

Computation of operators in wavelet coordinates

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Overview

- Linear operator equation $Lu = g$ with $L : \mathcal{H} \rightarrow \mathcal{H}'$
- Riesz basis $\Psi = \{\psi_\lambda\}$ of \mathcal{H} , e.g. $u = \sum_\lambda \mathbf{u}_\lambda \psi_\lambda$
- Infinite dimensional matrix-vector system $\mathbf{L}\mathbf{u} = \mathbf{g}$, with $\mathbf{u} = (\mathbf{u}_\lambda)_\lambda$ and $\mathbf{L} : \ell_2 \rightarrow \ell_2$
- Convergent iterations such as $\mathbf{u}^{(i+1)} = \mathbf{u}^{(i)} + \alpha[\mathbf{g} - \mathbf{L}\mathbf{u}^{(i)}]$
- We **can** approximate $\mathbf{L}\mathbf{u}^{(i)}$ by a finitely supported vector
- How cheap can we **compute** this approximation?
- The answer will depend on L and Ψ

Linear operator equations

- Let Ω be an n -dimensional domain or smooth manifold
- $H^t \subset H^t(\Omega)$ be a subspace, and H^{-t} be its dual space
- Consider the problem of finding u from

$$Lu = g$$

- where $L : H^t \rightarrow H^{-t}$ is a self-adjoint elliptic operator of order $2t$
- and $g \in H^{-t}$ is a linear functional

Differential operators

- Partial differential operators of order $2t$

$$\langle v, Lu \rangle = \sum_{|\alpha|, |\beta| \leq t} \langle \partial^\alpha v, a_{\alpha\beta} \partial^\beta u \rangle,$$

- Example: The reaction-diffusion equation ($t = 1$)

$$\langle v, Lu \rangle = \int_{\Omega} \nabla v \cdot \nabla u + \kappa^2 vu,$$

Singular integral operators

- Boundary integral operators

$$[Lu](x) = \int_{\Omega} K(x, y)u(y)d\Omega_y$$

with the kernel $K(x, y)$ singular at $x = y$

- Example: The single layer operator for the Laplace BVP in 3-d domain ($t = -\frac{1}{2}$)

$$K(x, y) = \frac{1}{4\pi|x - y|}$$

Multiresolution analysis

- $\mathcal{S}_0 \subset \mathcal{S}_1 \subset \dots \subset H^t$ and $\tilde{\mathcal{S}}_0 \subset \tilde{\mathcal{S}}_1 \subset \dots \subset H^{-t}$
- $\dim \mathcal{S}_j, \dim \tilde{\mathcal{S}}_j = \mathcal{O}(2^{jn})$ (dyadic refinements)
- \mathcal{S}_j contains all piecewise pols of degree $d - 1$
- $\tilde{\mathcal{S}}_j$ contains all piecewise pols of degree $\tilde{d} - 1$
- \mathcal{S}_j is globally C^r -smooth

