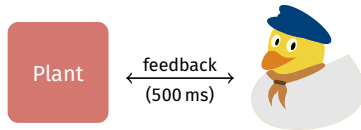


The Lanczos Tau Framework for Time-Delay Systems

Evert Provoost and Wim Michiels

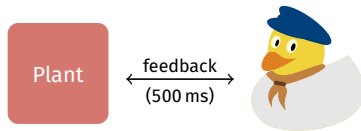
Preprint of same title available at <https://arxiv.org/abs/2403.03895>.

Why delay?



(More examples: Sipahi et al. [2].)

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RDDE state space

$$\begin{aligned}\dot{\mathbf{x}}(t) &= A_0 \mathbf{x}(t) + A_1 \mathbf{x}(t - \tau) + B \mathbf{u}(t), \\ \mathbf{y}(t) &= C \mathbf{x}(t).\end{aligned}$$

Delay differential systems

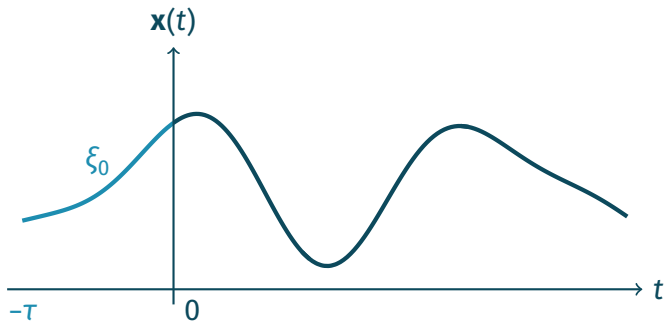
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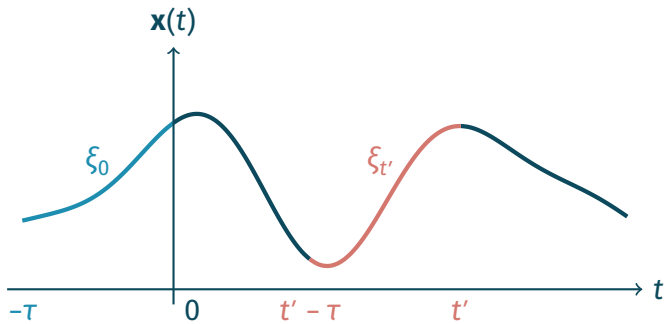
Transfer function

$$H(s) = C(sI_n - A_0 - A_1 e^{-\tau s})^{-1} B.$$

A functional differential equation



A functional differential equation



PDE formulation

$$\begin{cases} \dot{\xi}_t(0) = A_0 \xi_t(0) + A_1 \xi_t(-\tau) + B \mathbf{u}(t), \\ \dot{\xi}_t(\theta) = \frac{d}{d\theta} \xi_t(\theta), \\ \mathbf{y}(t) = C \xi_t(0), \end{cases}$$

where $\xi_t : [-\tau, 0] \rightarrow \mathbb{C}^n$,
 $\theta \mapsto \mathbf{x}(t + \theta)$.

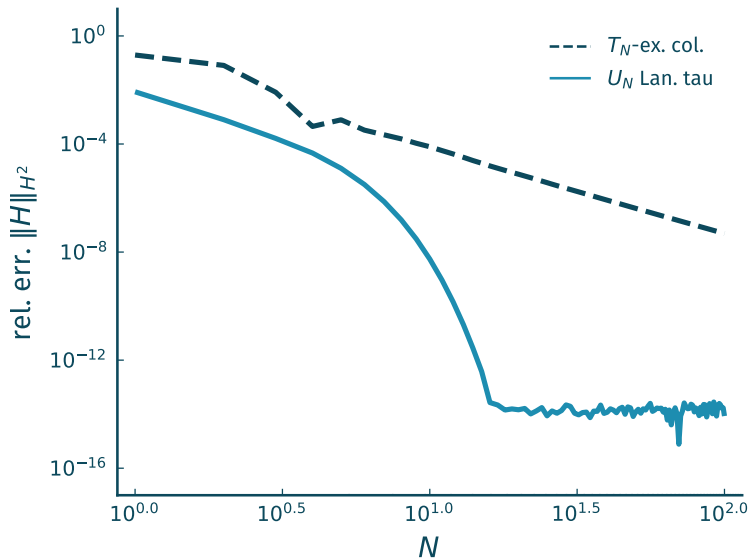
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where $\xi_t : [-\tau, 0] \rightarrow \mathbb{C}^n$,
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(See Breda et al. [3] for collocation.)

