

A PDE approach to spectral fractional diffusion

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Outline

Motivation: Fractional powers of an operator

Direct discretization approach

Best uniform rational approximation

The Balakrishnan formula

The Caffarelli-Silvestre extension

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Back to linear algebra I

- If A is symmetric, that is

$$A^T = A,$$

then it is **diagonalizable**.

- This means that there are Q orthogonal, and Λ diagonal, such that

$$A = Q^T \Lambda Q, \quad Q^T = Q^{-1}, \quad \Lambda = \text{diag}\{\lambda_1, \dots, \lambda_n\}.$$

- In this case, the action $\mathbf{w} = A\mathbf{v}$ can be described as follows:
 - $\tilde{\mathbf{v}} = Q\mathbf{v}$ is a change of basis.
 - $\tilde{\mathbf{v}} = \Lambda\tilde{\mathbf{v}}$ is a scaling in this new basis.
 - $\mathbf{w} = Q^T\Lambda\tilde{\mathbf{v}}$ is returning to the original basis.
- If, in addition, A is positive, that is

$$\mathbf{v}^T A \mathbf{v} > 0,$$

then *all its eigenvalues are positive* $\lambda_i > 0$.

Back to linear algebra II

Why do we care about this? If $A \in \mathbb{R}^{n \times n}$ is symmetric:

- With this we can define almost any **function of a matrix** via

$$f(A) = Q^T f(\Lambda) Q, \quad f(\Lambda) = \text{diag}\{f(\lambda_1), \dots, f(\lambda_n)\}.$$

- **Solution of ODEs:**

$$\dot{\mathbf{y}}(t) = A\mathbf{y}, \quad t > 0 \quad \mathbf{y}(0) = \mathbf{y}_0 \quad \implies \quad \mathbf{y}(t) = \exp(tA)\mathbf{y}_0.$$

- **Theory of iterative schemes:** To solve $A\mathbf{x} = \mathbf{f}$ we can use a two-layer implicit scheme

$$B \frac{\mathbf{x}^{k+1} - \mathbf{x}^k}{\alpha} + A\mathbf{x}^k = \mathbf{f}$$

with SPD preconditioner B . The analysis of such schemes can be reduced to that of the explicit one

$$\frac{\mathbf{v}^{k+1} - \mathbf{v}^k}{\alpha} + C\mathbf{v}^k = \mathbf{g}$$

where

$$\mathbf{v}^k = B^{1/2}\mathbf{x}^k, \quad C = B^{-1/2}AB^{-1/2}, \quad \mathbf{g} = B^{-1/2}\mathbf{f}.$$

- ...

Spectral theory 101

Question: What happens in infinite dimensions? In particular, for differential operators?

A (the?) basic partial differential operator that expresses **diffusion** is the *Laplacian*

$$-\Delta = -\sum_{i=1}^n \frac{\partial^2}{\partial x_i^2}$$

- Integration by parts shows that $-\Delta$ is **positive**

$$\int_{\Omega} -\Delta v v \, dx = \int_{\Omega} |\nabla v|^2 \, dx > 0, \quad \forall v \in C_0^\infty(\Omega).$$

- One can show that $(-\Delta)^{-1} : L^2(\Omega) \rightarrow L^2(\Omega)$ is **compact**:
 - There exist $\{\lambda_k, \varphi_k\}_{k \in \mathbb{N}} \subset \mathbb{R}^+ \times L^2(\Omega)$ such that:

$$-\Delta \varphi_k = \lambda_k \varphi_k, \quad \varphi_k|_{\partial\Omega} = 0$$

and $\{\varphi_k\}_{k \in \mathbb{N}}$ is an orthonormal basis of $L^2(\Omega)$.

- This means that if $w \in L^2(\Omega)$, then it has the following representation

$$w = \sum_{k=1}^{\infty} w_k \varphi_k \quad w_k = \int_{\Omega} w \varphi_k \, dx.$$

The spectral fractional Laplacian I

- In addition, if w is sufficiently nice, then we have that

$$-\Delta w = \sum_{k=1}^{\infty} w_k \lambda_k \varphi_k, \quad w_k = \int_{\Omega} w \varphi_k \, dx$$

which is an analogue of the matrix case:

- The term w_k is a change of basis.
- Multiplication by the eigenvalue λ_k is a diagonal scaling.
- The outer sum is returning to the original basis.
- We can now define functions of $-\Delta$. For instance, if $s \in (0, 1)$ and w is sufficiently nice,

$$(-\Delta)^s w = \sum_{k=1}^{\infty} w_k \lambda_k^s \varphi_k,$$

Questions: Why do we care? What is the domain of this operator? What is its range?

The spectral fractional Laplacian II

- The heat equation

$$\partial_t u - \Delta u = 0, \quad u|_{t=0} = v$$

smoothens and **smears** the initial condition v . This could be used, for instance, in image denoising. However, the effect of $-\Delta$ is **too strong**. Thus, it can be weakened by

$$\partial_t u + (-\Delta)^s u = 0, \quad u|_{t=0} = v.$$

- Some special cases of random walks also lead to the fractional heat equation^[1].
- Models in phase transition^[2]: fractional Allen Cahn ($\alpha = 0$, $\beta \in (0, 1)$) and Cahn Hilliard ($\alpha, \beta \in (0, 1)$) equations


$$\partial_t u + (-\Delta)^\alpha (\varepsilon^2 (-\Delta)^\beta u + F'(u)) = 0,$$

^[1]Valdinoci 2017

^[2]Ainsworth and Mao 2017, Antil and Bartels 2018

The spectral fractional Laplacian III



- Original, noisy, regularized images for L^2 and H^{-1} fidelity terms.
- Top: $s = 0.42$
- Bottom: $s = 0.35$
- Stolen from .

Spectral theory 102

- Let \mathcal{L} be a symmetric second order elliptic operator, i.e.,

$$\mathcal{L}w = -\nabla \cdot (a \nabla w) + cw$$

with $a \in L^\infty(\Omega, \mathbb{S}_+^d)$ uniformly positive definite and $0 \leq c \in L^\infty(\Omega)$.

- In a similar way we can define \mathcal{L}_0^s , the fractional powers of \mathcal{L} supplemented with homogeneous Dirichlet (or Neumann) boundary conditions.
- From now on, **and for simplicity only**, we will only deal with the Laplacian. **Everything** that we will say applies to \mathcal{L}_0^s .

Goal

- Given a suitable f find u such that

$$(-\Delta)^s u = f$$

in the sense described above.

- Where's the catch? The domain Ω can be quite general, so the spectrum of $-\Delta$ is not readily available.

Domain, range, and regularity I

- Because of the way that we defined the fractional Laplacian we have

$$(-\Delta)^s : \mathbb{H}^s(\Omega) \rightarrow \mathbb{H}^{-s}(\Omega)$$

where

$$\mathbb{H}^s(\Omega) = \left\{ w = \sum_{k=1}^{\infty} w_k \varphi_k : \sum_{k=1}^{\infty} \lambda_k^s |w_k|^2 < \infty \right\}$$

- It turns out that

$$\mathbb{H}^s(\Omega) = \begin{cases} H^s(\Omega), & s \in (0, \frac{1}{2}), \\ H_{00}^{1/2}(\Omega), & s = \frac{1}{2}, \\ H_0^s(\Omega), & s \in (\frac{1}{2}, 1), \end{cases}$$

where the zero subindices mean “zero boundary values”.

- The fact that the domain has fractional Sobolev regularity reinforces the idea that we are taking fractional order derivatives.

Domain, range, and regularity II

If we wish to develop a rigorous numerical approximation of u , then we **must** understand its regularity.

- From the definition it follows that, if $f \in \mathbb{H}^r(\Omega)$, then $u \in \mathbb{H}^{r+2s}(\Omega)$, for all $r \in \mathbb{R}$.
- If $r \geq -s$ this means that, at least for $\omega \Subset \Omega$,

$$u \in H^{r+2s}(\omega).$$

- What about near the boundary? For $x \in \bar{\Omega}$ let $\text{dist}(x, \partial\Omega)$ be the distance of x to $\partial\Omega$:
 - If $s \neq \frac{1}{2}$ then \blacksquare there is a smooth function v such that

$$u(x) \approx v(x) + \text{dist}(x, \partial\Omega)^{\min\{1, 2s\}}$$

- If $s = \frac{1}{2}$ then we have the exceptional case \blacksquare

$$u(x) \approx v(x) + \text{dist}(x, \partial\Omega) |\log \text{dist}(x, \partial\Omega)|.$$

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Direct discretization

Given $f \in \mathbb{H}^{-s}(\Omega)$,

$$f = \sum_{k=1}^{\infty} f_k \varphi_k : \quad (-\Delta)^s u = f \implies \quad u_k = f_k \lambda_k^{-s}$$

Algorithm:

- Compute a “sufficiently large” number of eigenpairs $\{\lambda_k, \varphi_k\}_{k=1}^N$.
- Compute the Fourier coefficients f_k .
- Find the solution: $u_k = f_k \lambda_k^{-s}$.

But

- How to choose N ?
- VERY time consuming!
- Error analysis?

Error analysis I

The eigenpairs can only be computed approximately (read, via finite elements). The error analysis in this case is as follows^[1]:

- Let X be a Hilbert space and A be a positive definite self-adjoint operator on X .
- Let $\{X_h\}_{h>0}$ be a family of closed subspaces of X and A_h is a positive definite bounded self-adjoint operator on X_h .
- **Inverse estimate:** There is $\varepsilon : \mathbb{R}_+ \rightarrow \mathbb{R}_+$ with $\lim_{h \rightarrow 0} \varepsilon(h) = 0$ such that

$$\|A_h\| \lesssim \frac{1}{\varepsilon(h)}$$

- **Approximability:** If P_h is the orthogonal projection onto X_h

$$\|(A_h^{-1}P_h - A^{-1})f\|_X \lesssim \varepsilon(h)\|f\|_X$$

- In this case, for $s \in (0, 1)$, we have

$$\|(A_h^{-s}P_h - A^{-s})f\|_X \lesssim \varepsilon(h)^s\|f\|_X$$

Error analysis II

In our case:

- $X = L^2(\Omega)$, X_h is a (piecewise linear) finite element space, $A = -\Delta$, and $A_h = -\Delta_h$.
- Since X_h consists of piecewise polynomials

$$\|A_h\| \lesssim \frac{1}{h^2}, \quad \implies \quad \varepsilon(h) = h^2.$$

- For $f \in L^2(\Omega)$ we have

$$u = (-\Delta)^{-1} f \in H^2(\Omega) \cap H_0^1(\Omega)$$

and, if $u_h \in X_h$ is its finite element approximation:

$u_h = (-\Delta_h)^{-1} P_h f$, then Aubin–Nitsche duality yields

$$\|u - u_h\|_{L^2(\Omega)} \lesssim h^2 |u|_{H^2(\Omega)} \lesssim h^2 \|f\|_{L^2(\Omega)}$$

- The previous theory then gives

$$\|(-\Delta)^{-s} f - (-\Delta_h)^{-s} P_h f\|_{L^2(\Omega)} \lesssim h^{2s} \|f\|_{L^2(\Omega)}.$$

We still need to compute $(-\Delta_h)^{-s}$!

