

Civil aircraft reduced order modelling, control and validation: Numerical and experimental challenges

... how much dynamical data science is a strong ally?

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Friday September 18, 2020

28th IEEE Mediterranean Conference on Control and Automation Conference

(e-) St Raphael, France

FrPL3 Plenary Session, ROOM SR1



Introduction

Mobility and civil aviation challenges

(Aircraft) mobility plays a central role...

in our **life style** (leisure travels) and **societal organisation** (commercial exchanges)
but suffers of
severe environmental changes (global warming).

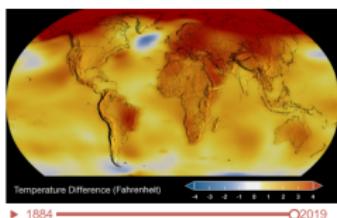


TIME SERIES: 1884 TO 2019

Data source: NASA GISS

Credit: NASA Scientific Visualization Studio

2019



Left : world view of civil aviation (August 19th, 2020)
[<https://www.flightradar24.com/>]

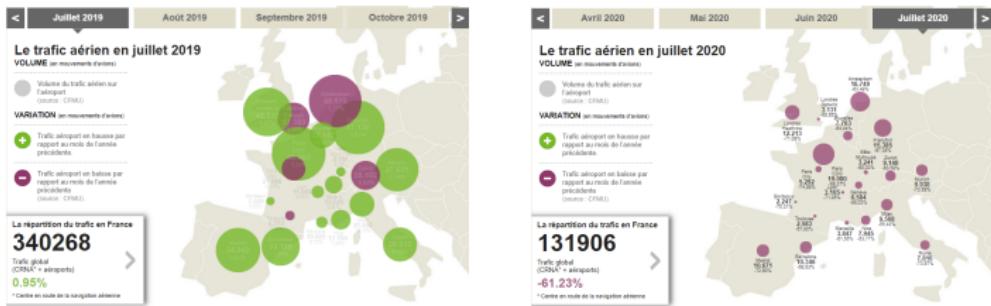
Right : 2019 temperature differences wrt. 1880
[<https://climate.nasa.gov/vital-signs/global-temperature/>]

Introduction

Mobility and civil aviation challenges

(Aircraft) mobility plays a central role...

in our life style (leisure travels) and societal organisation (commercial exchanges)
but suffers of
severe environmental changes (global warming) and pandemic issues (COVID19).



Civil aircraft traffic in the main French and border airports (July, 2019 vs. 2020)

[<http://salledelecture-ext.aviation-civile.gouv.fr/externe/mouvementsDavions/index.php>]

Introduction

Some civil aircraft challenges

Reducing footprint and emissions while remaining competitive?

- ▶ Enhance aviation flow (traffic optimisation)
- ▶ Enhance motors technology (combustion, electricity)
- ▶ Enhance aircraft efficiency (weight, structure)



Dynamical data science...

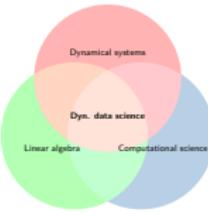
Dynamical systems, Linear algebra and Computational science provide an appropriate toolkit to address these problems, make decisions, enhance the development process and optimise aircraft systems, in a certified and competitive industrial context.

Introduction

Some civil aircraft challenges

Reducing footprint and emissions while remaining competitive?

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Dynamical data science...

Dynamical systems, Linear algebra and Computational science provide an appropriate toolkit to address these problems, make decisions, enhance the development process and **optimise aircraft systems, in a certified and competitive industrial context.**

Introduction

Today's talk outline

How to enhance aircraft efficiency and design process? Use dynamical data science

Application on a Generic Business Jet aircraft **models and experiments**
Gust load alleviation and Vibration reduction



1. Aircraft gust load and vibration **control** problems
2. Aircraft large-scale **modelling**, **model approximation** and **reduction**
3. Aircraft gust load and vibration **control design** in a industrial setup
4. Aircraft **control implementation** and **uncertainty analysis**
5. Aircraft ground and flight **experimental validations** and **data-driven modelling**

Introduction

Today's talk outline

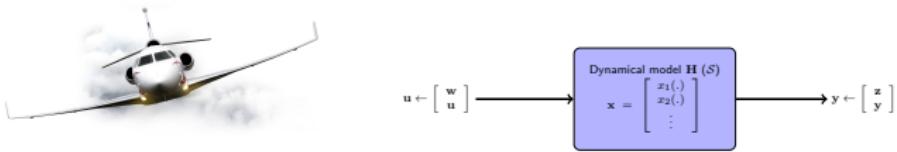
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Application on a Generic Business Jet aircraft models and experiments
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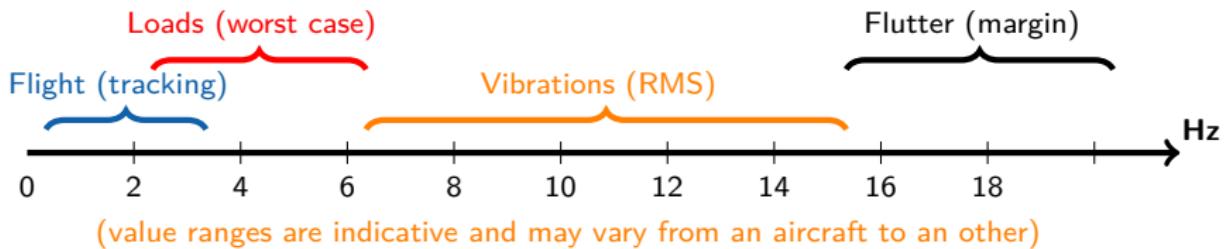
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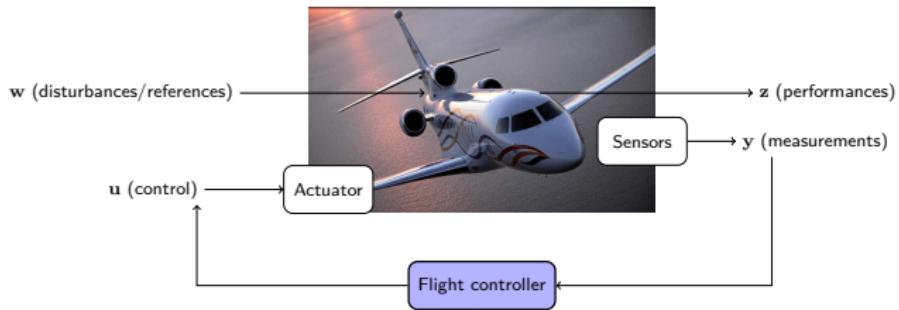
Aircraft control system problem

