



# Quadrature-based diffusive representation of the fractional derivative with applications

in aeroacoustics and eigenvalue methods for stability

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# Outline

- 1 **Introduction**
  - Introduction
- 2 Quadrature-based discretization of diffusive representations
- 3 Numerical comparisons and applications
- 4 Conclusion

# Motivation: fractional delay systems in aeroacoustics

**Context:** Noise regulations  $\Rightarrow$  research into sound absorption.

## Modeling of locally-reacting sound absorbing material

Passive LTI system:

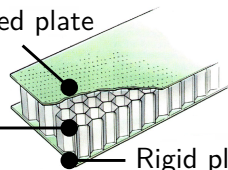
$$p(t, x) = [z \star_t \mathbf{u} \cdot \mathbf{n}(\cdot, x)](t)$$

with kernel  $z \in \mathcal{D}'_+(\mathbb{R}) \cap \mathcal{S}'(\mathbb{R})$ .

Perforated plate

Cavity

Rigid plate



**Key components** of  $\hat{z}$ : (Monteghetti et al. 2016)

$$\hat{z}(s) = 1 + \hat{h}(s) + e^{-sT}$$

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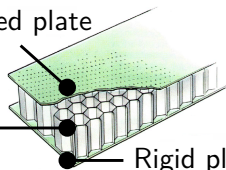
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**Key components** of  $z$ : (Monteghetti et al. 2016)

$$\hat{z}(s) = 1 + \hat{h}(s) + e^{-s\tau}$$

$\Rightarrow$  Boundary condition of a PDE on  $(p, u)$ .

$\Rightarrow$  Spatial discretization yields memory delay equation ( $x \in \mathbb{R}^n$ ):

$$M \cdot \dot{x}(t) + K \cdot x(t) = F_1 \cdot h \star x(t) + F_2 \cdot x(t - \tau).$$

**Objective:** discretization of a time-local representation of  $h \star x(t)$ .

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  - Diffusive representation
  - Quadrature method
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# Background on diffusive representation: definition

Definition (Desch et al. 1988; Staffans 1994; Mainardi 1997; Montseny 1998)

A causal kernel  $h \in L^1_{\text{loc}}([0, \infty))$  is *standard diffusive* if

$$h(t) = \int_0^\infty e^{-\xi t} \mu(\xi) d\xi, \quad \hat{h}(s) = \int_0^\infty \frac{1}{s + \xi} \mu(\xi) d\xi,$$

with weight  $\mu \in \mathcal{C}((0, \infty))$ .

**Computational interest.** Time-local computation of convolution through

$$h \star u(t) = \int_0^\infty \varphi(t, \xi) \mu(\xi) d\xi$$

with state  $\varphi(t, \cdot) \in L^2(0, \infty; \mu(\xi)d\xi)$  that obeys

$$\begin{cases} \partial_t \varphi(t, \xi) = -\xi \varphi(t, \xi) + u(t) \\ \varphi(0, \cdot) = 0. \end{cases}$$

# Background on diffusive representation: examples

Let  $\alpha \in (0, 1)$ .

Riemann-Liouville integral and Caputo derivative of order  $\alpha$

$$I^\alpha u := Y_\alpha \star u$$

with  $u \in L^2$ .

$$d^\alpha u := I^{1-\alpha} \dot{u}$$

with  $u \in H^1$ .

Fractional integral (Montseny 1998)

$$I^\alpha u(t) = \int_0^\infty \varphi(t, \xi) \mu_\alpha(\xi) d\xi,$$

$$\mu_\alpha(\xi) = \frac{\sin(\alpha\pi)}{\pi\xi^\alpha}.$$

Similar for Caputo derivative (Lombard et al. 2016)

$$d^\alpha u(t) = \int_0^\infty \underbrace{[-\xi \varphi(t, \xi) + u(t)]}_{=\partial_t \varphi(t, \xi)} \mu_{1-\alpha}(\xi) d\xi,$$

$$\varphi(0, \xi) = \frac{u(0)}{\xi}.$$

Also: Bessel function  $J_0$ . (Maignon 1998, § 3.3)

Discretization?



# Discretization of diffusive representation: state of the art

$$h(t) = \int_0^\infty e^{-\xi t} \mu(\xi) d\xi \simeq h_{\text{num}}(t) := \sum_{n=1}^N \mu_n e^{-\xi_n t}$$

Optimization-based

Quadrature-based

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## Optimization-based

- Linear least squares (Garcia et al. 1998; Hélie et al. 2006)
- Nonlinear least squares (Lombard et al. 2016)

## Quadrature-based

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## Optimization-based

- Linear least squares (Garcia et al. 1998; Hélie et al. 2006)
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## Quadrature-based

- Truncation on  $(10^a, 10^b)$  with Gauss-Legendre and Curtis-Clenshaw (Baranowski 2017)
- Fast time stepping: split of  $Y_\alpha \star u(t)$  between “local” and “historical” parts (Haddar et al. 2010; Li 2010)
- Gauss-Laguerre on  $(0, \infty)$  (Yuan et al. 2002)
- Gauss-Jacobi on  $(-1, 1)$  (Diethelm 2008) (Birk et al. 2010)

# Quadrature-based discretization: definition

$$h(t) = \int_0^{\infty} e^{-\xi t} \mu(\xi) d\xi \simeq h_{\text{num}}(t) := \sum_{n=1}^N \mu_n e^{-\xi_n t}$$

**Strategy?** (Davis et al. 1984, Chap. 3) (Atkinson 1989, § 5.6) (Shampine 2008, § 4.2)

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**Strategy?** (Davis et al. 1984, Chap. 3) (Atkinson 1989, § 5.6) (Shampine 2008, § 4.2) Change of variables  $\xi = \Psi(v)$

$$h(t) = \int_{-1}^1 \mu(\Psi(v)) e^{-\Psi(v)t} \Psi'(v) dv \Rightarrow \begin{cases} \xi_n := \Psi(v_n) \\ \mu_n := w_n \Psi'(v_n) \mu(\xi_n) \end{cases}$$

with  $(v_n, w_n)$  Gauss-Legendre nodes.

