



# Inverse coefficient problems and shape reconstruction

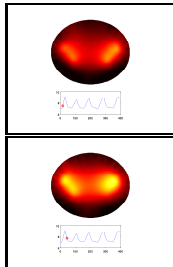
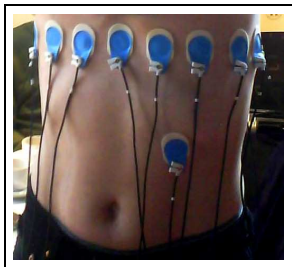
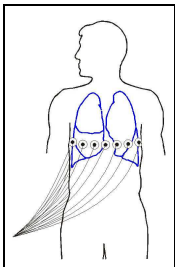
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# Electrical impedance tomography (EIT)



- ▶ Apply electric currents on subject's boundary
- ▶ Measure necessary voltages
- ↪ Reconstruct conductivity inside subject.

Images from BMBF-project on EIT

*(Hanke, Kirsch, Kress, Hahn, Weller, Schilcher, 2007-2010)*



## Mathematical Model

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Electrical potential  $u(x)$  solves

$$\nabla \cdot (\sigma(x) \nabla u(x)) = 0 \quad x \in \Omega$$

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$\Omega \subset \mathbb{R}^n$ : imaged body,  $n \geq 2$

$\sigma(x)$ : conductivity

$u(x)$ : electrical potential

Idealistic model for boundary measurements (**continuum model**):

$\sigma \partial_\nu u(x)|_{\partial\Omega}$ : applied electric current

$u(x)|_{\partial\Omega}$ : measured boundary voltage (potential)

## Calderón problem

Can we recover  $\sigma \in L_+^\infty(\Omega)$  in

$$\nabla \cdot (\sigma \nabla u) = 0, \quad x \in \Omega \quad (1)$$

from all possible Dirichlet and Neumann boundary values

$$\{(u|_{\partial\Omega}, \sigma \partial_\nu u|_{\partial\Omega}) : u \text{ solves (1)}\}?$$

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Equivalent: Recover  $\sigma$  from **Neumann-to-Dirichlet-Operator**

$$\Lambda(\sigma) : L_\diamond^2(\partial\Omega) \rightarrow L_\diamond^2(\partial\Omega), \quad g \mapsto u|_{\partial\Omega},$$

where  $u$  solves (1) with  $\sigma \partial_\nu u|_{\partial\Omega} = g$ .

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## Partial/local data

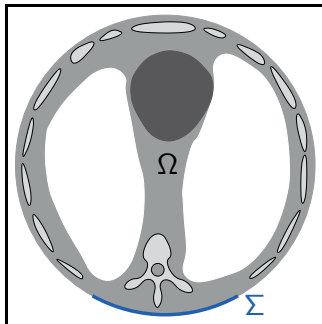
Measurements on open part of boundary  $\Sigma \subset \partial\Omega$ :  
 ( $\partial\Omega \setminus \Sigma$  is kept insulated.)

Recover  $\sigma$  from

$$\Lambda(\sigma) : L^2_{\diamond}(\Sigma) \rightarrow L^2_{\diamond}(\Sigma), \quad g \mapsto u|_{\Sigma},$$

where  $u$  solves  $\nabla \cdot (\sigma \nabla u) = 0$  with

$$\sigma \partial_{\nu} u|_{\Sigma} = \begin{cases} g & \text{on } \Sigma, \\ 0 & \text{else.} \end{cases}$$



## Challenges

Challenges in inverse coefficient problems such as EIT:

- ▶ Uniqueness
  - ▶ Is  $\sigma$  uniquely determined from the NtD  $\Lambda(\sigma)$ ?
- ▶ Non-linearity and ill-posedness
  - ▶ Reconstruction algorithms to determine  $\sigma$  from  $\Lambda(\sigma)$ ?
  - ▶ Local/global convergence results?
- ▶ Realistic data
  - ▶ What can we recover from real measurements?  
(Finite number of electrodes, realistic electrode models, ...)
  - ▶ Measurement and modelling errors? Resolution?

