

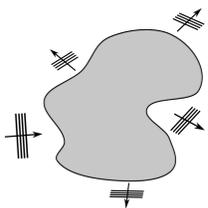
Regularization and Numerical Solution of the Inverse Scattering Problem Using Shearlet Frames

joint work with Gitta Kutyniok and Volker Mehrmann (both TU Berlin)

Introduction

Inverse Scattering

- The *inverse scattering problem* deals with reconstruction of objects from emitted waves.
- The waves can be, for instance, *acoustic* or *electromagnetic waves*.



- This problem is known to be an *ill-posed problem*. Therefore a regularization is necessary.

Ansatz

- Choose *cartoon-like images* as a model for scattering objects.
- Shearlet systems* allow for *sparse representations* of a large class of cartoon-like images.
- Employ the sparse representations of the scattering objects to *regularize* the problem.

Acoustic Scattering

Helmholtz equation in inhomogeneous media with wave number k_0 :

$$\Delta u + k_0^2(1-f)u = 0, \text{ with } u = u^s + u^{\text{inc}}$$

- Contrast function* $f \in L^2(\mathbb{R}^2)$ describes scattering object.
- Incident wave* u^{inc} is used to stimulate the medium.

The setup

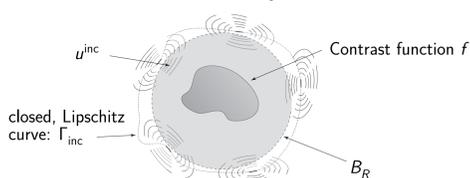


Figure 2: Setup with domain B_R contrast function f and single layer potentials as incident waves.

Solution operator \mathcal{S} :

$$\mathcal{S} : L^2(B_R) \times L^2(B_R) \rightarrow H_{\text{loc}}^2(B_R), \quad \mathcal{S}(f, u^{\text{inc}}) = u.$$

Measurement process and inverse problem

Let $\text{HS}(L^2(\Gamma_{\text{inc}}), L^2(\Gamma_{\text{meas}}))$ denote the space of *Hilbert-Schmidt Operators* from $L^2(\Gamma_{\text{inc}})$ to $L^2(\Gamma_{\text{meas}})$.

Definition 1 (Lechleiter, Kazimierski, Karamehmedovic; 2013).

For a Lipschitz curve Γ_{meas} the *multi-static measurement operator* \mathcal{N} is defined as

$$\mathcal{N} : L_{\text{Im} \geq 0}^2(B_R) \rightarrow \text{HS}(L^2(\Gamma_{\text{inc}}), L^2(\Gamma_{\text{meas}})), \quad f \mapsto N_f,$$

where

$$N_f : L^2(\Gamma_{\text{inc}}) \rightarrow L^2(\Gamma_{\text{meas}}), \quad \phi \mapsto \mathcal{S}(f, SL_{\Gamma_{\text{inc}}} \phi)|_{\Gamma_{\text{meas}}}.$$

The *inverse scattering problem* is that of reconstructing f from N_f .

Model Assumption

We assume, that the scatterer is a *cartoon-like function*.

Definition 2. A function $f \in L^2((0,1)^2)$ such that there exists $B \subset (0,1)^2$ such that ∂B twice differentiable, $f_1, f_2 \in C_c^2((0,1)^2)$ and

$$f = f_1 + \chi_B f_2$$

is called *cartoon-like function*.

Cartoon-like images and representation systems

Error of best N-term approximation: For $f \in L^2(\mathbb{R}^2)$ and a dictionary $(\phi_i)_{i \in I}$ the best N -term approximation error is

$$\min_{\substack{\Lambda \subset \mathbb{N}, |\Lambda|=N, \\ \tilde{f}_N = \sum_{i \in \Lambda} c_i \phi_i}} \|f - \tilde{f}_N\|_{L^2(\mathbb{R}^2)}.$$

Theorem 1 (Donoho; 2001). For an arbitrary representation system $(\phi_i)_{i \in I} \subset L^2(\mathbb{R}^2)$, the minimally achievable asymptotic error of best N -term approximation for the class of cartoon-like functions is $O(N^{-2})$.

Shearlets

Shearlet Systems are multiscale representation systems build upon *parabolic scaling* and *shearing*.

- Elements of the shearlet system:

$$\psi_{j,k,m} := 2^{3j} \psi(S_k A_j \cdot - cm), j \geq 0, k \in \mathbb{Z}, m \in \mathbb{Z}^2.$$

- Parabolic scaling* and *shearing matrices*:

$$A_j = \begin{pmatrix} 2^j & 0 \\ 0 & 2^{j/2} \end{pmatrix}, S_k = \begin{pmatrix} 1 & k \\ 0 & 1 \end{pmatrix}.$$

Theorem 2 (Kutyniok, Lim; 2011). There are shearlet systems such that the error of the best N -term approximation for any cartoon-like image decays as $O(N^{-2} \log(N)^3)$ as $N \rightarrow \infty$.

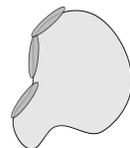


Figure 4: Shearlet elements overlapping the discontinuity curve of a cartoon-like function.

