigma_j^{(k)}, \hat{b}_j)) \text{ and } \text{span}(W(r_i^*, \sigma_j^{(k)}, \hat{c}_j^*)) \quad (29)$$

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

PROJECTION-BASED APPROXIMATION FRAMEWORK

Fluid flow dynamical model approximation - Re=7000 and Re=7500



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Rational interpolation

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Rational interpolation

Given $H(s)$, complex points $\sigma_1, \dots, \sigma_n$, and tangential directions $\hat{b}_1, \dots, \hat{b}_n$, $\hat{c}_1, \dots, \hat{c}_n$, one constructs $(\hat{E}, \hat{A}, \hat{B}, \hat{C})$ such that the transfer function $\hat{H}(s) = \hat{C}(s\hat{E} - \hat{A})^{-1}\hat{B}$ satisfies the tangential interpolation conditions :

$$\begin{aligned}
 H(\sigma_j)\hat{b}_j &= \hat{H}(\sigma_j)\hat{b}_j \\
 \hat{c}_j^T H(\sigma_j) &= \hat{c}_j^T \hat{H}(\sigma_j) \\
 \hat{c}_j^T \frac{d}{ds} H(s) \Big|_{s=\sigma_j} \hat{b}_j &= \hat{c}_j^T \frac{d}{ds} \hat{H}(s) \Big|_{s=\sigma_j} \hat{b}_j
 \end{aligned} \tag{30}$$

This is possible thanks to Loewner matrices.

RATIONAL INTERPOLATION LOEWNER FRAMEWORK

Loewner approach²²


The rational function $\hat{C}(s\hat{E} - \hat{A})^{-1}\hat{B}$ interpolates $H(s)$ at points σ_i and directions \hat{b}_i and \hat{c}_i iff.

$$(\hat{E})_{ij} = \begin{cases} -\frac{\hat{c}_i^T (H(\sigma_i) - H(\sigma_j)) \hat{b}_j}{\sigma_i - \sigma_j} & i \neq j \\ -\hat{c}_i^T H'(\sigma_i) \hat{b}_i & i = j \end{cases}$$

$$(\hat{A})_{ij} = \begin{cases} -\frac{\hat{c}_i^T (\sigma_i H(\sigma_i) - \sigma_j H(\sigma_j)) \hat{b}_j}{\sigma_i - \sigma_j} & i \neq j \\ -\hat{c}_i^T (sH(s))'|_{s=\sigma_i} \hat{b}_i & i = j \end{cases}$$

$$\hat{C} = [H(\sigma_1)\hat{b}_1, \dots, H(\sigma_r)\hat{b}_r] \quad \text{and} \quad \hat{B} = \begin{bmatrix} \hat{c}_1^T H(\sigma_1) \\ \vdots \\ \hat{c}_r^T H(\sigma_r) \end{bmatrix}.$$

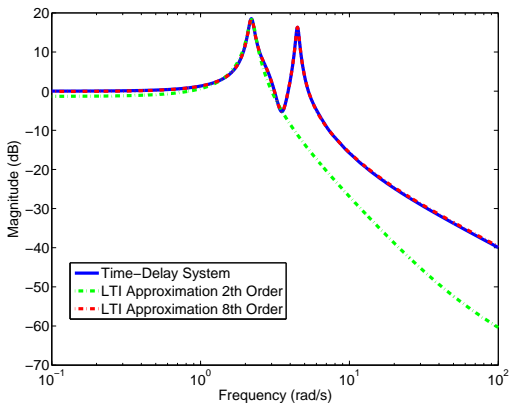
- An analogous to IRKA iterative method was proposed.

²²  A.J. Mayo and A.C. Antoulas, "A framework for the solution of the generalized realization problem", Linear Algebra and its Applications 425(2-3) 634 - 662, 2007.

- 1: **Initialization:** transfer function $H(s)$, dimension r , $\sigma^0 = \{\sigma_1^0, \dots, \sigma_r^0\} \in \mathbb{C}$ initial interpolation points and tangential directions $\{b_1, \dots, b_r\} \in \mathbb{C}^{n_u \times 1}$ and $\{c_1, \dots, c_r\} \in \mathbb{C}^{n_y \times 1}$.
- 2: **while** not convergence **do**
- 3: **Build** \hat{E} , \hat{A} , \hat{B} and \hat{C} using *Loewner Matrices*.
- 4: Solve the generalized eigenvalue problem $\hat{A}^{(k)} x_i^{(k)} = \lambda_i^{(k)} \hat{E}^{(k)} x_i^{(k)}$ and $y_i^{(k)}$ such that $y_i^{(k)*} \hat{E}^{(k)} x_j^{(k)} = \delta_{i,j}$.
- 5: **Set** $\sigma_i^{(k+1)} \leftarrow -\lambda_i^{(k)}$, $b_i^{(k+1)T} \leftarrow y_i^{(k)} \hat{B}^{(k)}$ and $c_i^{(k+1)} \leftarrow \hat{C}^{(k)} x_i^{(k)}$, for $i = 1, \dots, r$.
- 6: **end while**
- 7: **Ensure** conditions (31) are satisfied.
- 8: **Build** \hat{E} , \hat{A} , \hat{B} and \hat{C} .

