



Monotony based imaging in EIT

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Joint work with Marcel Ullrich, Universität Mainz, Germany

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Forward operator of EIT:

$$\Lambda : \sigma \mapsto \Lambda(\sigma), \quad \text{"conductivity" } \mapsto \text{"measurements"}$$

- ▶ Conductivity: $\sigma \in L_+^\infty(\Omega)$
- ▶ Continuum model: $\Lambda(\sigma)$: Neumann-Dirichlet-operator

$$\begin{aligned} \Lambda(\sigma) : g &\mapsto u|_{\partial\Omega}, \quad \text{"applied current" } \mapsto \text{"measured voltage"} \\ \nabla \cdot (\sigma \nabla u) &= 0 \quad \text{in } \Omega, \quad \sigma \partial_\nu u|_{\partial\Omega} = g \quad \text{on } \partial\Omega. \end{aligned} \tag{1}$$

- ▶ Linear elliptic PDE theory:

$$\Lambda(\sigma) : L_\diamond^2(\partial\Omega) \rightarrow L_\diamond^2(\partial\Omega) \text{ linear, compact, self-adjoint}$$

Non-linear forward operator of EIT

$$\Lambda : \sigma \mapsto \Lambda(\sigma), \quad L_+^\infty(\Omega) \rightarrow \mathcal{L}(L_\diamond^2(\partial\Omega))$$

Inverse problem of EIT:

$$\Lambda(\sigma) \mapsto \sigma?$$

- ▶ Uniqueness ("Calderón problem"): Is Λ injective?
- ▶ Convergent numerical methods to reconstruct σ ?

Convergent numerical methods to reconstruct σ ?

- ▶ Newton iteration: almost no theory

Dobson (1992): (Local) convergence for regularized EIT equation.

Lechleiter/Rieder(2008): (Local) convergence for discretized setting.

- ▶ D-bar method: convergent 2D-implementation for $\sigma \in C^2$

Knudsen, Lassas, Mueller, Siltanen (2008)

In practice:

- ▶ large jumps in conductivity
- ▶ large interest in detecting shapes / inclusions / anomalies

Inclusion/shape detection problem:

Reconstruct $\text{supp}(\sigma - \sigma_0)$, σ_0 : reference conductivity.

$$\int_{\Omega} (\sigma_1 - \sigma_2) |\nabla u_1|^2 \, dx \leq (g, (\Lambda(\sigma_2) - \Lambda(\sigma_1))g)$$

u_1 solution corresponding to σ_1 and boundary current g .

Simple consequence:

$$\sigma_1 \leq \sigma_2 \implies \Lambda(\sigma_1) \geq \Lambda(\sigma_2)$$

- ▶ True conductivity: $\sigma = 1 + \chi_D$, D : unknown inclusion
 - ~~> $\Lambda(\sigma)$: measured data
- ▶ Test conductivity: $\kappa = 1 + \chi_B$, B : small ball
 - ~~> $\Lambda(\kappa)$ can be simulated for different balls B

Monotony:

$$B \subseteq D \implies \Lambda(\sigma) \geq \Lambda(\kappa)$$

Monotony based reconstruction algo. for EIT

Tamburrino/Rubinacci (2002)

- ▶ For all balls B , calculate $\Lambda(\kappa)$ and test whether $\Lambda(\sigma) \geq \Lambda(\kappa)$
 - ~~> Result: upper bound of D .
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Monotony based reconstruction (*up to now...*)

- ▶ Simple theory, simple implementation
- ▶ Regularization seems straight-forward
- ▶ Only reconstructs upper bound
- ▶ Expensive, requires one forward solution for each test ball
- ▶ Needs definiteness assumption

Comparison: Factorization Method *Kirsch 1998, Hanke/Brühl 2000*

- ▶ Complicated theory and implementation
- ▶ No known convergent regularization strategies
- ▶ Reconstructs exact shape (if $\Omega \setminus \overline{D}$ connected)
- ▶ Cheap, requires only one homogeneous forward solution
- ▶ Needs definiteness assumption

Theorem (H./Ullrich, 2010)

$\Omega \setminus \overline{D}$ connected. $\sigma = 1 + \chi_D$, $\kappa = 1 + \chi_B$.

$$B \not\subseteq D \implies \Lambda(\kappa) \not\geq \Lambda(\sigma).$$

↔ Monotony method detects exact shape.

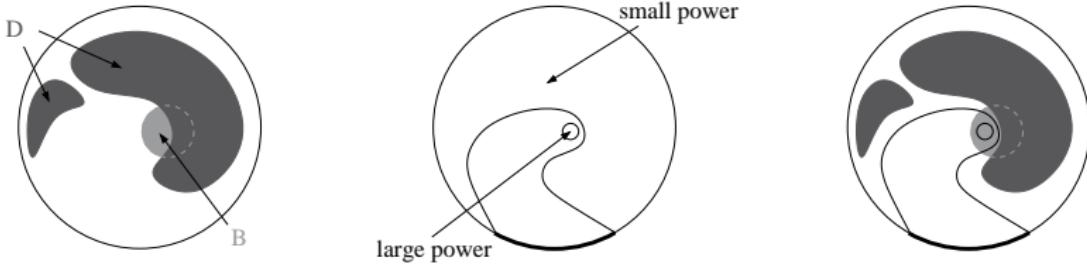
(Extensions possible for non-connected complement, inhomogeneous inclusions or background, continuous transitions between inclusion and background, . . .)

