

# A $\star$ -product approach to non-autonomous linear ordinary differential equations

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# Our teams

## Local team (Charles University)



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# Schrödinger equation

Nuclear Magnetic Resonance (NMR) spectroscopy analyses dynamics of nuclear spins in magnetic field:

$$\hbar \frac{d}{dt} \Psi(t) = -iH(t)\Psi(t),$$

with

wave function  $\Psi(t)$

Hamiltonian  $H(t)$

- $\ell$  spins  $\rightarrow$  size  $2^\ell \times 2^\ell$
- Sparse

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4 strongly coupled spins

$$\begin{bmatrix} \times & & & & & & & & & & & & & & & & \\ & \times & \times & & \times & & & & & & & & & & & & \\ & & \times & & \times & \times & & & & & \times & \times & & & & \\ & & & \times & \times & \times & \times & & & & & \times & \times & & & \\ & & & & \times & \times & \times & & & & \times & \times & \times & & & \\ & \times & \times & & \times & & & & & & & \times & \times & \times & & \\ & & \times & \times & & \times & \times & \times & & & & & \times & \times & \times & \\ & & & \times & \times & & \times & \times & \times & & & & & \times & \times & \\ & & & & \times & \times & & \times & \times & \times & & & & & \times & \times \\ & & & & & \times & \times & \times & \times & \times & \times & & & & & \times \\ & & & & & & \times & \times & \times & \times & \times & \times & & & & \\ & & & & & & & \times & \times & \times & \times & \times & \times & & & \\ & & & & & & & & \times & \times & \times & \times & \times & \times & & \\ & & & & & & & & & \times & \times & \times & \times & \times & \times & \\ & & & & & & & & & & \times & \times & \times & \times & \times & \times \\ & & & & & & & & & & & \times & \times & \times & \times & \times \\ & & & & & & & & & & & & \times & \times & \times & \times \\ & & & & & & & & & & & & & \times & \times & \times \\ & & & & & & & & & & & & & & \times & \times \\ & & & & & & & & & & & & & & & \times \end{bmatrix}$$

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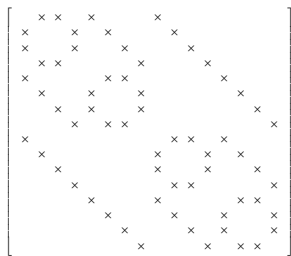
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4 uncoupled spins + pulse wave



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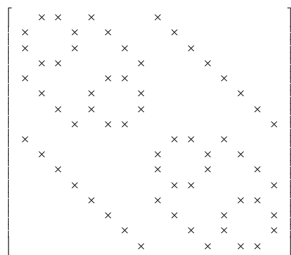
Why? Computer simulation

improves design

improves analysis

of practical experiments

4 uncoupled spins + pulse wave



# Numerical methods - references (incomplete list)

- [Moro, Freed, 1981]
- [Tal-Ezer, Kosloff, 1984]
- [Park, Light, 1985]
- [Leforestier et al., 1991]
- [Chin, Chen, 2002]
- [Hochbruck, Lubich, 2003]
- [Veshtort, Griffin, 2006]
- [Lauvergnat, Blasco, Chapuisat, 2007]
- [Kormann, Holmgren, Karlsson, 2008]
- [Lubich, 2008]
- [Blanes, Casas, Oteo, Ros, 2009]
- [Tošner, Vosegaard, Kehlet, Khaneja, Glaser, Nielsen, 2009]
- [Mazzi, 2010]
- [Sánchez, Casas, Fernández, 2011]
- [Iserles, Munthe-Kaas, Nørsett, Zanna, 2015]
- [Blanes, Casas, Murua, 2015]
- [Bader, Iserles, Kropielnicka, Singh, 2016]
- [Blanes, Casas, 2016]
- ...

Softwares: SIMPSON, SPINACH.

# Outline

- ▶ Solution of ODEs by  $\star$ -product.
- ▶ From the  $\star$ -product algebra to the matrix algebra.
- ▶ A new NLA approach for linear ODEs.



# Scalar ODE

For the sake of a simpler explanation, consider the scalar problem:

$$\frac{d}{dt}u(t) = \tilde{f}(t)u(t), \quad u(s) = 1, \quad t \geq s \in I \subset \mathbb{R}$$

with  $\tilde{f}(t)$  a smooth and bounded function over  $I$ . Then

$$u(t) = \exp\left(\int_s^t \tilde{f}(\tau) d\tau\right).$$

We are going to show an **alternative expression** for the solution.  
**Everything we are going to see can be easily extended to a system of ODEs.**

# Volterra composition

Let  $\tilde{f}_1(t, s), \tilde{f}_2(t, s)$  be bivariate matrices. The following operation

$$(\tilde{f}_2 \star_v \tilde{f}_1)(t, s) := \int_s^t \tilde{f}_2(t, \tau) \tilde{f}_1(\tau, s) d\tau.$$

is known as **Volterra composition**. Picard iteration:

$$\frac{d}{dt} u(t) = \tilde{f}(t)u(t), \quad u(s) = 1$$

↓ integration

$$u(t) = 1 + \int_s^t \tilde{f}(\tau)u(\tau)d\tau$$

↓ integration

$$\begin{aligned} u(t) &= 1 + \int_s^t \tilde{f}(\tau) \left( 1 + \int_s^\tau \tilde{f}(\rho)u(\rho)d\rho \right) d\tau \\ &= 1 + \int_s^t \tilde{f}(\tau) + \int_s^\tau \tilde{f}(\tau)\tilde{f}(\rho)u(\rho) d\rho d\tau \end{aligned}$$

↓ after many iterations

$$u(t) = 1 + \int_s^t \tilde{f}(\tau)d\tau + \int_s^t \tilde{f}^{\star_v 2}(\tau)d\tau + \dots = 1 + \int_s^t \sum_{j=1}^{\infty} \tilde{f}^{\star_v j}(\tau)d\tau.$$

## ★-product

We extend the Volterra composition defining the so-called ★-product

$$(f_2 \star f_1)(t, s) := \int_{-\infty}^{+\infty} f_2(t, \tau) f_1(\tau, s) d\tau.$$

that is an actual product for a specific class of distributions [Giscard, P., Ryckebusch].