Sequential feedback control (industrial) problem description and architecture



Aircraft control system problem

Sequential feedback control (industrial) problem description and architecture

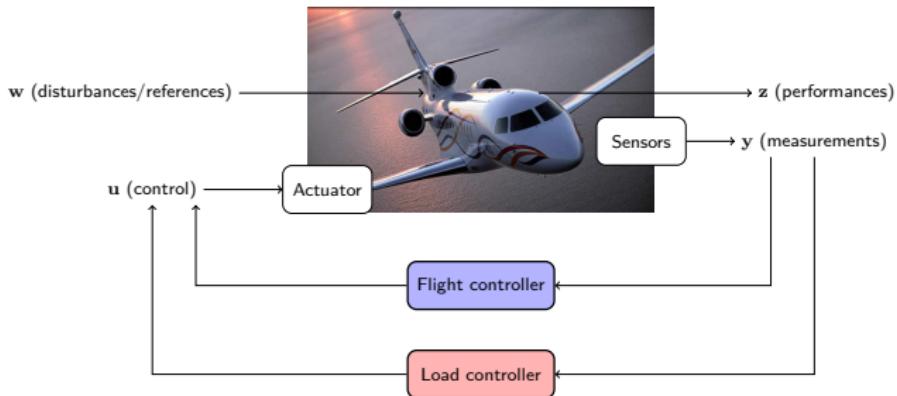


Flight qualities impact handling

⇒ aircraft manoeuvrability

Aircraft control system problem

Sequential feedback control (industrial) problem description and architecture

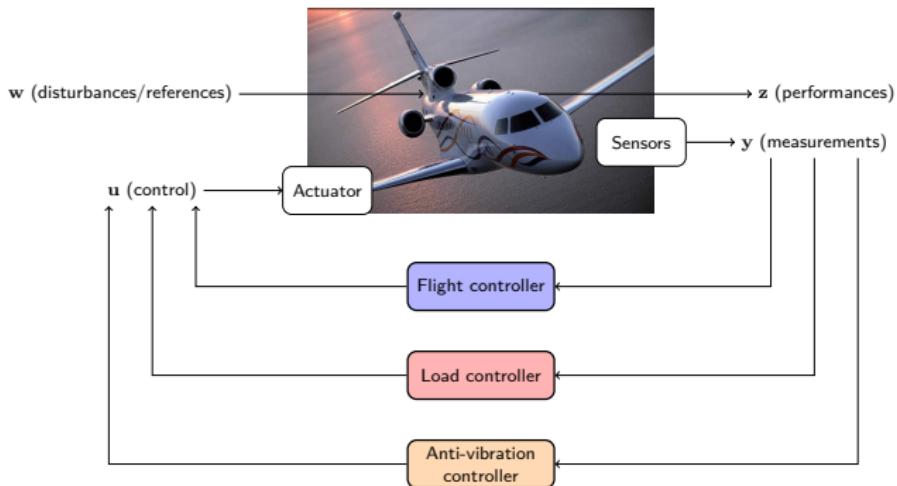


Flight qualities impact handling
Loads impact structure

⇒ aircraft manoeuvrability
⇒ aircraft weight reduction

Aircraft control system problem

Sequential feedback control (industrial) problem description and architecture

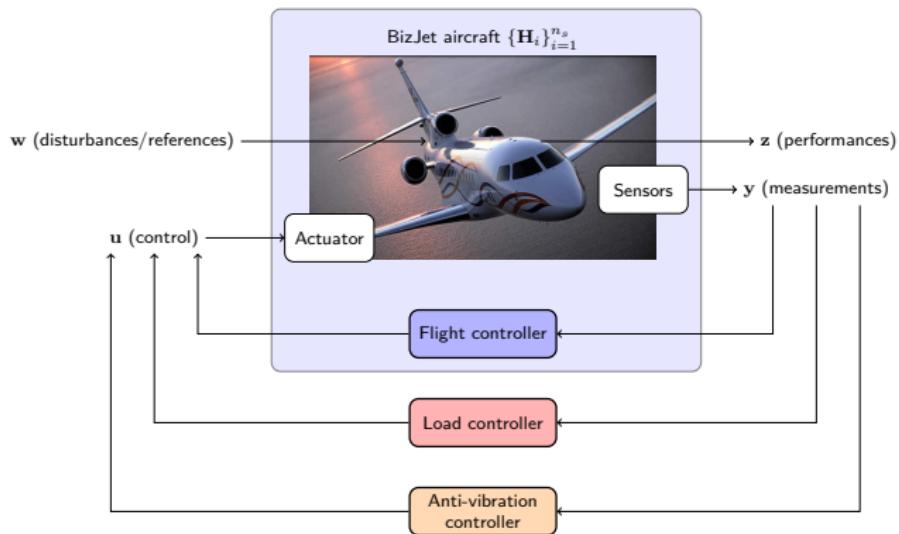


Flight qualities impact handling
Loads impact structure
Vibrations impact fatigue

⇒ aircraft manoeuvrability
⇒ aircraft weight reduction
⇒ aircraft life-time

Aircraft control system problem

Sequential feedback control (industrial) problem description and architecture



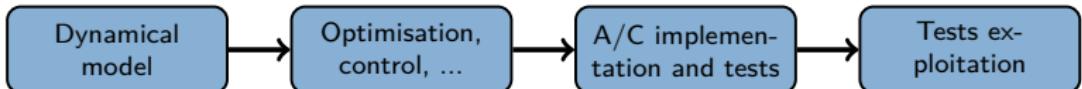
Aircraft are dynamical systems

How to model, design controllers, analyse and validate ?

Today: focus on **Gust Load Alleviation** and **Vibration Reduction** objectives

Aircraft control system problem

Find solutions fitting the industrial process, constraints and philosophy

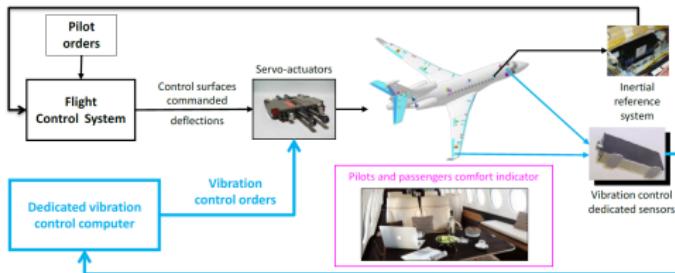
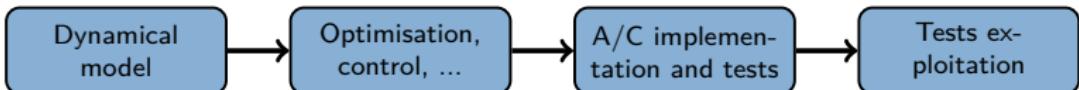


 C. Meyer, J. Prodigue, G. Broux, O. Cantinaud and C. P-V., "[Ground test for vibration control demonstrator](#)", in Proceedings of the International Conference on Motion and Vibration Control (MOVIC), Southampton, United Kingdom, July, 2016, pp. 1-12.

 D. Quero-Martin, P. Vuillemin and C. P-V., "[A Generalized State-Space Aeroservoelastic Model based on Tangential Interpolation](#)", in Aerospace, Special Issue Aeroelasticity, January 2019 (Open Access paper).

Aircraft control system problem

Find solutions fitting the industrial process, constraints and philosophy



Hidden constraints

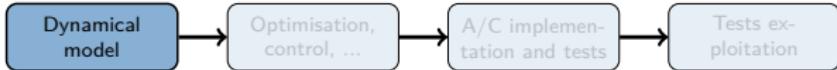
Find solution that fit the industrial workflow.

Sensors/Actuators

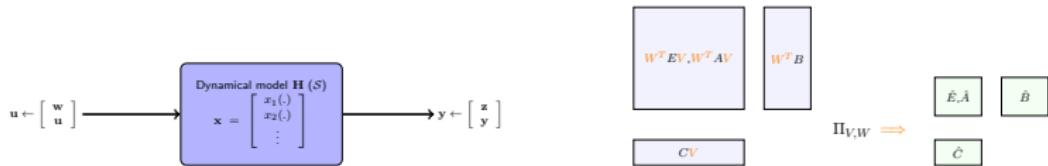
Use existing control layout.

C. Meyer, J. Prodigue, G. Broux, O. Cantinaud and C. P-V., "[Ground test for vibration control demonstrator](#)", in Proceedings of the International Conference on Motion and Vibration Control (MOVIC), Southampton, United Kingdom, July, 2016, pp. 1-12.

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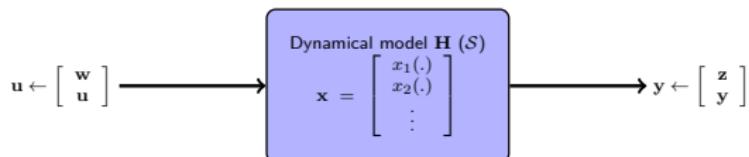
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Aircraft reduced order modelling

System and models

A dynamical model H is a function mapping input u to output y signals of a system Σ



$$\mathcal{S} : \begin{cases} \dot{x}(t) &= f(x(t), u(t)) & , \text{ differential equation} \\ 0 &= g(x(t), u(t)) & , \text{ algebraic equation} \\ y(t) &= h(x(t), u(t)) & , \text{ output equation} \end{cases}$$

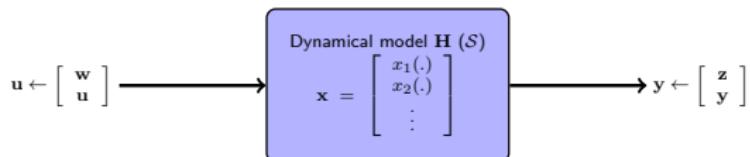
Nonlinear models

Mathematical and numerical tools **not simple** to deploy in industrial context

Aircraft reduced order modelling

System and linear models

A linear dynamical model H is a linear function mapping input u to output y signals of a system Σ



$$\mathcal{S} : \begin{cases} E\dot{x}(t) &= A(x(t)) + Bu(t) & , \text{ algebraic - differential equation} \\ y(t) &= C(t) & , \text{ output equation} \end{cases}$$

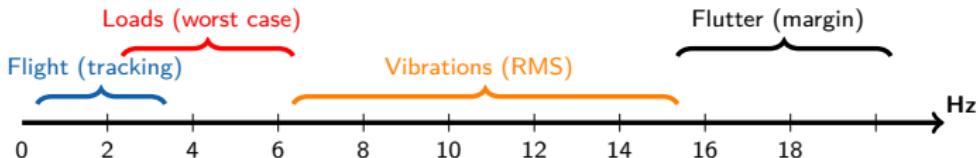
$$\begin{array}{ll} A(x(t)) &= Ax(t) \text{ or } Ax(t) + A_\tau x(t - \tau) \dots \\ E & \text{possibly singular} \end{array}$$

Linear models

Mathematical and numerical tools
ready to deploy in industrial context

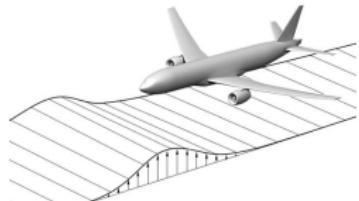
Aircraft reduced order modelling

Models tailored to phenomena and problems specificities



#1 gust load control models

- ▶ Multiple flight config. ($\approx \#1,000$)
- ▶ Large-scale (≈ 300 internal variables)
- ▶ Polynomial (due to pos./vel./accel. inputs)
- ▶ Irrational (due to internal delays)
- ▶ Band-limited phenomena [2 6]Hz



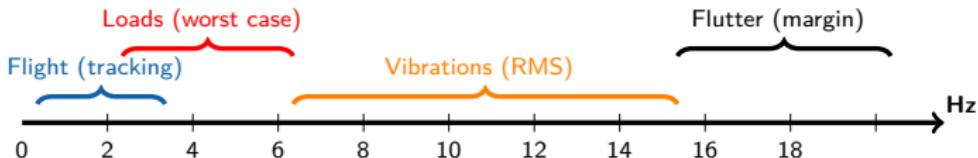
P. Lancelot et al., "Investigation of the unsteady flow over a wing gust excitation", IFASD 2017.