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Computing the discrete spectrum

Evaluating the eigenvalues of $-\Delta_h$ is **time consuming**: MTT, Lanczos, ...
Best uniform rational approximation (BURA)[■]: Assume we need to solve

$$\mathcal{A}^s \mathbf{u} = \mathbf{f}$$

where \mathcal{A} is a rescaled version of $(-\Delta_h)^s$ so that its spectrum lies in $(0, 1]$.

- Let r_s be analytic on $(0, 1]$ and, for some constant $\varepsilon > 0$ satisfies

$$\sup_{t \in (0, 1]} |r_s(t) - t^{1-s}| \leq \varepsilon,$$

then, for every $\gamma \in \mathbb{R}$ and $\mathbf{F} \in \mathbb{R}^N$ we have

$$\|(r_s(\mathcal{A}) - \mathcal{A}^{1-s})\mathbf{F}\|_{\mathcal{A}^\gamma} \leq \varepsilon \|\mathbf{F}\|_{\mathcal{A}^\gamma}$$

- The previous result implies that, if $\mathbf{u}_r = r_s(\mathcal{A})\mathcal{A}^{-1}\mathbf{f}$, then

$$\|\mathbf{u}_r - \mathbf{u}\|_{\mathcal{A}^\gamma} \leq \varepsilon \|\mathbf{f}\|_{\mathcal{A}^{-1}}$$

- Taking into account the discretization error, then ($\gamma = 0$)

$$\|u - u_{h,r}\|_{L^2(\Omega)} \lesssim h^{2s} + \varepsilon.$$

- **Question:** What is a suitable r_s ?

BURA

- We choose r_s as the **best uniform** (m, k) -approximation to t^{1-s} .
- Apply a partial fraction decomposition to $t^{-1}r_s(t)$:

$$t^{-1}r_s(t) = \sum_{j=0}^{m-k-1} b_j t^j + \frac{c_0}{t} + \sum_{j=1}^{p_1} \frac{c_j}{t - d_j} + \sum_{j=1}^{p_2} \frac{B_j t + C_j}{(t - F_j)^2 + D_j^2}$$

where $k = p_1 + 2p_2$.

- To compute $\mathbf{u}_r = \mathcal{A}^{-1}r_s(\mathcal{A})\mathbf{f}$ we need to evaluate

$$\begin{aligned} \mathcal{A}^{-1}r_s(\mathcal{A})\mathbf{f} &= \sum_{j=0}^{m-k-1} b_j \mathcal{A}^j \mathbf{f} + c_0 \mathcal{A}^{-1} \mathbf{f} + \sum_{j=1}^{p_1} c_j (\mathcal{A} - d_j \mathcal{I})^{-1} \mathbf{f} \\ &\quad + \sum_{j=1}^{p_2} (B_j \mathcal{A} + C_j \mathcal{I}) ((\mathcal{A} - F_j \mathcal{I})^2 + D_j^2 \mathcal{I})^{-1} \mathbf{f} \end{aligned}$$

- How do we choose m and k ? This is **classical** in rational approximation. For the optimal choice we have $m = k$ and

$$\varepsilon \lesssim 4^{2-s} |\sin \pi(1-s)| e^{-2\pi\sqrt{(1-s)k}}$$

so that, for this choice, the error decays **exponentially** in the polynomial degree.

Outlook

To solve

$$(-\Delta)^s u = f$$

with BURA we must:

- Solve $\mathcal{O}(|\log h|)$ problems of the type $(-\Delta_h + c\mathcal{I})\mathbf{w} = \mathbf{g}$.
- Embarrassingly parallelizable.
- Error estimate

$$\|u - u_{h,r}\|_{L^2(\Omega)} \lesssim h^{2s}.$$

Questions:

- Other norms?
- Other types of problems? Time-dependent? Nonlinear?
- Stability? It is known that rational approximations are very sensitive to numerical rounding.

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The Balakrishnan formula

- Notice that, for $\lambda > 0$ and $\theta \in (0, 1)$

$$\frac{\sin \pi \theta}{\pi} \int_0^\infty t^{\theta-1} (\lambda + t)^{-1} dt = \lambda^{\theta-1}.$$

- Functional calculus then says that, if X is a Hilbert space and A is a self-adjoint and positive operator on X :

$$A^\theta = AA^{\theta-1} = A \frac{\sin \pi \theta}{\pi} \int_0^\infty t^{\theta-1} (A + t\mathcal{I})^{-1} dt.$$

- Let $X = L^2(\Omega)$ and $A = -\Delta$, then

$$\begin{aligned} (-\Delta)^{-s} &= (-\Delta)^{-1} (-\Delta)^{1-s} \\ &= (-\Delta)^{-1} (-\Delta) \frac{\sin \pi(1-s)}{\pi} \int_0^\infty t^{1-s-1} (t\mathcal{I} - \Delta)^{-1} dt \\ &= \frac{\sin \pi s}{\pi} \int_0^\infty t^{-s} (t\mathcal{I} - \Delta)^{-1} dt \end{aligned}$$

where we used the previous formula with $\theta = 1 - s$.

Numerical scheme

Using

$$(-\Delta)^{-s} = \frac{\sin \pi \theta}{\pi} \int_0^\infty t^{-s} (t\mathcal{I} - \Delta)^{-1} dt,$$

we can formulate the following game plan to devise a numerical scheme^[1]:

- **Step 1:** Use a quadrature for the t variable:

$$(-\Delta)^{-s} f \approx \frac{\sin \pi s}{\pi} k \sum_{j=0}^J t_j^{-s} (t_j \mathcal{I} - \Delta)^{-1} f$$

- **Step 2:** Use standard finite element methods on the **same mesh** to approximate

$$w_j \in H_0^1(\Omega) : \quad t_j w_j - \Delta w_j = f \quad \text{in } \Omega,$$

i.e., $w_j = (t_j \mathcal{I} - \Delta)^{-1} f$.

- **Step 3:** Gather all contributions.

Step 1: Sinc quadrature

- **Change of variable:** Let $t = e^y$ to get

$$u = (-\Delta)^{-s} f = \frac{\sin(\pi s)}{\pi} \int_{-\infty}^{\infty} e^{(1-s)y} (e^y I - \Delta)^{-1} f \, dy.$$

- **Quadrature:** Given $N \in \mathbb{N}$, define $k = 1/\sqrt{N}$, $y_j = jk$ and the quadrature approximation

$$u^N = Q^N f = \frac{\sin(\pi s)}{\pi} k \sum_{j=-N}^N e^{(1-s)y_j} (e^{y_j} I - \Delta)^{-1} f.$$

- **Exponential convergence:** Let $s \in [0, 1)$ and $r \in [0, 1]$. If $f \in \mathbb{H}^r(\Omega)$, then

$$\|u - u^N\|_{\mathbb{H}^r(\Omega)} \lesssim e^{-c\sqrt{N}} \|f\|_{\mathbb{H}^r(\Omega)}.$$

Steps 2 and 3: Finite element approximation and parallelization

- Let X_h be a finite element space over Ω , and assume that the mesh is **quasiuniform**.
- $w_h^j \in X_h$ are the finite element solutions of

$$(e^{y_j} \mathcal{I} - \Delta)w = f.$$

- These can be solved independently (**embarrassingly parallelizable**) and then gathered to obtain

$$u_h^N = \frac{\sin(\pi s)}{\pi} k \sum_{j=-N}^N e^{(1-s)y_j} w_h^j$$

Error analysis

For simplicity, assume that Ω is convex.

- For $r \leq 2s$ define

$$\alpha_\star = \frac{1}{2} (\alpha + \min\{1 - r, \alpha\}), \quad \sigma = \max\{2\alpha_\star - 2s, 0\}.$$

If $f \in \mathbb{H}^\sigma(\Omega)$ then

$$\|u - u_h^N\|_{\mathbb{H}^r(\Omega)} \lesssim h^{2\alpha_\star} |\log h| \|f\|_{\mathbb{H}^\sigma(\Omega)}.$$

- Setting $r = s$ we get

$$\|u - u_h^N\|_{\mathbb{H}^s(\Omega)} \lesssim h^{2-s} \|f\|_{\mathbb{H}^{2-2s}(\Omega)},$$

which is “optimal” in order $2 - s$ and regularity $f \in \mathbb{H}^{2-2s}(\Omega)$.
However, this requires $u \in \mathbb{H}^2(\Omega)$, which is **not generic!**

Outlook

To solve

$$(-\Delta)^s u = f$$

with the Balakrishnan formula we must:

- Solve $\mathcal{O}(|\log h|)$ problems of the type $(e^y \mathcal{I} - \Delta)w = f$.
- Embarrassingly parallelizable.
- Error estimate

$$\|u - u_h^N\|_{\mathbb{H}^s(\Omega)} \lesssim h^{2-s} \|f\|_{\mathbb{H}^{2-2s}(\Omega)},$$

Questions:

- Other types of problems? Time-dependent? Nonlinear?
- Lower regularity on f ? How can we capture the boundary singularities of u ?

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Regularity

Discretization

Tensor Product FEMs

Outlook

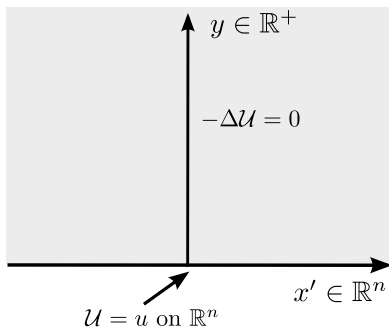
$(-\Delta)^{1/2}$: The Dirichlet to Neumann operator I

- Let $u : \mathbb{R}^n \rightarrow \mathbb{R}$.
- Extend it harmonically to \mathbb{R}_+^{n+1}

$$-\Delta \mathcal{U} = 0, \text{ in } \mathbb{R}_+^{n+1}, \quad \mathcal{U}(\cdot, 0) = u$$

- The Dirichlet to Neumann map is

$$\text{DtN} : u \mapsto -\partial_y \mathcal{U}(\cdot, 0).$$



$(-\Delta)^{1/2}$: The Dirichlet to Neumann operator II

The Dirichlet to Neumann map

$$\text{DtN} : u \mapsto -\partial_y \mathcal{U}(\cdot, 0).$$

has the following properties:

- $\text{DtN}^2 = -\Delta$: Indeed, since $-\Delta_{x',y} \mathcal{U} = -\Delta_{x'} \mathcal{U} - \partial_y^2 \mathcal{U} = 0$,

$$\text{DtN}^2 u = \partial_y (\partial_y \mathcal{U}(\cdot, 0)) = -\Delta_{x'} \mathcal{U}(\cdot, 0) = -\Delta_{x'} u.$$

- DtN is positive: Since \mathcal{U} is harmonic

$$0 = - \int_{\mathbb{R}_+^{n+1}} \Delta \mathcal{U} \mathcal{U} \, dx \, dy = \int_{\mathbb{R}_+^{n+1}} |\nabla \mathcal{U}|^2 \, dx \, dy + \int_{\mathbb{R}^n} \partial_y \mathcal{U} \mathcal{U} \, dx.$$