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## Assumption

$$\mu \text{ is non-oscillating} \quad \text{and} \quad \mu(\xi) \underset{0}{=} \mathcal{O}\left(\frac{1}{\xi^\alpha}\right)$$

## Definition of $Q_{\beta, N}$ method

$$\Psi_\beta(v) := \left(\frac{1+v}{1-v}\right)^{\frac{1}{\beta}} \quad \text{with} \quad \beta > 0$$

Properties? Choice of  $\beta$ ?

# Quadrature-based discretization: analysis

$$h(t) = \frac{2}{\beta} \int_{-1}^1 e^{-\Psi_{\beta}(v)t} (1-v)^{-1-\frac{1}{\beta}} (1+v)^{\frac{1}{\beta}-1} \mu(\Psi_{\beta}(v)) dv$$

Best  $\beta$ ?  $\Rightarrow$  Integrand regularity. (Atkinson 1989, Thm. 5.4) (Dunham 1930, Thm. I.VIII)

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Results for  $h = Y_{\alpha}$ ,  $\alpha \in (0, 1)$

$$\beta \leq 1 - \alpha \quad \Rightarrow \quad Y_{\alpha, \text{num}}(t) \rightarrow Y_{\alpha}(t) \quad (t > 0)$$



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$$\begin{aligned} \beta \leq 1 - \alpha &\Rightarrow Y_{\alpha, \text{num}}(t) \rightarrow Y_{\alpha}(t) \quad (t > 0) \\ \beta \leq \min(\alpha, 1 - \alpha) &\Rightarrow \|Y_{\alpha, \text{num}}\|_{L^1(0, T)} \rightarrow \|Y_{\alpha}\|_{L^1(0, T)} \end{aligned}$$

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$$(+)\ \frac{1}{\beta} \in \mathbb{N}, \quad \frac{\alpha}{\beta} \in \mathbb{N} \quad \Rightarrow \quad \text{Spectral convergence}$$

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Conclusion: two choices for  $\beta$

$$\alpha = n_1/n_2 \quad \text{or} \quad \alpha \simeq n_1/n_2$$

$$\Rightarrow \beta_1 := 1/n_2$$

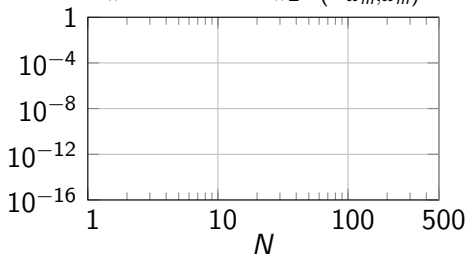
$$\beta_2 := \min(\alpha, 1 - \alpha)$$

Let's test this numerically!

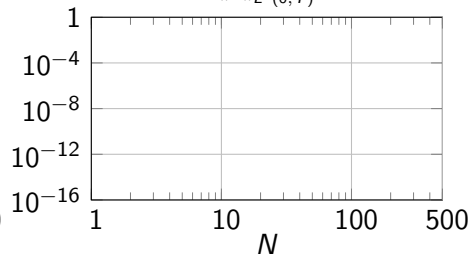
# Quadrature-based discretization: illustration

$$h = Y_{\frac{1}{2}}, \quad \omega_m = 10^4, \quad T = 10^4.$$

$$\left\| 1 - \frac{\hat{h}_{\text{num}}(i\omega)}{\hat{h}} \right\|_{L^\infty(-\omega_m, \omega_m)}$$

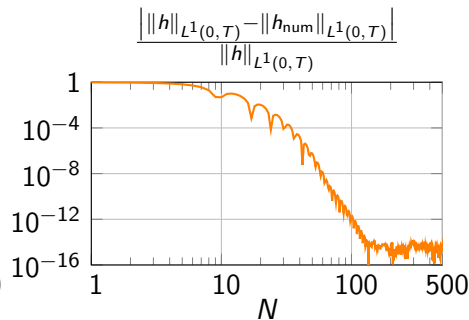
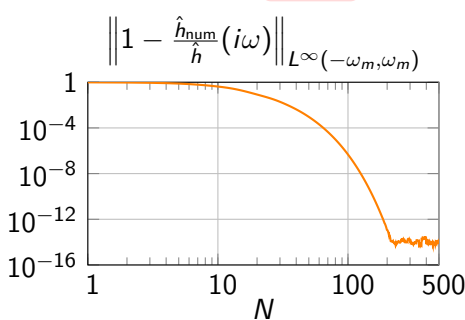


$$\frac{\left| \|h\|_{L^1(0, T)} - \|h_{\text{num}}\|_{L^1(0, T)} \right|}{\|h\|_{L^1(0, T)}}$$



