

Sparse spectral methods for fractional PDEs

ICIAM 2023: CT048



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José Carrillo²



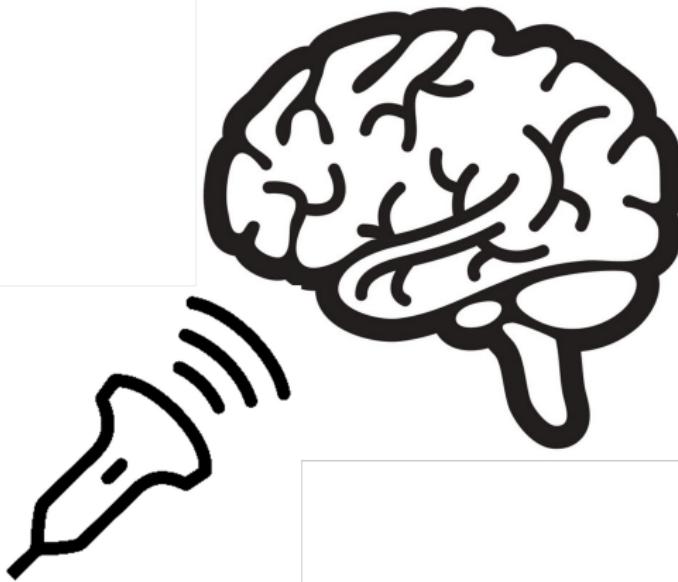
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Are fractional PDEs physical?



FPDEs describe wave absorption in the brain¹.

¹Images from <https://clipart.world/brain-clipart/black-and-white-brain-clipart/>,
https://www.kindpng.com/imgv/iRoiRR_sound-wave-clipart-ultrasound-ultrasound-clip-art-hd/.

Other applications?

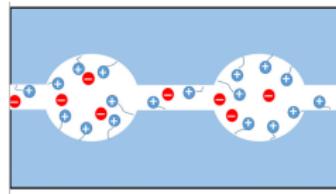
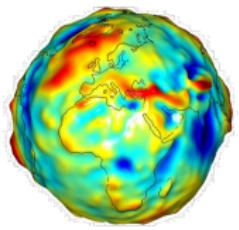
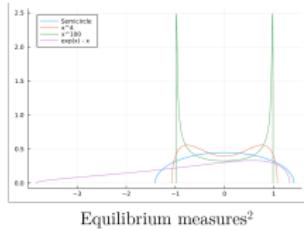
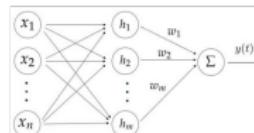


Image denoising⁴



Fully-layered connected neural networks⁶

² EquilibriumMeasures.jl

³ Alabi, Adetunji, et al. npj Clean Water 1.1 (2018): 10.

⁴ Ren, Zemin, Chuanjiang He, and Qifeng Zhang. Signal Processing 93.9 (2013): 2408-2421.

⁵ <https://planetary-science.org/planetary-science-3/geophysics/>

⁶ Zhang, Xuefeng, and Wenkai Huang. Fractal and Fractional 4.4 (2020): 50.

Observation

Solutions of fractional PDEs are “nonlocal” and may exhibit singularities.

Consequence

The solutions can be difficult to approximate numerically.

Challenge

How do we compute them with fast convergence?

Our proposal

A spectral method based on a so-called sum space.

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The problem

The PDE

Find $u \in H^s(\mathbb{R})$, $s \in (0, 1)$, that satisfies, for $\lambda \in \mathbb{R}$:

$$(\lambda I + (-\Delta)^s)u = f. \text{ (fractional Helmholtz)}$$

$H^s(\mathbb{R})$

We seek solutions u that decay sufficiently quickly as $|x| \rightarrow \infty$. In particular

$$\|u\|_{H^s(\mathbb{R})} := \left(\int_{\mathbb{R}} u^2 dx + \int_{\mathbb{R}} \int_{\mathbb{R}} \frac{|u(x) - u(y)|^2}{|x - y|^{1+2s}} dx dy \right)^{1/2} < \infty.$$

$\|\cdot\|_{H^s(\mathbb{R})}$ interpolates between $\|\cdot\|_{L^2(\mathbb{R})}$ and $\|\cdot\|_{H^1(\mathbb{R})}$.