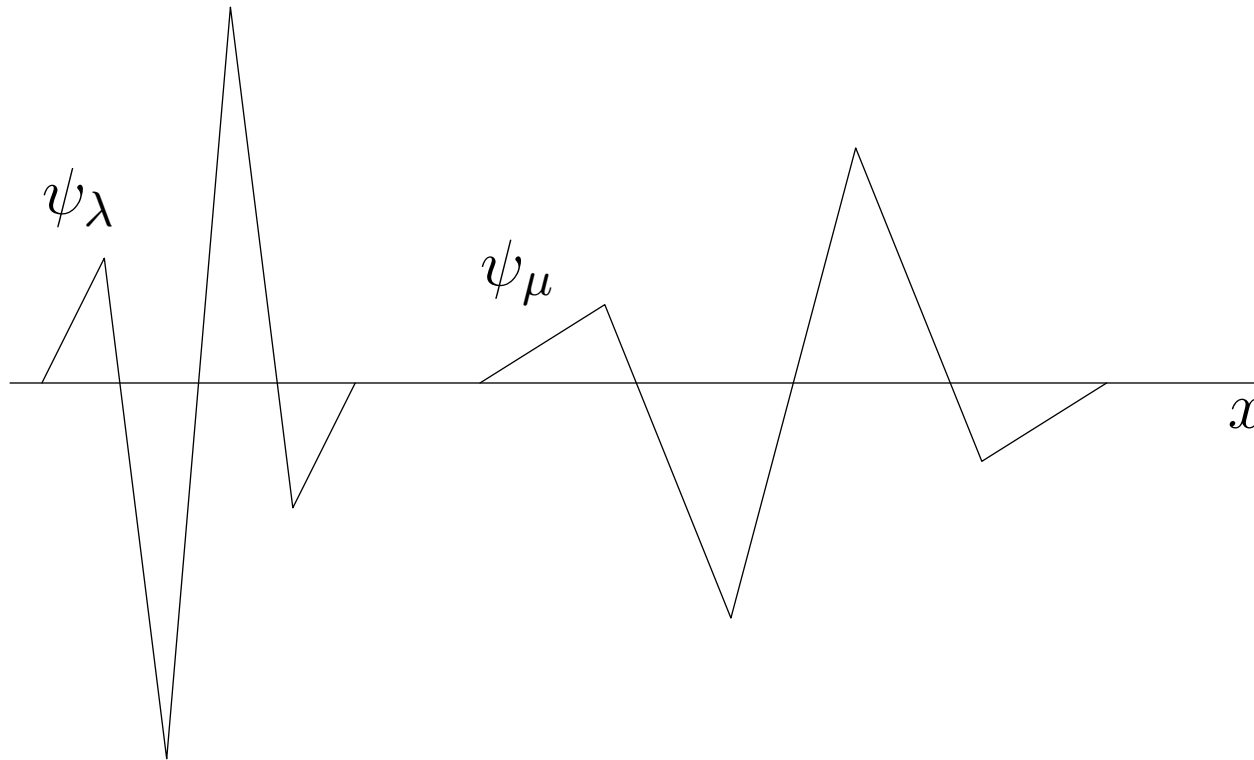
Wavelet bases

- $\Psi = \{\psi_\lambda : \lambda \in \Lambda\}$ is a Riesz basis for H^t
 - each $v \in H^t$ has a unique expansion

$$v = \sum_{\lambda \in \Lambda} \mathbf{v}_\lambda \psi_\lambda \quad \text{s.t.} \quad c \|\mathbf{v}\| \leq \|v\|_{H^t} \leq C \|\mathbf{v}\|$$

- For every index $\lambda \in \Lambda$, there is a number $|\lambda| \in \mathbb{N}_0$ called the **level** of ψ_λ
- $\text{span}\{\psi_\lambda : |\lambda| \leq j\} = \mathcal{S}_j$
- $\langle \psi_\lambda, v \rangle = 0$ for any $v \in \tilde{\mathcal{S}}_{|\lambda|-1}$
- $\text{diam}(\text{supp } \psi_\lambda) = \mathcal{O}(2^{-|\lambda|})$

Typical wavelets



- ψ_λ is a piecewise polynomial of degree $d - 1$
- $\int x^k \psi_\lambda(x) dx = 0$ for $k < \tilde{d}$ (\tilde{d} vanishing moments)

Galerkin methods

- Wavelet basis $\Psi_j := \{\psi_\lambda : |\lambda| \leq j\}$ of \mathcal{S}_j
- **Stiffness** $\mathbf{L}_{(j)} = \langle L\psi_\lambda, \psi_\mu \rangle_{|\lambda|, |\mu| \leq j}$
- **load** $\mathbf{g}_{(j)} = \langle g, \psi_\lambda \rangle_{|\lambda| \leq j}$
- Linear equation in \mathbb{R}^{N_j} ($N_j := \dim \mathcal{S}_j$)

$$\mathbf{L}_{(j)} \mathbf{u}_{(j)} = \mathbf{g}_{(j)}$$

- $\mathbf{L}_{(j)} : \mathbb{R}^{N_j} \rightarrow \mathbb{R}^{N_j}$ SPD and $\mathbf{g}_{(j)} \in \mathbb{R}^{N_j}$
- $u_{(j)} = \sum_\lambda [\mathbf{u}_{(j)}]_\lambda \psi_\lambda$ **approximates** the solution of $Lu = g$