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**In this talk:** A simple strategy to attack these challenges  
(monotonicity + localized potentials)

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# Uniqueness

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## Uniqueness results

- ▶ Measurements on complete boundary (full data):  
*Calderón (1980), Druskin (1982+85), Kohn/Vogelius (1984+85), Sylvester/Uhlmann (1987), Nachman (1996), Astala/Päivärinta (2006)*
- ▶ Measurements on part of the boundary (local data):  
*Bukhgeim/Uhlmann (2002), Knudsen (2006), Isakov (2007), Kenig/Sjöstrand/Uhlmann (2007), H. (2008), Imanuvilov/Uhlmann/Yamamoto (2009+10), Kenig/Salo (2012+13)*
- ▶  $L^\infty$  coefficients are uniquely determined from full data in 2D.
- ▶ In all cases, piecew.-anal. coefficients are uniquely determined.
- ▶ Sophisticated research on uniqueness for  $\approx C^2$ -coefficients (based on CGO-solutions for Schrödinger eq.  $-\Delta u + qu = 0$ ,  $q = \frac{\Delta\sqrt{\sigma}}{\sqrt{\sigma}}$ ).



## Monotonicity

For two conductivities  $\sigma_0, \sigma_1 \in L^\infty(\Omega)$ :

$$\sigma_0 \leq \sigma_1 \implies \Lambda(\sigma_0) \geq \Lambda(\sigma_1)$$

This follows from

$$\int_{\Omega} (\sigma_1 - \sigma_0) |\nabla u_0|^2 \geq \int_{\Sigma} g (\Lambda(\sigma_0) - \Lambda(\sigma_1)) g \geq \int_{\Omega} \frac{\sigma_0}{\sigma_1} (\sigma_1 - \sigma_0) |\nabla u_0|^2$$

for all solutions  $u_0$  of

$$\nabla \cdot (\sigma_0 \nabla u_0) = 0, \quad \sigma_0 \partial_\nu u_0|_{\Sigma} = \begin{cases} g & \text{on } \Sigma, \\ 0 & \text{else.} \end{cases}$$

(e.g., Kang/Seo/Sheen 1997, Ikehata 1998)

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Can we prove uniqueness by controlling  $|\nabla u_0|^2$ ?

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## Localized potentials

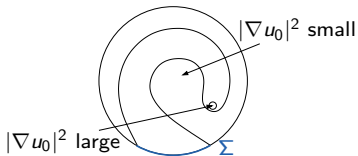
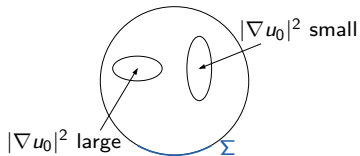
**Theorem** (H., *Inverse Probl. Imaging* 2008)

Let  $\sigma_0$  fulfill unique continuation principle (UCP),

$$\overline{D_1} \cap \overline{D_2} = \emptyset, \quad \text{and} \quad \Omega \setminus (\overline{D_1} \cup \overline{D_2}) \text{ be connected with } \Sigma.$$

Then there exist solutions  $u_0^{(k)}$ ,  $k \in \mathbb{N}$  with

$$\int_{D_1} |\nabla u_0^{(k)}|^2 dx \rightarrow \infty \quad \text{and} \quad \int_{D_2} |\nabla u_0^{(k)}|^2 dx \rightarrow 0.$$



## Consequences

- ▶ Back to Calderón: Let  $\Lambda(\sigma_0) = \Lambda(\sigma_1)$ ,  $\sigma_0$  fulfills (UCP).
- ▶ By monotonicity,

$$\int_{\Omega} (\sigma_1 - \sigma_0) |\nabla u_0|^2 \, dx \geq 0 \geq \int_{\Omega} \frac{\sigma_0}{\sigma_1} (\sigma_1 - \sigma_0) |\nabla u_0|^2 \, dx \quad \forall u_0$$

- ▶ Assume:  $\exists$  neighbourhood  $U$  of  $\Sigma$  where  $\sigma_1 \geq \sigma_0$  but  $\sigma_1 \neq \sigma_0$
- ↪ Potential with localized energy in  $U$  contradicts monotonicity

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Higher conductivity reachable by the bndry cannot be balanced out.