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$(-\Delta)^s$

Ten (or more) equivalent definitions of the fractional Laplacian over \mathbb{R}^d . E.g. for $s \in (0, 1)$,

$$(-\Delta)^s u(x) := c_{d,s} \int_{\mathbb{R}^d} \frac{u(x) - u(y)}{|x - y|^{d+2s}} dy$$

or

$$\mathcal{F}[(-\Delta)^s u](\omega) = |\omega|^{2s} \mathcal{F}[u](\omega).$$

We will focus on the special case $s = 1/2$.

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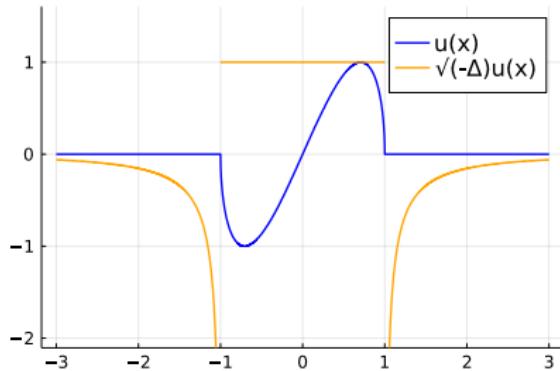
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The fractional Laplacian is not local. E.g.



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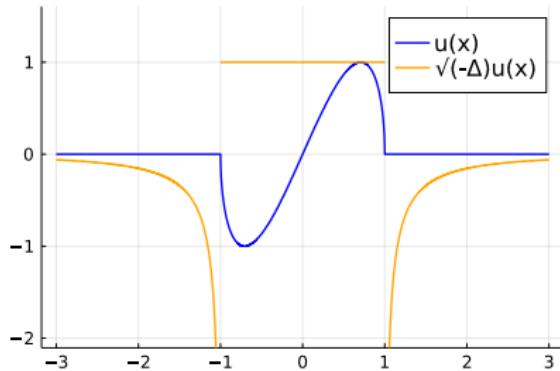
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As $x \downarrow 1$ and $x \uparrow -1$, then $|(-\Delta)^{1/2}u(x)| \rightarrow \infty$.

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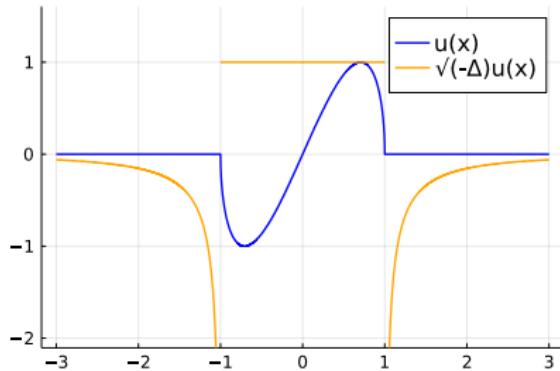
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Spectral methods

Consider the *ChebyshevT* polynomials, denoted $T_n(x)$. These satisfy

$$\int_{-1}^1 \frac{T_n(x) T_m(x)}{\sqrt{1-x^2}} dx = \delta_{nm}; \quad T_0(x) = 1, \quad T_1(x) = x, \quad T_{n+1}(x) = 2xT_n(x) - T_{n-1}(x).$$

For $x \in [-1, 1]$, consider the approximation: $e^{-x^2} \sin(x) \approx \sum_{k=0}^n f_k T_k(x)$.