Let  $\Theta(t - s)$  be the Heaviside function

$$\Theta(t - s) = \begin{cases} 1, & t \geq s, \\ 0, & t < s \end{cases}$$

Then

$$\tilde{f}_2(t - s)\Theta(t - s) \star \tilde{f}_1(t, s)\Theta(t - s) = \tilde{f}_2(t, s) \star_v \tilde{f}_1(t, s).$$

# Discontinuity

Definition:  $\star$ -product

- $f_1(t, s), f_2(t, s) \in$  specific class of distributions
- $(f_2 \star f_1)(t, s) := \int_{-1}^1 f_2(t, \tau) f_1(\tau, s) d\tau$

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The class consists of  $f(t, s) = \tilde{f}(t)\Theta(t - s)$

- $\tilde{f}(t)$  smooth
- Heaviside function  $\Theta(t - s) = \begin{cases} 1, & t \geq s \\ 0, & t < s \end{cases}$

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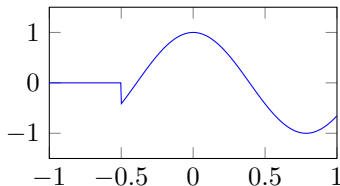
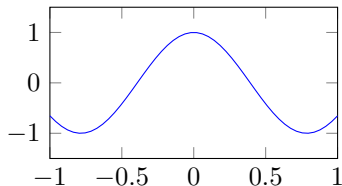
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Example:  $\tilde{f}(t) = \cos(4t)$

$$f(t, s) = \cos(4t)\Theta(t - s)|_{s=-0.5}$$



# Discontinuity

Definition:  $\star$ -product

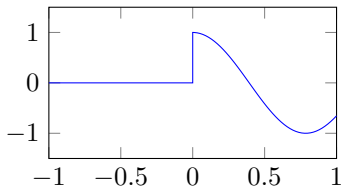
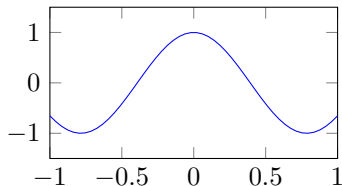
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Example:  $\tilde{f}(t) = \cos(4t)$

$$f(t, s) = \cos(4t)\Theta(t - s)|_{s=0}$$



# The $\star$ -product algebra

$$r(t, s) = f(t, s) \star g(t, s)$$

$$f + g$$

$$1_\star = \delta(t - s)$$

$$f^{\star-1}(t, s)$$

$$R_\star(f)(t, s) := (1_\star - f)^{\star-1}(t, s)$$

$$\tilde{f}\Theta \star \tilde{g}\Theta \rightarrow \tilde{r}\Theta, \text{ closed}$$

closed

Dirac delta

Dirac delta derivatives  $\delta', \delta'', \dots$

$\star$ -resolvent



## Solution by $\star$ -resolvent

Let us define  $f(t, s) = \tilde{f}(t)\Theta(t - s)$ . The  $\star$ -resolvent is defined as

$$R_{\star}(f) := (1_{\star} - f)^{\star-1} = 1_{\star} + \sum_{k \geq 1} f^{\star k}$$

with  $1_{\star} = \delta(t - s)$  the Dirac delta function. The  $\star$ -resolvent exists if  $f$  is bounded for  $t, s \in I$ . Then the Time-ordered exponential can be given as

$$U(t, s) = \int_s^t R_{\star}(f)(\tau, s) d\tau = \Theta(t - s) \star R_{\star}(f)(t, s),$$

where for a fixed  $s$ ,  $U(t, s)$  solves the ODE with starting time  $s$ ; [Giscard & al., 2015].

## Example

Consider the simple case  $\tilde{f}(t) = 1$ ,

$$\frac{d}{dt}u(t) = u(t), \quad u(s) = 1.$$

Then  $f(t, s) = \Theta(t - s)$  and

$$R_{\star}(f)(t, s) = \sum_{k=0}^{\infty} \Theta(t - s)^{\star k}.$$

Hence

$$U(t, s) = \Theta(t - s) \star R_{\star}(f)(t, s) = \sum_{k=0}^{\infty} (\Theta(t - s))^{\star(k+1)} = \sum_{k=0}^{\infty} \frac{(t - s)^k}{(k)!} \Theta(t - s).$$

As expected, for a fixed  $s$ , the solution is

$$u(t) = U(t, s) = \exp(t - s)\Theta(t - s).$$

## § 4. — Risoluzione generale di equazioni integrali.

9. Abbiassi una funzione analitica del tipo (1)

$$(1') \quad \mathbf{F}(z_1, z_2, \dots, z_n).$$

Scriviamo l'equazione

$$(4) \quad \mathbf{F}(z_1, z_2, \dots, z_n) = 0.$$

$$S(x, y) = R(x, y) - \frac{1}{2} R^2(x, y) + \frac{1}{3} R^3(x, y) - \dots + \frac{(-1)^n}{n} R^n(x, y) + \dots$$

ove

$$R^n(x, y) = \int_x^y R^{n-1}(x, \xi) R(\xi, y) d\xi.$$

e non dovremo porre alcuna limitazione per i valori assoluti di  $S(x, y)$ ,  $R(x, y)$ , purchè siano finiti.

10. Supponiamo in particolare che la (4') sia un polinomio razionale e

[Volterra, Rend Lincei, 1910]

## Vito Volterra - 91 years from "fascism oath" rejection

- ▶ In 1931, the Italian fascist regime imposed an oath of allegiance to the fascist government to all university professors.
- ▶ Only 12 professors refused to sign it. They lost their position.
- ▶ Vito Volterra was one of them. He was marginalized from the Italian scientific community (from '38, also because of the racial laws).
- ▶ First helped by the Vatican Science Academy, he then lived in Spain and France.