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**Corollary** (*Druskin 1982+85, Kohn/Vogelius, 1984+85*)

Calderón problem is uniquely solvable for piecw.-anal. conductivities.



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# Non-linearity

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## Non-linearity

Back to the non-linear forward operator of EIT

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

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Generic approach for inverting  $\Lambda$ : **Linearization**

$$\Lambda(\sigma) - \Lambda(\sigma_0) \approx \Lambda'(\sigma_0)(\sigma - \sigma_0)$$

$\sigma_0$ : known reference conductivity / initial guess / ...

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$\Lambda'(\sigma_0)$ : Fréchet-Derivative / sensitivity matrix.

$$\Lambda'(\sigma_0) : L_+^\infty(\Omega) \rightarrow \mathcal{L}(L_\diamond^2(\Sigma)).$$

$\rightsquigarrow$  Solve linearized equation for difference  $\sigma - \sigma_0$ .

**Often:**  $\text{supp}(\sigma - \sigma_0) \subset \Omega$  ("*shape*" / "*inclusion*")

# Linearization

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## Linear reconstruction method

e.g. NOSER (Cheney et al., 1990), GREIT (Adler et al., 2009)

Solve  $\Lambda'(\sigma_0)\kappa \approx \Lambda(\sigma) - \Lambda(\sigma_0)$ , then  $\kappa \approx \sigma - \sigma_0$ .

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- ▶ Multiple possibilities to measure residual norm and to regularize.
- ▶ No rigorous theory for single linearization step.
- ▶ Almost no theory for Newton iteration:
  - ▶ Dobson (1992): (Local) convergence for regularized EIT equation.
  - ▶ Lechleiter/Rieder(2008): (Local) convergence for discretized setting.
  - ▶ No (local) convergence theory for non-discretized case!  
Non-linearity condition (Scherzer / tangential cone cond.) still open problem
- ▶ D-bar method: convergent 2D-implementation for  $\sigma \in C^2$  and full bndry data (Knudsen, Lassas, Mueller, Siltanen 2008)



# Linearization

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## Linear reconstruction method

e.g. NOSER (Cheney et al., 1990), GREIT (Adler et al., 2009)

Solve  $\Lambda'(\sigma_0)\kappa \approx \Lambda(\sigma) - \Lambda(\sigma_0)$ , then  $\kappa \approx \sigma - \sigma_0$ .

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- ▶ **Seemingly**, no rigorous results possible for single lineariz. step.
- ▶ **Seemingly**, only justifiable for small  $\sigma - \sigma_0$  (local results).

**Here:** Rigorous and global(!) result about the linearization error.



## Linearization and shape reconstruction

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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

$$\text{supp}_{\Sigma}\kappa = \text{supp}_{\Sigma}(\sigma - \sigma_0)$$

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

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- ▶ Solution of lin. equation yields correct (outer) shape.
- ▶ No assumptions on  $\sigma - \sigma_0$ !
- ↪ Linearization error does not lead to shape errors.