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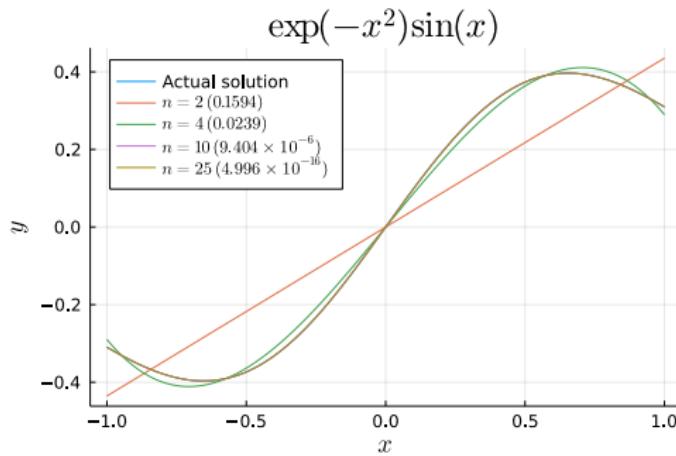
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Dense spectral methods

Many spectral methods for differential equations induce *dense* matrices $\textcolor{red}{X}$. Consider solving, on $[-1, 1]$,

$$-u'(x) = f(x), \quad u(-1) = 0.$$

A spectral method recipe

- ① Expand $f(x)$ in the ChebyshevT polynomial basis, truncate, and collect the coefficients in vector \mathbf{f} .
- ② Construct the derivative matrix D via a collocation method. D is **dense**.
- ③ Solve $D\mathbf{u} = \mathbf{f}$ for the coefficients \mathbf{u} in the ChebyshevT expansion of $u(x)$.

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Sparse spectral methods

Let $\{U_n\}$ denote the ChebyshevU polynomials ortho. to $\sqrt{1 - x^2}$.

An observation

For $n \geq 1$, $T'_n(x) = nU_{n-1}(x)$. Or in *quasimatrix* notation:

$$(T'_0(x) \ T'_1(x) \ T'_2(x) \ \dots) \begin{pmatrix} 0 & 1 & & \\ & 2 & & \\ & & \ddots & \end{pmatrix} = (U_0(x) \ U_1(x) \ U_2(x) \ \dots)$$

A sparse spectral method recipe (generalizable to ODEs)

- ➊ Expand $f(x)$ in the ChebyshevU polynomial basis, truncate, and collect the coefficients in vector \mathbf{f} .
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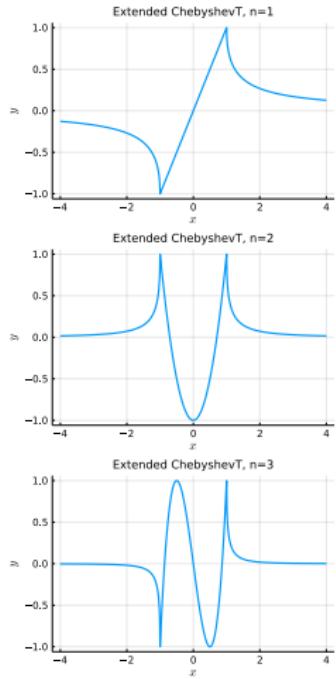
Extended Chebyshev functions

For $n \geq 1$,

$$\tilde{T}_n(x) := \begin{cases} T_n(x) & |x| \leq 1, \\ (x - \operatorname{sgn}(x)\sqrt{x^2 - 1})^n & |x| > 1. \end{cases}$$

$$\tilde{U}_n(x) := \begin{cases} U_n(x) & |x| \leq 1, \\ 2\tilde{T}_n(x) + \tilde{U}_{n-2}(x) & |x| > 1. \end{cases}$$

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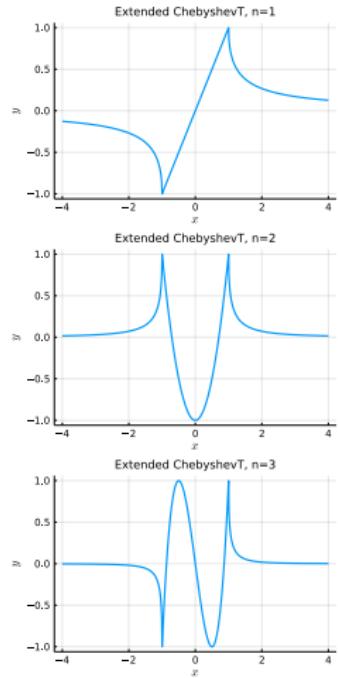
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A sparse spectral method for an FPDE

$$W_n(x) := (1 - x^2)_+^{1/2} U_n(x), \quad V_n(x) := (1 - x^2)_+^{-1/2} T_n(x).$$

$$(-\Delta)^{1/2}$$

$$(-\Delta)^{1/2} W_n(x) = (n+1) \tilde{U}_n(x),$$
$$(-\Delta)^{1/2} \tilde{T}_n(x) = n V_n(x).$$