Why Lanczos tau?



Lanczos tau discretization

$$\begin{pmatrix} \varepsilon_0 \\ I \end{pmatrix} \dot{\xi}_t = \begin{pmatrix} A_0 \varepsilon_0 + A_1 \varepsilon_{-\tau} \\ \mathcal{D} \end{pmatrix} \xi_t + \begin{pmatrix} B \\ \mathbf{0} \end{pmatrix} \mathbf{u}(t),$$
$$\mathbf{y}(t) = C \varepsilon_0 \xi_t,$$

with $\varepsilon_\theta \xi = \xi(\theta)$ and $(\mathcal{D}\xi)(\theta) = \frac{d}{d\theta} \xi(\theta)$.

Lanczos tau discretization

$$\begin{pmatrix} \varepsilon_0 \\ \mathcal{T}_{N-1} \end{pmatrix} \dot{\xi}_{tN} = \begin{pmatrix} A_0 \varepsilon_0 + A_1 \varepsilon_{-\tau} \\ \mathcal{D} \end{pmatrix} \xi_{tN} + \begin{pmatrix} B \\ \mathbf{0} \end{pmatrix} \mathbf{u}(t),$$
$$\mathbf{y}(t) \approx C \varepsilon_0 \xi_{tN},$$

with $\mathcal{T}_{N-1} \xi = \xi - \langle \xi, \phi_N \rangle \phi_N$.

(Initially presented by Ito and Teglas [4].)

$$\begin{aligned}\dot{\mathbf{x}}(t) &= A_0 \mathbf{x}(t) + A_1 \mathbf{x}(t - \tau) + B \mathbf{u}(t) \\ \mathbf{y}(t) &= C \mathbf{x}(t)\end{aligned}$$

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$$\xrightarrow{\xi_t(\theta) = \mathbf{x}(t + \theta)}$$

$$\begin{aligned}\mathcal{E} \dot{\xi}_t &= \mathcal{A} \xi_t + \mathcal{B} \mathbf{u}(t) \\ \mathbf{y}(t) &= \mathcal{C} \xi_t\end{aligned}$$

$$\begin{aligned}\dot{\mathbf{x}}(t) &= A_0 \mathbf{x}(t) + A_1 \mathbf{x}(t - \tau) + B \mathbf{u}(t) \\ \mathbf{y}(t) &= C \mathbf{x}(t)\end{aligned}$$