# Systems of ODEs

Let  $t \geq s \in I \subseteq \mathbb{R}$ ,  $A(t)$  a time dependent matrix. The **time-ordered exponential** is the unique solution  $U(t, s)$  of

$$\tilde{A}(t)U(t, s) = \frac{d}{dt}U(t, s), \quad U(s, s) = I_N.$$

If  $\tilde{A}(\tau_1)\tilde{A}(\tau_2) = \tilde{A}(\tau_2)\tilde{A}(\tau_1)$  for all  $\tau_1, \tau_2 \in I$ , then

$$U(t, s) = \exp\left(\int_s^t \tilde{A}(\tau) d\tau\right).$$

$U$  has generally **no explicit form**. Expression by ([Dyson, 1948])

$$U(t, s) = \mathcal{T} \exp\left(\int_s^t \tilde{A}(\tau) d\tau\right).$$

with  $\mathcal{T}$  the time-ordering operator.

# Time-ordered exponential

The time-ordering expression is more a notation as the action of the time-ordering operator is very difficult to evaluate.

- ▶ **Classical approaches** Perturbative methods (Floquet-based and Magnus series techniques), often prohibitively involved, e.g., [Blanes & al., 2009];
- ▶ **Path-sum approach**: The expression has a finite number of scalar integro-differential equations, but its complexity can be too large; [Giscard & al., 2015].
- ▶ **★-Lanczos + Path-sum**: [Giscard, P., 2020-2022]

## Solution by $\star$ -product

Nevertheless, the  $\star$ -resolvent expression for the solution remains:

$$U(t, s) = \Theta(t - s) \star R_\star(A),$$

with

$$R_\star(A)(t, s) = \sum_{k=0}^{\infty} \left( \tilde{A}(t) \Theta(t - s) \right)^{\star k},$$

where the  $\star$ -product here is extended in the matrix-product sense.

The problem is: **How do we compute  $R_\star(A)$ ?**

Symbolic expression: **Path-sum method** [Giscard & al., 2015] and  **$\star$ -Lanczos symbolic method** [Giscard, P., 2020-2022].

# Compute $u(t, s)$

Compute  $u(t, s)$

- Symbolically: slow
- Numerically: rest of the presentation

Numerical procedure:

1. Discretize  $f(t, s)$
2. Discrete analogue of  $\star$ -operations
3. Compute discretized resolvent
4. Express  $u(t, s)$  in terms of resolvent



## Discretize input $f(t, s)$

Function of interest

$$f(t, s) = \tilde{f}(t)\Theta(t - s)$$

Use Legendre polynomials  $\{p_k\}$

$$\int_{-1}^1 p_k(\tau)p_l(\tau)d\tau = \begin{cases} 0, & \text{if } k \neq l \\ 1, & \text{if } k = l \end{cases}$$

Legendre series

$$f(t, s) \approx \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} f_{i,j} p_i(t) p_j(s)$$

## Discretize input $f(t, s)$

Function of interest

$$f(t, s) = \tilde{f}(t)\Theta(t - s)$$

Use Legendre polynomials  $\{p_k\}_{k=0}^M$

$$\int_{-1}^1 p_k(\tau)p_l(\tau)d\tau = \begin{cases} 0, & \text{if } k \neq l \\ 1, & \text{if } k = l \end{cases}$$

Legendre series

$$f(t, s) \approx \sum_{i=0}^M \sum_{j=0}^M f_{i,j} p_i(t) p_j(s) =: f_M(t, s)$$

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$$f_M(t, s) = \begin{bmatrix} p_0(t) & p_1(t) & \dots & p_M(t) \end{bmatrix} \begin{bmatrix} f_{0,0} & f_{0,1} & \dots & f_{0,M} \\ f_{1,0} & f_{1,1} & \dots & f_{1,M} \\ \vdots & \vdots & & \vdots \\ f_{M,0} & f_{M,1} & \dots & f_{M,M} \end{bmatrix} \begin{bmatrix} p_0(s) \\ p_1(s) \\ \vdots \\ p_M(s) \end{bmatrix}$$

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Legendre series

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$$f_M(t, s) = [p_0(t) \quad p_1(t) \quad \dots \quad p_M(t)] F_M \begin{bmatrix} p_0(s) \\ p_1(s) \\ \vdots \\ p_M(s) \end{bmatrix}$$

# The matrix algebra

In Legendre basis:

$$f(t, s) \approx [p_0(t) \quad p_1(t) \quad \dots \quad p_M(t)] F_M \begin{bmatrix} p_0(s) \\ p_1(s) \\ \vdots \\ p_n(s) \end{bmatrix}$$

$f(t, s)$

$\star$ -operation

$$s(t, s) = f(t, s) \star g(t, s)$$

$f + g$

$$1_\star = \delta(t - s)$$

$$f^{\star^{-1}}(t, s)$$

$$R_\star(f)(t, s) := (1_\star - f)^{\star^{-1}}(t, s)$$

$F_M$

matrix operation

$$S_M = F_M G_M$$

$$F_M + G_M$$

$I_M$ , identity matrix

$$F_M^{-1}$$

$$R(F_M) := (I_M - F_M)^{-1}$$

# The matrix algebra

In Legendre basis:

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$$S_M = F_M G_M$$

$$F_M + G_M$$

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$$F_M^{-1}$$

$$R(F_M) := (I_M - F_M)^{-1}$$

Solution to  $\star$ -ODE:

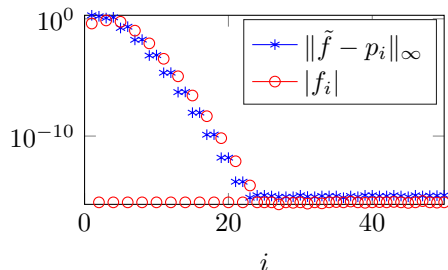
$U(t, s)$

$$\Theta(t - s) \star (1_\star - f)^{\star-1}(t, s) \left| \begin{array}{l} U_M \\ H_M(I_M - F_M)^{-1} \end{array} \right.$$