Models / data of the form

$$H(s) = C \left(sE - A_0 - A_1 e^{-\tau_1 s} - A_2 e^{-\tau_2 s} \right)^{-1} B$$

Aircraft reduced order modelling

Models tailored to phenomena and problems specificities



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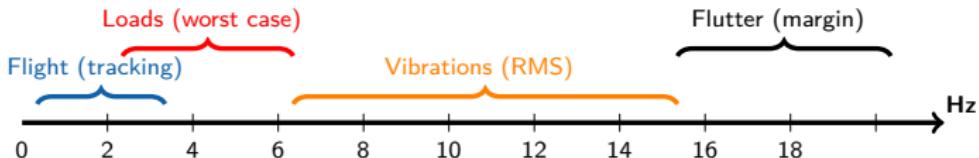
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Aircraft reduced order modelling

Models tailored to phenomena and problems specificities



#2 vibration control models

- ▶ Multiple flight config. ($\approx \#100$)
- ▶ Large-scale (≈ 800 internal variables)
- ▶ Rational (no polynomial part)
- ▶ Band-limited phenomena [6 15]Hz
- ▶ Cross validation using data in GVT

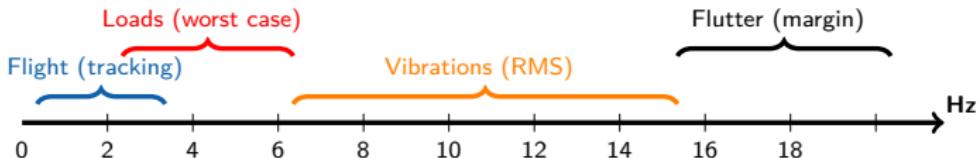


Models / data of the form

$$\mathbf{H}(s) = C(sI - A)^{-1} B \text{ (and } \{\mathbf{u}(t_k), \mathbf{y}(t_k)\} \text{ for tests)}$$

Aircraft reduced order modelling

Models tailored to phenomena and problems specificities



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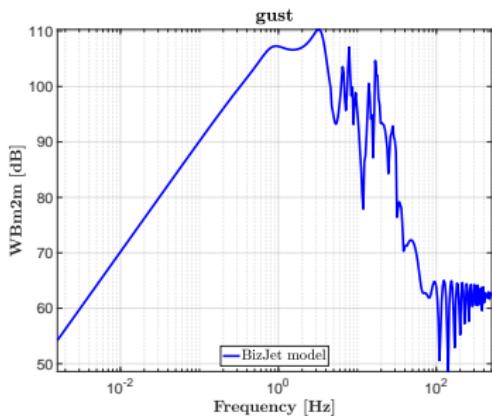
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Aircraft reduced order modelling

Models tailored to phenomena and problems specificities (the gust load case #1)

Large aeroelastic models

- ▶ ≈ 300 state variables
- ▶ Irrational and polynomial
- ▶ Not suitable for analysis and control



$$\mathbf{H}(s) = C(sE - A_0 - A_1 e^{\tau_1 s} - A_2 e^{\tau_2 s})^{-1} B$$

$$\mathbf{H}(s) = \mathbf{H}_{\text{ODE irrat.}} + \mathbf{P}_{\text{Poly.}}$$

Aircraft reduced order modelling

Models tailored to phenomena and problems specificities (the gust load case #1)

Large aeroelastic models

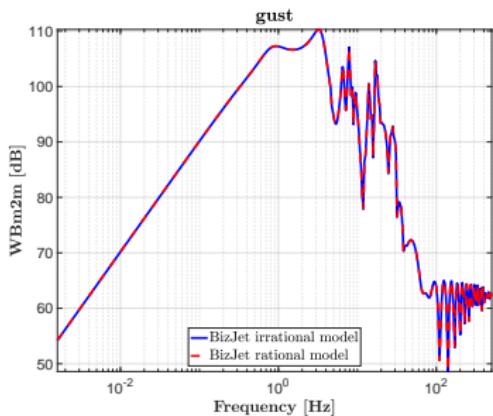
- ▶ ≈ 300 state variables
- ▶ Irrational and polynomial
- ▶ Not suitable for analysis and control



Model Approximation

Small and accurate models

- ▶ ≈ 400 state variables
- ▶ Rational (and polynomial)
- ▶ Suitable for analysis but not for control



$$\mathbf{H}(s) = C(sE - A_0 - A_1 e^{\tau_1 s} - A_2 e^{\tau_2 s})^{-1} B$$

$$\mathbf{H}(s) = \mathbf{H}_{\text{ODE irrat.}} + \mathbf{P}_{\text{Poly.}}$$

↓ (Rational approximation)

$$\mathbf{H}(s) = C(sE - A)^{-1} B.$$

Aircraft reduced order modelling

Models tailored to phenomena and problems specificities (the gust load case #1)

Large aeroelastic models

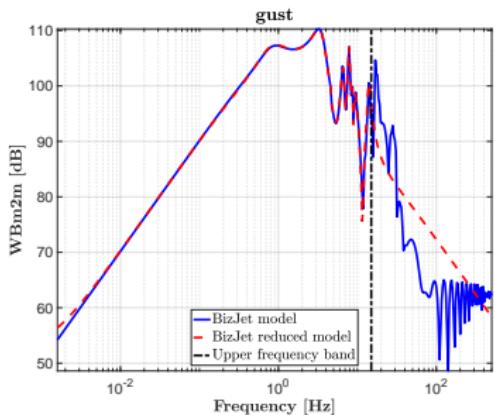
- ▶ ≈ 300 state variables
- ▶ Irrational and polynomial
- ▶ Not suitable for analysis and control



Small and accurate models

- ▶ ≈ 30 state variables
- ▶ Rational (over frequency range)
- ▶ Suitable for analysis and control

Model Order Reduction



$$\mathbf{H}(s) = C(sE - A_0 - A_1 e^{\tau_1 s} - A_2 e^{\tau_2 s})^{-1} B$$

$$\mathbf{H}(s) = \mathbf{H}_{\text{ODE irrat.}} + \mathbf{P}_{\text{Poly.}}$$

↓ (Rational approximation)

$$\mathbf{H}(s) = C(sE - A)^{-1} B.$$

↓ (Order reduction)

$$\hat{\mathbf{H}}(s) = \hat{C}(s\hat{E} - \hat{A})^{-1} \hat{B}.$$

Aircraft reduced order modelling

Focus on model approximation and reduction

Let us consider \mathbf{H} , a n_u inputs, n_y outputs linear dynamical system (evaluation) described by the complex-valued function from \mathbf{u} to \mathbf{y} , of order n^1

$$\mathbf{H} : \mathbb{C} \rightarrow \mathbb{C}^{n_y \times n_u}$$

model approximation consists in finding $\hat{\mathbf{H}}$, mapping \mathbf{u} to $\hat{\mathbf{y}}$, of order $r \ll n$

$$\hat{\mathbf{H}} : \mathbb{C} \rightarrow \mathbb{C}^{n_y \times n_u},$$

that well reproduces the input-output behaviour

¹ n large or ∞ , or just data.