Galerkin approximation

- If $u \in H^s$ for some $s \in [t, d]$

$$\varepsilon_{(j)} := \|u_{(j)} - u\|_{H^t} \leq \mathcal{O}(2^{-j(s-t)})$$

- $N_j = \dim \mathcal{S}_j = \mathcal{O}(2^{jn})$

- $\varepsilon_{(j)} \leq \mathcal{O}(N_j^{-\frac{s-t}{n}})$

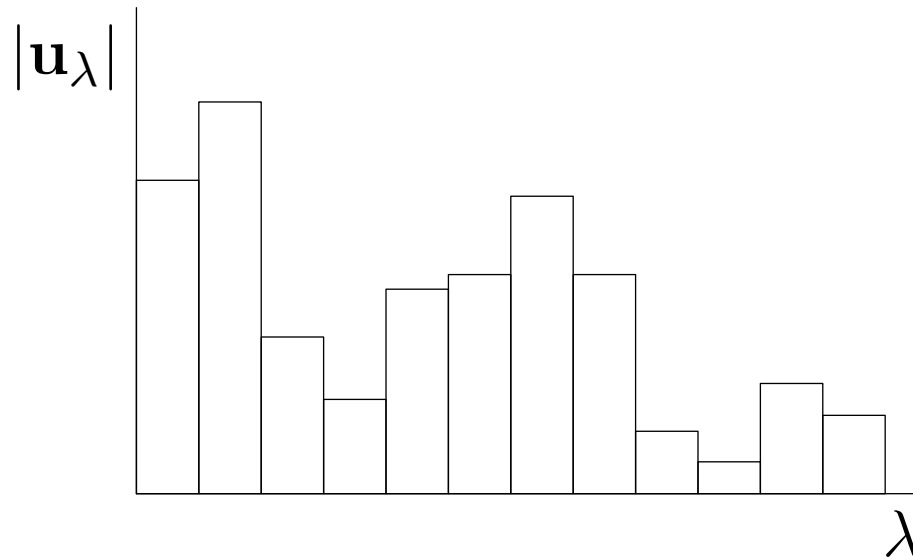
- Solve $\mathbf{L}_{(j)} \mathbf{u}_{(j)} = \mathbf{g}_{(j)}$ with CG \rightsquigarrow complexity $\mathcal{O}(N_j)$

- Similar estimates for FEM

- Better convergence? Adaptive methods?

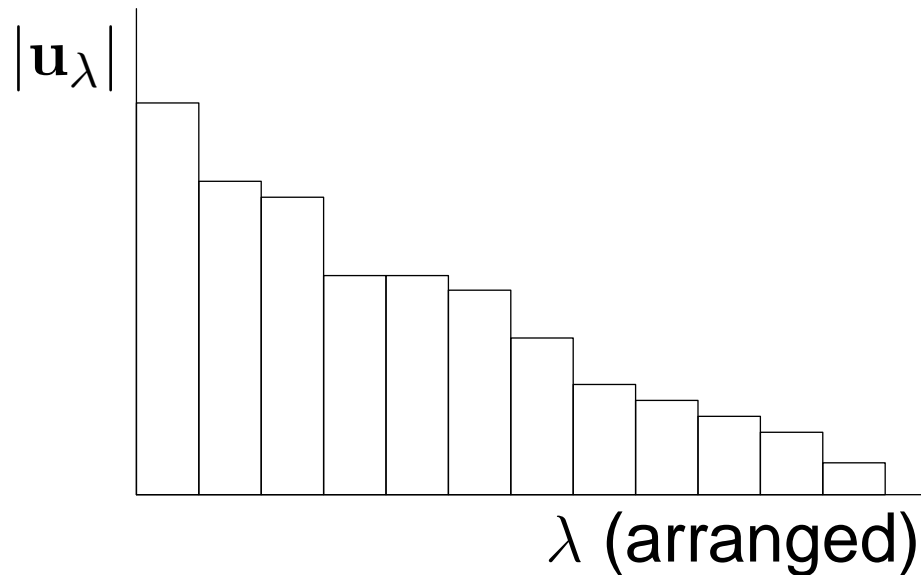
Nonlinear approximation

- Given $\mathbf{u} = (\mathbf{u}_\lambda)_\lambda \in \ell_2$
- Approximate \mathbf{u} using N coeffs