# Legendre approximation

Analytic  $\rightarrow$  geometric convergence

$$\tilde{f}(t) = \cos(4t) \approx \sum_{i=0}^M \tilde{f}_i p_i(t)$$
$$M = 23$$



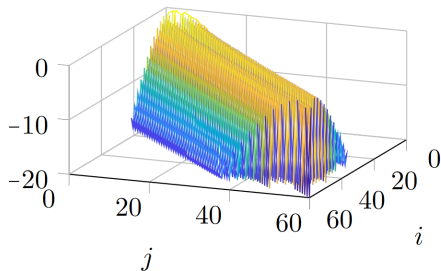
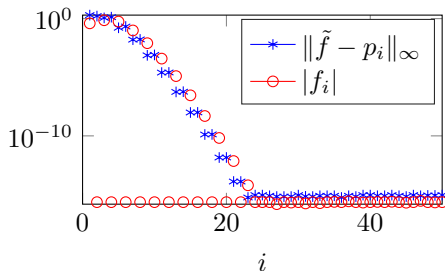
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$$M = 23$$

$\star$ -framework  $\Rightarrow$  discontinuous

$$f(t, s) = \cos(4t)\Theta(t - s)$$
$$\approx \sum_{i=0}^M \sum_{j=0}^M \tilde{f}_{i,j} p_i(t) p_j(s)$$





# Legendre approximation

Analytic  $\rightarrow$  geometric convergence

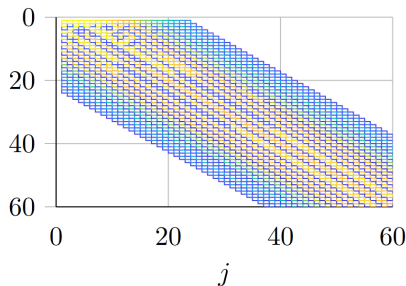
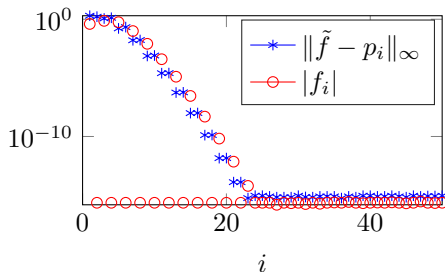
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\*-framework  $\Rightarrow$  discontinuous

$$f(t, s) = \cos(4t)\Theta(t - s)$$
$$\approx \sum_{i=0}^M \sum_{j=0}^M f_{i,j} p_i(t) p_j(s)$$

$F_M$  is (numerically) banded!

Note:  $f(t, -1) = f(t)$



# Reconstruction and Gibbs phenomenon

Even if  $\tilde{f}(t)$  is analytic,  $f(t, s) = \tilde{f}(t)\Theta(t - s)$  is discontinuous. Therefore, the expansion

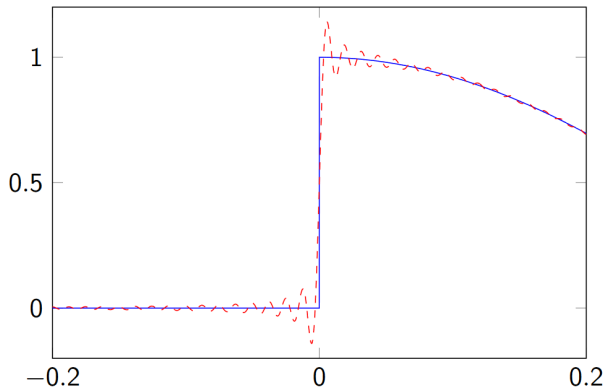
$$f(t, s) = \tilde{f}(t)\Theta(t - s) \approx \sum_{k=0}^{m-1} \sum_{\ell=0}^{m-1} \alpha_{k,\ell} p_k(t) p_\ell(s),$$

is affected by the so-called **Gibbs phenomenon**. Assuming to only know a matrix  $F$ , we may not be able to accurately reconstruct  $f(t, s)$ .

# Gibbs phenomenon

Gibbs phenomenon:

- ▶ Overshoot near discontinuity
- ▶ Spurious oscillations in whole domain



## Is it really a problem?

For  $s = -1$  the function  $f(t, -1) = \tilde{f}(t)$ , for  $t \in [-1, 1]$ . Therefore,

$$\tilde{f}(t) \approx f(t, -1) = [p_0(t) \quad p_1(t) \quad \dots \quad p_M(t)] F_M \begin{bmatrix} p_0(-1) \\ p_1(-1) \\ \vdots \\ p_M(-1) \end{bmatrix}.$$

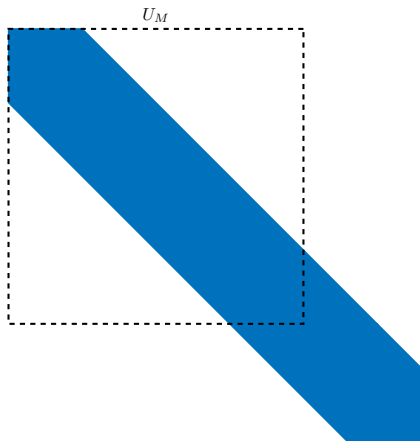
Therefore  $F_M [p_0(-1) \quad p_1(-1) \quad \dots \quad p_M(-1)]^T$  is a vector containing the Legendre coefficients of the  $1D$  expansion

$$\tilde{f}(t) = \sum_{k=0}^{m-1} \tilde{\alpha}_k p_k(t).$$

However, we still need to analyze the **truncation error**.