On the other hand

$$\int_{\mathbb{R}^n} u \, \text{DtN} \, u \, dx = - \int_{\mathbb{R}^n} \partial_y \mathcal{U} \mathcal{U} \, dx > 0.$$

Thus, we define

$$\text{DtN} = (-\Delta_x)^{\frac{1}{2}}, \quad (-\Delta_x)^{\frac{1}{2}} u = \partial_\nu \mathcal{U}.$$

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Regularity

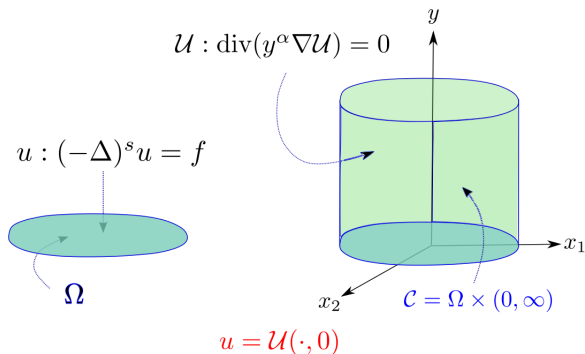
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Tensor Product FEMs

Outlook

The α -harmonic extension I

The previous extension property can be generalized to any $s \in (0, 1)$ 



- $s \in (0, 1)$ and $\alpha = 1 - 2s \in (-1, 1)$.
- $\partial_\nu^\alpha \mathcal{U} = -\lim_{y \downarrow 0} y^\alpha \partial_y \mathcal{U} = d_s f$ on $\Omega \times \{0\}$.
- $d_s = 2^\alpha \Gamma(1 - s) / \Gamma(s)$.

The α -harmonic extension II

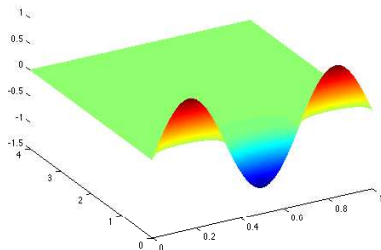
Fractional powers of $-\Delta$ can be realized as a generalization of the Dirichlet to Neumann operator:

$$\begin{cases} \partial_{yy}^2 \mathcal{U} + \frac{\alpha}{y} \partial_y \mathcal{U} + \Delta_x \mathcal{U} = 0 & \text{in } \mathcal{C} \\ \mathcal{U} = 0 & \text{on } \partial_L \mathcal{C} \\ \partial_{\nu^\alpha} \mathcal{U} = d_s f & \text{on } \Omega \times \{0\} \end{cases} \iff (-\Delta)^s u = f \text{ in } \Omega$$

$$u = \mathcal{U}(\cdot, 0).$$

Here:

- $\mathcal{C} = \Omega \times (0, \infty)$.
- $\alpha = 1 - 2s \in (-1, 1)$.
- $\partial_{\nu^\alpha} \mathcal{U} = -\lim_{y \downarrow 0} y^\alpha \partial_y \mathcal{U} = d_s f$.
- $d_s = 2^\alpha \Gamma(1 - s) / \Gamma(s)$.



The α -harmonic extension III

Why does this make sense?

- For $\lambda > 0$ and $g \in \mathbb{R}$ consider the ODE:

$$\begin{cases} \psi'' + \frac{1-2s}{y} \psi' - \lambda \psi = 0, & \text{in } (0, \infty), \\ -\lim_{y \downarrow 0} y^{1-2s} \psi' = d_s, & \lim_{y \uparrow \infty} \psi(y) = 0. \end{cases}$$

- This is a **Bessel equation** with solution

$$\psi(y) = C_s \lambda^{-s} \left(\sqrt{\lambda} y \right)^s K_s(\sqrt{\lambda} y)$$

where K_s is the **modified Bessel function** of the second kind.

- It is well known that $K_s(z) = az^{-s} + o(z^{-s})$, with $a > 0$ as $z \downarrow 0$. Thus


$$\psi(y) = c_s \lambda^{-s} \left(\sqrt{\lambda} y \right)^s \left(a(\sqrt{\lambda} y)^{-s} \right) \rightarrow ac_s \lambda^{-s}, \quad y \downarrow 0.$$

- Choosing C_s appropriately we get $\psi(0) = \lambda^{-s}$.

The α -harmonic extension IV

- Recall that

$$f = \sum_{k=1}^{\infty} f_k \varphi_k \in \mathbb{H}^{-s}(\Omega), \quad (-\Delta)^s u = f, \quad \implies u = \sum_{k=1}^{\infty} \lambda_k^{-s} f_k \varphi_k$$

- Applying separation of variables to the extension problem 

$$u(x) = \sum_{k=1}^{\infty} u_k \varphi_k(x) \implies \mathcal{U}(x, y) = \sum_{k=1}^{\infty} u_k \varphi_k(x) \psi_k(y),$$

where the functions ψ_k solve

$$\psi_k'' + \frac{\alpha}{y} \psi_k' = \lambda_k \psi_k, \quad \text{in } (0, \infty), \quad \psi_k(0) = 1, \quad \lim_{y \rightarrow \infty} \psi_k(y) = 0.$$

so that, as before,

$$\psi_k(y) = c_s \left(\sqrt{\lambda_k y} \right)^s K_s(\sqrt{\lambda_k y}),$$

Weak formulation

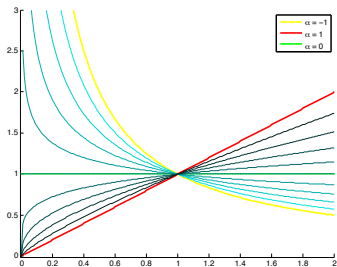
- Multiply $\nabla \cdot (y^\alpha \nabla U)$ by a test function ϕ and integrate over the cylinder \mathcal{C} to obtain a possible **weak formulation**

$$\int_{\mathcal{C}} y^\alpha \nabla U \cdot \nabla \phi \, dx \, dy = d_s \int_{\Omega} f \phi(x, 0) \, dx, \quad \forall \phi \in \mathring{H}_L^1(y^\alpha, \mathcal{C}),$$

- Where the energy space is

$$L^2(y^\alpha, \mathcal{C}) = \left\{ w : \int_{\mathcal{C}} |w|^2 y^\alpha \, dx \, dy < \infty \right\}$$

$$\mathring{H}_L^1(y^\alpha, \mathcal{C}) = \{ w \in L^2(y^\alpha, \mathcal{C}) : \nabla w \in L^2(y^\alpha, \mathcal{C}), w|_{\partial_L \mathcal{C}} = 0 \}.$$



The weight y^α is degenerate ($\alpha > 0$) or singular ($\alpha < 0$)!

Muckenhoupt weights

For every $a, b \in \mathbb{R}$, with $a < b$,

$$\frac{1}{b-a} \int_a^b |y|^\alpha dy \cdot \frac{1}{b-a} \int_a^b |y|^{-\alpha} dy \lesssim 1$$

which means y^α belongs to the **Muckenhoupt class** A_2 .

This condition, essentially, means that y^α behaves like a constant **at every scale!**

Since $y^\alpha \in A_2$:

- The Hardy-Littlewood maximal operator is continuous on $L^2(y^\alpha, \mathcal{C})$.
- Singular integral operators are continuous on $L^2(y^\alpha, \mathcal{C})$.
- $L^2(y^\alpha, \mathcal{C}) \hookrightarrow L^1_{loc}(\mathcal{C})$.
- $H^1(y^\alpha, \mathcal{C})$ is Hilbert and $\mathcal{C}_b^\infty(\mathcal{C})$ is dense.
- Traces on $\partial_L \mathcal{C}$ are well defined.

Weighted Sobolev spaces

- Weighted Poincaré inequality:

$$\int_{\mathcal{C}} y^\alpha |w|^2 dx dy \lesssim \int_{\mathcal{C}} y^\alpha |\nabla w|^2 dx dy \quad \forall w \in \mathring{H}_L^1(y^\alpha, \mathcal{C}).$$

- Surjective trace operator $\text{tr}_\Omega : \mathring{H}_L^1(y^\alpha, \mathcal{C}) \rightarrow \mathbb{H}^s(\Omega)$.
- Lax-Milgram \Rightarrow existence and uniqueness for every $f \in \mathbb{H}^{-s}(\Omega)$.

Also

$$\|\mathcal{U}\|_{\mathring{H}_L^1(y^\alpha, \mathcal{C})}^2 = \|u\|_{\mathbb{H}^s(\Omega)}^2 = d_s \|f\|_{\mathbb{H}^{-s}(\Omega)}^2.$$

We will discretize the α -harmonic extension!

$$\mathcal{U} \in \mathring{H}_L^1(y^\alpha, \mathcal{C}) : \quad \begin{cases} \nabla \cdot (y^\alpha \nabla \mathcal{U}) = 0 & \text{in } \mathcal{C} \\ \mathcal{U} = 0 & \text{on } \partial_L \mathcal{C} \\ \partial_{\nu^\alpha} \mathcal{U} = d_s f & \text{on } \Omega \times \{0\} \end{cases}$$

Advantages and disadvantages

Advantages:

- Implementation requires **standard** numerical PDE components.
- It is **very flexible** as we will see later.

Disadvantages:

- **One extra dimension!** We have **efficient solvers**, and we will see later how to **minimize** the effect of y .
- **Singular/degenerate weight y^α ?** The weight $y^\alpha \in A_2$ for which there is a very well developed theory.

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Solution representation

- Recall that we found, via [separation of variables](#)

$$u(x) = \sum_{k=1}^{\infty} \lambda_k^{-s} f_k \varphi_k(x) \implies \mathcal{U}(x, y) = \sum_{k=1}^{\infty} \lambda_k^{-s} f_k \varphi_k(x) \psi_k(y),$$

- The pairs $\{\lambda_k, \varphi_k\}_{k=1}^{\infty}$ are the eigenpairs of the Laplacian.
- The ψ_k are

$$\psi_k(y) = c_s \left(\sqrt{\lambda_k y} \right)^s K_s(\sqrt{\lambda_k y}),$$

where K_s is the [modified Bessel function of the second kind](#).