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that well reproduces the input-output behaviour and equipped with realisation e.g.

$$\hat{\mathbf{H}} = \hat{\mathbf{C}} \left(s\hat{\mathbf{E}} - \hat{\mathbf{A}} \right)^{-1} \hat{\mathbf{B}}$$

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$$\hat{\mathbf{H}} = \hat{\mathbf{C}} \left(s\hat{\mathbf{E}} - \hat{\mathbf{A}} \right)^{-1} \hat{\mathbf{B}}$$

"Well reproduce..."? : $\hat{\mathbf{H}}$ is a "good" approximation of \mathbf{H} if for any driving $\mathbf{u}(t)$, $(\mathbf{H} - \hat{\mathbf{H}})\mathbf{u}(t) = \mathbf{y}(t) - \hat{\mathbf{y}}(t)$ is "small"

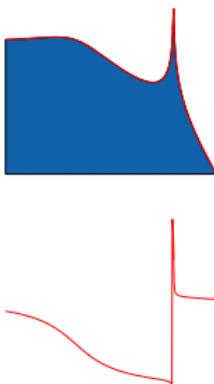
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Aircraft reduced order modelling

Focus on model approximation and reduction (\mathcal{H}_2 problem)

\mathcal{H}_2 model approximation

$$\hat{\mathbf{H}} := \arg \min_{\substack{\mathbf{G} \in \mathcal{H}_2 \\ \text{rank}(\mathbf{G}) = r \ll n}} \|\mathbf{H} - \mathbf{G}\|_{\mathcal{H}_2}$$



\mathcal{H}_2 -norm of $\mathbf{H} - \hat{\mathbf{H}}$

Frequency-domain interpretation:

$$\|\mathbf{H}\|_{\mathcal{H}_2}^2 := \frac{1}{2\pi} \int_{-\infty}^{\infty} \text{tr}(\overline{\mathbf{H}(\nu)} \mathbf{H}^T(\nu)) d\nu$$

Time-domain interpretation:

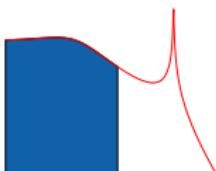
$$\|\mathbf{y}(t) - \hat{\mathbf{y}}(t)\|_{L_\infty} \leq \|\mathbf{H} - \hat{\mathbf{H}}\|_{\mathcal{H}_2} \|\mathbf{u}(t)\|_{L_2}$$

Aircraft reduced order modelling

Focus on model approximation and reduction ($\mathcal{H}_{2,\Omega}$ problem)

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

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



$\mathcal{H}_{2,\Omega}$ -norm of $\mathbf{H} - \hat{\mathbf{H}}$

(finite support) Frequency-domain interpretation:

$$\|\mathbf{H}\|_{\mathcal{H}_{2,\Omega}}^2 := \frac{1}{\pi} \int_{\Omega} \text{tr}(\overline{\mathbf{H}(\nu)} \mathbf{H}^T(\nu)) d\nu$$



P. Vuillemin, C. P-V. and D. Alazard, " [\$\mathcal{H}_2\$ optimal and frequency-limited approximation methods for large-scale LTI dynamical systems](#)", in Proc. of the IFAC SSSC, Grenoble, France, 2013, pp. 719-724.

Aircraft reduced order modelling

Focus on model approximation and reduction by interpolation (\mathcal{H}_2)

If $\hat{\mathbf{H}}$ is solution of the \mathcal{H}_2 approximation problem, then

$$\begin{aligned}\mathbf{H}(-\hat{\lambda}_l) &= \hat{\mathbf{H}}(-\hat{\lambda}_l) \\ \mathbf{H}'(-\hat{\lambda}_l) &= \hat{\mathbf{H}}'(-\hat{\lambda}_l)\end{aligned}$$

where $\hat{\lambda}_l := \text{eig}(\hat{E}, \hat{A})^a$ and

$$\mathbf{H}(s) = \sum_{j=1}^n \frac{\phi_j}{s - \lambda_j}$$

By considering

$$\begin{aligned}\hat{\mathbf{H}}(s) &= \sum_{k=1}^r \frac{\hat{\phi}_k}{s - \hat{\lambda}_k} \\ &= \hat{\mathbf{c}} (s\hat{E} - \hat{A})^{-1} \hat{\mathbf{b}}\end{aligned}$$

^a $l = 1, \dots, r$

Stands as the central result \Rightarrow but how to determine $\hat{\lambda}_l$?

-
-  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, 2008, Vol. 30(2), 2008, pp. 609-638.
 -  K. A. Gallivan, A. Vanderope, and P. Van-Dooren, "Model reduction of MIMO systems via tangential interpolation", SIAM Journal of Matrix Analysis and Application, 2004, Vol. 26(2), pp. 328-349.

Aircraft reduced order modelling

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$$\mathbf{H}(s) = \sum_{j=1}^n \frac{\phi_j}{s - \lambda_j}$$

If a realisation of \mathbf{H} exists,

$$\mathbf{H}(s) = \mathbf{c} (sE - A)^{-1} \mathbf{b}$$

^a $l = 1, \dots, r$

Aircraft reduced order modelling

Focus on model approximation and reduction by interpolation (\mathcal{H}_2)

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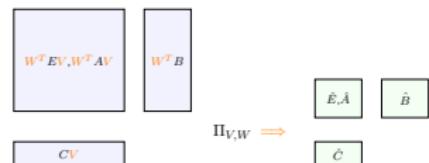
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If a realisation of \mathbf{H} exists,

$$\mathbf{H}(s) = \mathbf{c} (sE - A)^{-1} \mathbf{b}$$

↓ Petrov-Galerkin (interp.)

$$\hat{\mathbf{H}}(s) = \mathbf{c} \mathbf{V} \left(s \mathbf{W}^T E \mathbf{V} - \mathbf{W}^T A \mathbf{V} \right)^{-1} \mathbf{W}^T \mathbf{b}$$



For some $\sigma_l \in \mathbb{C}$ and $\mathbf{W}^T \mathbf{V} = I_r$

$$\begin{aligned}\text{span}(\mathbf{V}) &\subseteq \left[(\sigma_1 E - A)^{-1} \mathbf{b}, \dots, (\sigma_r E - A)^{-1} \mathbf{b} \right] \\ \text{span}(\mathbf{W}) &\subseteq \left[(\sigma_1 E - A)^{-T} \mathbf{c}^T, \dots, (\sigma_r E - A)^{-T} \mathbf{c}^T \right]\end{aligned}$$

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where $\hat{\lambda}_l := \text{eig}(\hat{E}, \hat{A})^a$ and

$$\mathbf{H}(s) = \sum_{j=1}^n \frac{\phi_j}{s - \lambda_j}$$

^a $l = 1, \dots, r$

For $\sigma_l \leftarrow -\hat{\lambda}_l$ and $\mathbf{W}^T \mathbf{V} = I_r$

$$\begin{aligned}\text{span}(\mathbf{V}) &\subseteq \left[(-\hat{\lambda}_1 E - A)^{-1} \mathbf{b}, \dots, (-\hat{\lambda}_r E - A)^{-1} \mathbf{b} \right] \\ \text{span}(\mathbf{W}) &\subseteq \left[(-\hat{\lambda}_1 E - A)^{-T} \mathbf{c}^T, \dots, (-\hat{\lambda}_r E - A)^{-T} \mathbf{c}^T \right]\end{aligned}$$

 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, 2008, Vol. 30(2), pp. 609-638.

If a realisation of \mathbf{H} exists,

$$\mathbf{H}(s) = \mathbf{c} (sE - A)^{-1} \mathbf{b}$$

↓ Petrov-Galerkin (interp.)

$$\hat{\mathbf{H}}(s) = \mathbf{c} \mathbf{V} \left(s \mathbf{W}^T E \mathbf{V} - \mathbf{W}^T A \mathbf{V} \right)^{-1} \mathbf{W}^T \mathbf{b}$$

↓ IRKA or TF-IRKA

$\hat{\mathbf{H}}$ solves the \mathcal{H}_2 model approx. pb.

Aircraft reduced order modelling

Focus on model approximation and reduction by interpolation (rational interpolation)

Given model \mathbf{H} , seek $\hat{\mathbf{H}}$
s.t.

$$\begin{aligned}\hat{\mathbf{H}}(\mu_j) &= \mathbf{H}(\mu_j) \\ \hat{\mathbf{H}}(\lambda_i) &= \mathbf{H}(\lambda_i)\end{aligned}$$

$i = 1, \dots, k; j = 1, \dots, q.$



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.

Aircraft reduced order modelling

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$i = 1, \dots, k; j = 1, \dots, q.$

$$\mathbb{L} = \begin{bmatrix} \frac{\mathbf{H}(\mu_1) - \mathbf{H}(\lambda_1)}{\mu_1 - \lambda_1} & \dots & \frac{\mathbf{H}(\mu_1) - \mathbf{H}(\lambda_k)}{\mu_1 - \lambda_k} \\ \vdots & \ddots & \vdots \\ \frac{\mathbf{H}(\mu_q) - \mathbf{H}(\lambda_1)}{\mu_q - \lambda_1} & \dots & \frac{\mathbf{H}(\mu_q) - \mathbf{H}(\lambda_k)}{\mu_q - \lambda_k} \end{bmatrix}$$

$$\mathbb{L}_\sigma = \begin{bmatrix} \frac{\mu_1 \mathbf{H}(\mu_1) - \mathbf{H}(\lambda_1) \lambda_1}{\mu_1 - \lambda_1} & \dots & \frac{\mu_1 \mathbf{H}(\mu_1) - \mathbf{H}(\lambda_k) \lambda_k}{\mu_1 - \lambda_k} \\ \vdots & \ddots & \vdots \\ \frac{\mu_q \mathbf{H}(\mu_q) - \mathbf{H}(\lambda_1) \lambda_1}{\mu_q - \lambda_1} & \dots & \frac{\mu_q \mathbf{H}(\mu_q) - \mathbf{H}(\lambda_k) \lambda_k}{\mu_q - \lambda_k} \end{bmatrix}$$

$$\mathbf{W} = \begin{bmatrix} \mathbf{H}(\lambda_1) & \dots & \mathbf{H}(\lambda_k) \end{bmatrix}$$

$$\mathbf{V}^T = \begin{bmatrix} \mathbf{H}(\mu_1) & \dots & \mathbf{H}(\mu_q) \end{bmatrix}$$

$$\hat{\mathbf{H}}(s) = \mathbf{W}(-s\mathbb{L} + \mathbb{L}_\sigma)^{-1}\mathbf{V} \Rightarrow \text{Rational interpolation}$$



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.