Converse monotony relation

Proof $(\sigma = 1 + \chi_D, \kappa = 1 + \chi_B)$

$$\int_{\Omega} (\kappa - \sigma) |\nabla u_{\kappa}|^2 \, dx \leq (g, (\Lambda(\sigma) - \Lambda(\kappa))g)$$

Apply localized potentials (H 2008) to control power term $|\nabla u_{\kappa}|^2$.



$$\rightsquigarrow \exists g : (g, (\Lambda(\sigma) - \Lambda(\kappa))g) \geq 0 \implies \Lambda(\sigma) \not\leq \Lambda(\kappa)$$

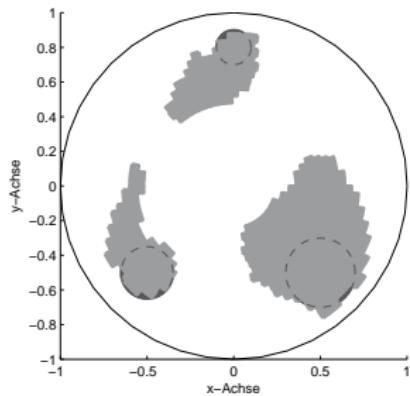
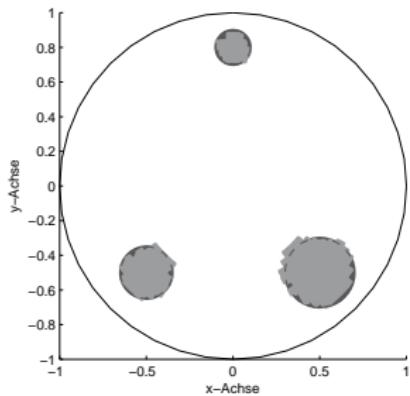
Computation costs:

- ▶ Using linear approx. of $\Lambda(\kappa)$ still fulfills monotony relation
(still exact, no linearization error)
- ~~> Fast implementation, requires only homogeneous forward solution
- ▶ Comp. cost equivalent to linearized methods or FM

Indefinite inclusions (larger and smaller than background conductivity)

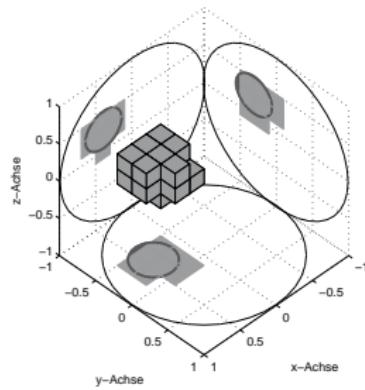
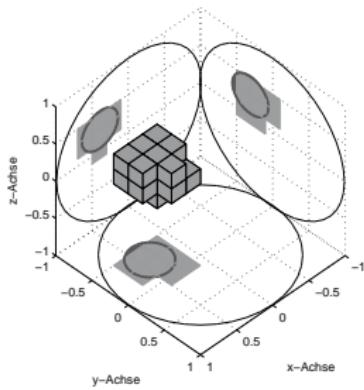
- ▶ can be treated by step-wise shrinking of larger test domains.

Numerical results



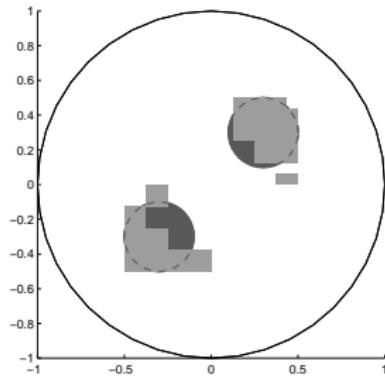
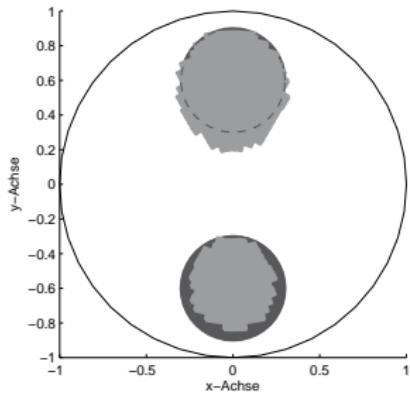
Reconstructions with exact data and with 0.1% noise.

Numerical results



Reconstructions with exact data and with 0.1% noise.

Numerical results



Reconstructions for smooth transitions between inclusion and background and for the indefinite case.

New results on the monotonicity method of Tamburrino and Rubinacci

- ▶ Method yields the exact shape not just an upper bound
- ▶ Method can be efficiently implemented by linearization
(while still reconstructing the exact shape)

Possible advantages

- ▶ Rigorous treatment of indefinite inclusions seems possible
- ▶ Convergent implementation of testing criteria seems possible

Goal

- ▶ Enhance linearized/iterative methods by exact shape reconstruction (*H./Seo SIMA 2010, H./Seo/Woo IEEE TMI 2010*)