$$\xi_{tN}(\theta) \approx \mathbf{x}(t + \theta)$$

~~~~~→

$$\begin{aligned}\mathcal{E}_N \dot{\xi}_{tN} &= \mathcal{A}_N \xi_{tN} + \mathcal{B}_N \mathbf{u}(t) \\ \mathbf{y}_N(t) &= \mathcal{C}_N \xi_{tN}\end{aligned}$$

## Interpretation in frequency domain

$$\begin{aligned}\dot{\mathbf{x}}(t) &= A_0 \mathbf{x}(t) + A_1 \mathbf{x}(t - \tau) + B \mathbf{u}(t) \\ \mathbf{y}(t) &= C \mathbf{x}(t)\end{aligned}$$

$$\xi_{tN}(\theta) \approx \mathbf{x}(t + \theta)$$


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Laplace



$$H(s) = C(sI_n - A_0 - A_1 e^{-\tau s})^{-1} B$$

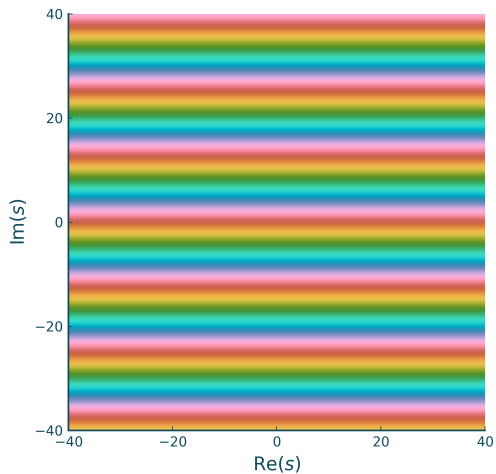


$$H_N(s) = C(sI_n - A_0 - A_1 r_N(s, -\tau))^{-1} B$$

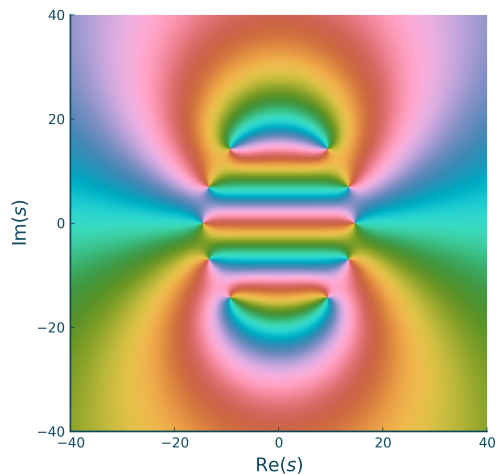
(See Vanbiervliet et al. [5].)



## In the complex plane

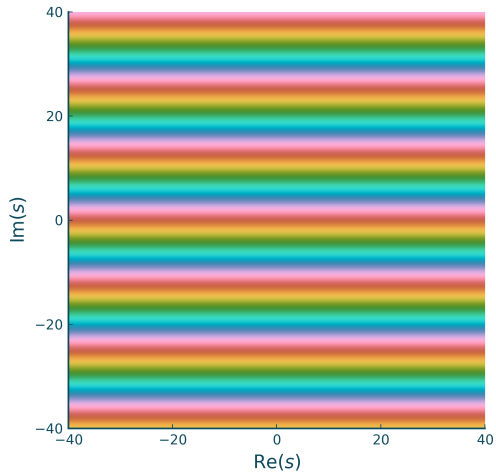


$$s \mapsto e^{-\tau s} \text{ with } \tau = 1$$

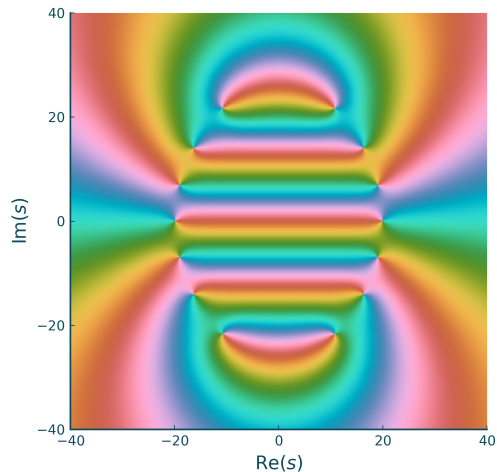


$$s \mapsto r_N(s, -\tau) \text{ with } \phi_k = P_k, \tau = 1, \text{ and } N = 5$$

## In the complex plane



$$s \mapsto e^{-\tau s} \text{ with } \tau = 1$$



$$s \mapsto r_N(s, -\tau) \text{ with } \phi_k = P_k, \tau = 1, \text{ and } N = 7$$

### Theorem

For the choice of  $\{P_k\}_{k=0}^N$  as basis,  $r_N(s, -\tau)$  is an  $(N, N)$  Padé approximant of  $s \mapsto e^{-\tau s}$  around zero.

## Theorem

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## Corollary

Then  $H_N$  is  $H$  with  $s \mapsto e^{-\tau s}$  replaced by an  $(N, N)$  Padé approximant around zero.

## Sparse, self-nesting discretizations

Clearly  $\text{span}\{\phi_k\}_{k=0}^{N_1} \subset \text{span}\{\phi_k\}_{k=0}^{N_2}$  for  $N_1 < N_2$ , hence easy self-nesting.

(See Jarlebring et al. [6] and Olver and Townsend [7].)