Aircraft reduced order modelling

Focus on model approximation and reduction by interpolation (Hermite interpolation)

Given model \mathbf{H} , seek $\hat{\mathbf{H}}$
s.t.

$$\begin{aligned}\hat{\mathbf{H}}(\sigma_i) &= \mathbf{H}(\sigma_i) \\ \hat{\mathbf{H}}'(\sigma_i) &= \mathbf{H}'(\sigma_i)\end{aligned}$$

$$i = 1, \dots, r.$$

$$\mathbb{L} = \begin{bmatrix} \mathbf{H}'(\sigma_1) & \dots & \frac{\mathbf{H}(\sigma_1) - \mathbf{H}(\sigma_r)}{\sigma_1 - \sigma_r} \\ \vdots & \ddots & \vdots \\ \frac{\mathbf{H}(\sigma_r) - \mathbf{H}(\sigma_1)}{\sigma_r - \sigma_1} & \dots & \mathbf{H}'(\sigma_r) \end{bmatrix}$$

$$\mathbb{L}_\sigma = \begin{bmatrix} (s\mathbf{H}(s))'_{s=\sigma_1} & \dots & \frac{\sigma_1 \mathbf{H}(\sigma_1) - \sigma_r \mathbf{H}(\sigma_r)}{\sigma_1 - \sigma_r} \\ \vdots & \ddots & \vdots \\ \frac{\sigma_r \mathbf{H}(\sigma_r) - \sigma_1 \mathbf{H}(\sigma_1)}{\sigma_r - \sigma_1} & \dots & (s\mathbf{H}(s))'_{s=\sigma_r} \end{bmatrix}$$

$$\mathbf{W} = \begin{bmatrix} \mathbf{H}(\sigma_1) & \dots & \mathbf{H}(\sigma_r) \end{bmatrix}$$

$$\mathbf{V}^T = \begin{bmatrix} \mathbf{H}(\sigma_1) & \dots & \mathbf{H}(\sigma_r) \end{bmatrix}$$

$$\hat{\mathbf{H}}(s) = \mathbf{W}(-s\mathbb{L} + \mathbb{L}_\sigma)^{-1}\mathbf{V} \Rightarrow \text{Hermite interpolation}$$



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.

Aircraft reduced order modelling

Back to the gust load models #1

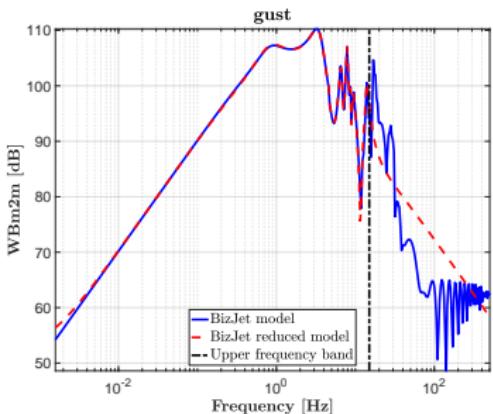
Large aeroelastic models

- ▶ ≈ 300 state variables
- ▶ Irrational and polynomial
- ▶ Not suitable for analysis and control



Small and accurate models

- ▶ ≈ 30 state variables
- ▶ Rational (red over frequency range)
- ▶ Suitable for analysis and control



$$\mathbf{H}(s) = C(sE - A_0 - A_1 e^{\tau_1 s} - A_2 e^{\tau_2 s})^{-1} B$$

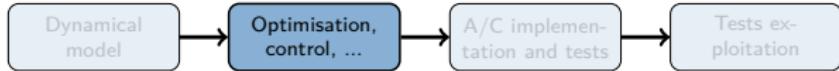
$$\mathbf{H}(s) = \mathbf{H}_{\text{ODE irrat.}} + \mathbf{P}_{\text{Poly.}}$$

⇓ (Rational approximation)

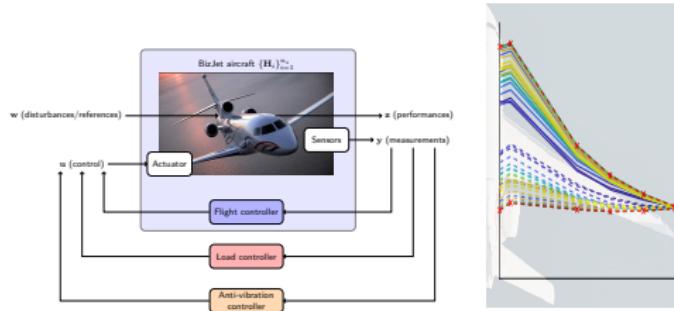
$$\mathbf{H}(s) = C(sE - A)^{-1} B.$$

⇓ (Reduction)

$$\hat{\mathbf{H}}(s) = \hat{C}(s\hat{E} - \hat{A})^{-1} \hat{B}.$$

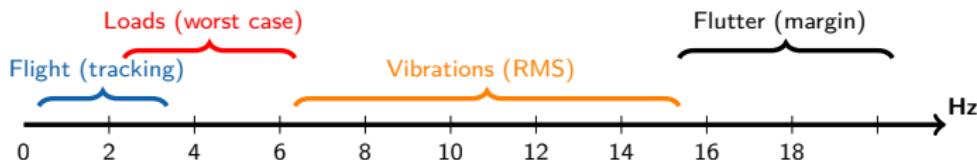


1. Aircraft gust load and vibration control problems
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Aircraft feedback control design

Control objectives



#1 Gust load objective

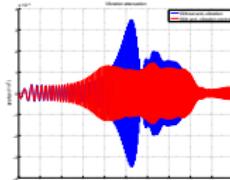
- ▶ Reduce the load enveloppe in response to a time-domain gust disturbance family
- ▶ Do not modify the flight controller performances



[Airbus image]

#2 Vibration attenuation objective

- ▶ Reduce the RMS vibrations around
- ▶ Do not modify the flight and load controllers performances

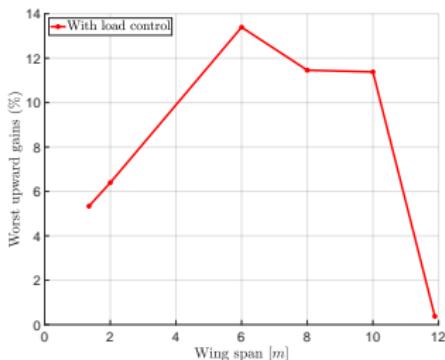


Aircraft feedback control design

Some results

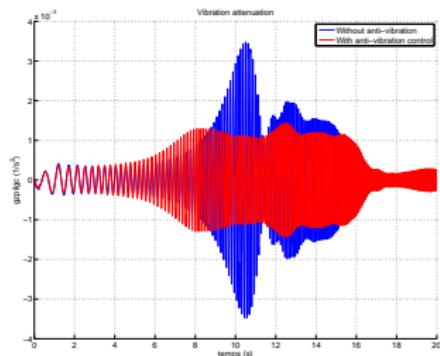
#1 Gust load control

From 5% to 14% of load attenuation



#2 Vibration control

Around 45% of vibration reduction



Notable reductions in loads and vibrations



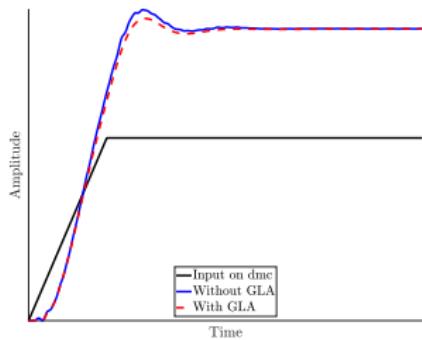
C. P-V., P. Vuillemin, O. Cantinaud and F. Sèze, "Interpolatory Methods for Business Jet Gust Load Alleviation Function: Modelling, Control Design and Numerical Validation", under submission.

Aircraft feedback control design

Some results

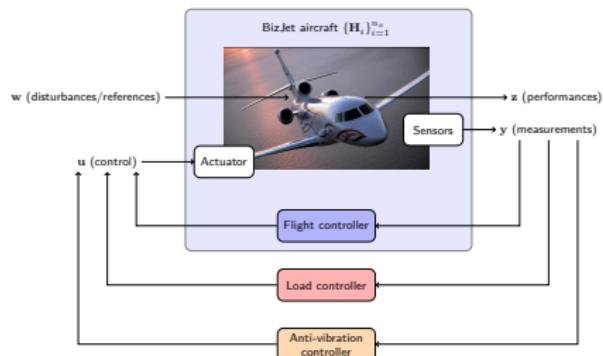
#1 Gust load control

From 5% to 14% of load attenuation



#2 Vibration control

Around 45% of vibration reduction



... are obtained without affecting the Flight control performances



C. P-V., P. Vuillemin, O. Cantinaud and F. Sèze, "Interpolatory Methods for Business Jet Gust Load Alleviation Function: Modelling, Control Design and Numerical Validation", under submission.