- The function ψ_k satisfies, as $y \rightarrow \infty$,

$$\psi_k(y) \approx \left(\sqrt{\lambda_k y} \right)^{s-1/2} e^{-\sqrt{\lambda_k y}}.$$

- The function ψ_k satisfies, as $y \rightarrow 0$,

$$\psi_k'(y) \approx y^{-\alpha}, \quad \psi_k''(y) \approx y^{-\alpha-1},$$

Global Sobolev Regularity

- **Compatible data:** Let $f \in \mathbb{H}^{1-s}(\Omega)$, which means that f has a vanishing trace for $s < \frac{1}{2}$.
- **Space regularity:**

$$\|\Delta_x \mathcal{U}\|_{L^2(y^\alpha, \mathcal{C})}^2 + \|\partial_y \nabla_x \mathcal{U}\|_{L^2(y^\alpha, \mathcal{C})}^2 = d_s \|f\|_{\mathbb{H}^{1-s}(\Omega)}^2$$

- **Regularity in extended variable y :** If $s \neq \frac{1}{2}$ and $\beta > 2\alpha + 1$ then

$$\|\partial_{yy} \mathcal{U}\|_{L^2(y^\beta, \mathcal{C})} \lesssim \|f\|_{L^2(\Omega)}.$$

If $s = \frac{1}{2}$, then

$$\|\mathcal{U}\|_{H^2(\mathcal{C})} \lesssim \|f\|_{\mathbb{H}^{1/2}(\Omega)}.$$

- **Elliptic pick-up regularity:** If Ω convex, then

$$\|w\|_{H^2(\Omega)} \lesssim \|\Delta_x w\|_{L^2(\Omega)} \quad \forall w \in H^2(\Omega) \cap H_0^1(\Omega).$$

Under this assumption, we further have

$$\|D_x^2 \mathcal{U}\|_{L^2(y^\alpha, \mathcal{C})} \lesssim \|f\|_{\mathbb{H}^{1-s}(\Omega)}.$$

Analytic Regularity

- Behavior of $\psi(z) = c_s z^s K_s(z)$ near $z = 0$:

$$\left| \frac{d^\ell}{dz^\ell} \psi(z) \right| \leq C d_s \ell! z^{2s-\ell},$$

where $d_s = 2^{1-2s} \Gamma(1-s) / \Gamma(s)$.

- Behavior of $\psi(z)$ for z large:

$$\left| \frac{d^\ell}{dz^\ell} \psi(z) \right| \leq C_{\epsilon,s} \ell! \epsilon^{-\ell} z^{s-\ell-\frac{1}{2}} e^{-(1-\epsilon)z}$$

- Global** regularity of \mathcal{U} : If $0 \leq \tilde{\nu} < s$ and $0 \leq \nu < 1 + s$, then there exists $\kappa > 1$ such that

$$\begin{aligned} \|\partial_y^{\ell+1} \mathcal{U}\|_{L^2(\omega_{\alpha+2\ell-2\tilde{\nu},\gamma}, \mathcal{C})} &\lesssim \kappa^{\ell+1} (\ell+1)! \|f\|_{\mathbb{H}^{-s+\tilde{\nu}}(\Omega)}, \\ \|\nabla_x \partial_y^{\ell+1} \mathcal{U}\|_{L^2(\omega_{\alpha+2(\ell+1)-2\nu,\gamma}, \mathcal{C})} &\lesssim \kappa^{\ell+1} (\ell+1)! \|f\|_{\mathbb{H}^{-s+\nu}(\Omega)}, \\ \|\Delta_x \partial_y^{\ell+1} \mathcal{U}\|_{L^2(\omega_{\alpha+2(\ell+1)-2\nu,\gamma}, \mathcal{C})} &\lesssim \kappa^{\ell+1} (\ell+1)! \|f\|_{\mathbb{H}^{1-s+\nu}(\Omega)}, \end{aligned}$$

with weight $\omega_{\beta,\gamma}(y) = y^\beta e^{\gamma y}$, $0 \leq \gamma < 2\sqrt{\lambda_1}$.

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Domain truncation

The domain \mathcal{C} is **infinite**. We need to consider a **truncated** problem.

Theorem (exponential decay)

For every $\gamma > 0$

$$\|\mathcal{U}\|_{\mathring{H}_L^1(y^\alpha, \Omega \times (\gamma, \infty))} \lesssim e^{-\sqrt{\lambda_1}\gamma/2} \|f\|_{\mathbb{H}^{-s}(\Omega)}.$$

Let v solve

$$\begin{cases} \nabla \cdot (y^\alpha \nabla v) = 0 & \text{in } \mathcal{C}_\gamma = \Omega \times (0, \gamma), \\ v = 0 & \text{on } \partial_L \mathcal{C}_\gamma \cup \Omega \times \{\gamma\}, \\ \partial_{\nu^\alpha} v = d_s f & \text{on } \Omega \times \{0\}. \end{cases}$$

Theorem (exponential convergence)

For all $\gamma > 0$,

$$\|\mathcal{U} - v\|_{\mathring{H}_L^1(y^\alpha, \mathcal{C}_\gamma)} \lesssim e^{-\sqrt{\lambda_1}\gamma/4} \|f\|_{\mathbb{H}^{-s}(\Omega)}.$$

Finite element method I: Mesh

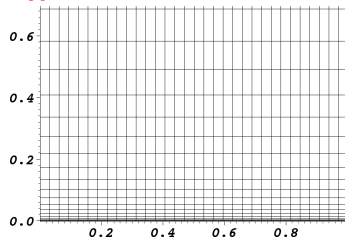
Let $\mathcal{T}_\Omega = \{K\}$ be triangulation of Ω (simplices or cubes)

- \mathcal{T}_Ω is conforming and shape regular.

Let $\mathcal{T}_\mathcal{Y} = \{T\}$ be a triangulation of $\mathcal{C}_\mathcal{Y}$ into cells of the form

$$T = K \times I, \quad K \in \mathcal{T}_\Omega, \quad I = (a, b).$$

$U_{yy} \approx y^{-\alpha-1}$ as $y \approx 0+$ so we consider anisotropic elements



Shape regularity condition
does NOT hold!

Finite element method II: Spaces

We **only** require that if $T = K \times I$ and $T' = K' \times I'$ are **neighbors**

$$\frac{|I|}{|I'|} \approx 1,$$

This weak condition allows us to consider **anisotropic meshes**

Define

$$\mathbb{V}(\mathcal{T}_{\mathcal{Y}}) = \{W \in \mathcal{C}^0(\bar{\mathcal{C}}_{\mathcal{Y}}) : W|_T \in \mathcal{P}_1(K) \otimes \mathbb{P}_1(I), W|_{\Gamma_D} = 0\}$$

with $\Gamma_D = \partial_L \mathcal{C} \cup \Omega \times \{\mathcal{Y}\}$, and

$$\mathbb{U}(\mathcal{T}_{\Omega}) = \text{tr}_{\Omega} \mathbb{V}(\mathcal{T}_{\mathcal{Y}}) = \{W \in \mathcal{C}^0(\bar{\Omega}) : W|_K \in \mathcal{P}_1(K), W|_{\partial\Omega} = 0\}.$$

Here $\mathcal{P}_1 = \mathbb{P}_1$ if K is a simplex and $\mathcal{P}_1 = \mathbb{Q}_1$ if is a “brick”.

Finite element method III: Discrete problem

- **Galerkin method** for the extension: Find $V_{\mathcal{T}_y} \in \mathbb{V}(\mathcal{T}_y)$ such that

$$\int_{\mathcal{C}_y} y^\alpha \nabla V_{\mathcal{T}_y} \nabla W \, dx \, dy = d_s \int_{\Omega} f W(x, 0) \, dx, \quad \forall W \in \mathbb{V}(\mathcal{T}_y).$$

- Define

$$U_{\mathcal{T}_\Omega} = V_{\mathcal{T}_y}(\cdot, 0) \in \mathbb{U}(\mathcal{T}_\Omega).$$

- A trace estimate and C ea's Lemma imply quasi-best approximation:

$$\|u - U_{\mathcal{T}_\Omega}\|_{\mathbb{H}^s(\Omega)} \lesssim \|v - V_{\mathcal{T}_y}\|_{\mathring{H}_L^1(y^\alpha, \mathcal{C}_y)} = \inf_{W \in \mathbb{V}(\mathcal{T}_y)} \|v - W\|_{\mathring{H}_L^1(y^\alpha, \mathcal{C}_y)}$$

We reduced the **error analysis** to a question of **approximation theory** in weighted spaces. Usually we set $W = \Pi v \in \mathbb{V}(\mathcal{T}_y)$ where Π is a suitable **interpolation** operator.

The quasi-interpolation operator

We introduce an **averaged** interpolation operator Π

$$\Pi\phi(z) = Q_z^m\phi(z).$$

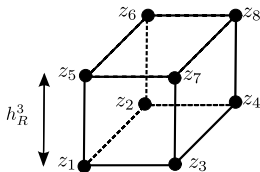
where $Q_z^m\phi$ is an averaged Taylor polynomial of ϕ of degree m .

Notice that:

- This is defined for all polynomial degree m and any element shape (simplices or rectangles).
- We do not go back to the reference element — This is important for anisotropic estimates.

If the mesh is rectangular and Cartesian If R and S are neighbors

$$h_R^i/h_S^i \lesssim 1, \quad i = \overline{1, N}.$$



Error estimates on rectangles

Theorem

If $\varpi \in A_p(\mathbb{R}^N)$, and $\phi \in W_p^1(\varpi, S_R)$

$$\|\phi - \Pi\phi\|_{L^p(\varpi, R)} \lesssim \sum_{i=1}^N h_R^i \|\partial_i \phi\|_{L^p(\varpi, S_R)}.$$

If $\phi \in W_p^2(\varpi, S_R)$

$$\|\partial_j(\phi - \Pi\phi)\|_{L^p(\varpi, R)} \lesssim \sum_{i=1}^N h_R^i \|\partial_i \partial_j \phi\|_{L^p(\varpi, S_R)},$$

$$\|\phi - \Pi\phi\|_{L^p(\varpi, R)} \lesssim \sum_{i,j=1}^N h_R^i h_R^j \|\partial_i \partial_j \phi\|_{L^p(\varpi, S_R)}.$$

- **Directional estimates:** note the products of the form

$$h_R^i h_R^j \|\partial_i \partial_j \phi\|_{L^p(\varpi, S_R)}.$$

- Estimates on simplicial elements, different metrics and applications.

Error estimates. Quasiuniform meshes

On quasiuniform meshes $h_T \approx h_K \approx h_I$ for all $T \in \mathcal{T}_y$, then

Theorem (error estimates)

The following estimate holds for all $\epsilon > 0$

$$\begin{aligned}\|\nabla(v - V_{\mathcal{T}_y})\|_{L^2(y^\alpha, \mathcal{C}_y)} &\lesssim h_K \|\partial_y \nabla_{x'} v\|_{L^2(y^\alpha, \mathcal{C})} + h_I^{s-\epsilon} \|\partial_{yy} v\|_{L^2(y^\beta, \mathcal{C})} \\ &\lesssim h^{s-\epsilon} \|f\|_{\mathbb{H}^{1-s}(\Omega)}.\end{aligned}$$

Consequently,

$$\|u - U_{\mathcal{T}_\Omega}\|_{\mathbb{H}^s(\Omega)} \lesssim h^{s-\epsilon} \|f\|_{\mathbb{H}^{1-s}(\Omega)}.$$

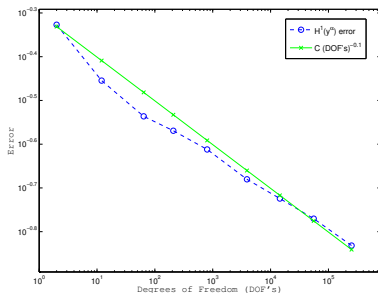
- This is **suboptimal** in terms of order (only order $s - \epsilon$)
- Is it **sharp**?