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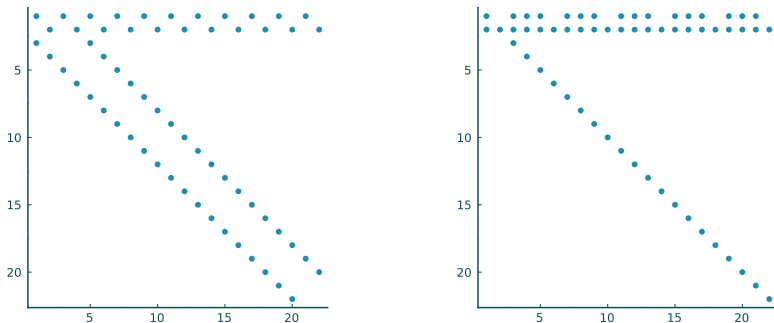
Choose  $\phi_k = U_k$ , but represent input with respect to  $\{T_k\}_{k=0}^N$ ,

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Choose  $\phi_k = U_k$ , but represent input with respect to  $\{\tau_k\}_{k=0}^N$ , then



Simple example of an  $(\mathcal{E}_N, \mathcal{A}_N)$  pencil at  $n(N+1) = 22$ .

(See Jarlebring et al. [6] and Olver and Townsend [7].)

What is the energy of  
the impulse response?



What is the energy of  
the impulse response?

What is the steady-state power of  
the output response to unit white noise?

### Definition (stable system)

$$\|H\|_{H^2} = \left( \frac{1}{2\pi} \int_{-\infty}^{+\infty} \|H(i\omega)\|_F^2 d\omega \right)^{\frac{1}{2}},$$

where  $H(s) = C(sI_n - A_0 - A_1 e^{-\tau s})^{-1} B$ .

## Computing the $H^2$ -norm of an ODE

$$\begin{aligned} E\dot{\mathbf{x}}(t) &= A\mathbf{x}(t) + B\mathbf{u}(t), \\ \mathbf{y}(t) &= C\mathbf{x}(t), \end{aligned}$$

with  $E$  invertible.

$$\|H\|_{H^2} = \sqrt{\text{tr}(CVC^T)}, \text{ where}$$

$$AVE^T + EVA^T = -BB^T.$$

(See Zhou et al. [8, Lemma 4.6].)

## Approximating the $H^2$ -norm of a RDDE

$$\begin{aligned}\mathcal{E}_N \dot{\mathbf{x}}(t) &= \mathcal{A}_N \mathbf{x}(t) + \mathcal{B}_N \mathbf{u}(t), \\ \mathbf{y}(t) &\approx \mathcal{C}_N \mathbf{x}(t),\end{aligned}$$

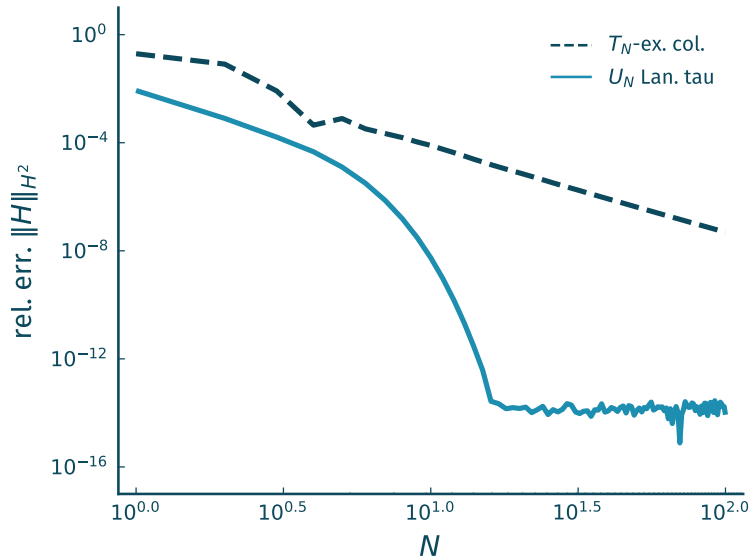
with  $\mathcal{E}_N$  invertible.

$$\|H\|_{H^2} \approx \sqrt{\text{tr}(\mathcal{C}_N V \mathcal{C}_N^T)}, \text{ where}$$

$$\mathcal{A}_N V \mathcal{E}_N^T + \mathcal{E}_N V \mathcal{A}_N^T = -\mathcal{B}_N \mathcal{B}_N^T.$$

(See Vanbiervliet et al. [5].)