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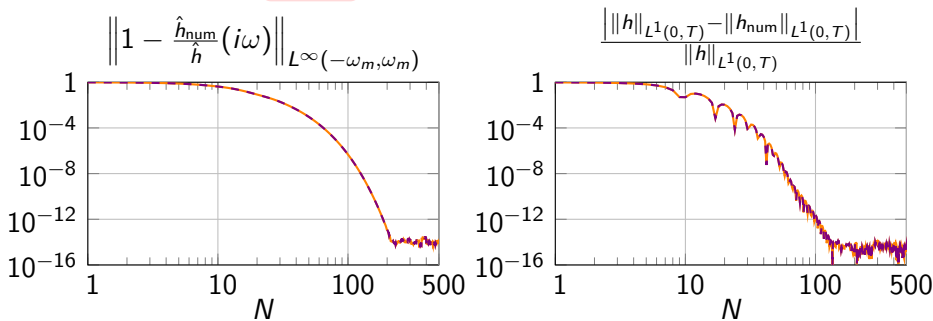
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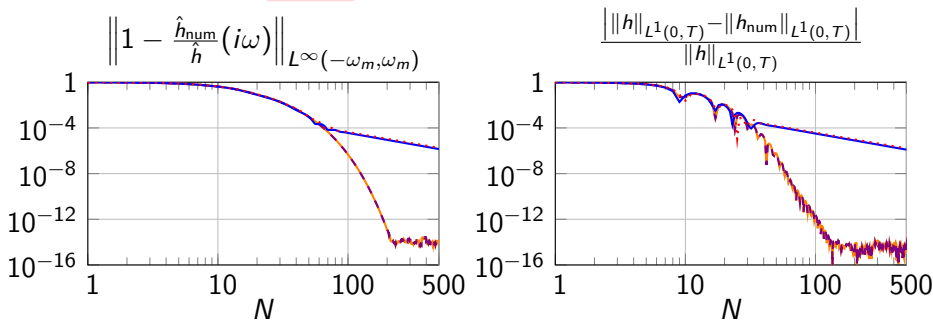
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$Q_{\beta, N}$  discretization with  $(\text{---}) \beta = \beta_1 \Leftrightarrow (\text{---}) \beta = \beta_2 = \frac{1}{2}$

$(\text{---}) \beta = \beta_2 \times 0.99$   
 $(\text{...}) \beta = \beta_2 \times 1.01$

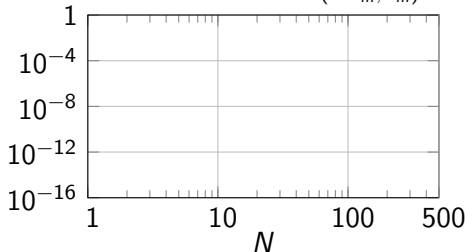
Sensitivity to  $\beta$

# Quadrature-based discretization: illustration (2)

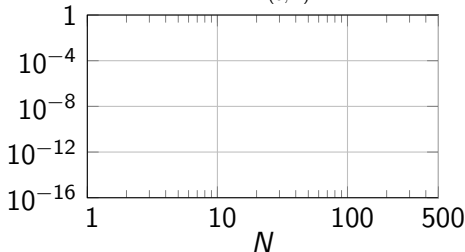
If  $\alpha \neq \frac{1}{2}$ ,  $\beta_1 \neq \beta_2$ . Let

$$h = Y_{\frac{5}{8}}, \quad \omega_m = 10^4, \quad T = 10^4.$$

$$\left\| 1 - \frac{\hat{h}_{\text{num}}}{\hat{h}}(i\omega) \right\|_{L^\infty(-\omega_m, \omega_m)}$$



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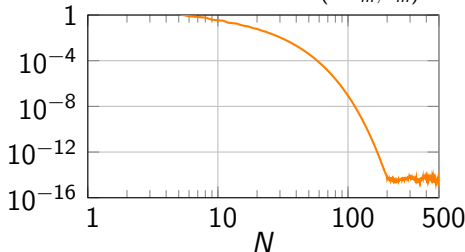


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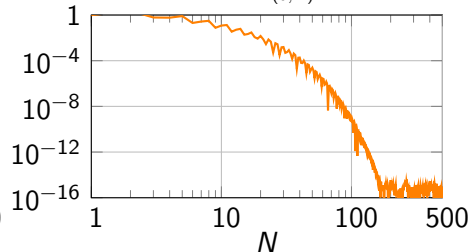
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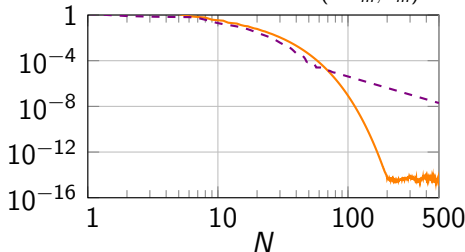
$Q_{\beta,N}$  discretization with (—)  $\beta = \beta_1 = \frac{1}{8}$

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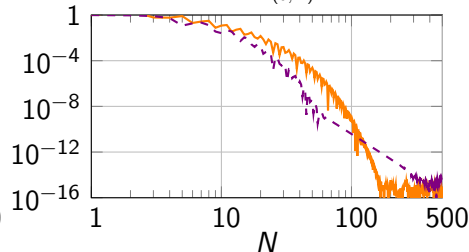
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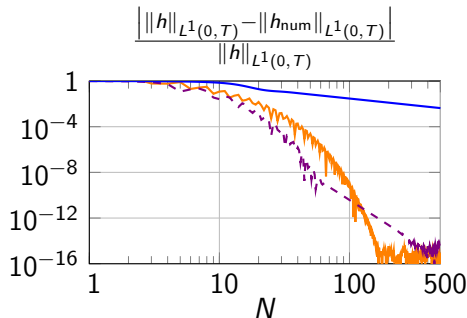
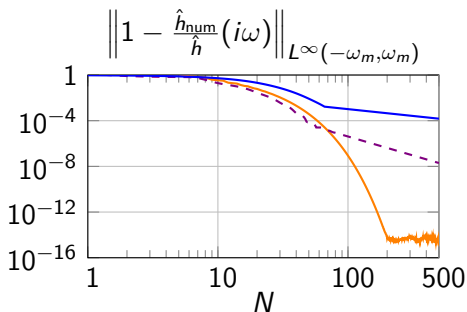


$Q_{\beta,N}$  discretization with (—)  $\beta = \beta_1 = \frac{1}{8}$   
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(- - -)  $\beta = \beta_2 \Rightarrow$  Choice from now on

(—)  $\beta = \max(\alpha, 1 - \alpha)$

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  - Spectral accuracy of quadrature methods
  - Comparison with optimization method
  - Application in duct aeroacoustics
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# Eigenvalue approach to stability: overview

