

# Sparse spectral methods for fractional PDEs

ICIAM 2023: CT048



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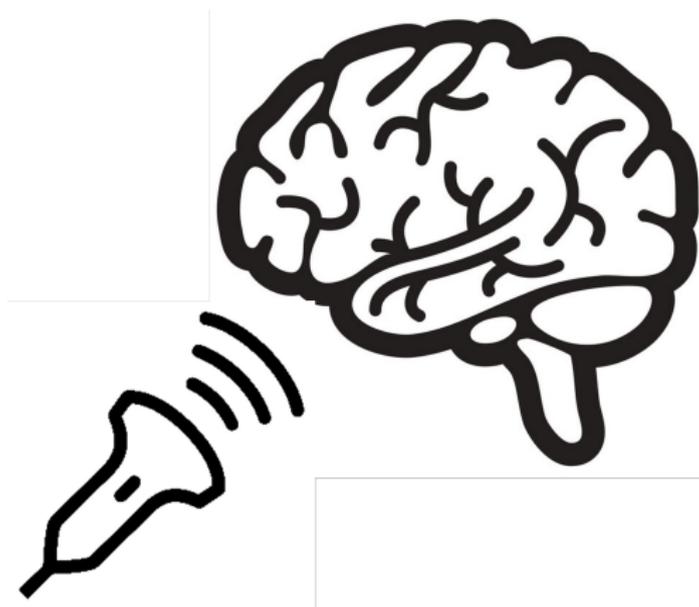
Timon Gutleb<sup>2</sup>



Bradley Treeby<sup>3</sup>

<sup>1</sup>Imperial College London; <sup>2</sup>University of Oxford; <sup>3</sup>UCL

# Are fractional PDEs physical?

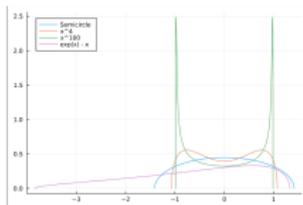


FPDEs describe wave absorption in the brain<sup>1</sup>.

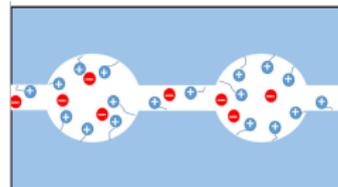
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<sup>1</sup>Images from <https://clipart.world/brain-clipart/black-and-white-brain-clipart/>,  
[https://www.kindpng.com/imgv/iRoiRR\\_sound-wave-clipart-ultrasound-ultrasound-clip-art-hd/](https://www.kindpng.com/imgv/iRoiRR_sound-wave-clipart-ultrasound-ultrasound-clip-art-hd/).

# Other applications?



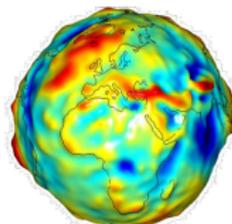
Equilibrium measures<sup>2</sup>



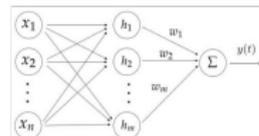
Dispersive transport of ions<sup>4</sup>



Image denoising<sup>4</sup>



Long-range geophysical effects<sup>5</sup>



Fully-layer connected neural networks<sup>6</sup>

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<sup>2</sup>EquilibriumMeasures.jl

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

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

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

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

# Fractional PDEs

## 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  $\lambda \in \mathbb{R}$ :

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

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

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  $s \in (0, 1)$ ,

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

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

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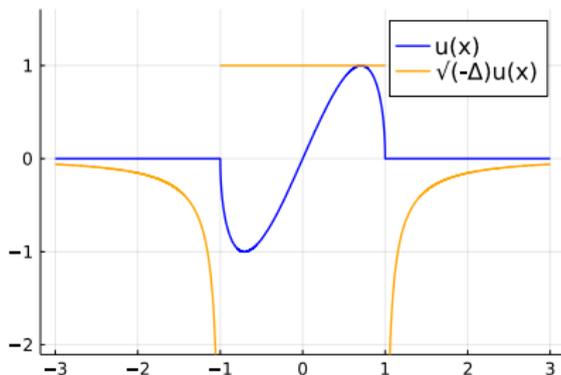
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The fractional Laplacian is not local. E.g.