# Convergence



## Polynomial formulation

$$\text{As } \mathbb{P}_N^n \cong \mathbb{C}^{nN}$$

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$$U(\theta, \theta') = \sum_{j,k} V_{jk} \phi_j(\theta) \phi_k(\theta') \in \mathbb{C}^{n \times n}.$$

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$$\|H\|_{H^2} \approx \|H_N\|_{H^2} = \sqrt{\text{tr}(C \varepsilon_0 U \varepsilon_0^T C^T)}, \text{ where}$$

$$\begin{cases} \mathcal{D}U\varepsilon_0^T + \mathcal{T}_{N-1}U(\varepsilon_0^T A_0^T + \varepsilon_{-\tau}^T A_1^T) = \mathbf{0}, \\ \mathcal{D}U\mathcal{T}_{N-1}^T + \mathcal{T}_{N-1}U\mathcal{D}^T = \mathbf{0}, \\ \varepsilon_\theta U \varepsilon_{\theta'}^T = (\varepsilon_{\theta'} U \varepsilon_\theta^T)^T, \\ (A_0 \varepsilon_0 + A_1 \varepsilon_{-\tau}) U \varepsilon_0^T + \varepsilon_0 U (\varepsilon_0^T A_0^T + \varepsilon_{-\tau}^T A_1^T) = -BB^T. \end{cases}$$

## Polynomial formulation

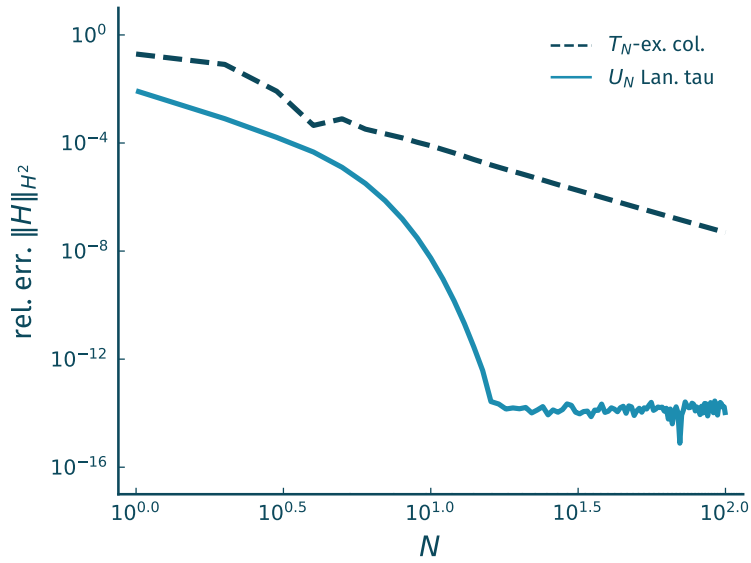
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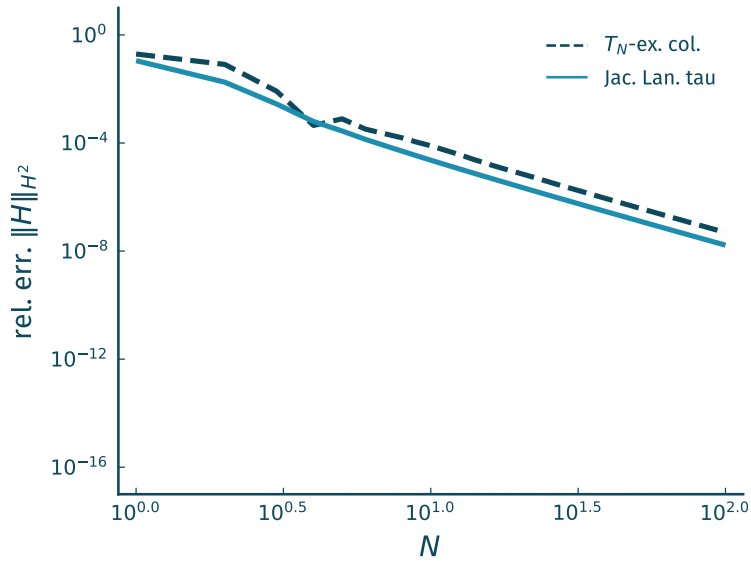
$$\|H\|_{H^2} \approx \|H_N\|_{H^2} = \sqrt{\text{tr}(C \varepsilon_0 U \varepsilon_0^T C^T)}, \text{ where}$$

some constraints on  $U$ .

# Symmetry is important



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### Assumption

$$\phi_k(-\tau - \theta) = (-1)^k \phi_k(\theta), \quad \forall \theta \in [-\tau, 0].$$

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### Proposition

Under this assumption

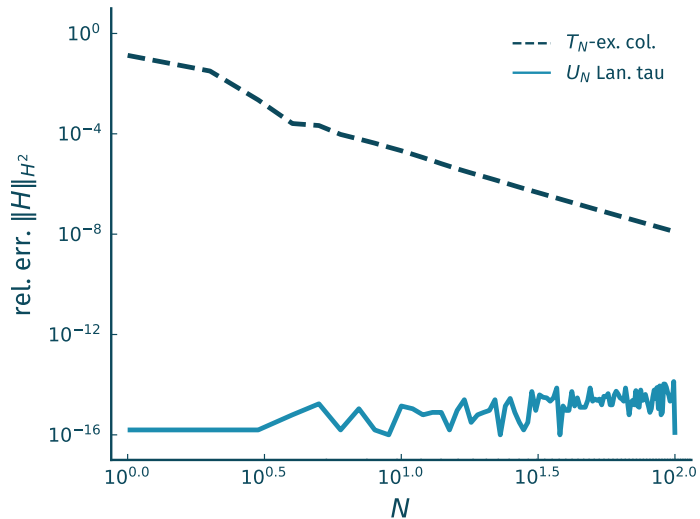
$$|r_N(i\omega, -\tau)| = 1, \quad \forall \omega \in \mathbb{R}.$$

## Super convergence

Let  $A_0 = A_1 = a < 0$ ,

## Super convergence

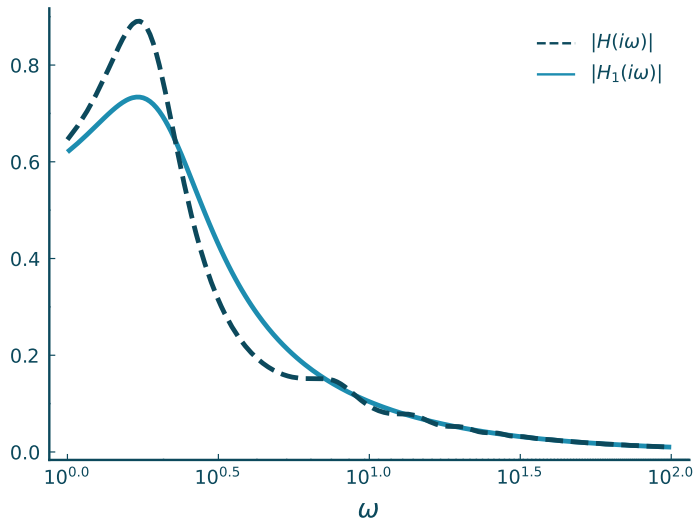
Let  $A_0 = A_1 = a < 0$ , then