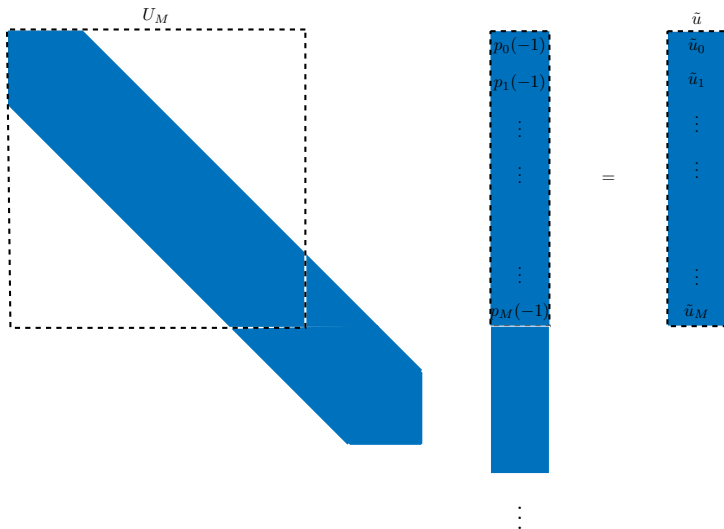
## Solution

$$u(t, s) = \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} u_{i,j} p_i(t) p_j(s) \approx \sum_{i=0}^M \sum_{j=0}^M u_{i,j} p_i(t) p_j(s)$$



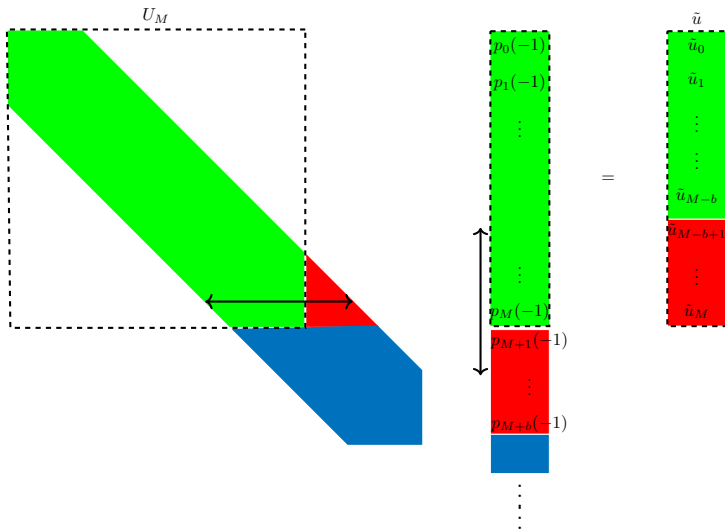
# Solution

$$\tilde{u}(t) = u(t, -1) \approx \sum_{i=0}^M \sum_{j=0}^M u_{i,j} p_i(t) p_j(-1) = \sum_{i=0}^M \tilde{u}_i p_i(t)$$



# Solution

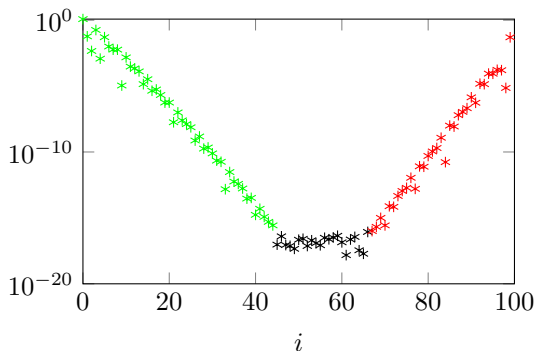
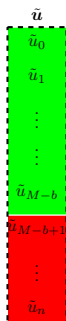
$$\tilde{u}(t) = u(t, -1) \approx \sum_{i=0}^M \sum_{j=0}^M u_{i,j} p_i(t) p_j(-1) = \sum_{i=0}^{M-b} \tilde{u}_i p_i(t)$$



# Coefficient vector $\tilde{\mathbf{u}}$

Approximate solution

$$U(t, -1) \approx \sum_{i=0}^{M-b} \tilde{u}_i p_i(t), \quad \tilde{\mathbf{u}} = U_M p, \quad \text{with } b \approx 40$$

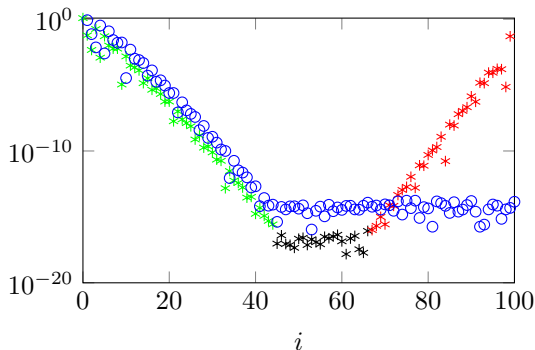
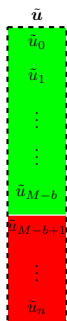




# Coefficient vector $\tilde{\mathbf{u}}$

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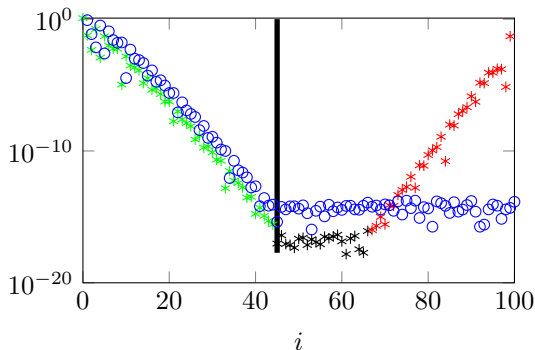
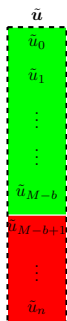
Analytic solution:

$$u(t) = \exp\left(\frac{1}{4} \sin(4t)\right) = \sum_{i=0}^{\infty} \alpha_i p_i(t)$$

# Coefficient vector $\tilde{\mathbf{u}}$

Approximate solution

$$U(t, -1) \approx \sum_{i=0}^{M-b} \tilde{u}_i p_i(t), \quad \tilde{\mathbf{u}} = U_M p, \quad \text{with } b \approx 40$$