Vector-valued fractional delay system: (Monteghetti et al. 2017)

$$\dot{x}(t) = Ax(t) + Bx(t - \tau) - Gd^{1-\alpha}x(t) \quad (\tau \geq 0).$$

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Hyperbolic realisation (PDE)

$$\theta \in (-\tau, 0)$$

$$\partial_t \psi = \partial_\theta \psi, \quad \psi(t, \theta = 0) = x(t)$$

$$x(t - \tau) = \psi(t, -\tau)$$

Parabolic realisation (ODE)

$$\xi \in (0, \infty)$$

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$$d^{1-\alpha}x = \int_0^\infty (\dots) \mu_\alpha(\xi) d\xi$$

⇒ Abstract Cauchy problem on  $H$

$$\dot{X}(t) = \mathcal{A}X(t), \quad \text{with } X := (x, \psi, \varphi).$$

Stability  $\iff$  spectrum  $\sigma(\mathcal{A})$

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Discretization

High-order DG-FEM  $(N_p)$

$Q_{\beta, N}$  or optimization  $(N)$

$$\dot{X}_h(t) = \mathcal{A}_h X_h(t) \quad \text{with } X_h := (x, \underbrace{\psi_h}_{N_p}, \underbrace{\varphi_h}_N)$$

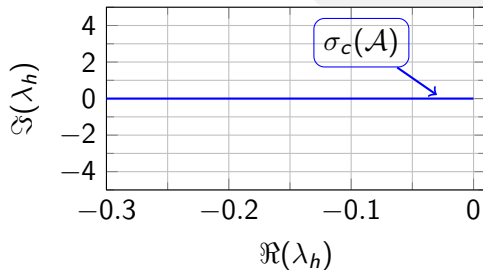
**Challenge** Ensure  $\sigma(\mathcal{A}_h)$  is “meaningful”.

# Eigenvalue approach to stability: spectral accuracy (1)

$A$  and  $B$  so that  $\mathcal{A}$  is asymptotically stable (Monteghetti et al. 2017).

Theoretically

$$S(\mathcal{A}) := \sup_{\lambda \in \sigma(\mathcal{A})} \Re(\lambda) = 0$$



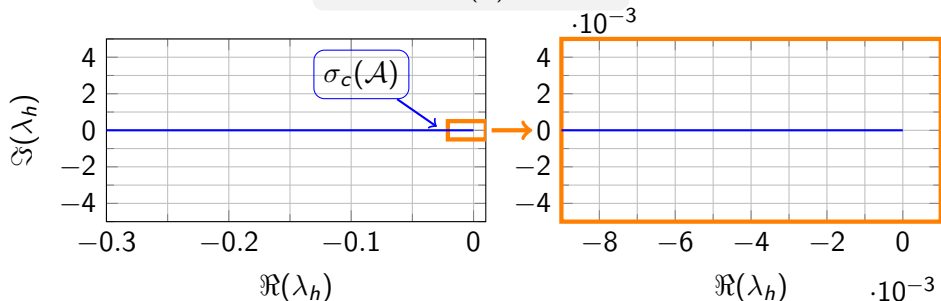


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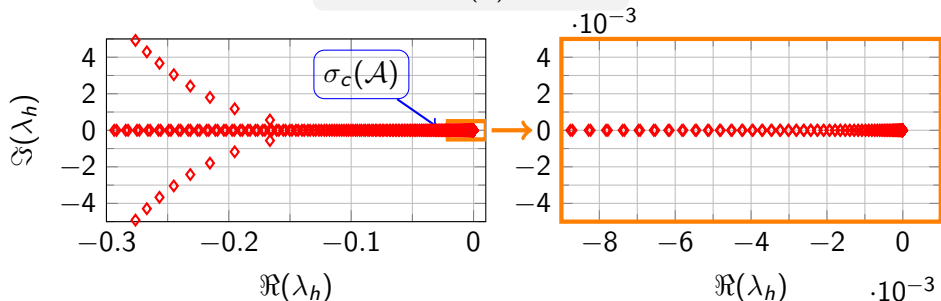
|                                      |  |  |  |
|--------------------------------------|--|--|--|
| $Q_{\beta,N}$ with $\beta = \beta_2$ |  |  |  |
| $N$                                  |  |  |  |
| $S(\mathcal{A}_h)$                   |  |  |  |

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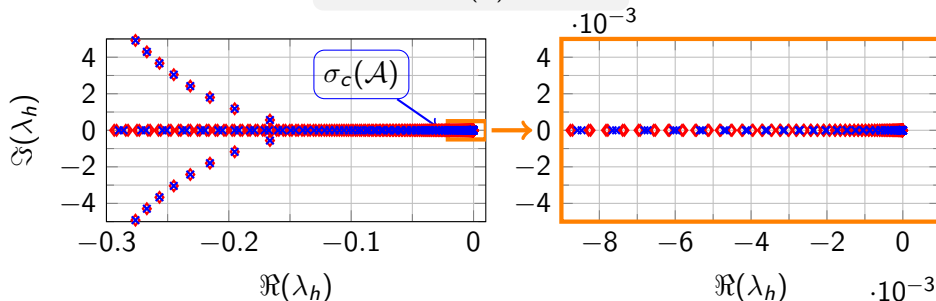
|                                       |                        |  |  |
|---------------------------------------|------------------------|--|--|
| $Q_{\beta, N}$ with $\beta = \beta_2$ | ( $\diamond$ )         |  |  |
| $N$                                   | 400                    |  |  |
| $S(\mathcal{A}_h)$                    | $-8.1 \times 10^{-11}$ |  |  |