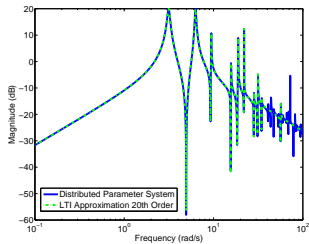
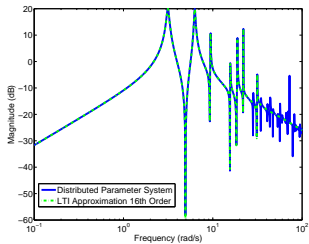
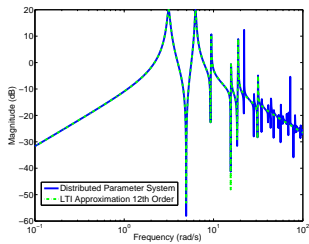
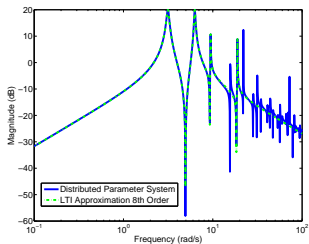
²³  C. Beattie and S. Gugercin, "*Realization-independent \mathcal{H}_2 -approximation*", Proceedings of the 51st IEEE Conference on Decision and Control, 2012.

Example TDS-#1 for $k = 1$ and $\tau = 3$



RATIONAL INTERPOLATION LOEWNER FRAMEWORK

Example DPS



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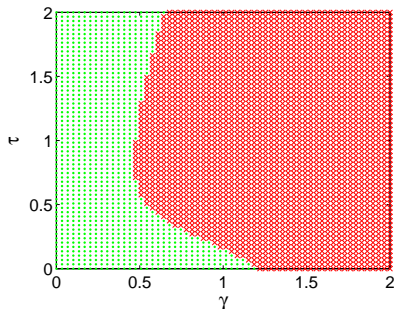
Proposed Strategy - Approximation & Eigenvalues (accompanied with proofs)

- ▶ **Procedure:** Estimating stability regions using model approximation & eigenvalues
- ▶ **Arguments for proof:** Provide some arguments why this procedure is valid

Proposed Strategy - Approximation & Eigenvalues (accompanied with proofs)

- ▶ **Procedure:** Estimating stability regions using model approximation & eigenvalues
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Example: Let us consider the model described by the transfer function $H(s) = \frac{1}{1+e^{\tau s}+2e^{\gamma s}}$, with $\tau, \gamma \in [0, 2]$. After discretizing $[0, 2]$ and finding LTI approximation via TF-IRKA, the stability of the reduced model is plotted



STABILITY REGIONS ESTIMATION

Results about stability approximation in \mathcal{L}_2

- ▶ $\mathcal{L}_2(i\mathbb{R})$ the Hilbert space of matrix-valued functions $\mathbb{C} \rightarrow \mathbb{C}^{n_y \times n_u}$ satisfying $\int_{\mathbb{R}} \text{trace}[\overline{F(i\omega)}F(i\omega)^T]d\omega < \infty$.
- ▶ $\langle \mathbf{H}, \mathbf{G} \rangle_{\mathcal{L}_2} = \frac{1}{2\pi} \int_{-\infty}^{\infty} \text{trace}(\overline{H(i\omega)}G(i\omega)^T) d\omega$.
- ▶ $\mathcal{H}_2(\mathbb{C}^+)$ ($\mathcal{H}_2(\mathbb{C}^-)$) closed subspace of $\mathcal{L}_2(i\mathbb{R})$ containing the matrix functions $F(s)$ analytic in the open right-half plane (open left-half plane).
- ▶ $\mathcal{L}_2(i\mathbb{R}) = \mathcal{H}_2(\mathbb{C}^-) \oplus \mathcal{H}_2(\mathbb{C}^+)$
- ▶ $\mathcal{L}_2(i\mathbb{R}) \setminus \mathcal{H}_2(\mathbb{C}^+)$ set of unstable LTI systems
- ▶ **Remark: TF-IRKA** allows us to obtain a system of order r which satisfies the interpolation conditions :

$$H(-\hat{\lambda}_k)\hat{b}_k = \hat{H}(-\hat{\lambda}_k)\hat{b}_k, \quad \hat{c}_k^T H(-\hat{\lambda}_k) = \hat{c}_k^T \hat{H}(-\hat{\lambda}_k) \quad (31)$$

$$\hat{c}_k^T \left. \frac{dH}{ds} \right|_{s=-\hat{\lambda}_k} \hat{b}_k = \hat{c}_k^T \left. \frac{d\hat{H}}{ds} \right|_{s=-\hat{\lambda}_k} \hat{b}_k, \quad (32)$$