Analytic solution:

$$u(t) = \exp\left(\frac{1}{4} \sin(4t)\right) = \sum_{i=0}^{\infty} \alpha_i p_i(t)$$

Error:

$$\|\hat{u}(t) - u(t)\|_{\infty} = 8e - 16$$

# Proposed procedure

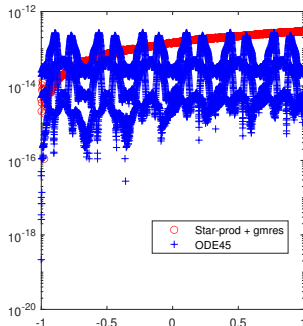
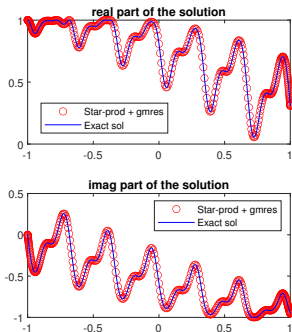
Solve  $\frac{d}{dt}\tilde{u}(t) = \tilde{f}(t)\tilde{u}(t)$ :

1. Represent  $\tilde{f}(t)\Theta(t-s)$  as Legendre series  $\Rightarrow F_M$
2. Compute  $\tilde{\mathbf{u}}_M = H_M(I_M - F_M)^{-1}p$  (GMRES!)
3. Choose right amount of coefficients to keep  $\tilde{\mathbf{u}}_{M-b}$
4. Solution is  $\tilde{u}(t) \approx \sum_{i=0}^{M-b} \tilde{u}_i p_i(t)$

## Example

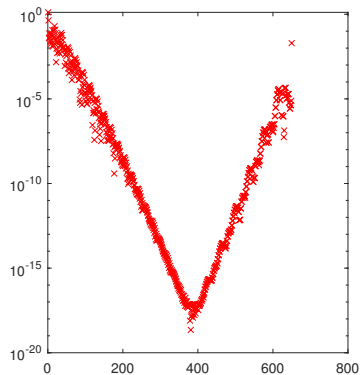
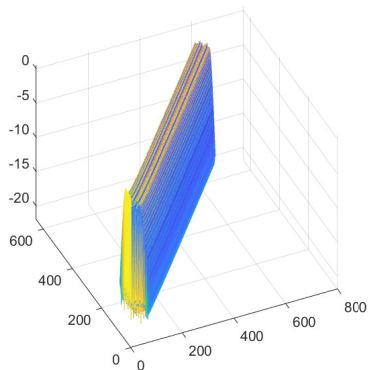
$$\frac{d}{dt}u(t) = \tilde{f}(t)u(t), \quad u(-1) = 1, \quad t \in [-1, 1]$$

$$\tilde{f}(t) = -2\pi i(0.1 + \cos(6\pi(t + 1)) + \cos(12\pi(t + 1)))$$



GMRES: 13 iterations for a  $800 \times 800$  matrix.

# Example



Left: Element magnitude of the Legendre coefficient matrix  $F$ .

Right: Element magnitude of the computed Legendre coefficients  $c$ .

# Proposed procedure

Compute  $w^H \tilde{U}(t)v$ , with  $\frac{d}{dt} \tilde{U}(t) = \tilde{A}(t) \tilde{U}(t)$ :

1. Represent  $\tilde{A}(t)\Theta(t-s)$  as Legendre series  $\Rightarrow \mathcal{F}_n$  block matrix  $[F_{i,j}]_{i,j=1}^m$

$$\begin{bmatrix} \tilde{a}_{1,1}(t) & \dots & \tilde{a}_{1,m}(t) \\ \vdots & & \vdots \\ \tilde{a}_{m,1}(t) & \dots & \tilde{a}_{m,m}(t) \end{bmatrix} \Theta(t-s) \xrightarrow{\text{discretize}} \begin{bmatrix} \boxed{F_{1,1}} & \dots & \boxed{F_{1,m}} \\ \vdots & & \vdots \\ \boxed{F_{m,1}} & \dots & \boxed{F_{m,n}} \end{bmatrix}$$

2. Compute  $w^H \tilde{U}_M v = (w \otimes I_M)^H \mathcal{H}_M (\mathcal{I}_M - \mathcal{F}_M)^{-1} (v \otimes I_n) p$  (GMRES!)
3. Choose right amount of coefficients to keep  $\tilde{u}_{M-b}$
4. Solution is  $w^H \tilde{U}(t)v \approx \sum_{i=0}^{M-b} \tilde{u}_i p_i(t)$

# Kronecker structure

In many applications, the function  $\tilde{A}(t)$  is given as a sum of Kronecker products:

$$\tilde{A}(t) = \sum_{j=0}^s A_j \otimes \tilde{f}_j(t),$$

with  $A_j$  sparse matrices, and  $\tilde{f}_j(t)$  analytic functions. Our approach reformulates the problem as the linear system

$$(Id - \sum_{j=0}^s A_j \otimes F_j) \text{vec}(X) = v \otimes \varphi_m(-1),$$

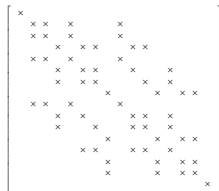
with  $F_j$  the Legendre discretization matrices, or, equivalently, as the **matrix equation** with a low-rank rhs

$$X - \sum_{j=0}^s F_j X A_j^T = \varphi_m(-1) v^T.$$

# Experiment 1 [Baligács, Bonhomme]

Setup:

- $\ell = 4$  spins  $\rightarrow 16 \times 16$
- homonuclear dipolar coupling
- Magic angle spinning  $\nu \in [5e3, 120e3]$
- $H(t) = D + B \cos(2\pi\nu t) + C \cos(4\pi\nu t)$
- time  $t \in [0, 1e-4]$  (practical:  $\mathcal{O}(e-2)$ )
- Structure  $B$  and  $C$



Compute:

- bilinear form  $w^H \tilde{U}(t)v$
- Initial state  $v = [1 \ \dots \ 1]^T \in \mathbb{C}^m$
- Measurement  $w = [1 \ \dots \ 1]^T \in \mathbb{C}^m$

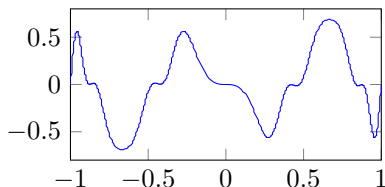
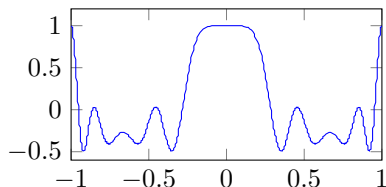


# Experiment 1: numerical

Numerical:

- $\tilde{\mathbf{u}} := U_n \mathbf{p} = H_n (I_n - F_n)^{-1} \mathbf{p} \rightarrow$  GMRES
- Reference solution  $\hat{u}$ : ode45 in Matlab
- Error:  $\text{err} := \|\hat{u}(t) - \sum_{i=0}^n \tilde{u} p_i(t)\|_\infty$

Solution  $\hat{u}(t)$



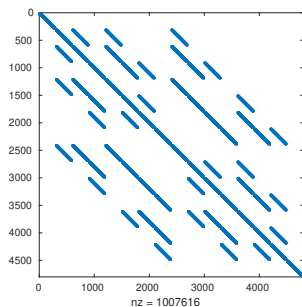
## Result $M = 300$ & cutoff at $k = 137$

Approximation for  $\tilde{u}(t) \approx \sum_{i=0}^k \tilde{u}_i p_i(t)$

Solve system  $(I_M - F_M)y = p$

Multiplication  $\tilde{u} = H_M y$

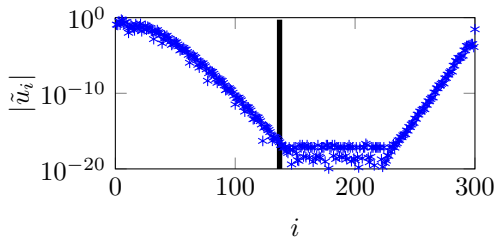
Size  $nM \times nM$ ,  $nM = 4800$



GMRES

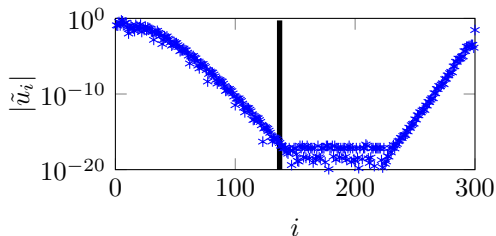
44 iterations

cutoff at  $k = 137$

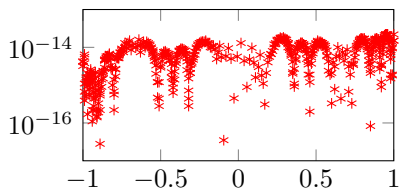
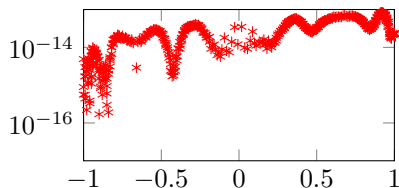


# Quality of approximation $M = 300$ , cutoff $k = 137$

Coefficients:



For  $\sum_{i=0}^{137} \tilde{u}_i p_i(t)$ : err =  $4.1e - 14$



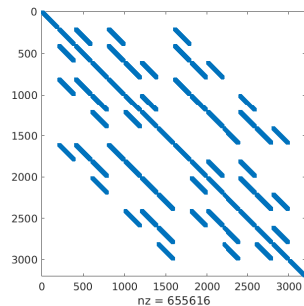
## Result $M = 200$ & cutoff at $k = 137$

Approximation for  $\tilde{u}(t) \approx \sum_{i=0}^k \tilde{u}_i p_i(t)$

Solve system  $(I_M - F_M)y = p$

Multiplication  $\tilde{u} = H_M y$

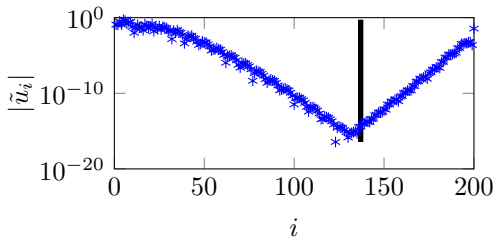
Size  $nM \times nM$ ,  $nM = 3200$



GMRES

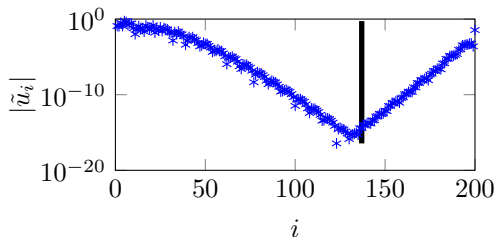
44 iterations

cutoff at  $k = 137$

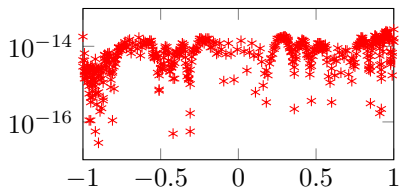
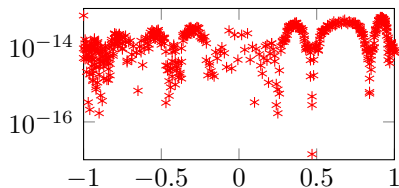


# Quality of approximation $M = 200$ , cutoff $k = 137$

Coefficients:

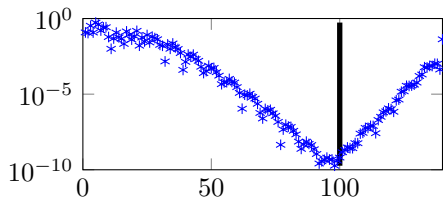


For  $\sum_{i=0}^{137} \tilde{u}_i p_i(t)$ :  $\text{err} = 4.7e - 14$

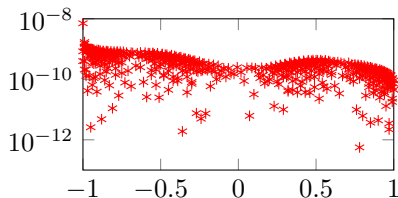
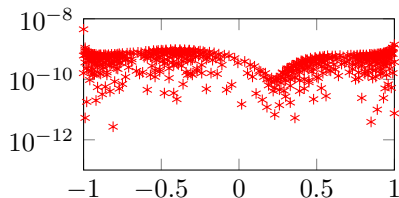


# Quality of approximation $M = 140$ , cutoff $k = 100$

Coefficients:



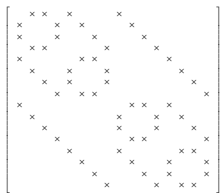
For  $\sum_{i=0}^{100} \tilde{u}_i p_i(t)$ :  $\text{err} = 7.1e - 9$



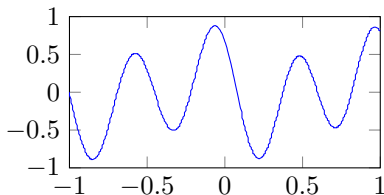
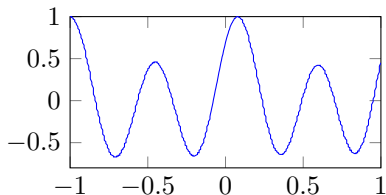
## Experiment 2 [Baligács, Bonhomme]

Setup:

- $\ell = 4$  spins  $\rightarrow 16 \times 16$
- uncoupled spins under a pulse wave
- $H(t) = A + B(0.5 + \cos(4t) + \sin(10t) - 0.4 \sin(16t)) + C(\sin(4t) + \cos(8t) + 2 \sin(12t))$
- time  $t \in [0, 1e-2]$
- Structure  $B$  and  $C$



Solution  $\hat{u}(t)$



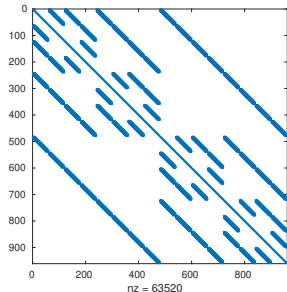
## Result $M = 60$ & cutoff at $k = 44$

Approximation for  $\tilde{u}(t) \approx \sum_{i=0}^k \tilde{u}_i p_i(t)$

Solve system  $(I_M - F_M)y = p$

Multiplication  $\tilde{u} = H_M y$

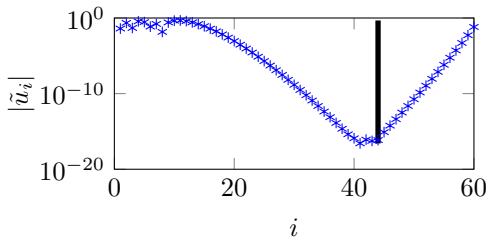
Size  $nM \times nM$ ,  $nM = 3200$



GMRES

59 iterations

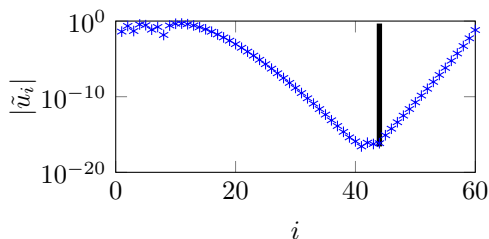
cutoff at  $i = 44$



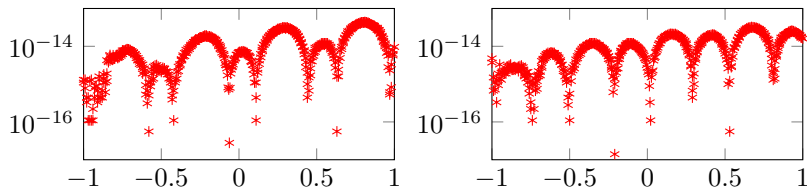


# Quality of approximation $M = 60$ , cutoff $k = 44$

Coefficients:



For  $\sum_{i=0}^{44} \tilde{u}_i p_i(t)$ : err =  $4.4e - 14$



# Recap

ODE:  $\frac{d}{dt}U(t) = A(t)U(t)$

- New expression for  $U(t) = U(t, -1)$ :  
 $U(t, s) = \Theta(t - s) \star (1_\star - A(t)\Theta(t - s))^{\star-1}$
- Legendre series:  $\tilde{\mathbf{u}} = U_M p$ , with  $U_M = H_M(I - F_M)^{-1}$
- Bandedness  $\Rightarrow$  recover decay of coefficients
- Efficient if:
  - o fast solution linear system (Krylov + preconditioner!)
  - o good estimation of  $M$

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Future work

# References

- S. P., N. V-B, paper in preparation
- P-L. Giscard, S. Pozza., *Lanczos-like method for the time-ordered exponential*, arXiv:1909.03437 [math.NA].
- P-L. Giscard, S. Pozza., *Lanczos-like algorithm for the time-ordered exponential: The  $*$ -inverse problem*, Applications of Mathematics, 2020.
- P-L. Giscard, S. Pozza, *Tridiagonalization of systems of coupled linear differential equations with variable coefficients by a Lanczos-like method*, Linear Algebra and its Applications, 2020.
- E. Baligács and C. Bonhomme, [github.com/BaligacsEni/TOMEexamples.git](https://github.com/BaligacsEni/TOMEexamples.git), 2022.

## Projects

- Charles University PRIMUS research project: *A Lanczos-like Method for the Time-Ordered Exponential*, [www.starlanczos.cz](http://www.starlanczos.cz).
- French ANR research project: *MAGICA (MAGnetic resonance techniques and Innovative Combinatorial Algebra)*.

Thank you for your attention!