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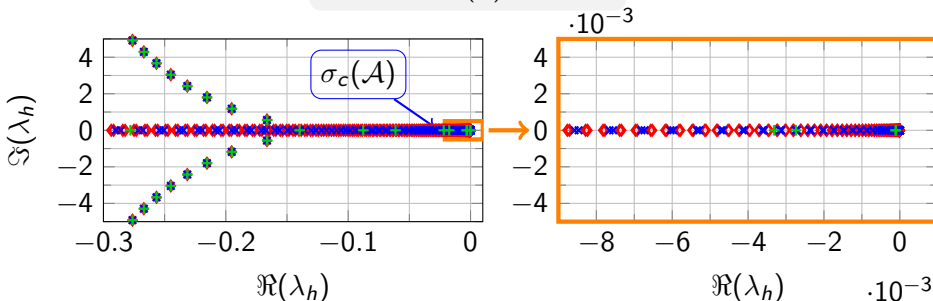
| $Q_{\beta,N}$ with $\beta = \beta_2$ | ( $\diamond$ )         | ( $\times$ )          |  |
|--------------------------------------|------------------------|-----------------------|--|
| $N$                                  | 400                    | 200                   |  |
| $S(\mathcal{A}_h)$                   | $-8.1 \times 10^{-11}$ | $-1.3 \times 10^{-9}$ |  |

# Eigenvalue approach to stability: spectral accuracy (1)

$A$  and  $B$  so that  $\mathcal{A}$  is asymptotically stable (Monteghetti et al. 2017).

Theoretically

$$S(\mathcal{A}) := \sup_{\lambda \in \sigma(\mathcal{A})} \Re(\lambda) = 0$$



| $Q_{\beta, N}$ with $\beta = \beta_2$ | (◇)                    | (×)                   | (+)                 |
|---------------------------------------|------------------------|-----------------------|---------------------|
| $N$                                   | 400                    | 200                   | 11                  |
| $S(\mathcal{A}_h)$                    | $-8.1 \times 10^{-11}$ | $-1.3 \times 10^{-9}$ | $-1 \times 10^{-4}$ |

What about the optimization method?

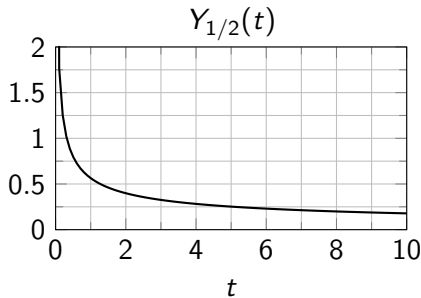
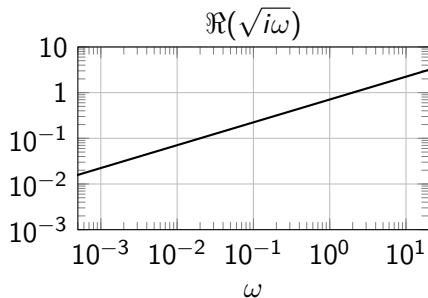
# Optimization-based discretization: principle

**Principle** Minimization of (Hélie et al. 2006)

$$(\xi_n, \mu_n) \mapsto \sum_{k=1}^K \left| \hat{h}(i\omega_k) - \sum_{n=1}^N \frac{\mu_n}{i\omega_k + \xi_n} \right|^2.$$

**Inputs**  $N$  poles  $\xi_n$  in  $[\xi_{\min}, \xi_{\max}]$

**Illustration**



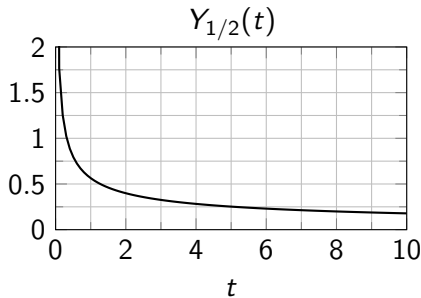
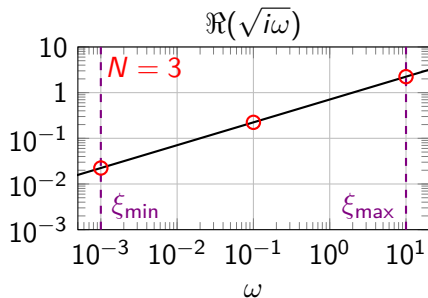
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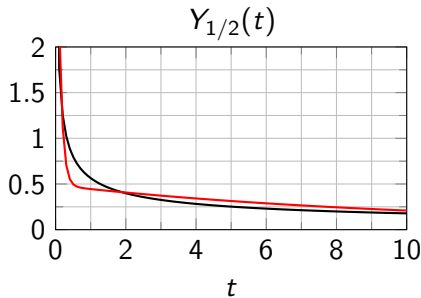
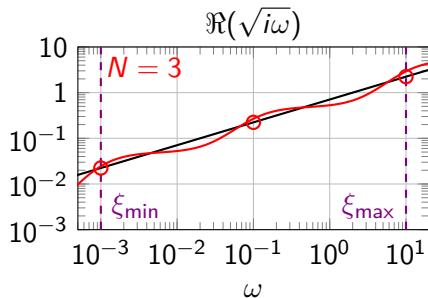
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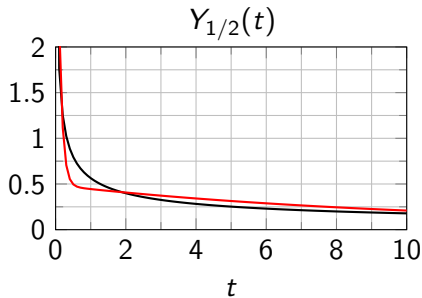
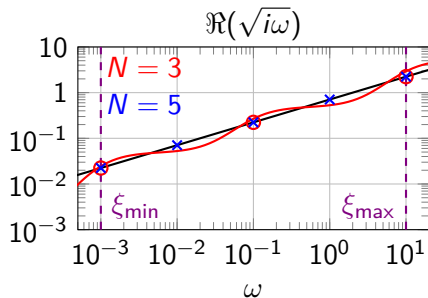
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**Illustration**