Proposition 1

If $\mathbf{H} \in \mathcal{H}_2(\mathbb{C}^+)$ and there exists a global minimizer $\hat{\mathbf{H}} \in \mathcal{L}_2(i\mathbb{R})$ of the \mathcal{L}_2 approximation problem, then $\hat{\mathbf{H}} \in \mathcal{H}_2(\mathbb{C}^+)$. Similarly, if $\mathbf{H} \in \mathcal{H}_2(\mathbb{C}^-)$ and there exists a global minimizer $\hat{\mathbf{H}} \in \mathcal{L}_2(i\mathbb{R})$ of the \mathcal{L}_2 approximation problem, then $\hat{\mathbf{H}} \in \mathcal{H}_2(\mathbb{C}^-)$.

Proof.

Let $\hat{\mathbf{H}} \in \mathcal{L}_2(i\mathbb{R})$ be the global minimizer of \mathcal{L}_2 approximation problem. Since $\mathbf{H} \in \mathcal{H}_2(\mathbb{C}^+)$, one has $\mathbf{H}^- = \mathbf{0}$. Seeing that $\mathcal{L}_2(i\mathbb{R}) = \mathcal{H}_2(\mathbb{C}^-) \oplus \mathcal{H}_2(\mathbb{C}^+)$ and this an orthogonal decomposition, thus

$$\|\mathbf{H} - \hat{\mathbf{H}}\|_{\mathcal{L}_2}^2 = \|\mathbf{H}^+ - \hat{\mathbf{H}}^+\|_{\mathcal{L}_2}^2 + \|\mathbf{0} - \hat{\mathbf{H}}^-\|_{\mathcal{L}_2}^2 \quad (33)$$

Thus, $\hat{\mathbf{H}}^- = \mathbf{0}$, otherwise $\hat{\mathbf{H}}$ is not a global minimizer. □

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Thus, $\hat{\mathbf{H}}^- = \mathbf{0}$, otherwise $\hat{\mathbf{H}}$ is not a global minimizer. □

Results about stability approximation in \mathcal{L}_2

Proposition 2

For every unstable system \mathbf{H} , there exists a neighborhood V of \mathbf{H} such that if $\mathbf{G} \in V$, \mathbf{G} is also unstable. In other words, the set of unstable systems $(\mathcal{L}_2(i\mathbb{R}) \setminus \mathcal{H}_2(\mathbb{C}^+))$ is open for the \mathcal{L}_2 norm.

Proof.

Since $\mathcal{H}_2(\mathbb{C}^+)$ is a closed set, its complement $(\mathcal{H}_2(\mathbb{C}^+))^c = \mathcal{L}_2(i\mathbb{R}) \setminus \mathcal{H}_2(\mathbb{C}^+)$ is open. \square

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Theorem 1

Given a unstable system $\mathbf{H} \in \mathcal{L}_2(i\mathbb{R}) \setminus \mathcal{H}_2(\mathbb{C}^+)$, there exists $n \in \mathbb{N}^*$ for which the minimizer G_k of order $k \in \mathbb{N}^*$, $k > n$, obtained from the \mathcal{L}_2 -approximation problem is also unstable.

Proof.

Proposition 2 states that if a system is sufficiently close to a unstable system in the $\mathcal{L}_2(i\mathbb{R})$ norm, it is also unstable. Furthermore, the subspace of rational functions which represents the finite LTI systems is dense in $\mathcal{L}_2(i\mathbb{R})$. Hence, for a given LTI unstable system $\mathbf{H} \in \mathcal{L}_2(i\mathbb{R}) \setminus \mathcal{H}_2(\mathbb{C}^+)$, a sequence G_k of systems of order k which satisfies the tangential interpolation conditions, will converge to \mathbf{H} . Thus, due to Proposition 2, there exists an order $n \in \mathbb{N}^*$ such that if $k \geq n$, G_k will be unstable as well. \square

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Results about stability approximation in \mathcal{L}_2

Proposition 3

For every stable system $\mathbf{H} \in \mathcal{H}_2(\mathbb{C}^+)$, there exists a sequence of unstable systems $\mathbf{G}_k \in \mathcal{L}_2(i\mathbb{R}) \setminus \mathcal{H}_2(\mathbb{C}^+)$, $k \in \mathbb{N}^*$, such that

$$\|\mathbf{H} - \mathbf{G}_k\|_{\mathcal{L}_2(i\mathbb{R})} \rightarrow 0, \quad \text{when } k \rightarrow \infty \quad (34)$$

In other words, the set $\mathcal{H}_2(\mathbb{C}^+)$ is not an open set of $\mathcal{L}_2(i\mathbb{R})$.