Numerical experiment. Quasiuniform mesh

Let $\Omega = (0, 1)$ and $f = \pi^{2s} \sin(\pi x)$, then

$$\mathcal{U} = \frac{2^{1-s} \pi^s}{\Gamma(s)} \sin(\pi x) y^s K_s(\pi y)$$

If $s = 0.2$, then



The energy error behaves like $DOFS^{-0.1} \approx h^{0.2}$, as predicted!

Error estimates. Graded meshes

We use the principle of **error equilibration**. We use a **graded mesh** on $(0, \mathcal{Y})$

$$y_j = \mathcal{Y} \left(\frac{j}{M} \right)^\gamma, \quad j = \overline{0, M}, \quad \gamma > 1$$

$\mathcal{U}_{yy} \approx y^{-\alpha-1} \implies$ energy equidistribution for $\gamma > 3/(1-\alpha)$.

Theorem (error estimates⁴)

If $f \in \mathbb{H}^{1-s}(\Omega)$ and $\mathcal{Y} \approx |\log \# \mathcal{T}_{\mathcal{Y}}|$,

$$\|u - U_{\mathcal{T}_{\Omega}}\|_{\mathbb{H}^s(\Omega)} = \|\nabla(\mathcal{U} - V_{\mathcal{T}_{\mathcal{Y}}})\|_{L^2(y^{\alpha}, c)} \lesssim |\log \# \mathcal{T}_{\mathcal{Y}}|^s \# \mathcal{T}_{\mathcal{Y}}^{-\frac{1}{n+1}} \|f\|_{\mathbb{H}^{1-s}(\Omega)},$$

or equivalently

$$\|u - U_{\mathcal{T}_{\Omega}}\|_{\mathbb{H}^s(\Omega)} \lesssim |\log \mathcal{T}_{\Omega}|^s \mathcal{T}_{\Omega}^{-1/n} \|u\|_{\mathbb{H}^{1+s}(\Omega)}.$$

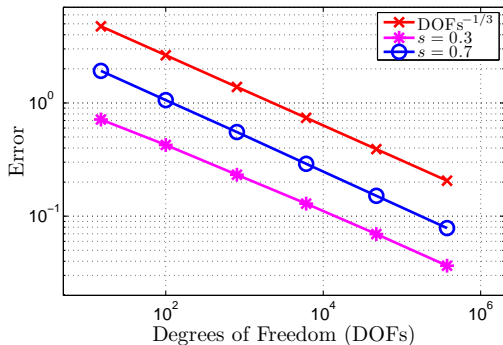
- This is **near optimal** in terms of regularity of $u \in \mathbb{H}^{1+s}(\Omega)$ and almost linear decay rate in h .
- This is **suboptimal** in terms of total number of degrees of freedom $\# \mathcal{T}_{\mathcal{Y}} \approx \# \mathcal{T}_{\Omega}^{1+\frac{1}{n}} \gg \# \mathcal{T}_{\Omega}$ with respect to the degrees of freedom in Ω .

Numerical experiment

Experimental rates for circle and $s = 0.3$ and $s = 0.7$.

Set $\Omega = D(0, 1) \subset \mathbb{R}^2$, $f = j_{1,1}^{2s} J_1(j_{1,1}r)(A_{1,1} \cos(\theta) + B_{1,1} \sin(\theta))$.

With graded meshes:



The experimental convergence rate $-1/3$ is optimal!

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Diagonalization I

- **Discretization in y :** Let \mathcal{G}^M be an arbitrary mesh in $(0, \mathcal{Y})$ with $M = \#\mathcal{G}^M$ and let $S^{\mathbf{r}}(0, \mathcal{Y}; \mathcal{G}^M)$ be a FE space of polynomial degree \mathbf{r} in y .

- Define

$$\mathbb{V}_M^{\mathbf{r}}(\mathcal{C}_{\mathcal{Y}}) = H_0^1(\Omega) \otimes S^{\mathbf{r}}(0, \mathcal{Y}; \mathcal{G}^M).$$

FE in y , continuous in x .

- **Semidiscrete solution:** $\mathcal{U}_M \in \mathbb{V}_M^{\mathbf{r}}(\mathcal{C}_{\mathcal{Y}})$ satisfies

$$\int_{\mathcal{C}_{\mathcal{Y}}} y^\alpha \nabla \mathcal{U}_M \nabla \phi \, dx \, dy = d_s \int_{\Omega} f \phi(x, 0) \, dx \quad \forall \phi \in \mathbb{V}_M^{\mathbf{r}}(\mathcal{C}_{\mathcal{Y}}).$$

- **Exponential convergence:** Let $f \in \mathbb{H}^{-s+\nu}(\Omega)$ for $0 < \nu < s$. If $\mathcal{Y} \approx M$, the mesh \mathcal{G}^M is geometric towards $y = 0$, and the polynomial degree \mathbf{r} grows linearly from $y = 0$, then there exists $b > 0$ such that

$$\|\nabla(\mathcal{U} - \mathcal{U}_M)\|_{L^2(y^\alpha, \mathcal{C})} \lesssim e^{-bM} \|f\|_{\mathbb{H}^{-s+\nu}(\Omega)}.$$

Diagonalization II

- **Eigenvalue problem:** Let $\mathcal{M} = \dim S^{\mathbf{r}}(0, \mathcal{Y}; \mathcal{G}^M)$ and $(\mu_i, v_i)_{i=1}^{\mathcal{M}}$ be the (normalized) eigenpairs of

$$\mu \int_0^{\mathcal{Y}} y^\alpha v'(y) w'(y) dy = \int_0^{\mathcal{Y}} y^\alpha v(y) w(y) dy \quad \forall w \in S^{\mathbf{r}}(0, \mathcal{Y}; \mathcal{G}^M).$$

- **Representation:** If $\mathcal{U}_M(x, y) = \sum_{j=1}^{\mathcal{M}} U_j(x) v_j(y)$ with $U_j \in H_0^1(\Omega)$, then

$$a_{\mu_i, \Omega}(U_i, V) = d_s v_i(0) \int_{\Omega} f V dx \quad \forall V \in H_0^1(\Omega),$$

where $a_{\mu_i, \Omega}$ are the singularly perturbed bilinear forms

$$a_{\mu_i, \Omega}(U, V) := \int_{\Omega} (\mu_i \nabla_x U \nabla_x V dx + UV) dx$$

Tensor product discretization

- Ritz projections:** $\Pi_i u \in S_0^q(\mathcal{T}_\Omega)$ satisfies

$$a_{\mu_i, \Omega}(u - \Pi_i u, v) = 0 \quad \forall v \in S_0^q(\mathcal{T}_\Omega),$$

where $S_0^q(\mathcal{T}_\Omega) \subset H_0^1(\Omega)$ is the FE space of piecewise polynomials of degree $\leq q$ over \mathcal{T}_Ω .

- Discrete solution:** Let $U_{h,M} \in S_0^q(\mathcal{T}_\Omega) \otimes S^r(0, \mathcal{Y}; \mathcal{G}^M)$ satisfy

$$\int_{\mathcal{C}_y} y^\alpha \nabla U_{h,M} \nabla V \, dx \, dy = d_s \int_{\Omega} f V(x, 0) \, dx, \quad \forall V \in S_0^q(\mathcal{T}_\Omega) \otimes S^r(0, \mathcal{Y}; \mathcal{G}^M)$$

and note that it can be represented as follows

$$U_{h,M}(x, y) = \sum_{i=1}^{\mathcal{M}} \Pi_i U_i(x) v_i(y).$$

- Parallelization:** This corresponds to solving \mathcal{M} decoupled elliptic problems with the singularly perturbed bilinear form $a_{\mu_i, \Omega}$ for $i = 1, \dots, \mathcal{M}$.

Tensor \mathbb{P}_1 -FEM

- Assume that $f \in L^2(\Omega)$ where $\Omega \subset \mathbb{R}^2$ is a polygon with corners \mathbf{c} .
- The solution to

$$-\Delta_x w = f, \text{ in } \Omega \quad w = 0, \text{ on } \partial\Omega \implies$$

$$\|w\|_{H_\beta^2(\Omega)} \lesssim \|f\|_{L^2(\Omega)}, \quad |w|_{H_\beta^2(\Omega)}^2 = \int_{\Omega} \prod_{\mathbf{c}} |x' - \mathbf{c}|^{2\beta} |D^2 w|^2 dx.$$

- This type of singularity can be captured by using a **graded mesh in Ω** : Let \mathcal{T}_Ω be graded towards the re-entrant corners so that, if $N = \#\mathcal{T}_\Omega$ and $h = N^{-1/2}$, for any $w \in S_0^1(\mathcal{T}_\Omega)$

$$N \|w - \Pi w\|_{L^2(\Omega)}^2 \lesssim \|w\|_{H^1(\Omega)}^2, \quad N^2 \|w - \Pi w\|_{L^2(\Omega)}^2 \lesssim \|w\|_{H_\beta^2(\Omega)}^2.$$

- With this construction we obtain that, if \mathcal{G}_η^M is a suitably graded **radical mesh** $\{y_i = (\frac{i}{M})^\eta \mathcal{Y}\}_{i=0}^M$, with $\eta s > 1$ and $M \approx N^{\frac{1}{2}} = (\#\mathcal{T}_\Omega)^{\frac{1}{2}}$, the discrete solution $U_{h,M}$ satisfies

$$\|u - \text{tr}_\Omega U_{h,M}\|_{\mathbb{H}^s(\Omega)} \leq h \|f\|_{\mathbb{H}^{1-s}(\Omega)}$$

and

$$\dim \mathbb{V}_{h,M}^{1,1}(\mathcal{T}_\Omega, \mathcal{G}^M) \approx h^{-3} \log |\log h| \approx N_\Omega^{1+\frac{1}{2}} \log \log N_\Omega.$$

Sparse grid FEM

- Complexity of tensor product: $N_\Omega^{1+\frac{1}{2}}$ is **suboptimal**.
- To overcome this we use a **sparse grid space**. Let

$$\mathbb{V}_L^{1,1}(\mathcal{C}_y) = \sum_{\ell, \ell' \geq 0, \ell + \ell' \leq L} S_0^1(\mathcal{T}_\Omega^\ell) \otimes S^1(0, y; \mathcal{G}_\eta^{2^{\ell'}}),$$

where \mathcal{T}_Ω^ℓ and $\mathcal{G}_\eta^{2^{\ell'}}$ are nested meshes of levels ℓ and ℓ' graded towards corners \mathbf{c} of Ω and $y = 0$, respectively.