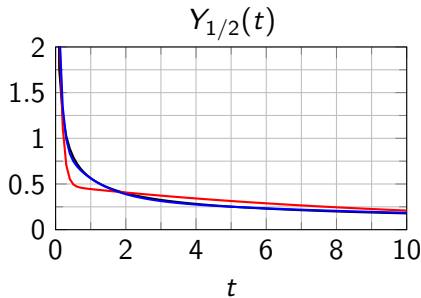
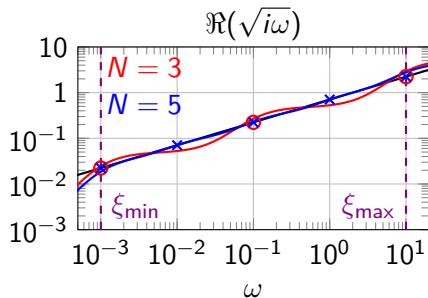
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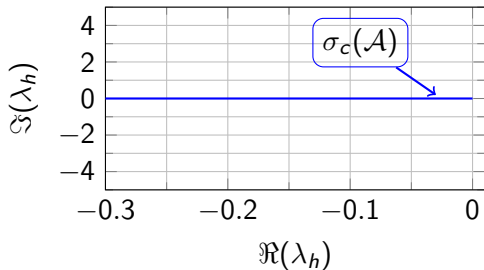
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**Illustration**



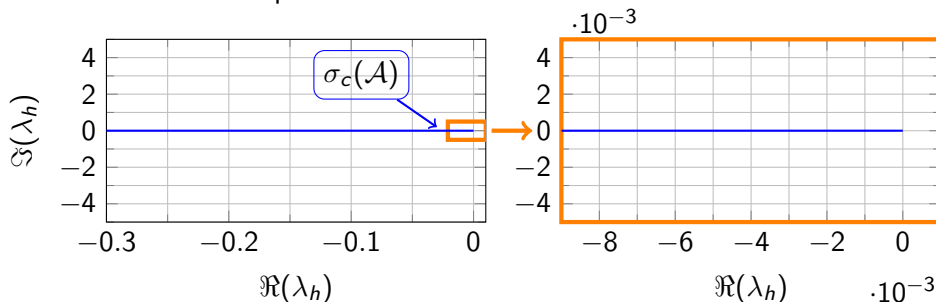
# Eigenvalue approach to stability: spectral accuracy (2)

What about the optimization method?



# Eigenvalue approach to stability: spectral accuracy (2)

What about the optimization method?

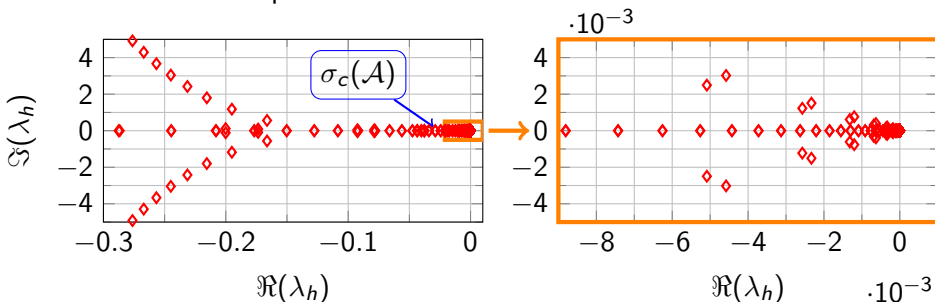


Let  $\xi_{\min} = 10^{-16}$ .

|                    | $\xi_{\max} = 10^4$ |  | $\xi_{\max} = 10^6$ |  |
|--------------------|---------------------|--|---------------------|--|
| $N$                |                     |  |                     |  |
| $S(\mathcal{A}_h)$ |                     |  |                     |  |

# Eigenvalue approach to stability: spectral accuracy (2)

What about the optimization method?

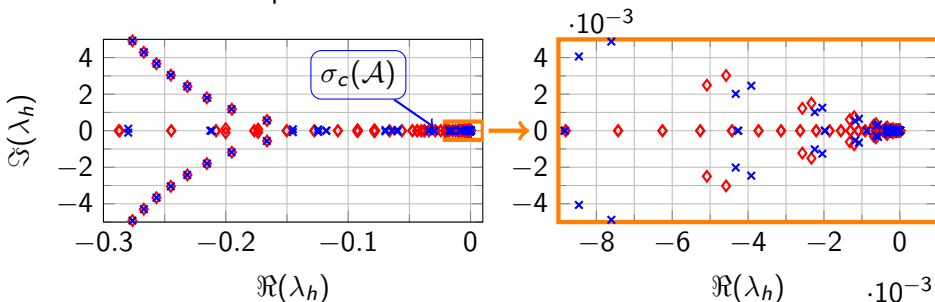


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|                    | $\xi_{\max} = 10^4$   |  | $\xi_{\max} = 10^6$ |  |
|--------------------|-----------------------|--|---------------------|--|
|                    | (◇)                   |  |                     |  |
| $N$                | 400                   |  |                     |  |
| $S(\mathcal{A}_h)$ | $3.4 \times 10^{-13}$ |  |                     |  |

# Eigenvalue approach to stability: spectral accuracy (2)

What about the optimization method?