Schrödinger Scattering

- Schrödinger equation:* $f \in L^2(\mathbb{R}^2)$ and a wave number $k > 0, d \in \mathbb{S}^1$, we search for $u = u^s + e^{ik(x,d)}$

$$\Delta u + (f + k^2)u = 0, \quad \lim_{r \rightarrow \infty} r^{1/2} \left(\frac{\partial u^s(x)}{\partial r} - ik u^s(x) \right) = 0.$$

- Backscattering data:* Let $\theta \in [0, 2\pi]$, $\tau_\theta = (\cos(\theta), \sin(\theta))$, the backscattering amplitude is defined as

$$A(k, \theta) = \int_{\mathbb{R}^2} e^{ik \langle \tau_\theta, y \rangle} f(y) u(y) dy.$$

- Born Approximation*

$$f_B(x) = \int_{\mathbb{R}^2} e^{-i \langle \xi, x \rangle} A\left(\frac{|\xi|}{2}, \theta\right) d\xi, \quad \xi = |\xi| \tau_\theta.$$

Theorem 3 (Kutyniok, Mehrmann, Petersen; 2015). Let $\varepsilon > 0$, let $s \in \mathbb{N}, s \geq 2$, and let, for some $x_0 \in \mathbb{R}^2$, $f \in L^2(\mathbb{R}^2) \cap H^{s+\varepsilon}(x_0)$ be compactly supported and real valued, then $f_B \in H^s(x_0)$.

Corollary 1 (Kutyniok, Mehrmann, Petersen; 2015). If f is a cartoon-like function, then f_B is almost cartoon-like.

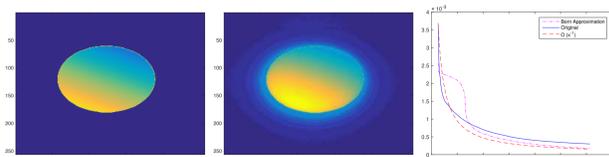


Figure: Decay of the shearlet coefficients (right) of the original function f (left) and its Born approximation f_B (middle). The decay is asymptotically of order $O(x^{-1})$.

Sparse Regularization

We employ the sparse approximation properties of shearlets to regularize the acoustic scattering problem.

- Noisy measurements* $N_{\text{meas}}^\varepsilon \in \text{HS}(L^2(\Gamma_{\text{inc}}), L^2(\Gamma_{\text{meas}}))$, s.t.

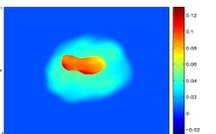
$$\|N_{\text{meas}}^\varepsilon - N_{f^\dagger}\|_{\text{HS}(L^2(\Gamma_{\text{inc}}), L^2(\Gamma_{\text{meas}}))} < \varepsilon$$

- For a shearlet system $(\psi_{j,k,m})$, a *regularization parameter* $\alpha > 0$ and $1 \leq p \leq 2$, we minimize the following *Tikhonov functional*:

$$\mathcal{J}_\alpha^\varepsilon(f) := \|\mathcal{N}(f) - N_{\text{meas}}^\varepsilon\|_{\text{HS}}^2 + \alpha \|\langle f, \psi_{j,k,m} \rangle\|_{\ell^p}^p. \quad (1)$$

Numerical Examples

Task: Reconstruct a cartoon-like contrast function (right) by minimizing (1).



Comparison results

Compare with solutions of the L^1 Tikhonov functional

$$\mathcal{J}_\alpha^\varepsilon(f) := \|\mathcal{N}(f) - N_{\text{meas}}^\varepsilon\|_{\text{HS}}^2 + \alpha \|f\|_{L^1(B_R)}. \quad (2)$$

Experimental results

We solve (1) and (2) by *non-linear Landweber iteration*.

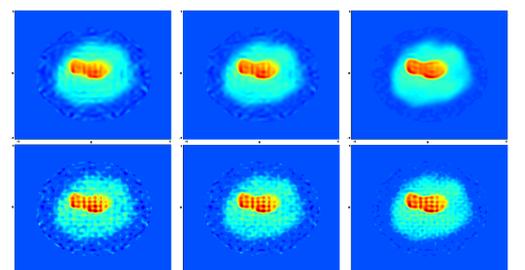


Figure: Reconstructed scatterers using shearlet regularization (top) and L^1 regularization (bottom). Noise levels $\varepsilon = 0.01, 0.005, 0.002$

$k_0 = 40$	Regularization method	Noise level	Rel. error	#iterations
1.	L^1 Tikhonov	0.02	0.1853	15
2.	Shearlets	0.02	0.1450	10
3.	No Penalty	0.02	0.2031	14
4.	L^1 Tikhonov	0.01	0.1204	40
5.	Shearlets	0.01	0.0845	24
6.	No Penalty	0.01	0.1471	47
7.	L^1 Tikhonov	0.005	0.1023	74
8.	Shearlets	0.005	0.0603	43
9.	No Penalty	0.005	0.1127	161
10.	L^1 Tikhonov	0.002	0.0797	303
11.	Shearlets	0.002	0.0381	100
12.	No Penalty	0.002	0.1014	978

Benefits of our approach

- Superior reconstructions to other penalties both *visually* and in terms of *relative errors*.
- Less iterations* necessary to meet stopping criterion for the shearlet based regularization compared to other methods.



Philipp Petersen

- Education
 - Master thesis at TU Berlin
 - Research Assistant in DFG Research Center TRR 109
- Research interests
 - Applied harmonic analysis (Wavelets, Shearlets,...)
 - Inverse problems
 - Numerical analysis of PDEs
 - Signal and image processing