Identity

$$W_n(x) = \frac{1}{2}[V_n(x) - V_{n+2}(x)],$$
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Observation: The relationships are banded!

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A sparse spectral method for an FPDE

Key idea: use the sum space $\{\tilde{T}_n\} \cup \{W_n\}$.

$$\lambda\mathcal{I} + (-\Delta)^{1/2}$$

$$\underbrace{\{\tilde{T}_n\} \cup \{W_n\}}_{\text{sum space, } S} \xrightarrow{\lambda\mathcal{I} + (-\Delta)^{1/2}} \underbrace{\{\tilde{U}_n\} \cup \{V_n\}}_{\text{dual sum space, } S^*}.$$

A sparse spectral method recipe 📋

- ① Expand f in the **dual sum space** $f(x) \approx S^*(x)\mathbf{f}$.
- ② Assemble the **sparse** matrix D induced by $(\lambda\mathcal{I} + (-\Delta)^{1/2})$.
- ③ Solve $D\mathbf{u} = \mathbf{f}$ for the coefficients \mathbf{u} , $u(x) \approx S(x)\mathbf{u}$.

A sparse spectral method for an FPDE

Key idea: use the sum space $\{\tilde{T}_n\} \cup \{W_n\}$.

$$\lambda\mathcal{I} + (-\Delta)^{1/2}$$

$$\underbrace{\{\tilde{T}_n\} \cup \{W_n\}}_{\text{sum space, } S} \xrightarrow{\lambda\mathcal{I} + (-\Delta)^{1/2}} \underbrace{\{\tilde{U}_n\} \cup \{V_n\}}_{\text{dual sum space, } S^*}.$$

A sparse spectral method recipe 📋

- ① Expand f in the dual sum space $f(x) \approx S^*(x)\mathbf{f}$.
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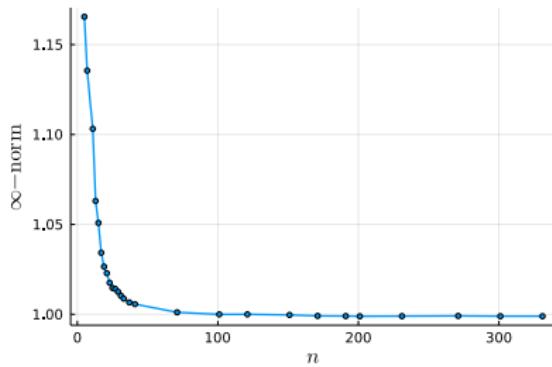
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Expansion of the right-hand side f

- Only expand f in $V_n^i(x)$ or W_n^i via the DCT.
- Solve a least squares collocation problem via a truncated SVD. [Backed by *frame theory*].



$$\| \infty\text{-norm of the coefficient vector for } (1 - 2x)e^{-x^2} - \frac{i}{x} \left(e^{-x^2} |x| \operatorname{erf}(i|x|) \right) + \frac{2}{\sqrt{\pi}} {}_1F_1(1; 1/2; -x^2)$$

Example: the Gaussian

$$(\mathcal{I} + (-\Delta)^{1/2})u(x) = e^{-x^2} + \frac{2}{\sqrt{\pi}} {}_1F_1(1; 1/2; -x^2).$$

${}_1F_1$ is the Kummer confluent hypergeometric function.