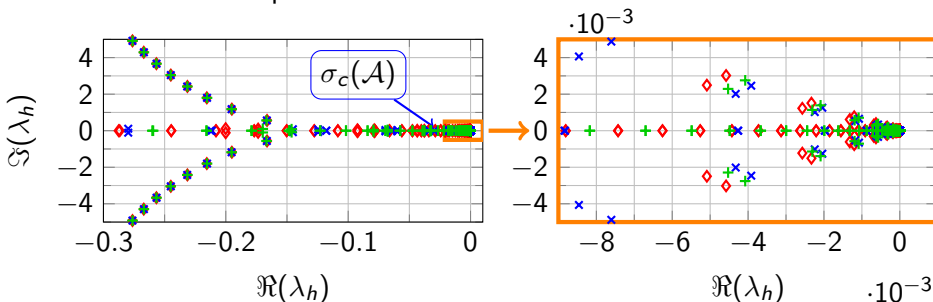


Let  $\xi_{\min} = 10^{-16}$ .

|                    | $\xi_{\max} = 10^4$   |                       | $\xi_{\max} = 10^6$ |  |
|--------------------|-----------------------|-----------------------|---------------------|--|
|                    | (◇)                   | (×)                   |                     |  |
| $N$                | 400                   | 200                   |                     |  |
| $S(\mathcal{A}_h)$ | $3.4 \times 10^{-13}$ | $3.8 \times 10^{-14}$ |                     |  |

# Eigenvalue approach to stability: spectral accuracy (2)

What about the optimization method?

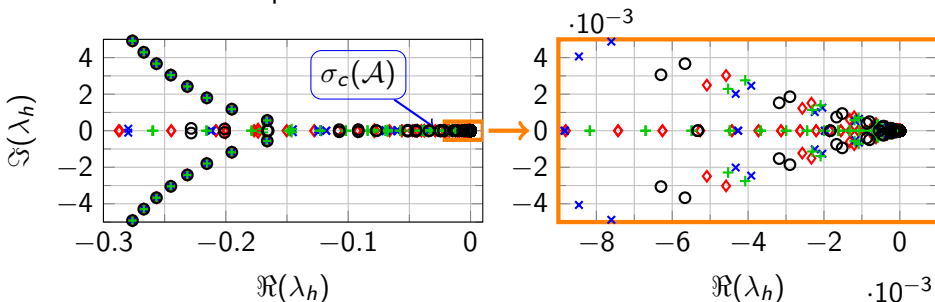


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|                    | $\xi_{\max} = 10^4$   |                       | $\xi_{\max} = 10^6$ |  |
|--------------------|-----------------------|-----------------------|---------------------|--|
|                    | (◇)                   | (×)                   | (+)                 |  |
| $N$                | 400                   | 200                   | 400                 |  |
| $S(\mathcal{A}_h)$ | $3.4 \times 10^{-13}$ | $3.8 \times 10^{-14}$ | $9 \times 10^{-12}$ |  |

# Eigenvalue approach to stability: spectral accuracy (2)

What about the optimization method?



Let  $\xi_{\min} = 10^{-16}$ .

|                    | $\xi_{\max} = 10^4$   |                       | $\xi_{\max} = 10^6$ |                       |
|--------------------|-----------------------|-----------------------|---------------------|-----------------------|
|                    | (◇)                   | (×)                   | (+)                 | (○)                   |
| $N$                | 400                   | 200                   | 400                 | 200                   |
| $S(\mathcal{A}_h)$ | $3.4 \times 10^{-13}$ | $3.8 \times 10^{-14}$ | $9 \times 10^{-12}$ | $1.6 \times 10^{-12}$ |

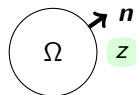
**Conclusion**  $Q_{\beta,N}$  best suited for stability studies.

# Duct aeroacoustics: overview

## Linearized Euler equations in $\Omega$

$$\partial_t \begin{bmatrix} \mathbf{u} \\ p \end{bmatrix} (t, \mathbf{x}) + \mathcal{A} \begin{bmatrix} \mathbf{u} \\ p \end{bmatrix} (t, \mathbf{x}) = 0$$

with  $p(t, \mathbf{x}) = [z \star_t \mathbf{u} \cdot \mathbf{n}(\cdot, \mathbf{x})](t)$  on  $\partial\Omega$ .



Physical impedance model (Monteghetti et al. 2016)

$$\hat{Z}_{\text{phys}}(s) = a_0 + a_{1/2} \sqrt{s} + a_1 s + \coth(a_0 + a_{1/2} \sqrt{s} + a_1 s)$$

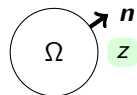


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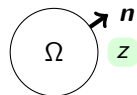
$$\begin{aligned} \hat{Z}_{\text{phys}}(s) &= a_0 + a_{1/2} \sqrt{s} + a_1 s + \coth(a_0 + a_{1/2} \sqrt{s} + a_1 s) \\ &= 1 + a_0 + a_{1/2} \hat{h}_1(s) + e^{-\tau s} \hat{h}_2(s). \end{aligned}$$

# Duct aeroacoustics: overview

## Linearized Euler equations in $\Omega$

$$\partial_t \begin{bmatrix} \mathbf{u} \\ p \end{bmatrix} (t, \mathbf{x}) + \mathcal{A} \begin{bmatrix} \mathbf{u} \\ p \end{bmatrix} (t, \mathbf{x}) = 0$$

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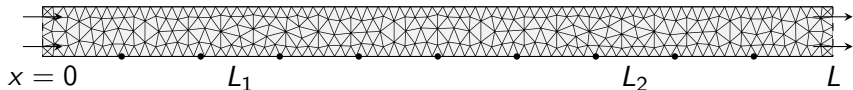
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Discretization process  $h_i \simeq h_{i,\text{num}}$ :