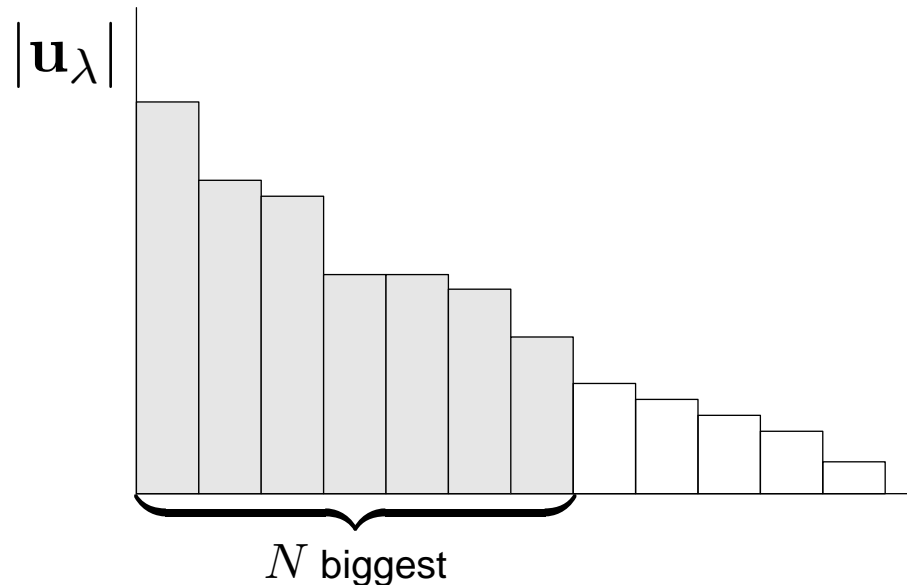
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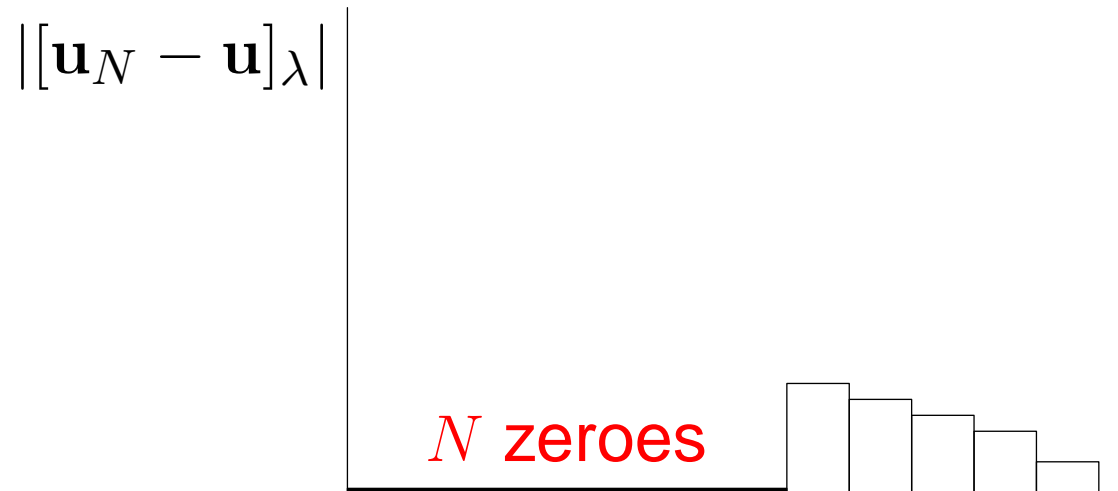
- Given $\mathbf{u} = (\mathbf{u}_\lambda)_\lambda \in \ell_2$
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- \mathbf{u}_N best approximation of \mathbf{u} with $\#\text{supp } \mathbf{u}_N \leq N$

Nonlinear approximation

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- \mathbf{u}_N best approximation of \mathbf{u} with $\#\text{supp } \mathbf{u}_N \leq N$

Nonlinear vs. linear approximation

- If $u \in B_{\tau, \tau}^s$ with $\frac{1}{\tau} = \frac{1}{2} + \frac{s-t}{n}$ for some $s < d$