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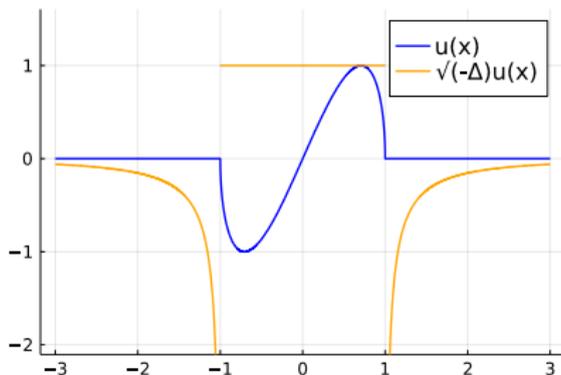
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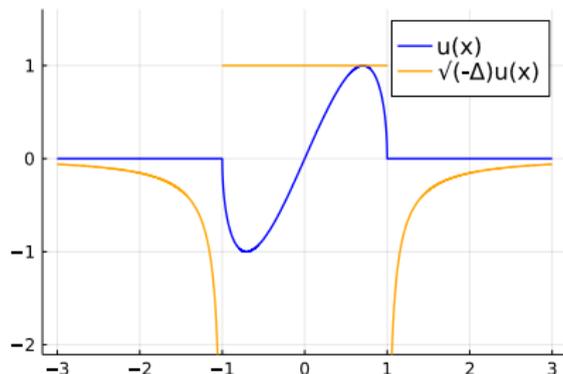
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## Spectral methods

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

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

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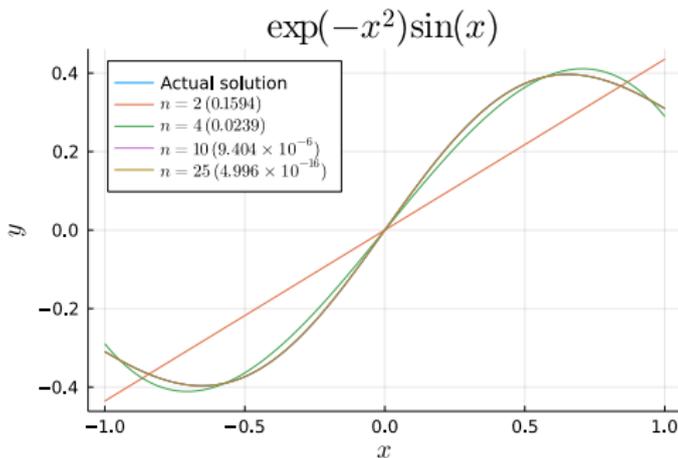
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## Dense spectral methods

Many spectral methods for differential equations induce *dense* matrices **×**. 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**.
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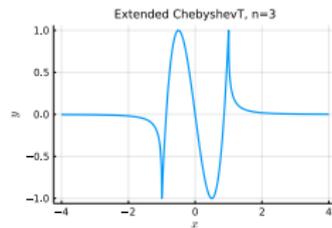
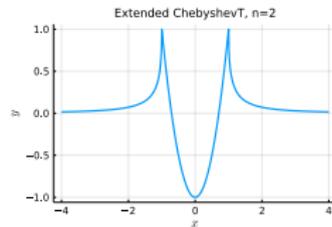
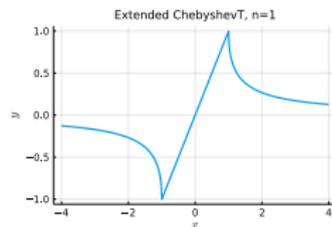
## Extended Chebyshev functions

For  $n \geq 1$ ,

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

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

$$\text{where } \tilde{U}_{-1}(x) := \begin{cases} 0 & |x| \leq 1, \\ -\frac{\operatorname{sgn}(x)}{\sqrt{x^2 - 1}} & |x| > 1, \end{cases}$$



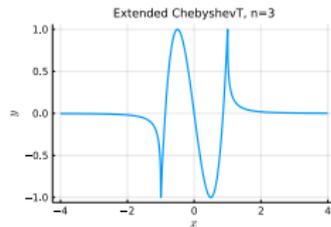
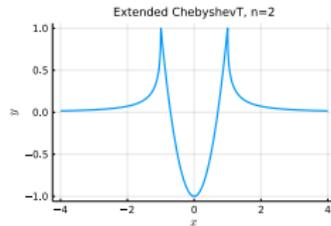
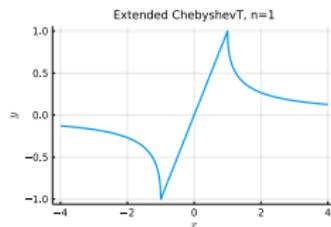
## Extended Chebyshev functions

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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## A sparse spectral method for an FPDE

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

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

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## Identity

$$W_n(x) = \frac{1}{2} [V_n(x) - V_{n+2}(x)],$$

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

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A sparse spectral method recipe 

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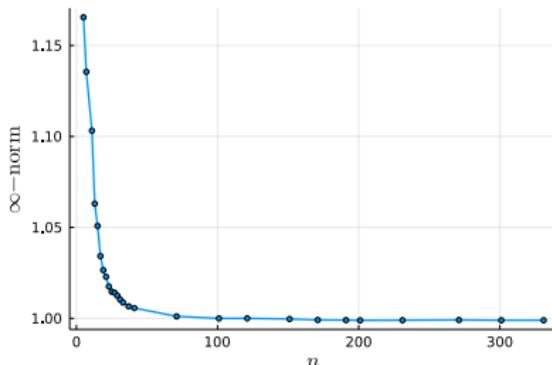
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Expansion of the right-hand side  $f$ 

- 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].