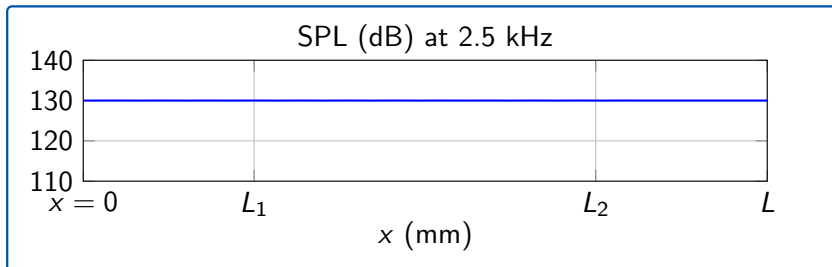
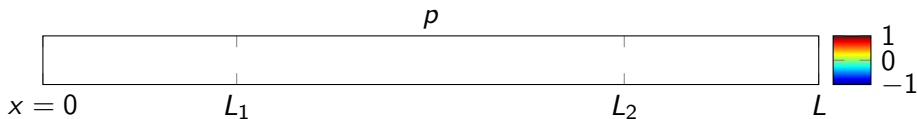
- ① Quadrature or optimization  $\Rightarrow$  Initial guess for  $h_{i,\text{num}} \Rightarrow z_{\text{num}}$
- ② Nonlinear least squares against experimental data

$$\|\hat{Z}_{\text{num}}(i\omega) - \hat{Z}_{\text{exp}}(i\omega)\|$$

# Duct aeroacoustics: application

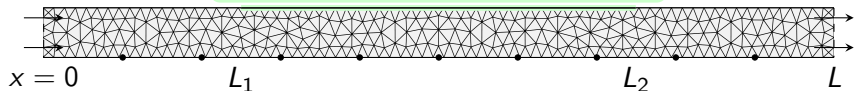


ArtimonDG(4). 552 triangles. CFL = 0.85. LSERK (8,4) (Toulorge et al. 2012)

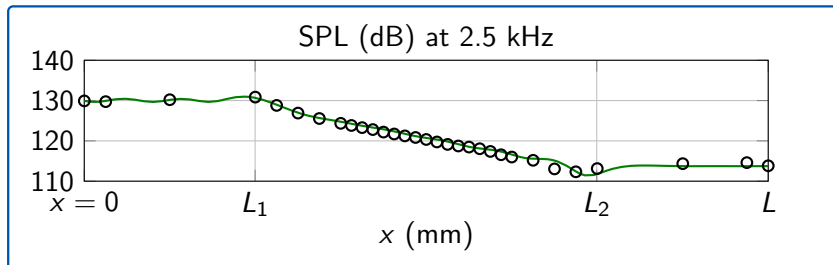
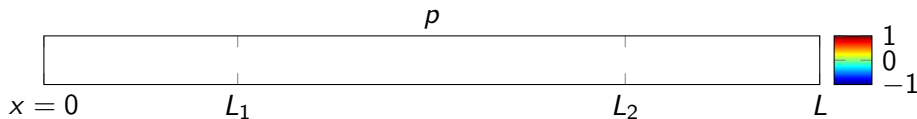


# Duct aeroacoustics: application

$\hat{\beta}_a(s)$  (liner CT57 – NASA Langley)

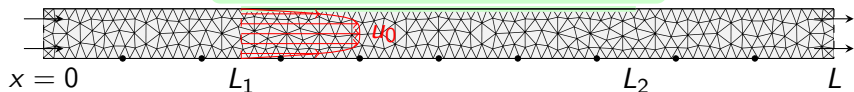


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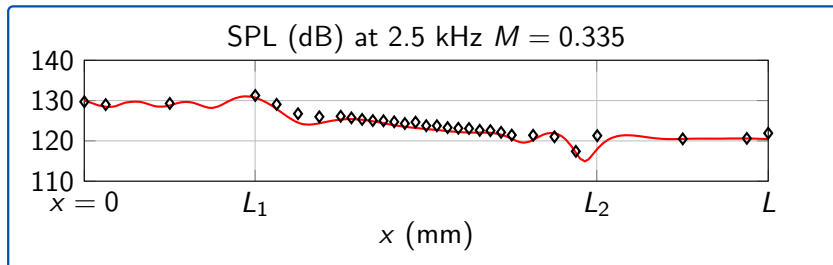
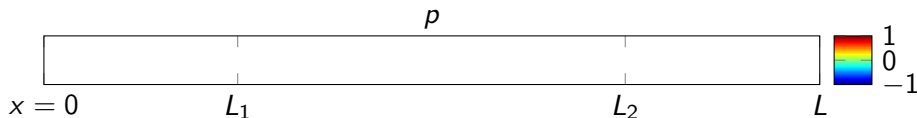


# Duct aeroacoustics: application

$\hat{\beta}_a(s)$  (liner CT57 – NASA Langley)



ArtimonDG(4). 552 triangles. CFL = 0.85. LSERK (8,4) (Toulorge et al. 2012)



# Outline

- 1 Introduction
- 2 Quadrature-based discretization of diffusive representations
- 3 Numerical comparisons and applications
- 4 Conclusion**
  - Conclusion

# Conclusion

## $Q_{\beta,N}$ method

- ✓ One parameter  $\beta(\alpha)$
- ✓ No spectral pollution
- ✗  $\xi_{\max} \propto N^{\frac{2}{\beta}}$

## Optimization-based method

- ✗ At least three parameters
- ✗ Polluted spectrum
- ✓ Control of  $\xi_{\max}$
- ✓ Suited for wave propagation

## Perspectives

- Applications to fractional PDE (e.g. fractional Schrödinger (Garrappa et al. 2015))
- Extension to diffusive operators with singular or sharply-varying weight  $\mu$  (e.g. Webster-Lokshin, Cavity impedance)
- Enhancement of the optimization method

# Table of contents

[▶ Main TOC](#)[▶ Additional slides TOC](#)

## Quadrature-based diffusive representation of the fractional derivative with applications in aeroacoustics and eigenvalue methods for stability







- 1 Introduction
- 2 Quadrature-based discretization of diffusive representations
- 3 Numerical comparisons and applications
- 4 Conclusion

Thanks for your attention. Any questions?






Contact: [florian.monteghetti@onera.fr](mailto:florian.monteghetti@onera.fr)



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