Proof.

Given $\mathbf{H} \in \mathcal{H}_2(\mathbb{C}^+)$, let $\mathbf{h} \in \mathcal{H}_2(\mathbb{C}^-)$ be an element such that $\|\mathbf{h}\|_{\mathcal{L}_2(i\mathbb{R})} = 1$. The system $\mathbf{G}_k = \mathbf{H} + \frac{1}{k}\mathbf{h} \in \mathcal{L}_2(i\mathbb{R}) \setminus \mathcal{H}_2(\mathbb{C}^+)$ and $\|\mathbf{H} - \mathbf{G}_k\|_{\mathcal{L}_2(i\mathbb{R})} = \frac{1}{k}\|\mathbf{h}\|_{\mathcal{L}_2(i\mathbb{R})} \rightarrow 0$ when $k \rightarrow \infty$. \square

Proposition 3

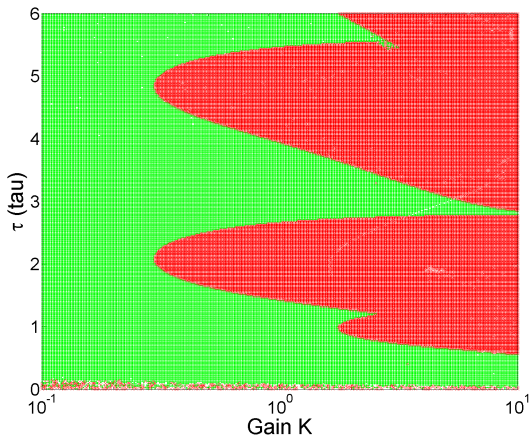
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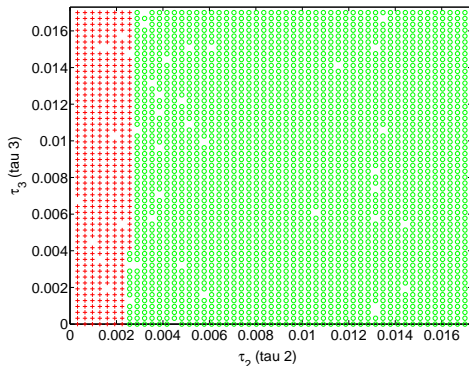
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- ▶ ≈ 0.13 s for each approximation
- ▶ Approximation of order $r = 6$.

STABILITY REGIONS ESTIMATION

Benchmark TDS-#2



- ▶ τ_1 fixed as 17ms.
- ▶ ≈ 30 s for approximation
- ▶ Approximation of order $r = 12$.

Introduction

Optimal model approximation

Projection-based approximation framework

Rational interpolation Loewner framework

Stability regions estimation

Conclusions

Conclusion

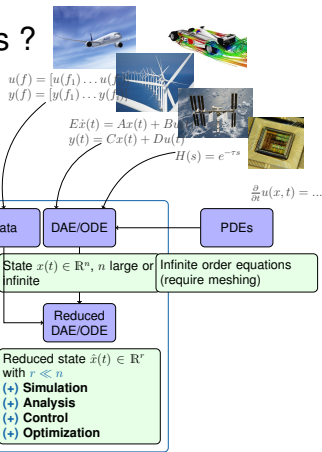
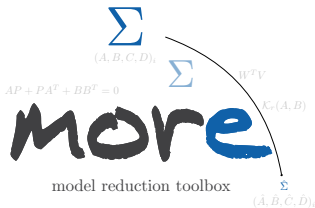
- ▶ Projection model approximation method using realization.
- ▶ Loewner interpolation method using transfer function.
- ▶ Method to estimate the stability of large-scale TDS and PDE is proposed.
- ▶ Some arguments are given to justify this method
- ▶ No borne of estimation error.

Perspectives

- ▶ Algorithm 'branch and bound' to find borders.
- ▶ \mathcal{H}_2 -LPV model reduction.
- ▶ LSS reduction to TDS-system.

- MORE toolbox ²⁴

Thanks for your Attention. Questions ?



²⁴  C. Pousot-Vassal and P. Vuillemin, "Introduction to MORE: a Model Reduction Toolbox", IEEE Multi Systems Conference, pp. 776-781.