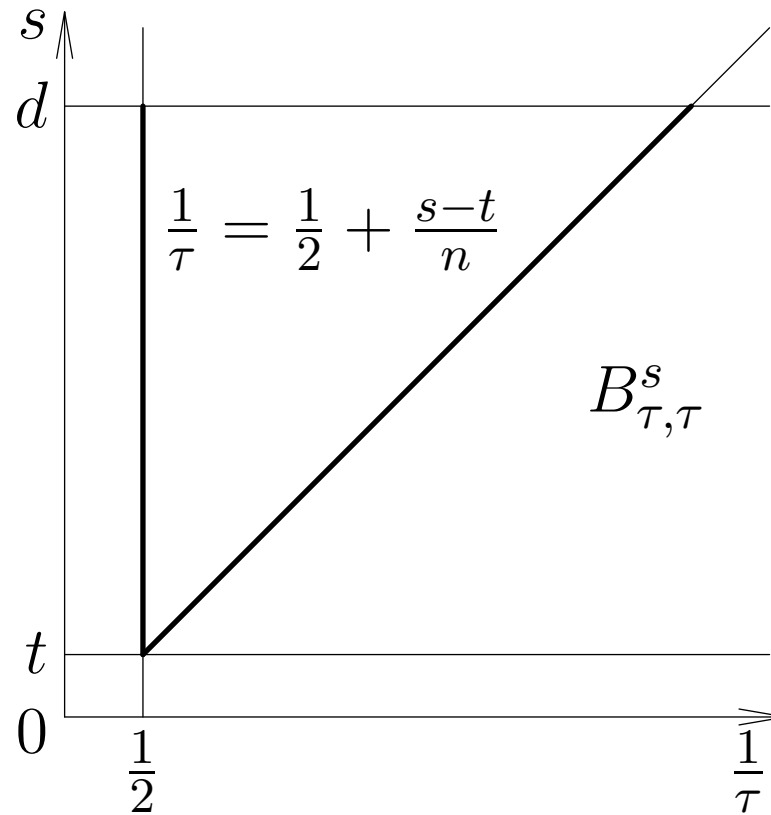
$$\varepsilon_N = \|\mathbf{u}_N - \mathbf{u}\| \leq \mathcal{O}(N^{-\frac{s-t}{n}})$$

- If $u \in H^s$ for some $s \leq d$, uniform refinement

$$\varepsilon_{(j)} = \|\mathbf{u}_{(j)} - \mathbf{u}\| \leq \mathcal{O}(N_j^{-\frac{s-t}{n}})$$

- $B_{\tau, \tau}^s$ is bigger than H^s

Besov vs. Sobolev regularity



- [Dahlke, DeVore]: $u \in B_{\tau, \tau}^d$ with $\frac{1}{\tau} = \frac{1}{2} + \frac{d-t}{n}$ "often"

Equivalent problem in ℓ_2

[Cohen, Dahmen, DeVore '02]

- Wavelet basis $\Psi = \{\psi_\lambda : \lambda \in \Lambda\}$
- **Stiffness** $\mathbf{L} = \langle L\psi_\lambda, \psi_\mu \rangle_{\lambda,\mu}$ and **load** $\mathbf{g} = \langle g, \psi_\lambda \rangle_\lambda$
- Linear equation in $\ell_2(\Lambda)$

$$\mathbf{L}\mathbf{u} = \mathbf{g}$$

- $\mathbf{L} : \ell_2(\Lambda) \rightarrow \ell_2(\Lambda)$ SPD and $\mathbf{g} \in \ell_2(\Lambda)$
- $u = \sum_\lambda \mathbf{u}_\lambda \psi_\lambda$ is **the solution** of $Lu = g$

Richardson iterations in ℓ_2

- $\mathbf{u}^{(0)} = \mathbf{0}$
- $\mathbf{u}^{(i+1)} = \mathbf{u}^{(i)} + \alpha[\mathbf{g} - \mathbf{L}\mathbf{u}^{(i)}] \quad i = 0, 1, \dots$
- \mathbf{g} and $\mathbf{L}\mathbf{u}^{(i)}$ are **infinitely** supported
- Approximate them by **finitely** supported sequences
- Algorithm **SOLVE** $[N, \mathbf{L}, \mathbf{g}] \rightarrow \mathbf{u}_{[N]}$ (N operations)
- $\#\text{supp } \mathbf{u}_{[N]} \leq \mathcal{O}(N)$ and

$$\varepsilon_{[N]} = \|\mathbf{u}_{[N]} - \mathbf{u}\| \rightarrow 0 \quad \text{as } N \rightarrow \infty$$

- $\varepsilon_{[N]}$ speed of convergence?