Aircraft feedback control design

Glimpse of the considered control design approach (step 1)

The control problem paradigm

$$\mathbf{K}^*(s) := \arg \min_{\mathbf{K} \in \mathcal{K}} \left\| \mathcal{F}_l(\mathbf{T}(s), \mathbf{K}(s)) \right\|_{\mathcal{H}_{\infty}}$$

- ▶ $\mathbf{T}(s) = \mathbf{W}_i(s)\hat{\mathbf{H}}(s)\mathbf{W}_o(s)$ is the generalised plant
- ▶ $\mathbf{K} \in \mathcal{K} \subseteq \mathcal{H}_{\infty}$ is the controller dynamical operator to be found.

$$\mathbf{K}(s) := \mathcal{F}_u \left(\textcolor{red}{K}, \frac{1}{s} I_{n_K} \right),$$

Provide the "best" achievable performances

$$\textcolor{red}{K} = \begin{bmatrix} A_K & B_u \\ C_y & D_{yu} \end{bmatrix}$$

... is any matrix

 P. Apkarian and D. Noll, "*Nonsmooth \mathcal{H}_{∞} synthesis*", IEEE Transactions on Automatic Control, 2006, Vol. 51(1), pp. 71-86..

Aircraft feedback control design

Glimpse of the considered control design approach (step 2)

Handle controllers "intersection" (given $\mathbf{K}^*(s)$)

$$\mathbf{K}_*(s) := \arg \min_{\mathbf{K} \in \mathcal{K}} \left| \left| \mathcal{F}_l(\mathbf{T}(s), \mathbf{K}^*(s)) - \mathcal{F}_l(\mathbf{T}(s), \mathbf{K}(s)) \right| \right|_{\mathcal{H}_\infty}$$

- $\mathcal{F}_l(\mathbf{T}(s), \mathbf{K}^*(s))$ is the closed-loop obtained in step 1
- $\mathbf{K} \in \mathcal{K} \subseteq \mathcal{H}_\infty$ is the controller dynamical operator to be found.

$$\mathbf{K}(s) := \mathcal{F}_u \left(\mathbf{K}, \frac{1}{s} I_{n_K} \right),$$

$$\mathbf{K} = \begin{bmatrix} A_K & B_u \\ C_y & D_{yu} \end{bmatrix}$$

... is NOT any matrix...

$$A_K = \begin{pmatrix} \times & \times & \dots & \dots & 0 \\ \times & \times & 0 & \dots & 0 \\ 0 & \times & \times & 0 & \dots \\ \vdots & & \ddots & \ddots & 0 \\ \vdots & 0 & \ddots & \ddots & 0 \\ 0 & \dots & 0 & \times & 0 \\ 0 & \dots & 0 & \times & \times \\ 0 & \dots & & 0 & \times \end{pmatrix}$$

 P. Apkarian and D. Noll, "Nonsmooth \mathcal{H}_∞ synthesis", IEEE Transactions on Automatic Control, 2006, Vol. 51(1), pp. 71-86..

Aircraft feedback control design

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$$\mathbf{K} = \begin{bmatrix} A_K & B_u \\ C_y & D_{yu} \end{bmatrix}$$

... is NOT any matrix...

$$A_K = \begin{pmatrix} x & x & \dots & \dots & 0 \\ x & x & 0 & \dots & 0 \\ 0 & x & x & 0 & \dots \\ \vdots & & \ddots & \ddots & \vdots \\ \vdots & 0 & \ddots & \ddots & 0 \\ 0 & \dots & 0 & \times & 0 \\ 0 & \dots & 0 & \times & \times \\ 0 & \dots & 0 & \times & \times \end{pmatrix}$$

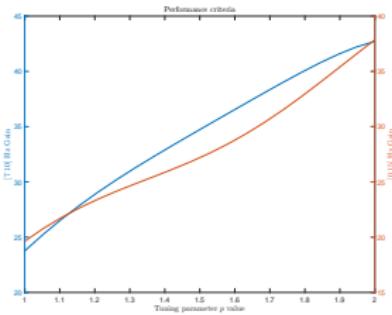
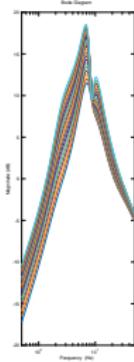
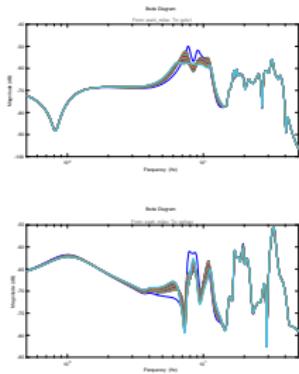
 P. Apkarian and D. Noll, "Nonsmooth \mathcal{H}_∞ synthesis", IEEE Transactions on Automatic Control, 2006, Vol. 51(1), pp. 71-86..

Aircraft feedback control design

Glimpse of the considered control design approach (go to parametric, for the vibration case #2)

The approach can be extended to parametric

Find $\mathbf{K}_*(s, \textcolor{orange}{p})$, a parametric law, that can be adjusted on-line.



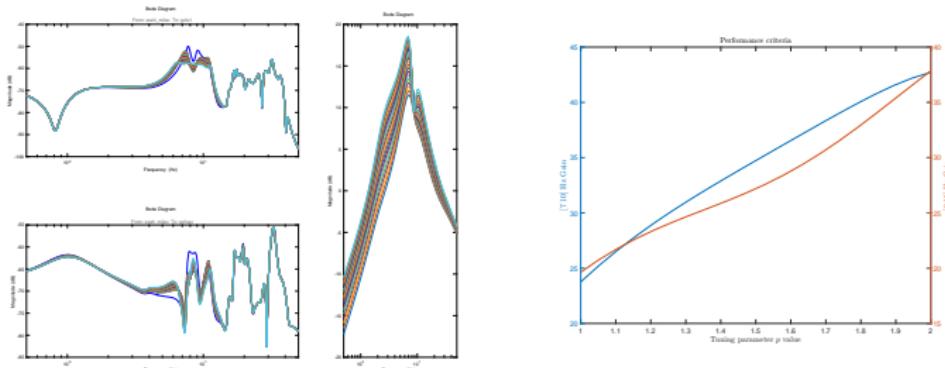
C. P-V., C. Leclercq and D. Sipp, "Structured linear fractional parametric controller \mathcal{H}_∞ design and its applications", in Proceedings of the European Control Conference (ECC), Limassol, Cyprus, June, 2018, pp. 2629-2634.

Aircraft feedback control design

Glimpse of the considered control design approach (go to parametric, for the vibration case #2)

The approach can be extended to parametric

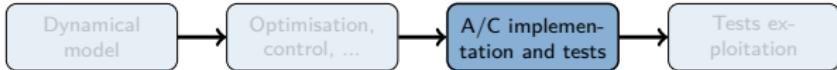
Find $\mathbf{K}_*(s, p)$, a parametric law, that can be adjusted on-line.



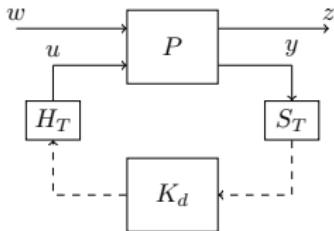
As a function of p , the vibration attenuation is better and better, but requires more actuator activity (used for testing)



C. P-V., C. Leclercq and D. Sipp, "Structured linear fractional parametric controller \mathcal{H}_∞ design and its applications", in Proceedings of the European Control Conference (ECC), Limassol, Cyprus, June, 2018, pp. 2629-2634.

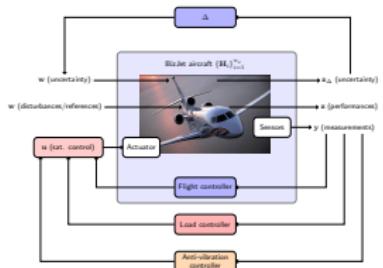
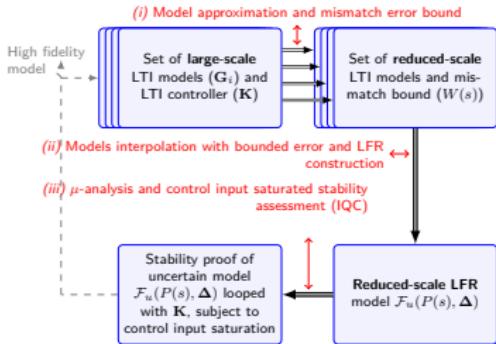


