

# Fast shape-reconstruction in electrical impedance tomography

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(joint work with Marcel Ullrich)

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Forward operator of EIT:

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

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

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

- ▶ Linear elliptic PDE theory:

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

## Inverse problem

Non-linear forward operator of EIT

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

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Inverse problem of EIT:

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

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- ▶ Uniqueness ("Calderón problem"): Is  $\Lambda$  injective?
- ▶ Convergent numerical methods to reconstruct  $\sigma$ ?

Convergent numerical methods to reconstruct  $\sigma$ ?

- ▶ 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

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Inclusion/shape detection problem:

$$\Lambda(\sigma) \mapsto \text{supp}(\sigma - \sigma_0)?, \quad \sigma_0: \text{reference conductivity.}$$

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# Shape detection

Promising approach: Factorization method (*Kirsch 1998*)

- ▶ *FM for EIT (1999–): Brühl, Hakula, Hanke, H., Hyvönen, Kirsch, Lechleiter, Nachman, Päivärinta, Pursiainen, Schappel, Schmitt, Seo, Teirilä*

Typical result:

$$z \notin \text{supp}(\sigma - \sigma_0) \quad \text{iff} \quad \lim_{\alpha \rightarrow 0} I_\alpha(z) = \infty.$$

( $I_\alpha(z)$ : indicator function)

Unsolved problems since 1998:

- ▶ Convergent regularization strategies for "infinity test"?
- ▶ Theory needs definiteness assumption, e.g.,  $\sigma \geq \sigma_0$  everywhere

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In this talk: A monotonicity based sampling method.

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

$u_1$  solution corresponding to  $\sigma_1$  and boundary current  $g$ .

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Simple consequence:

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

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# Monotony based imaging

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

Monotony:

$$B \subseteq D \implies 1 + \chi_B \leq 1 + \chi_D = \sigma \implies \Lambda(1 + \chi_B) \geq \Lambda(\sigma)$$

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Monotony based reconstruction algo. for EIT (*Tamburrino/Rubinacci 02*)

- ▶ For all  $B$ , calculate  $\Lambda(1 + \chi_B)$  & test whether  $\Lambda(1 + \chi_B) \geq \Lambda(\sigma)$
  - ↪ Result: upper bound of  $D$ .
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*Only an upper bound? Converse monotony relation?*

## Converse montony relation

Sample result (H./Ullrich)

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

$$B \not\subseteq D \implies \Lambda(1 + \chi_B) \not\cong \Lambda(\sigma).$$

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$\rightsquigarrow$  Monotony method detects exact shape.

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

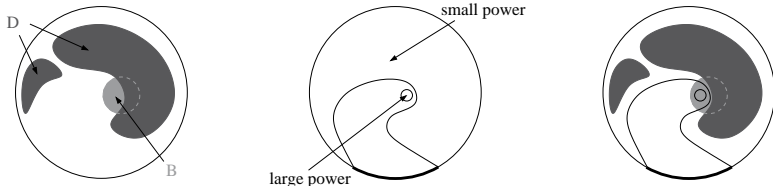


## Converse montony relation

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  $|\nabla u_{\kappa}|^2$ .



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

- ▶ Testing  $\Lambda(1 + \chi_B) \geq \Lambda(\sigma)$  is expensive. One forw. prob. per  $B$ .

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**Theorem** (H./Seo, SIAM J. Math. Anal. 2010)

Let  $\kappa, \sigma, \sigma_0$  piecewise analytic and  $\Lambda'(\sigma_0)\kappa = \Lambda(\sigma) - \Lambda(\sigma_0)$ . Then

$$\text{supp}_{\partial\Omega}\kappa = \text{supp}_{\partial\Omega}(\sigma - \sigma_0).$$

$\text{supp}_{\partial\Omega}$ : outer support (= support, if support is compact and has conn. complement)

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↪ Replacing  $\Lambda(1 + \chi_B)$  by its linear approx. should still recover the exact shape (linearization error does not affect the shape!)