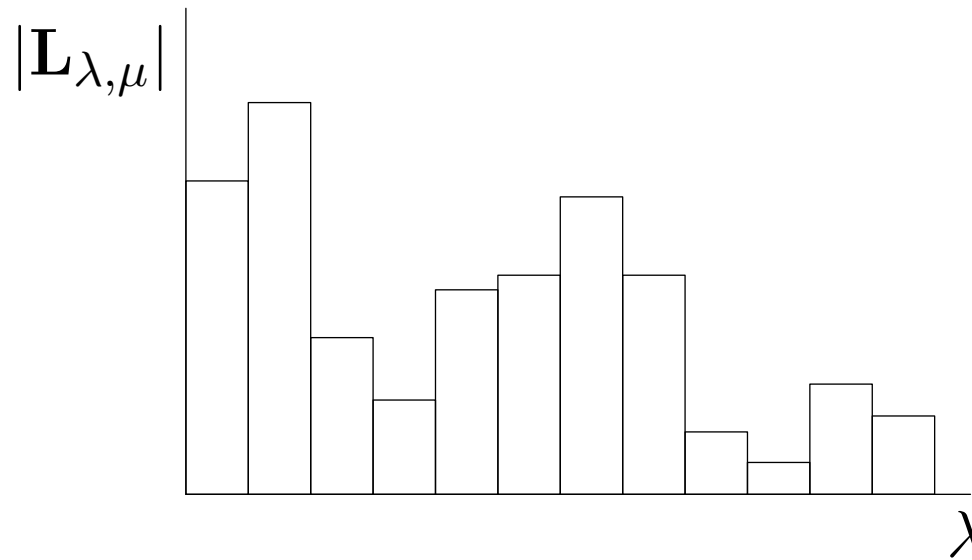
Complexity of SOLVE

- Matrix \mathbf{L} is called q^* -computable, when for each N one can construct an infinite matrix \mathbf{L}_N s.t.
 - for any $q < q^*$, $\|\mathbf{L}_N - \mathbf{L}\| \leq \mathcal{O}(N^{-q})$
 - having in each column $\mathcal{O}(N)$ non-zero entries
 - whose computation takes $\mathcal{O}(N)$ operations
 - [CDD'02]: Suppose that
 - $\|\mathbf{u}_N - \mathbf{u}\| \leq \mathcal{O}(N^{-s})$ $[s < \frac{d-t}{n}]$
 - \mathbf{L} is q^* -computable with $q^* > s$
- then for suitable \mathbf{g} , $\mathbf{u}_{[N]} = \text{SOLVE}[N, \mathbf{L}, \mathbf{g}]$ satisfies

$$\|\mathbf{u}_{[N]} - \mathbf{u}\| \leq \mathcal{O}(N^{-s})$$

Computability

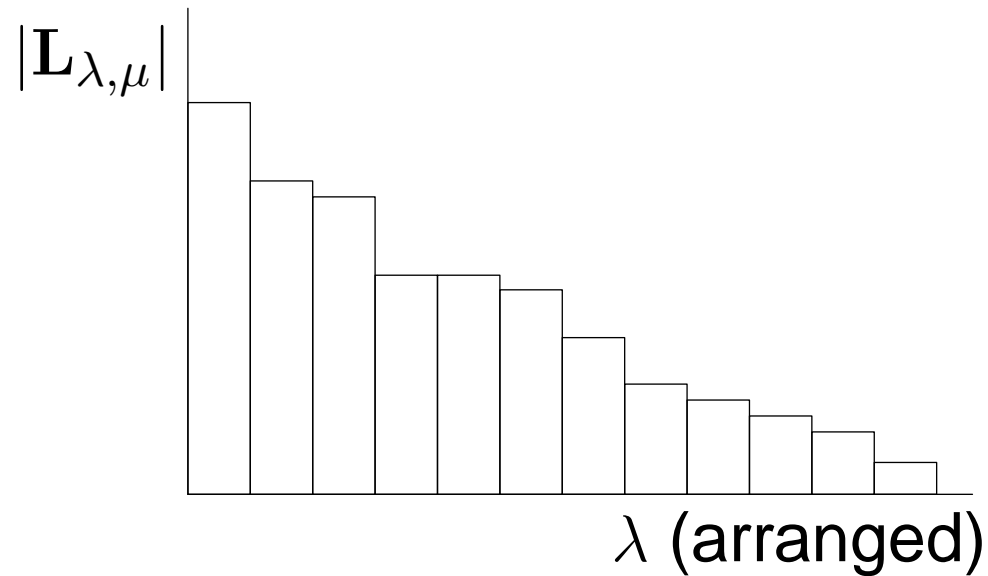
- $[\mathbf{L}_{\lambda,\mu}]_{\lambda \in \Lambda}$ – μ -th column



- Approximate by N entries?