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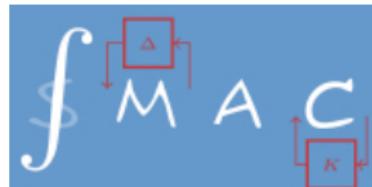
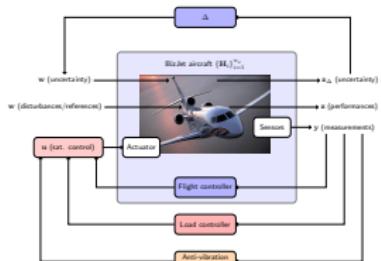
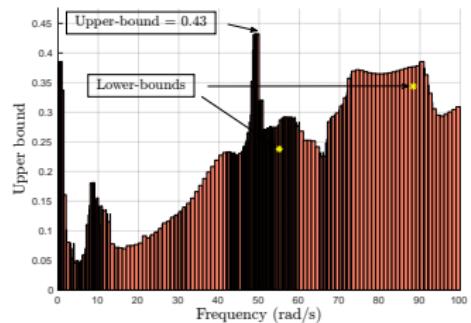
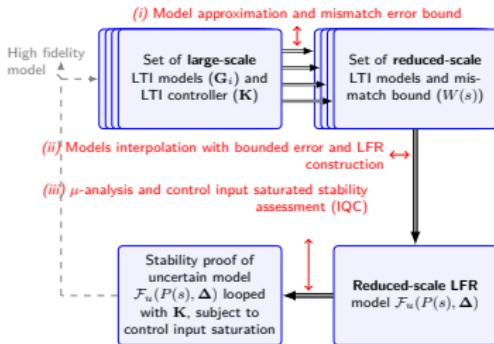
Aircraft implementation and analysis

Uncertainty and actuator saturation analysis



Aircraft implementation and analysis

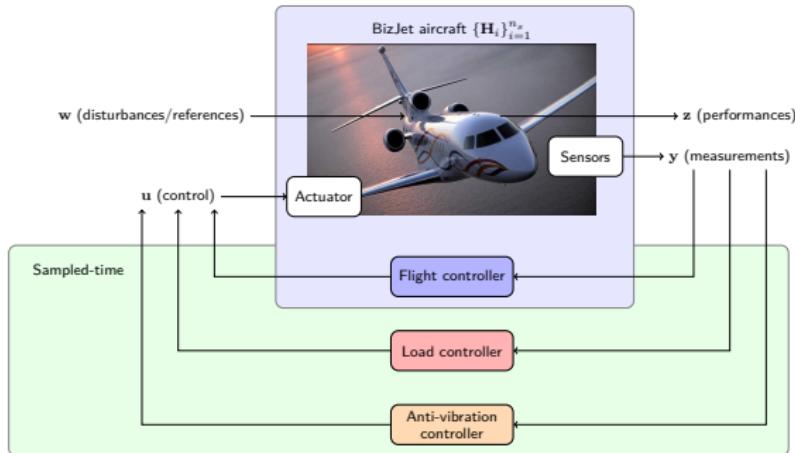
Uncertainty and actuator saturation analysis



<https://w3.onera.fr/smac/>

Aircraft implementation and analysis

Digital control design



$\mathbf{K}(s) \rightarrow \mathbf{K}_d(z)$ and implementation constrains

 P. Vuillemin and C. P-V., *"Discretisation of continuous-time linear dynamical model with the Loewner interpolation framework"*, initial version available as arXiv:1907.10956.

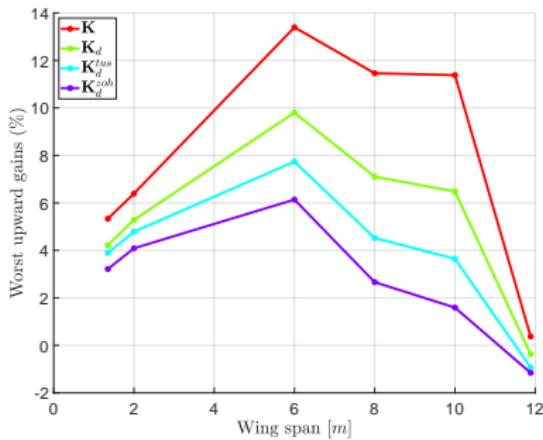
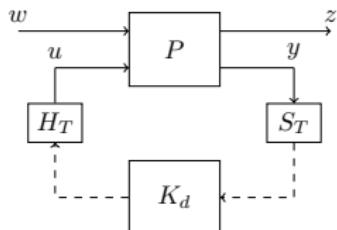
 C. P-V., P. Vuillemin, F. Sève and O. Cantinaud, *"Interpolatory Methods for Business Jet Gust Load Alleviation Function: Modelling, Control Design and Numerical Validation"*, under submission.

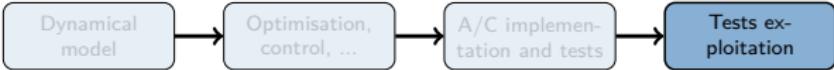
Aircraft implementation and analysis

Digital control design

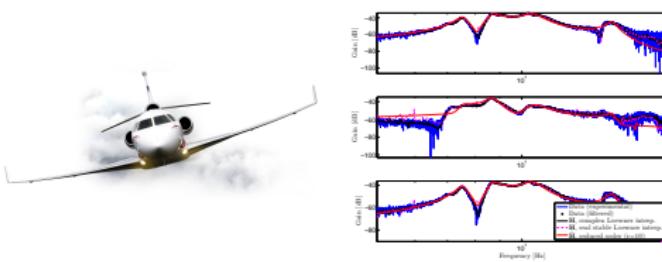
Control laws are discretised

LTI controller is discretised at fixed period with Tustin, ZOH and Loewner K_d .



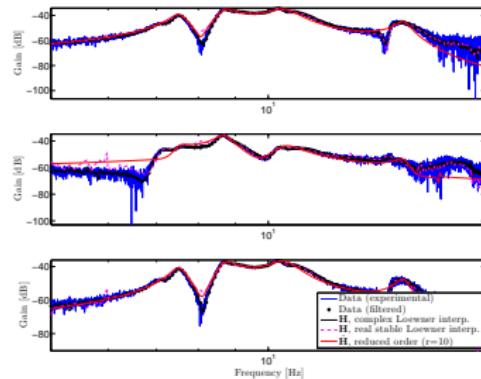


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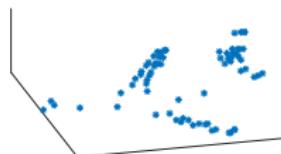


Aircraft experimental validation

Ground test validation by Dassault-Aviation, toward data-driven interpolation



- ▶ Ground test at Dassault-Aviation
- ▶ Measurements at #100 points
- ▶ Application of **data-driven interpolatory methods**



Aircraft experimental validation

Model approximation/reduction by interpolation (rational interpolation reminder)

Given model \mathbf{H} , seek $\hat{\mathbf{H}}$
s.t.

$$\begin{aligned}\hat{\mathbf{H}}(\mu_j) &= \mathbf{H}(\mu_j) \\ \hat{\mathbf{H}}(\lambda_i) &= \mathbf{H}(\lambda_i)\end{aligned}$$

$i = 1, \dots, k; j = 1, \dots, q.$

$$\mathbb{L} = \begin{bmatrix} \frac{\mathbf{H}(\mu_1) - \mathbf{H}(\lambda_1)}{\mu_1 - \lambda_1} & \dots & \frac{\mathbf{H}(\mu_1) - \mathbf{H}(\lambda_k)}{\mu_1 - \lambda_k} \\ \vdots & \ddots & \vdots \\ \frac{\mathbf{H}(\mu_q) - \mathbf{H}(\lambda_1)}{\mu_q - \lambda_1} & \dots & \frac{\mathbf{H}(\mu_q) - \mathbf{H}(\lambda_k)}{\mu_q - \lambda_k} \end{bmatrix}$$

$$\mathbb{L}_\sigma = \begin{bmatrix} \frac{\mu_1 \mathbf{H}(\mu_1) - \mathbf{H}(\lambda_1) \lambda_1}{\mu_1 - \lambda_1} & \dots & \frac{\mu_1 \mathbf{H}(\mu_1) - \mathbf{H}(\lambda_k) \lambda_k}{\mu_1 - \lambda_k} \\ \vdots & \ddots & \vdots \\ \frac{\mu_q \mathbf{H}(\mu_q) - \mathbf{H}(\lambda_1) \lambda_1}{\mu_q - \lambda_1} & \dots & \frac{\mu_q \mathbf{H}(\mu_q) - \mathbf{H}(\lambda_k) \lambda_k}{\mu_q - \lambda_k} \end{bmatrix}$$

$$\mathbf{W} = \begin{bmatrix} \mathbf{H}(\lambda_1) & \dots & \mathbf{H}(\lambda_k) \end{bmatrix}$$

$$\mathbf{V}^T = \begin{bmatrix} \mathbf{H}(\mu_1) & \dots & \mathbf{H}(\mu_q) \end{bmatrix}$$

$$\hat{\mathbf{H}}(s) = \mathbf{W}(-s\mathbb{L} + \mathbb{L}_\sigma)^{-1}\mathbf{V} \Rightarrow \text{Rational interpolation}$$



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.