## Super convergence

Let  $A_0 = A_1 = a < 0$ , then



## References

- [1] E. Provoost and W. Michiels. *The Lanczos Tau Framework for Time-Delay Systems: Padé Approximation and Collocation Revisited*. 2024. [arXiv: 2403.03895 \[math.NA\]](#).
- [2] R. Sipahi et al. “Stability and Stabilization of Systems with Time Delay”. In: *IEEE Control Systems* 31.1 (2011).
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- [4] K. Ito and R. Teglás. “Legendre–Tau Approximations for Functional-Differential Equations”. In: *SIAM Journal on Control and Optimization* 24.4 (1986).
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- [6] E. Jarlebring et al. “A Krylov Method for the Delay Eigenvalue Problem”. In: *SIAM Journal on Scientific Computing* 32.6 (2010).
- [7] S. Olver and A. Townsend. “A Fast and Well-Conditioned Spectral Method”. In: *SIAM Review* 55.3 (2013).
- [8] K. Zhou et al. *Robust and Optimal Control*. Pearson, 1995.

## Contributions

Operator formulation of the Lanczos tau method for time-delay systems.

Equivalence to rational approximation in frequency domain, with explicit expressions.\*

Construction of sparse, self-nesting discretizations.

Equivalence to pseudospectral collocation, with the non-zero collocation points the zeroes of  $\phi_N$ .\*

Equivalence to Padé approximation when using a Legendre basis.

Illustrated super-geometric convergence, and proved for some cases super convergence,\* for the  $H^2$ -norm.

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\*Not in this talk, see [E. Provoost and W. Michiels](#). *The Lanczos Tau Framework for Time-Delay Systems: Padé Approximation and Collocation Revisited*. 2024. [arXiv: 2403.03895 \[math.NA\]](#).