- We have the error estimate: Let $1 < \nu < 1 + s$, $\eta(\nu - 1) \geq 1$, and $y \approx |\log h_L|$. If $f \in \mathbb{H}^{-s+\nu}(\Omega)$, then $\mathcal{U}_L \in \mathbb{V}_L^{1,1}(\mathcal{C}_y)$ satisfies

$$\begin{aligned} \|\mathcal{U} - \mathcal{U}_L\|_{L^2(y^\alpha, \mathcal{C})} &\lesssim h_L |\log h_L| \|f\|_{\mathbb{H}^{-s+\nu}(\Omega)}, \\ \dim \mathbb{V}_L^{1,1}(\mathcal{C}_y) &\lesssim N_\Omega \log \log N_\Omega. \end{aligned}$$

- The complexity of sparse grids is **quasi-optimal** in terms of N_Ω .

hp -FEM in y and \mathbb{P}_1 -FEM in Ω

- **Graded geometric mesh:** Let $\mathcal{G}_\sigma^M = \{\mathcal{Y}\sigma^{M-i}\}_{i=1}^M$ with $\sigma < 1$.
- **Data regularity:** $f \in \mathbb{H}^{1-s}(\Omega)$ and $\Omega \subset \mathbb{R}^2$ is a polygon with corners \mathbf{c} .
- **FE space:** $\mathbb{V}_{h,M}^{1,\mathbf{r}}(\mathcal{T}_\Omega, \mathcal{G}_\sigma^M)$ is the space of piecewise polynomials of degree one over \mathcal{T}_Ω and piecewise polynomials of degree \mathbf{r} growing linearly from 1 over \mathcal{G}_σ^M .
- **Error estimates:** Let \mathcal{T}_Ω be a suitably graded mesh towards the re-entrant corners \mathbf{c} . If $\mathcal{Y} \approx |\log h|$ and $U_{h,M} \in \mathbb{V}_{h,M}^{1,\mathbf{r}}(\mathcal{T}_\Omega, \mathcal{G}_\sigma^M)$ is the Galerkin solution, then

$$\begin{aligned}\|\nabla(\mathcal{U} - U_{h,M})\|_{L^2(y^\alpha, \mathcal{C})} &\lesssim h \|f\|_{\mathbb{H}^{1-s}(\Omega)} \\ \dim \mathbb{V}_{h,M}^{1,\mathbf{r}}(\mathcal{T}_\Omega, \mathcal{G}_\sigma^M) &\approx h^{-2} |\log h|^2 \approx N_\Omega |\log N_\Omega|\end{aligned}$$

- **Complexity:** This is **quasi-optimal** in terms of N_Ω .

hp -FEM in y and Ω

- **Data regularity:** The domain $\Omega \subset \mathbb{R}^2$ and f are analytic.
- **Graded mesh in Ω :** The mesh \mathcal{T}_Ω is anisotropic and graded towards $\partial\Omega$ so that it resolves the smallest scale μ_M of the singularly perturbed problems originating from the diagonalization.
- **Graded mesh in y :** Let $\mathcal{G}_\sigma^M = \{y\sigma^{M-i}\}_{i=1}^M$ with $\sigma < 1$.
- **Error estimate:** If $y \approx M$, \mathbf{r} grows linearly from $y = 0$, then the Galerkin solution $U_{h,M} \in S_0^q(\mathcal{T}_\Omega) \otimes S^{\mathbf{r}}(\mathcal{G}_\sigma^M)$ and the total number $N_{\Omega,y}$ of degrees of freedom satisfy

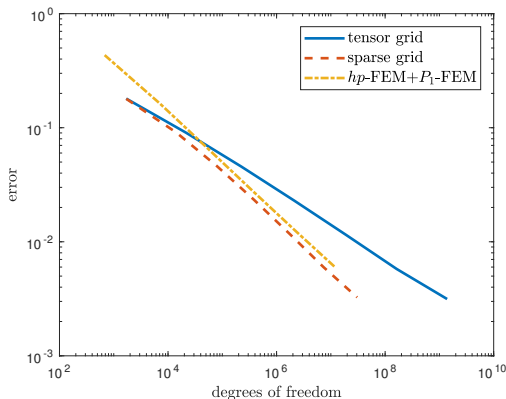
$$\begin{aligned}\|\nabla(U - U_{h,M})\|_{L^2(y^\alpha, \mathcal{C})} &\lesssim M^2 e^{-bq} + e^{-bM} \\ N_{\Omega,y} &\approx q^2 M^3.\end{aligned}$$

- **Exponential rate of convergence:** If $q \approx M$, then

$$\|\nabla(U - U_{h,M})\|_{L^2(y^\alpha, \mathcal{C})} \lesssim e^{-b' N_{\Omega,y}^{1/5}}.$$

Numerical experiment. Performance of tensor FEMs

- **Data:** Ω L-shaped domain in \mathbb{R}^2 ; $f = 1$; $s = 3/4$.
- **Error:** It is always measured in the energy space $\mathbb{H}^s(\Omega)$.



- **Conclusions:** Both sparse grid FEM and hp -FEM reduced substantially the DOFs relative to tensor FEM and deliver quasi-optimal complexity.

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Outlook I

- **PDE approach.** The extension converts the nonlocal problem into a local PDE problem in one higher dimension. This is very flexible:
 - Parabolic problems  [Details](#).
 - Stationary  [Details](#) and time dependent  [Details](#) obstacle problems.
- We have a complete and quasi-optimal **a priori error analysis** over anisotropic meshes. The complexity, in terms of total degrees of freedom, is:
 - $\mathbb{P}_1 - \mathbb{P}_1$ -elements: suboptimal complexity and linear rate for Ω convex and compatible data. Extension to non-convex domains.
 - *Sparse tensor* $\mathbb{P}_1 - \mathbb{P}_1$ -elements: quasi-optimal complexity and linear rate for Ω polygonal with compatible data.
 - *hp*-elements: quasi-optimal complexity and exponential rate for analytic but incompatible data.
 - We also have multigrid methods  [Details](#), a posteriori error estimators  [Details](#).

 Nochetto, Otárola, AJS 2016

 Nochetto, Otárola, AJS 2015

 Otárola, AJS 2016

 Chen, Nochetto, Otárola, AJS 2016

 Chen, Nochetto, Otárola, AJS 2015

Outlook II

Questions:

- **Adaptivity:** Convergence and optimality is still open (issue is anisotropic meshes and lack of shape regularity).
- *3d*-computations: A virtual implementation of extended variable is open.
- Theory and implementation of *3d hp*-FEM are open.

Outline

A posteriori error analysis

- Motivation

- A fundamental difficulty

- Anisotropic error estimation

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Adaptivity

Adaptivity is motivated by:

- Computational efficiency: **extra $n + 1$ -dimension**.
- The a priori theory requires:
 - Regularity of the datum: $f \in \mathbb{H}^{1-s}(\Omega)$.
 - Regularity of the domain: Ω is $C^{1,1}$ or a **convex polygon**.
- If one of these conditions is violated, the solution \mathcal{U} may have **singularities** in Ω which lead to **fractional regularity**.
- Quasiuniform refinement of Ω would **not result** in an efficient solution technique.
- We need **anisotropic a posteriori error estimators**.

Adaptive Loop

We consider an *almost* standard adaptive loop:

SOLVE \rightarrow ESTIMATE \rightarrow MARK \rightarrow REFINES

except for the statements in **red** below:

- **SOLVE**: Finds the Galerkin solution $V_{\mathcal{T}_y}$.
- **ESTIMATE**: Computes a **star-indicator** $\mathcal{E}_{z'}$ for every node $z' \in \Omega$.
- **MARK**: For $\theta \in (0, 1)$ choose a **minimal** subset of nodes \mathcal{M} :

$$\mathcal{E}_{\mathcal{M}}^2 = \sum_{z' \in \mathcal{M}} \mathcal{E}_{z'}^2 \geq \theta^2 \mathcal{E}_{\mathcal{T}}^2.$$

- **REFINE**: Given a set of **marked nodes** \mathcal{M}
 - Refine the **cells** $K \ni z'$ for all $z' \in \mathcal{M}$ to get $\widetilde{\mathcal{T}}_{\Omega}$.
 - Create an **anisotropic mesh** $\{y_j\}_{j=1}^M$ so that grading $y_j = \mathcal{Y} \left(\frac{j}{M}\right)^{\gamma}$ holds.
 - The refined mesh is $\widetilde{\mathcal{T}}_{\mathcal{Y}} = \widetilde{\mathcal{T}}_{\Omega} \times \{\widetilde{I}\}$ with $\widetilde{I} = [y_{j-1}, y_j]$.

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Isotropic a posteriori error indicators

- **Residual error indicator:** If we were to integrate by parts the discrete problem over an element $T \in \mathcal{T}_y$, we would get

$$\int_T y^\alpha \nabla V \nabla W = \int_{\partial T} y^\alpha W \nabla V \cdot \nu - \int_T \nabla \cdot (y^\alpha \nabla V) W$$

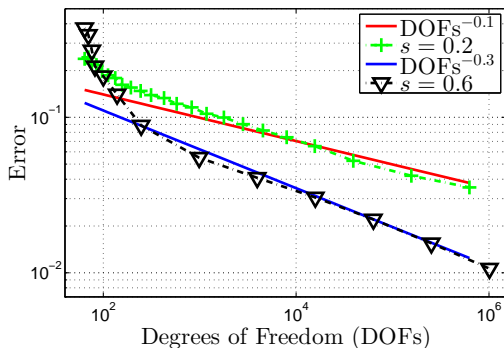
Since $\alpha \in (-1, 1)$, the **boundary integral is meaningless for $y = 0$** .

- **Alternative error indicators:** Residual indicators are not the only possibility:
 - Local problems on stars: $\mathcal{E}_z^2 = \int_{S_z} y^\alpha |\nabla Z|^2$ (Z solution of a BVP in S_z).
 - Zienkiewicz-Zhu estimators.
 - Hypercircle estimators.
- **Local problems on stars:** We prove for all nodes $z \in \mathcal{N}$

$$\mathcal{E}_z^2 \lesssim \|\nabla(v - V)\|_{L^2(y^\alpha, S_z)}^2 \lesssim \mathcal{E}_z^2 + \text{osc}(y^\alpha, V, f, S_z)^2$$

Numerical Experiment with Isotropic Refinement

- Set $\mathcal{C}_\gamma = (0, 1) \times (0, 4)$ and $u = \sin(\pi x)$
- Experimental convergence rates:



- The error decays like $(\#\mathcal{T}_\gamma)^{-(1-|\alpha|)/4}$ as in uniform/isotropic refinement!
- Does adaptivity help?