Aircraft experimental validation

Model approximation/reduction by interpolation (data-driven interpolation)

Given $\{\mu_j, \mathbf{v}_j\}$, $\{\lambda_i, \mathbf{w}_i\}$
data, seek $\hat{\mathbf{H}}$, s.t.

$$\begin{aligned}\hat{\mathbf{H}}(\mu_j) &= \mathbf{v}_j \\ \hat{\mathbf{H}}(\lambda_i) &= \mathbf{w}_i\end{aligned}$$

$i = 1, \dots, k; j = 1, \dots, q.$

$$\mathbb{L} = \begin{bmatrix} \frac{\mathbf{H}(\mu_1) - \mathbf{H}(\lambda_1)}{\mu_1 - \lambda_1} & \dots & \frac{\mathbf{H}(\mu_1) - \mathbf{H}(\lambda_k)}{\mu_1 - \lambda_k} \\ \vdots & \ddots & \vdots \\ \frac{\mathbf{H}(\mu_q) - \mathbf{H}(\lambda_1)}{\mu_q - \lambda_1} & \dots & \frac{\mathbf{H}(\mu_q) - \mathbf{H}(\lambda_k)}{\mu_q - \lambda_k} \end{bmatrix}$$

$$\mathbb{L}_\sigma = \begin{bmatrix} \frac{\mu_1 \mathbf{H}(\mu_1) - \mathbf{H}(\lambda_1) \lambda_1}{\mu_1 - \lambda_1} & \dots & \frac{\mu_1 \mathbf{H}(\mu_1) - \mathbf{H}(\lambda_k) \lambda_k}{\mu_1 - \lambda_k} \\ \vdots & \ddots & \vdots \\ \frac{\mu_q \mathbf{H}(\mu_q) - \mathbf{H}(\lambda_1) \lambda_1}{\mu_q - \lambda_1} & \dots & \frac{\mu_q \mathbf{H}(\mu_q) - \mathbf{H}(\lambda_k) \lambda_k}{\mu_q - \lambda_k} \end{bmatrix}$$

$$\mathbf{W} = \left[\begin{array}{ccc} \mathbf{H}(\lambda_1) & \dots & \mathbf{H}(\lambda_k) \end{array} \right]$$

$$\mathbf{V}^T = \left[\begin{array}{ccc} \mathbf{H}(\mu_1) & \dots & \mathbf{H}(\mu_q) \end{array} \right]$$



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.

Aircraft experimental validation

Model approximation/reduction by interpolation (data-driven interpolation)

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$$\begin{aligned}\hat{\mathbf{H}}(\mu_j) &= \mathbf{v}_j \\ \hat{\mathbf{H}}(\lambda_i) &= \mathbf{w}_i\end{aligned}$$

$i = 1, \dots, k; j = 1, \dots, q$.

$$\mathbb{L} = \begin{bmatrix} \frac{\mathbf{v}_1 - \mathbf{w}_1}{\mu_1 - \lambda_1} & \cdots & \frac{\mathbf{v}_1 - \mathbf{w}_k}{\mu_1 - \lambda_k} \\ \vdots & \ddots & \vdots \\ \frac{\mathbf{v}_q - \mathbf{w}_1}{\mu_q - \lambda_1} & \cdots & \frac{\mathbf{v}_q - \mathbf{w}_k}{\mu_q - \lambda_k} \end{bmatrix}$$

$$\mathbb{L}_\sigma = \begin{bmatrix} \frac{\mu_1 \mathbf{v}_1 - \mathbf{w}_1 \lambda_1}{\mu_1 - \lambda_1} & \cdots & \frac{\mu_1 \mathbf{v}_1 - \mathbf{w}_k \lambda_k}{\mu_1 - \lambda_k} \\ \vdots & \ddots & \vdots \\ \frac{\mu_q \mathbf{v}_q - \mathbf{w}_1 \lambda_1}{\mu_q - \lambda_1} & \cdots & \frac{\mu_q \mathbf{v}_q - \mathbf{w}_k \lambda_k}{\mu_q - \lambda_k} \end{bmatrix}$$

$$\mathbf{W} = \begin{bmatrix} \mathbf{w}_1 & \cdots & \mathbf{w}_k \end{bmatrix}$$

$$\mathbf{V}^T = \begin{bmatrix} \mathbf{v}_1 & \cdots & \mathbf{v}_q \end{bmatrix}$$

$$\hat{\mathbf{H}}(s) = \mathbf{W}(-s\mathbb{L} + \mathbb{L}_\sigma)^{-1}\mathbf{V} \Rightarrow \text{Rational interpolation (data driven)}$$



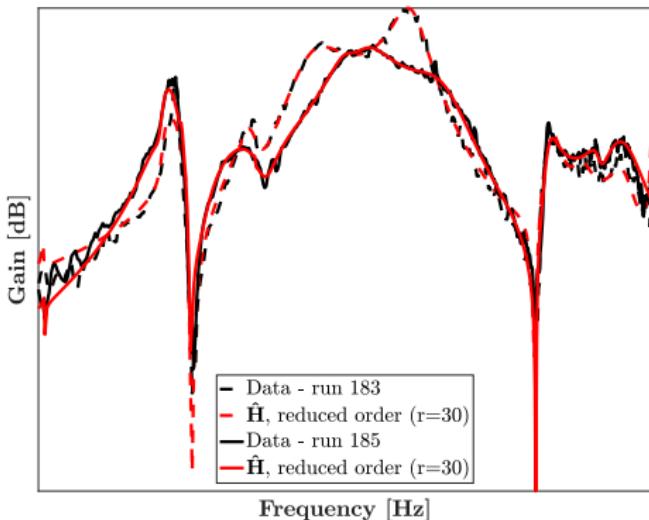
A.J. Mayo and A.C. Antoulas, "A framework for the solution of the generalized realization problem", Linear Algebra and its Applications, 425(2-3), 2007, pp 634-662.

Aircraft experimental validation

Ground vibration test interpolatory results

Aircraft experimental validation

Ground vibration test interpolatory results with control

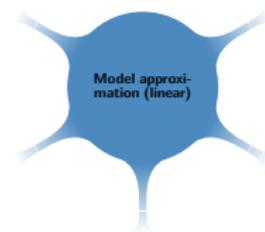


Data vs. **Model** & open-loop (dashed) vs. closed-loop (solid)
Vibration attenuation validated in GVT... then in flight test \Rightarrow Video

Conclusions

What to keep in mind...

Importance of dynamical data science within industry
Illustration on a Dassault Aviation Business Jet aircraft



C. P-V., "[Large-scale dynamical model approximation and its applications](#)", HDR thesis, INP Toulouse, Université de Toulouse, ONERA, July 2019 (on-line available).

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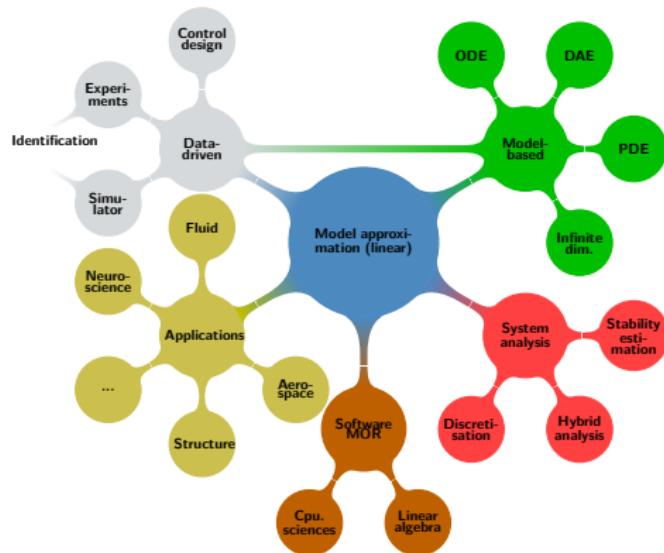


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Conclusions

... a versatile tool

Interpolation-based methods are remarkably versatile and indicated for

- ▶ model reduction and approximation,
- ▶ but also complex function analysis ...

→ direct impact in aircraft engineers life, but not only

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... a versatile tool

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- ▶ MOR Toolbox integrated tool at
<http://mordigitalsystems.fr/>
- ▶ More references at
<https://sites.google.com/site/charlespoussotvassal/>



“Merge data and physics using computational sciences and engineering, for enhanced dynamical systems experience”

Conclusions

Acknowledgements and on-going projects

Onera research collaborators

- ▶ P. Vuillemin [approx.]
- ▶ J-M. Biannic [uncertainty]
- ▶ E. Coustols [cs1 & 2]
- ▶ O. Kocan [learning control]

Dassault-Aviation collaborators

- ▶ Control team [control]
- ▶ Aeroelastic team [aeroelasticity]
- ▶ Test team [ground/flight tests]

Research collaborators

- ▶ I. Pontes Duff [approx.]
at MPI Magdebourg
- ▶ P. Kergus [data-driven]
at LTH control dpt.

Organisers

- ▶ D. Theilliol
- ▶ K. Valavanis
- ▶ J-P. Georges



Civil aircraft reduced order modelling, control and validation: Numerical and experimental challenges

... how much dynamical data science is a strong ally?

Charles Poussot-Vassal

Friday September 18, 2020

28th IEEE Mediterranean Conference on Control and Automation Conference

(e-) St Raphael, France

FrPL3 Plenary Session, ROOM SR1