Sample result (*H./Ullrich, submitted*)

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

$$\begin{aligned} B \subseteq D &\iff \Lambda(1 + \chi_B) \geq \Lambda(\sigma) \\ &\iff \Lambda(1) + \frac{1}{2}\Lambda'(1)\chi_B \geq \Lambda(\sigma). \end{aligned}$$

- ↪ Fast, requires only homogeneous forward solution
- ▶ Comp. cost equivalent to linearized methods or FM

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

Sample result (*H./Ullrich, submitted*)

Indefinite inclusions: Let  $\sigma = 1 + \chi_{D^+} - \frac{1}{2}\chi_{D^-}$  where  $D^+, D^-$  are disjoint and their union has conn. complement

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$$\begin{aligned} D^+ \cup D^- \subseteq B &\iff \Lambda(1 + \chi_B) \geq \Lambda(\sigma) \geq \Lambda(1 - \frac{1}{2}\chi_B) \\ &\iff \Lambda(1) + k\Lambda'(1)\chi_B \leq \Lambda(\sigma) \leq \Lambda(1) - \Lambda'(1)\chi_B. \end{aligned}$$

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- ▶ General (e.g., indefinite) cases can be treated by step-wise shrinking of larger test domains.

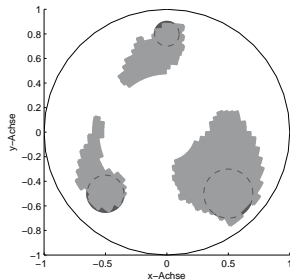
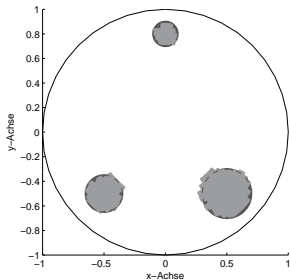
**Theorem** (H./Ullrich, submitted)

Let  $\sigma \in L_+^\infty(\Omega)$  be piecewise analytic. Then the (outer) support of  $\sigma - 1$  is the intersection of all sets  $B \subseteq \Omega$  (with no holes) s.t.

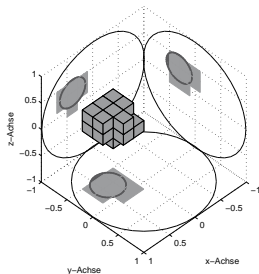
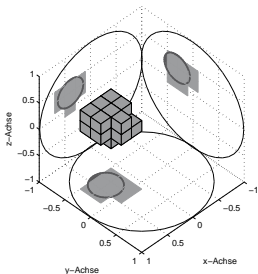
$$\exists \alpha > 0 : \quad \Lambda(1 + \alpha \chi_B) \leq \Lambda(\sigma) \leq \Lambda(1 - \frac{1}{\alpha} \chi_B).$$

or, equivalently, such that

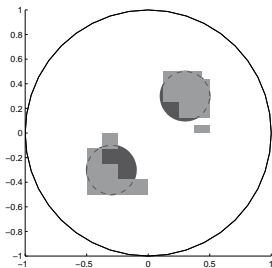
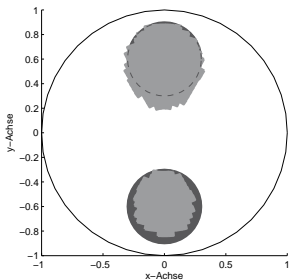
$$\exists \alpha > 0 : \quad \Lambda(1) + \alpha \Lambda'(1) \chi_B \leq \Lambda(\sigma) \leq \Lambda(1) - \alpha \Lambda'(1) \chi_B.$$



Reconstructions with exact data and with 0.1% noise.



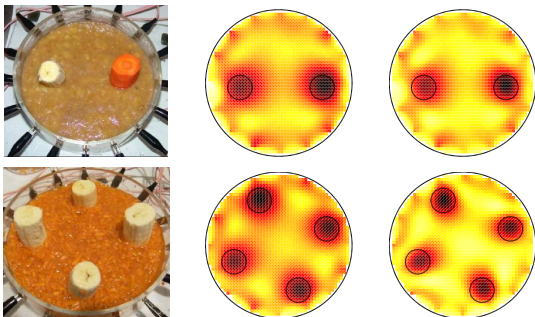
Reconstructions with exact data and with 0.1% noise.



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



Goal: Enhance linearized methods



Standard linearized method vs. heuristic combination with FM  
for frequency-difference EIT without ref. measurements

(H./Seo/Woo, IEEE Trans. Med. Imaging 2010)

# More goals / collaborations?

- ▶ Monotony/linearization arguments recover the shape.  
Also the coefficient on the outer boundary?  
*(Detect flaws/corrosions of steel in concrete?)*
- ▶ Detect inclusions in unknown background?  
Use bounds on contrast and monotony?
- ▶ Sampling/Factorization Method origins from inverse scattering.  
Extend our results to inverse scattering?  
Positivity up to finite-dimensional subspaces?
- ▶ More applications: complex conductivities, diffuse optical tomography, eddy currents, ...