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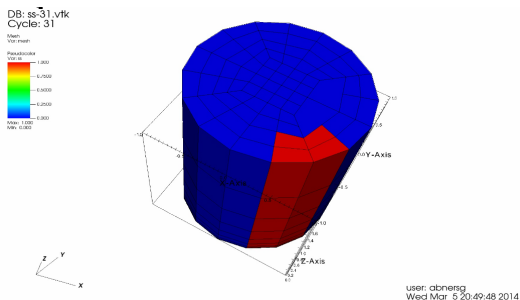
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Anisotropic Error Estimation

- Anisotropic a posteriori error estimator: we need to distinguish the behavior on the extended variable y from the rest.
- The theory of a posteriori error estimation (and adaptivity) on anisotropic discretizations is still in its infancy.
- Cylindrical stars: We propose an error estimator based on solving local problems on sets $\mathcal{C}_{z'} = S_{z'} \times (0, \mathcal{Y})$ as depicted in red in the figure:



An Ideal A Posteriori Error Estimator

- **Local space:** For $z' \in \Omega$ a node, let $\mathcal{C}_{z'} = S_{z'} \times (0, \mathcal{Y})$ and define

$$\mathcal{W}(\mathcal{C}_{z'}) = \{w \in H^1(y^\alpha, \mathcal{C}_{z'}) : w = 0 \text{ on } \partial\mathcal{C}_{z'} \setminus \Omega \times \{0\}\}.$$

- **Local star indicator:** The error indicator $\eta_{z'} \in \mathcal{W}(\mathcal{C}_{z'})$ is given by

$$\int_{\mathcal{C}_{z'}} y^\alpha \nabla \eta_{z'} \nabla w \, dx \, dy = d_s \int_{\Omega} f w(x, 0) \, dx - \int_{\mathcal{C}_{z'}} y^\alpha \nabla V \nabla w \, dx \, dy,$$

for every $w \in \mathcal{W}(\mathcal{C}_{z'})$.

- **Global error estimator:**

$$\mathcal{E}_{\mathcal{T}_\Omega} = \left(\sum_{z'} \mathcal{E}_{z'}^2 \right)^{1/2}, \quad \mathcal{E}_{z'} = \|\nabla \eta_{z'}\|_{L^2(y^\alpha, \mathcal{C}_{z'})}.$$

Anisotropic a posteriori error analysis

- **Efficiency:** For every node $z' \in \Omega$ we have

$$\mathcal{E}_{z'} \leq \|\nabla e\|_{L^2(y^\alpha, \mathcal{C}_{z'})}.$$

- **Data oscillation:** If $f_{z'|K} = \frac{1}{|K|} \int_K f \, dx$ for every element $K \subset S_{z'}$, then

$$\text{osc}_{\mathcal{T}_\Omega}(f)^2 = \sum_{z'} \text{osc}_{z'}(f)^2, \quad \text{osc}_{z'}(f)^2 = d_s h_{z'}^{2s} \|f - f_{z'}\|_{L^2(S_{z'})}^2$$

- **Reliability:**

$$\|\nabla e\|_{L^2(y^\alpha, \mathcal{C}_y)}^2 \lesssim \mathcal{E}_{\mathcal{T}_\Omega}^2 + \text{osc}_{\mathcal{T}_\Omega}(f)^2.$$

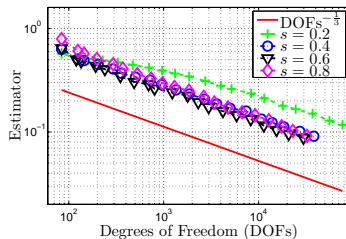
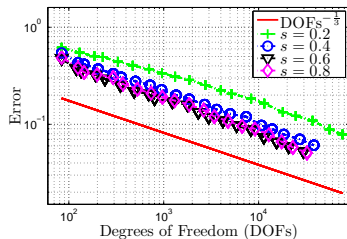
- **Computable estimator:** Restrict $\mathcal{W}(\mathcal{C}_{z'})$ to a discrete subspace

$$\{W \in \mathcal{W}(\mathcal{C}_{z'}) : W|_T \in \mathcal{P}_2(K) \otimes \mathbb{P}_2(I), \forall T = K \times I\}$$

$\mathcal{P}_2(K) = \mathbb{Q}_2(K)$ for rectangles, $\mathcal{P}_2(K) = \mathbb{P}_2(K) \oplus \mathbb{B}_3(K)$ for simplices.

Numerical experiment I

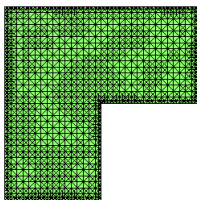
- Ω is the standard L-shaped domain in 2d.
- $f = 1$ which, for $s < \frac{1}{2}$, is **incompatible** with the problem and creates a **boundary layer**.
- **Experimental error and estimator:** error computed against a very fine discrete solution.



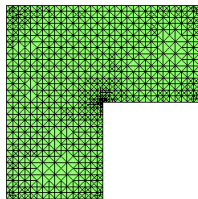
- **Optimal decay rate:** We get $DOF^{-1/3}$ for all s .

Numerical experiment II: Meshes

- **Meshes:** For $s < 1/2$ the solution exhibits a boundary layer.



$s = 0.2$



$s = 0.8$

- **Question:** Is there any theory on anisotropic adaptive approximation?

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Multilevel methods

If you do not diagonalize, **How do you solve the equations?**

We use **multilevel methods**.

- We have a sequence of nested meshes $\mathcal{T}_0 \preceq \mathcal{T}_1 \preceq \cdots \preceq \mathcal{T}_J$ which induces a sequence of nested FE spaces

$$\mathbb{V}_0 \subset \mathbb{V}_1 \subset \cdots \subset \mathbb{V}_J = \mathbb{V}.$$

- Introduce the space **macro** and **micro** decomposition

$$\mathbb{V} = \sum_{k=0}^J \mathbb{V}_k = \sum_{k=0}^J \sum_{j=1}^{\mathcal{M}_k} \mathbb{V}_{k,j}.$$

- Define a multigrid algorithm as a standard SSC² over this decomposition.
- This setting allows for **point** and **line** smoothers.

Properties of the decomposition

Lemma (stability and inverse inequality)

Let $v \in \mathbb{V}$ and $v = \sum_{i=1}^{\mathcal{N}} v_i$ be the *line* decomposition of v . Then we have the norm equivalence

$$\sum_{i=1}^{\mathcal{N}} \|v_i\|_{L^2(y^\alpha, \mathcal{C})}^2 \lesssim \|v\|_{L^2(y^\alpha, \mathcal{C})}^2 \lesssim \sum_{i=1}^{\mathcal{N}} \|v_i\|_{L^2(y^\alpha, \mathcal{C})}^2.$$

Moreover, for every $K \in \mathcal{T}_\Omega$ we have

$$\|\nabla v\|_{L^2(y^\alpha, K \times (0, \mathcal{Y}))} \lesssim h_K^{-1} \|v\|_{L^2(y^\alpha, K \times (0, \mathcal{Y}))}.$$

In both inequalities the hidden constant is independent of J and depends on y^α only through C_{2, y^α} .

- The proof relies *fundamentally* on the fact that $y^\alpha \in A_2$.

Convergence rate

Theorem (convergence of multigrid)

The contraction rate of the multigrid algorithm is

$$\delta \leq 1 - \frac{1}{1 + CJ}$$

where the constant C is independent of the mesh size, and it depends on y^α only through C_{2,y^α} .

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Space-time fractional parabolic problem

Let $T > 0$ be some positive time. Given $f : \Omega \rightarrow \mathbb{R}$ and $u_0 : \Omega \rightarrow \mathbb{R}$ find u such that

$$\partial_t^\gamma u + (-\Delta)^s u = f \text{ in } \Omega \times (0, T] \quad u|_{t=0} = u_0 \text{ in } \Omega.$$

Here $\gamma \in (0, 1]$.

For $\gamma = 1$ this is the usual time derivative, if $\gamma < 1$ we consider the **Caputo** derivative

$$\partial_t^\gamma u(x, t) = \frac{1}{\Gamma(1-\gamma)} \int_0^t \frac{\partial_r u(x, r)}{(t-r)^\gamma} dr = [I^{1-\gamma} \partial_r u(x, \cdot)](t),$$

where I^σ is the *Riemann-Liouville* fractional integral of order σ .

Nonlocality in space and time!

We will overcome the nonlocality in space using the **Caffarelli-Silvestre extension**.

Extended evolution problem

The Caffarelli-Silvestre extension turns our problem into a **quasistationary elliptic problem with dynamic boundary condition**

$$\begin{cases} -\nabla \cdot (y^\alpha \nabla \mathcal{U}) = 0, & \text{in } \mathcal{C}, t \in (0, T), \\ \mathcal{U} = 0, & \text{on } \partial_L \mathcal{C}, t \in (0, T), \\ d_s \partial_t^\gamma \mathcal{U} + \frac{\partial \mathcal{U}}{\partial \nu^\alpha} = d_s f, & \text{on } \Omega \times \{0\}, t \in (0, T), \\ \mathcal{U} = \mathbf{u}_0, & \text{on } \Omega \times \{0\}, t = 0. \end{cases}$$

Connection: $\mathbf{u} = \mathcal{U}(x, 0)$, $\alpha = 1 - 2s$.

Nonlocality just in time!

Weak formulation: seek $\mathcal{U} \in \mathbb{V}$ such that for a.e. $t \in (0, T)$,

$$\begin{cases} \int_{\Omega} \partial_t^\gamma \mathcal{U}(x, 0) \phi(x, 0) dx + a(w, \phi) = \int_{\Omega} f \phi(x, 0) dx, \\ \mathcal{U}|_{t=0} = \mathbf{u}_0 \end{cases}$$

for all $\phi \in \mathring{H}_L^1(y^\alpha, \mathcal{C})$, where

$$a(w, \phi) = \frac{1}{d_s} \int_{\mathcal{C}} y^\alpha \nabla w \cdot \nabla \phi dx dy.$$

Discretization

- As in the elliptic case \mathcal{C} is **infinite**, but we have **exponential decay**.
- This allows us to consider a **truncated problem**.
- In doing so we commit only an **exponentially small error**

$$I^{1-\gamma} \|tr_{\Omega}(\mathcal{U} - v)\|_{L^2(\Omega)}^2 + \|\nabla(\mathcal{U} - v)\|_{L^2(0,T;L^2(y^\alpha, \mathcal{C}_y))}^2 \lesssim e^{-\sqrt{\lambda_1} \gamma}.$$

- For $\gamma = 1$ [▣], we consider **backward Euler**:
 - We initialize by setting $V^0(x, 0) = u_0$.
 - For $k = 0, \dots, \mathcal{K} - 1$, we find $V^{k+1} \in \mathring{H}_L^1(y^\alpha, \mathcal{C}_y)$ solution of

$$\tau^{-1}(V^{k+1}(\cdot, 0) - V^k(\cdot, 0), W(\cdot, 0))_{L^2(\Omega)} + a(V^{k+1}, W) = (f^{k+1}, W(\cdot, 0))_{L^2(\Omega)}$$

for all $W \in \mathring{H}_L^1(y^\alpha, \mathcal{C}_y)$, where $f^{k+1} = f(t^{k+1})$.