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

## 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.

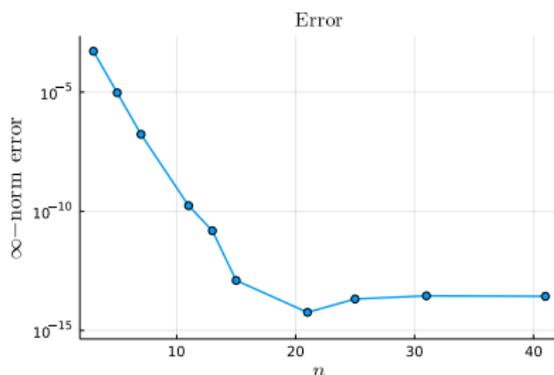
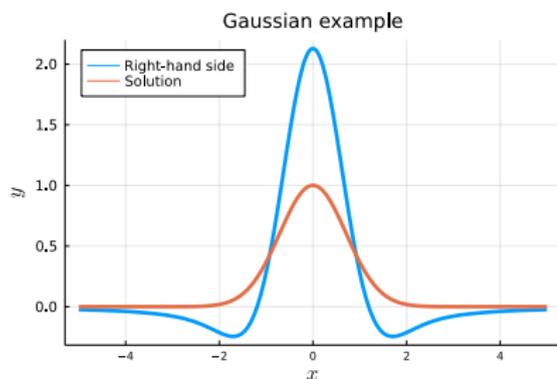
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$$[(-\Delta)^{1/2} + \mathcal{H} + \frac{\partial^2}{\partial t^2}]u(x, t) = (1 - x^2)_+^{1/2} U_4(x) e^{-t^2}.$$

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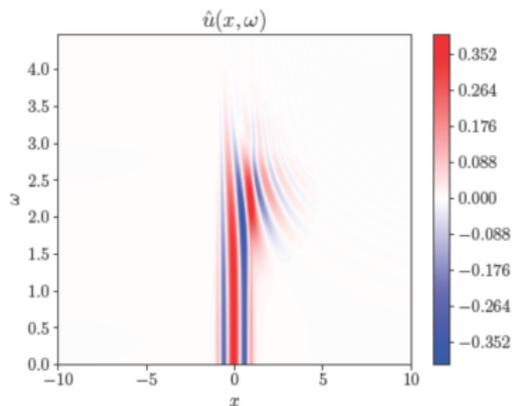
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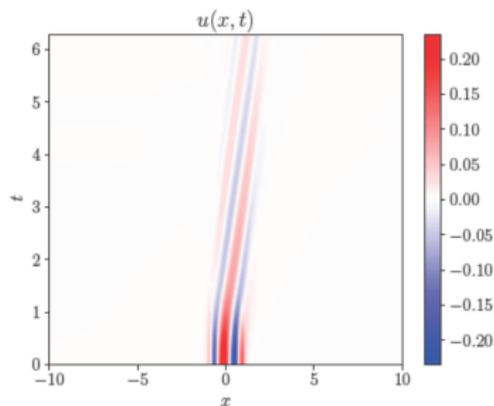
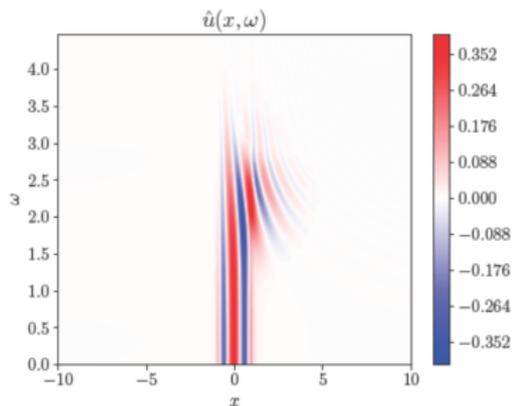
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- A sparse spectral method for solving the identity + sqrt-Laplacian;
- Based on a carefully chosen sum space;
- Implementation written in Julia  see <https://github.com/ioannisPApapadopoulos/SumSpaces.jl>.

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

I.P., S. Olver, 2022, arXiv preprint arXiv:2210.08247

### Ongoing work

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# Thank you for listening!

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