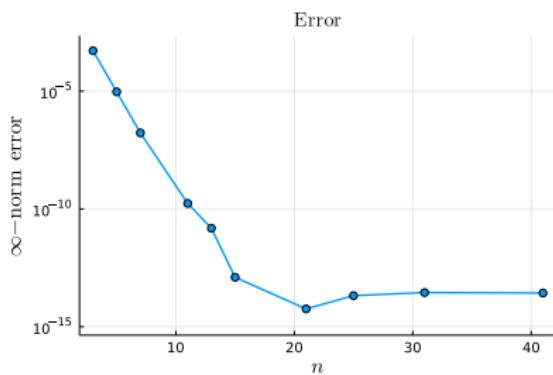
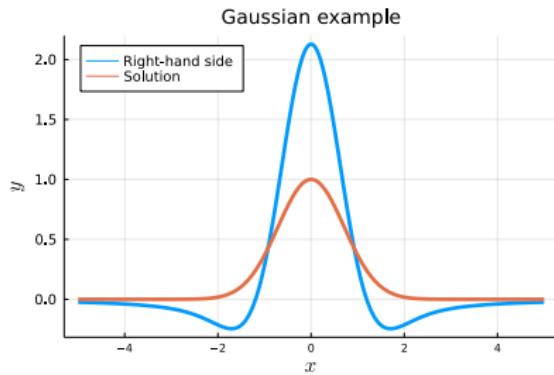
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Example: wave propagation

Consider the FPDE ($u(x, t) \rightarrow 0$ as $|x| \rightarrow \infty$):

$$[(-\Delta)^{1/2} + \mathcal{H} + \frac{\partial^2}{\partial t^2}]u(x, t) = (1 - x^2)_+^{1/2} U_4(x) e^{-t^2}.$$

A Fourier transform in time gives ($\hat{u}(x, \omega) \rightarrow 0$ as $|x| \rightarrow \infty$):

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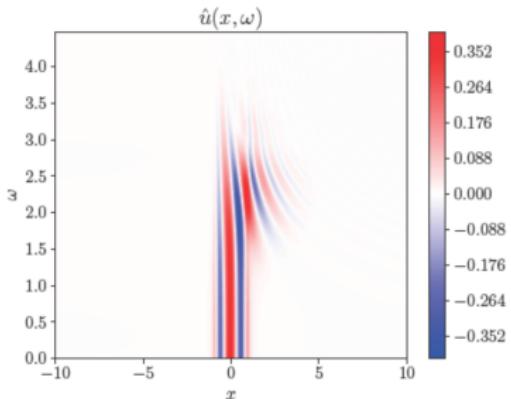
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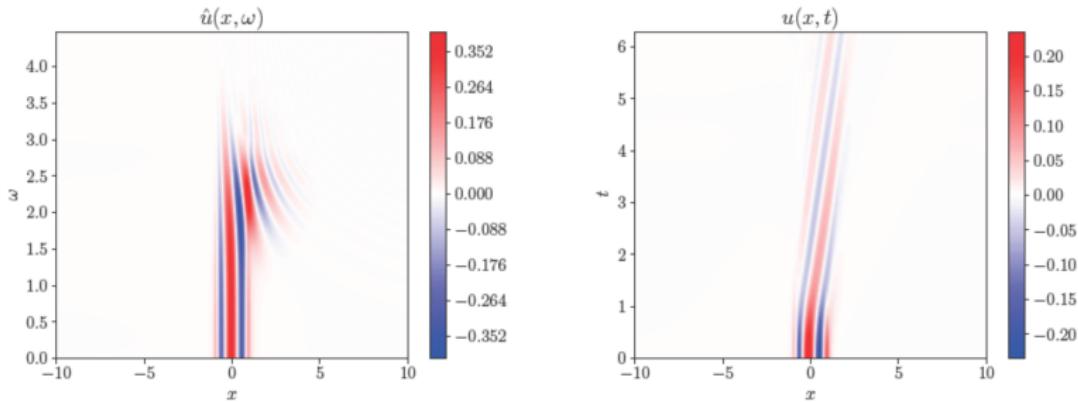
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- Based on a carefully chosen sum space;
- Implementation written in Julia • see <https://github.com/ioannisPapapadopoulos/SumSpaces.jl>.

A sparse spectral method for fractional differential equations in one-spacial dimension

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Thank you for listening!

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