- **Unconditional stability**:

$$\|V^\tau(\cdot, 0)\|_{\ell^\infty(L^2(\Omega))}^2 + \|V^\tau\|_{\ell^2(\mathring{H}_L^1(y^\alpha, \mathcal{C}_y))}^2 \lesssim \|u_0\|_{L^2(\Omega)}^2 + \|f^\tau\|_{\ell^2(\mathbb{H}^{-s}(\Omega))}^2.$$

Error estimates for fully discrete schemes

Discretization in time and space: stability + consistency yield

- Error estimates for \mathcal{U} : $s \in (0, 1)$ and $\gamma \in (0, 1)$

$$\begin{aligned} [I^{1-\gamma} \|tr_{\Omega}(v^{\tau} - V_{\mathcal{T}_y}^{\tau})\|_{L^2(\Omega)}(T)]^{\frac{1}{2}} &\lesssim \tau^{\theta} + |\log \# \mathcal{T}_y|^{2s} \# \mathcal{T}_y^{\frac{-(1+s)}{n+1}} \\ \|v^{\tau} - V_{\mathcal{T}_y}^{\tau}\|_{\ell^2(\mathring{H}_L^1(y^{\alpha}, \mathcal{C}_y))} &\lesssim \tau^{\theta} + |\log \# \mathcal{T}_y|^s \# \mathcal{T}_y^{\frac{-1}{n+1}}. \end{aligned}$$

- Error estimates for u : $s \in (0, 1)$ and $\gamma \in (0, 1)$

$$\begin{aligned} [I^{1-\gamma} \|u^{\tau} - U^{\tau}\|_{L^2(\Omega)}(T)]^{\frac{1}{2}} &\lesssim \tau^{\theta} + |\log \# \mathcal{T}_y|^{2s} \# \mathcal{T}_y^{\frac{-(1+s)}{n+1}} \\ \|u^{\tau} - U^{\tau}\|_{\ell^2(\mathbb{H}^s(\Omega))} &\lesssim \tau^{\theta} + |\log \# \mathcal{T}_y|^s \# \mathcal{T}_y^{\frac{-1}{n+1}}, \end{aligned}$$

where $\theta < \frac{1}{2}$.

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Formulation

- Given $f \in \mathbb{H}^{-s}(\Omega)$ and an obstacle $\psi \in \mathbb{H}^s(\Omega) \cap C(\bar{\Omega})$ with $\psi \leq 0$ on $\partial\Omega$.
- Find $u \in \mathcal{K}$ such that

$$\langle (-\Delta)^s u, u - w \rangle \leq \langle f, u - w \rangle \quad \forall w \in \mathcal{K}$$

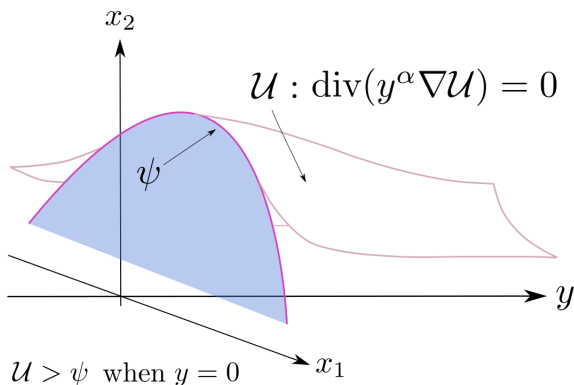
where

$$\mathcal{K} := \{w \in \mathbb{H}^s(\Omega) : w \geq \psi \text{ a.e. in } \Omega\}.$$

- **Nonlinear** and (because of $(-\Delta)^s$) **nonlocal** problem!
- Use the **Caffarelli-Silvestre extension**.

Thin obstacle problem

- We convert the fractional obstacle problem into a **thin obstacle** problem.



- The restriction $\mathcal{U} > \psi$ only applies when $y = 0$ (thin obstacle).

Truncation

- The domain \mathcal{C} is **infinite**.
- The energy of the solution **decays exponentially** in y .
- We truncate the cylinder $\mathcal{C}_y = \Omega \times (0, y)$ and consider a truncated problem.
- In doing this we only commit an **exponentially small** error

$$\|\nabla(\mathcal{U} - \mathcal{V})\|_{L^2(y^\alpha, \mathcal{C}_y)} \lesssim e^{-\sqrt{\lambda_1}y/8}.$$

Discretization

Discretize the truncation over an anisotropic mesh.

Theorem

If \mathcal{U} is the exact solution and $V_{\mathcal{T}_y}$ the discrete solution, then


$$\|\mathcal{U} - V_{\mathcal{T}_y}\|_{\mathring{H}_L^1(y^\alpha, \mathcal{C})} \lesssim |\log(\#\mathcal{T}_y)|^s (\#\mathcal{T}_y)^{-1/(n+1)},$$


where C depends on the Hölder moduli of smoothness of \mathcal{U} and \mathcal{V} , $\|f\|_{\mathbb{H}^{-s}(\Omega)}$ and $\|\psi\|_{\mathbb{H}^s(\Omega)}$.

- Optimal regularity in Ω^{\boxplus} : $u \in C^{1,s}$.
- This implies that $\partial_\nu^\alpha \mathcal{U}(\cdot, 0) \in C^{0,1-s}$.
- For y “small” use that \boxplus : $s \leq \frac{1}{2} \Rightarrow \mathcal{V} \in C^{0,2s}(\mathcal{C}_y)$ and $s > \frac{1}{2} \Rightarrow \mathcal{V} \in C^{1,2s-1}(\mathcal{C}_y)$.
- For y “big” use \boxplus $\mathcal{V} \in H^2(y^\beta, \mathcal{C}_y)$ with $\beta > 1 + 2\alpha$.

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 Nochetto, Otárola, AJS 2015

 Caffarelli, Salsa and Silvestre 2008

 Allen, Lindgren, and Petrosyan 2014

 Nochetto, Otárola, AJS 2015

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Formulation

- Define the energy

$$\mathcal{J}(v) = \frac{1}{2} \|v\|_{\mathbb{H}^s(\Omega)}^2 + \mathbf{1}_{\mathcal{K}}(v).$$

- We will study the (sub)gradient flow

$$u_t + \partial\mathcal{J}(u) \ni f \quad u|_{t=0} = u_0.$$

- Equivalently we have the evolution variational inequality

$$(u_t, u - \phi)_{L^2(\Omega)} + \langle (-\Delta)^s u, u - \phi \rangle \leq (f, u - \phi)_{L^2(\Omega)} \quad \forall \phi \in \mathcal{K}.$$

- Or the complementarity conditions

$$\min \{u_t + (-\Delta)^s u - f, u - \psi\} = 0.$$

The Caffarelli-Silvestre extension and truncation

- We will again overcome the **nonlocality** with the **Caffarelli-Silvestre extension** and consider

$$(\mathcal{U}_t(\cdot, 0), (\mathcal{U} - \phi)(\cdot, 0))_{L^2(\Omega)} + \frac{1}{d_s} \int_{\mathcal{C}} y^\alpha \nabla \mathcal{U} \nabla (\mathcal{U} - \phi) \, dx \, dy \leq (f, (\mathcal{U} - \phi)(\cdot, 0))_{L^2(\Omega)}$$

for all $\phi \in \mathring{H}_L^1(y^\alpha, \mathcal{C})$ with $\phi(\cdot, 0) \in \mathcal{K}$.

- We consider, again, a truncated problem over \mathcal{C}_γ :

$$\|(\mathcal{U} - \mathcal{V})(\cdot, 0)\|_{L^\infty(0, T; L^2(\Omega))} + \|\mathcal{U} - \mathcal{V}\|_{L^2(0, T; \mathring{H}_L^1(y^\alpha, \mathcal{C}_\gamma))} \lesssim e^{-\sqrt{\lambda_1} \gamma / 8}$$

Time discretization

- The energy \mathcal{J} is convex and lower semicontinuous $\implies \partial\mathcal{J}$ is maximal monotone.
- We use the implicit Euler method:

$$\left(\frac{V^{k+1} - V^k}{\tau}(\cdot, 0), (V^{k+1} - \phi)(\cdot, 0) \right)_{L^2(\Omega)} + \frac{1}{d_s} \int_{\mathcal{C}_y} y^\alpha \nabla V^{k+1} \nabla (V^{k+1} - \phi) \, dx \, dy \leq (f^{k+1}, (V^{k+1} - \phi)(\cdot, 0))_{L^2(\Omega)}$$

for all $\phi \in \mathring{H}_L^1(y^\alpha, \mathcal{C})$ with $\phi(\cdot, 0) \in \mathcal{K}$.

Time discretization

The general theory of gradient flows^[1] yields:

- If $u_0 \in \mathcal{K}$ and $f \in L^2(0, T; L^2(\Omega))$

$$\|(\mathcal{V} - V)(\cdot, 0)\|_{L^\infty(0, T; L^2(\Omega))} + \|\mathcal{V} - V\|_{L^2(0, T; \mathring{H}_L^1(y^\alpha, \mathcal{C}_y))} \lesssim \tau^{1/2}.$$

- If $u_0 \in \mathcal{K} \cap \mathbb{H}^{2s}(\Omega)$ and $f \in BV(0, T; L^2(\Omega))$

$$\|(\mathcal{V} - V)(\cdot, 0)\|_{L^\infty(0, T; L^2(\Omega))} + \|\mathcal{V} - V\|_{L^2(0, T; \mathring{H}_L^1(y^\alpha, \mathcal{C}_y))} \lesssim \tau.$$

These estimates are sharp!

Space discretization I: Minimal regularity

- Discretize in space using finite elements over an anisotropic mesh \mathcal{T}_y .
- If the discrete initial condition $V_{\mathcal{T}_y}^0$ satisfies

$$\|\nabla V_{\mathcal{T}_y}^0\|_{L^2(y^\alpha, \mathcal{C}_y)} \lesssim \|u_0\|_{\mathbb{H}^s(\Omega)}.$$

then 

$$\begin{aligned} \|(V - V_{\mathcal{T}_y})(\cdot, 0)\|_{L^\infty(0, T; L^2(\Omega))} + \|V - V_{\mathcal{T}_y}\|_{L^2(0, T; \dot{H}_L^1(y^\alpha, \mathcal{C}_y))} &\lesssim \\ \tau^\theta + \|\mathcal{V} - \Pi\mathcal{V}\|_{L^2(0, T; \dot{H}_L^1(y^\alpha, \mathcal{C}_y))}^{1/2} & \end{aligned}$$

where $\theta \in \{1/2, 1\}$ depends on the smoothness of f and u_0


- No regularity assumptions!

Space discretization II: Analysis with regularity

- Under certain conditions we have that 

$$u_t, (-\Delta)^s u \in \text{logLip}((0, T], C^{1-s}(\bar{\Omega})) \quad s \leq \frac{1}{3},$$

$$u_t, (-\Delta)^s u \in C^{\frac{1-s}{2s}}((0, T], C^{1-s}(\bar{\Omega})) \quad s > \frac{1}{3}.$$

- With this regularity 

$$\begin{aligned} & \| (V - V_{\mathcal{T}_y})(\cdot, 0) \|_{L^\infty(0, T; L^2(\Omega))} + \| V - V_{\mathcal{T}_y} \|_{L^2(0, T; \dot{H}_L^1(y^\alpha, \mathcal{C}_y))} \lesssim \\ & \tau + |\log \# \mathcal{T}_y|^s \left(\# \mathcal{T}_y^{-\frac{1}{n+1}} + \frac{\# \mathcal{T}_y^{-\frac{1+s}{n+1}}}{\tau^{1/2}} \right) \\ & \quad + \| \mathcal{V} - \Pi \mathcal{V} \|_{L^2(0, T; \dot{H}_L^1(y^\alpha, \mathcal{C}_y))} \end{aligned}$$

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