Computability

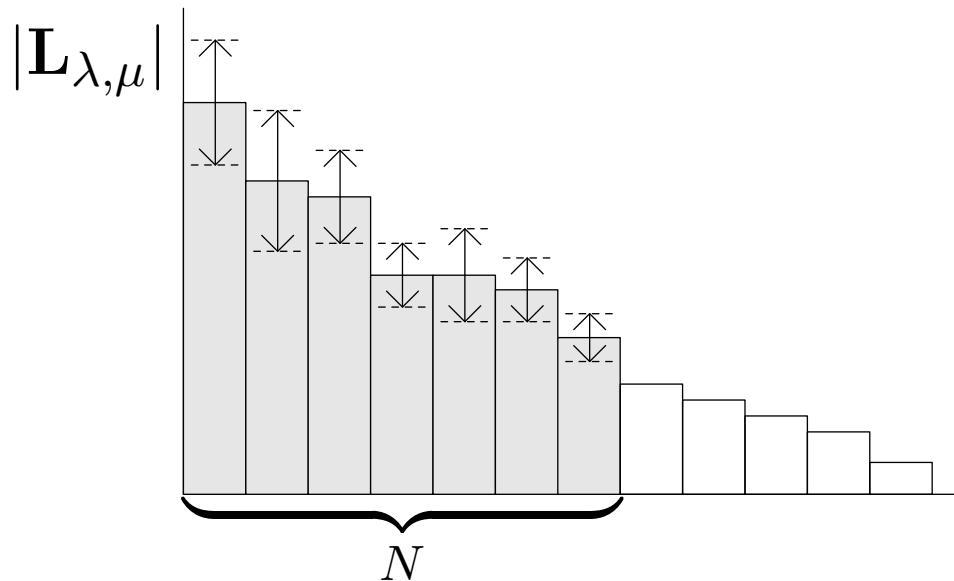
- $[\mathbf{L}_{\lambda,\mu}]_{\lambda \in \Lambda}$ – μ -th column arranged by modulus



- N biggest entries?

Computability

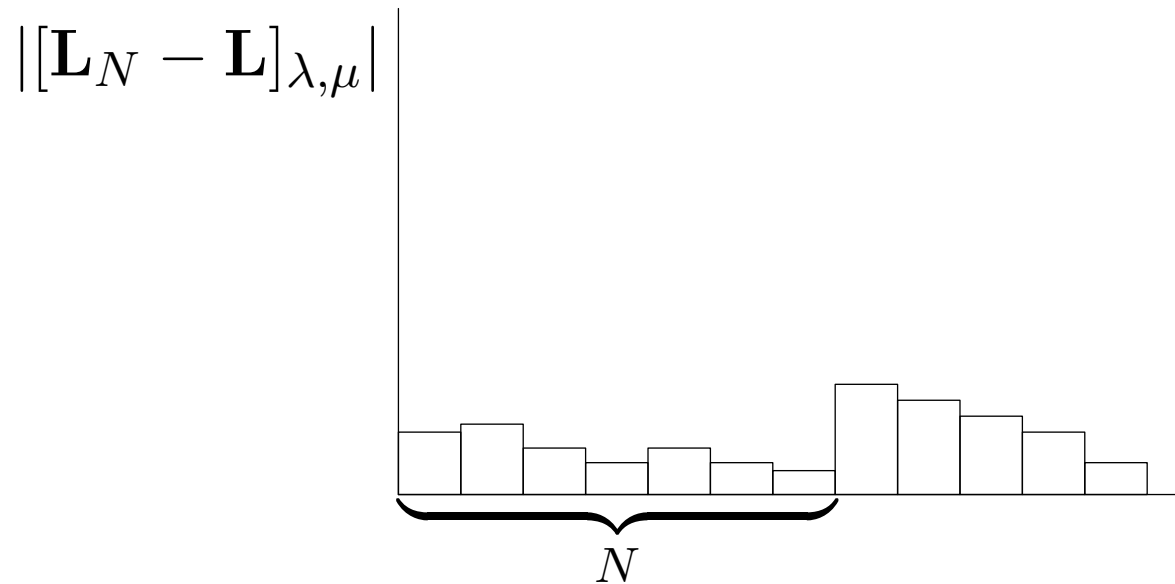
- $[\mathbf{L}_{\lambda,\mu}]_{\lambda \in \Lambda}$ – μ -th column



- Compute the N biggest entries

Computability

- The μ -th column of the difference



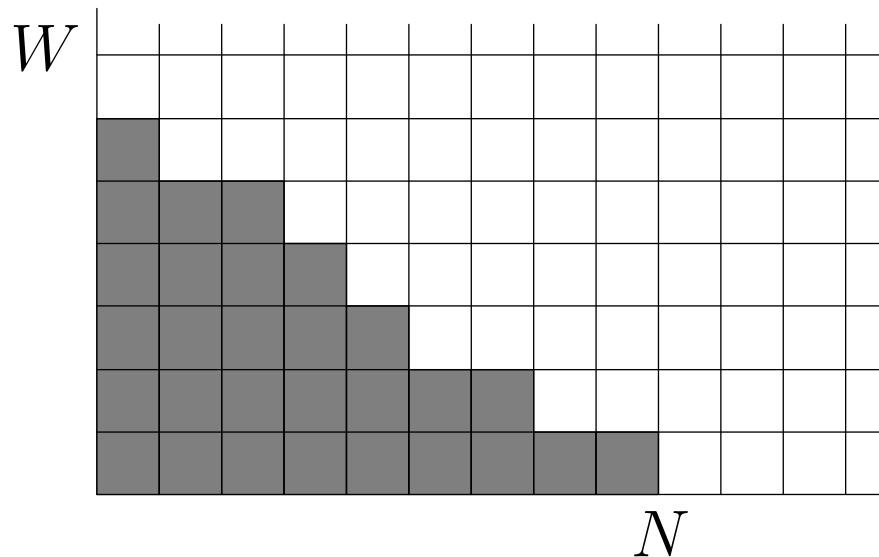
- Need to locate the biggest entries **a priori**

Compressibility

- \mathbf{L} is called q^* -compressible, when \mathbf{L} is q^* -computable assuming each entry of \mathbf{L} is available at unit cost
 - [CDD'01], [Stevenson '04]: Suppose
 - $\{\psi_\lambda\}$ are piecewise polynomial wavelets that
 - are sufficiently smooth and
 - have sufficiently many vanishing moments
 - \mathbf{L} is either differential or singular integral operator
- then \mathbf{L} is q^* -compressible for some $q^* \geq \frac{d-t}{n}$ ($> s$)

Computability

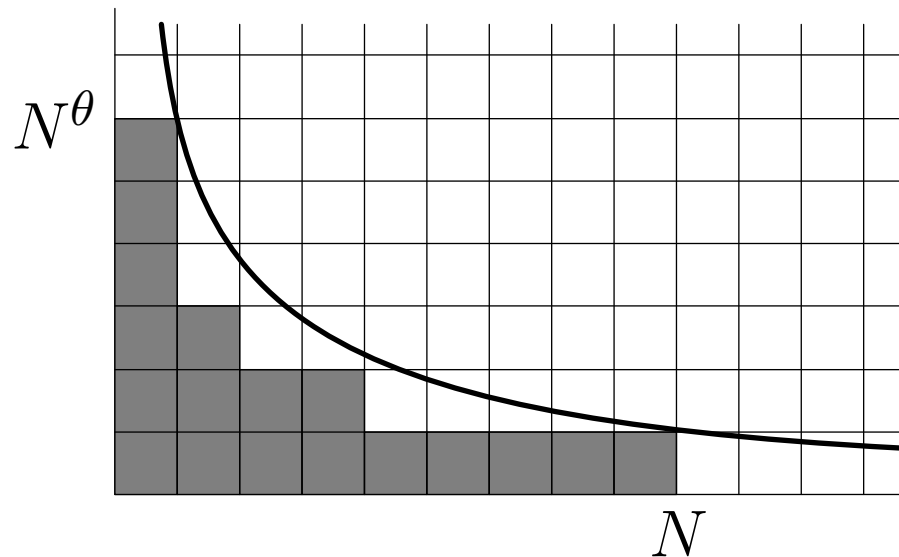
- Distribute computational works over the entries



- Require: shaded area = $\mathcal{O}(N)$

Computability

- Distribute computational works over the entries



- $p(x) = N^\theta x^{-\varrho} \rightsquigarrow$ when $\theta \leq \varrho < 1$

$$\int_0^N N^\theta x^{-\varrho} dx = \frac{1}{1-\varrho} N^{1+\theta-\varrho} = \mathcal{O}(N)$$

Computability

[T.G., Stevenson '04], [T.G.'04]: Suppose

- $\{\psi_\lambda\}$ are piecewise polynomial wavelets that
 - are sufficiently smooth and
 - have sufficiently many vanishing moments
- \mathbf{L} is either differential or singular integral operator

then \mathbf{L} is q^* -computable for some $q^* \geq \frac{d-t}{n} \quad (> s)$

So the adaptive wavelet method has the optimal convergence rate and optimal computational complexity

References

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- R.P. Stevenson. On the compressibility of operators in wavelet coordinates. *SIAM J. Math. Anal.*, 35(5):1110–1132, 2004.
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