*Taking the (wrong) reference current paths for reconstruction  
still yields the correct shape information!*



## Proof

- ▶ Linearization:  $\Lambda'(\sigma_0)\kappa = \Lambda(\sigma) - \Lambda(\sigma_0)$
- ▶ **Monotonicity**: For all "reference solutions"  $u_0$ :

$$\begin{aligned} & \int_{\Omega} (\sigma - \sigma_0) |\nabla u_0|^2 \, dx \\ & \geq \underbrace{\int_{\Sigma} g (\Lambda(\sigma_0) - \Lambda(\sigma)) g}_{=} \geq \int_{\Omega} \frac{\sigma_0}{\sigma} (\sigma - \sigma_0) |\nabla u_0|^2 \, dx. \\ & = \int_{\Sigma} g (\Lambda'(\sigma_0)\kappa) g = \int_{\Omega} \kappa |\nabla u_0|^2 \, dx \end{aligned}$$

- ▶ Use **localized potentials** to control  $|\nabla u_0|^2$

$\rightsquigarrow \text{supp}_{\Sigma} \kappa = \text{supp}_{\Sigma} (\sigma - \sigma_0)$

□

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In shape reconstruction problems we can avoid non-linearity.

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# Reconstruction from realistic data

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## Monotonicity based imaging

- ▶ Monotonicity:

$$\tau \leq \sigma \implies \Lambda(\tau) \geq \Lambda(\sigma)$$

- ▶ Idea: Simulate  $\Lambda(\tau)$  for test cond.  $\tau$  and compare with  $\Lambda(\sigma)$ .  
(*Tamburrino/Rubinacci 02, Lionheart, Soleimani, Ventre, ...*)
- ▶ Inclusion detection: For  $\sigma = 1 + \chi_D$  with unknown  $D$ , use  $\tau = 1 + \chi_B$ , with small ball  $B$ .

$$B \subseteq D \implies \tau \leq \sigma \implies \Lambda(\tau) \geq \Lambda(\sigma)$$

- ▶ Algorithm: Mark all balls  $B$  with  $\Lambda(1 + \chi_B) \geq \Lambda(\sigma)$
- ▶ Result: upper bound of  $D$ .

*Only an upper bound? Converse monotonicity relation?*



## Converse monotonicity relation

**Theorem** (H./Ullrich, *SIAM J. Math. Anal.*, to appear)

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

$$B \subseteq D \iff \Lambda(1 + \chi_B) \geq \Lambda(\sigma).$$

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$\rightsquigarrow$  Monotonicity method detects exact shape.

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For faster implementation:

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

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$\rightsquigarrow$  Linearized monotonicity method detects exact shape.

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**Proof:** Monotonicity + localized potentials

## General case

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**Theorem** (H./Ullrich). Let  $\sigma \in L_+^\infty(\Omega)$  be piecewise analytic.  
The intersection of all *hole-free*  $C \subseteq \overline{\Omega}$  with

$$\exists \alpha > 1 : \Lambda(1 + \alpha \chi_C) \leq \Lambda(\sigma) \leq \Lambda(1 - \chi_C/\alpha)$$

is identical to the (*outer*) support of  $\sigma - 1$ .

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- ▶ Result also holds with linearized condition

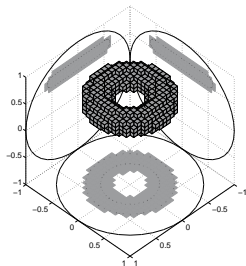
$$\exists \alpha > 1 : \Lambda(1) + \alpha \Lambda'(1) \chi_C \leq \Lambda(\sigma) \leq \Lambda(1) - \alpha \Lambda'(1) \chi_C.$$

- ▶ Result covers **indefinite case**,  
e.g.,  $\sigma = 1 + \chi_{D_1} - \frac{1}{2} \chi_{D_2}$

# Monotonicity based shape reconstruction

## Monotonicity based reconstruction

- ▶ is intuitive, yet rigorous
- ▶ is stable (no infinity or range tests)
- ▶ works for pcw. anal. conductivities (no definiteness conditions)
- ▶ requires only the reference solution



Theoretical results rely on idealized measurements (NtD-operators)

*Can we use these ideas for realistic measurements?*

*(finitely many electrodes, electrode models/shunting effects, uncertainties/noise, ...)*

## Realistic data & Uncertainties

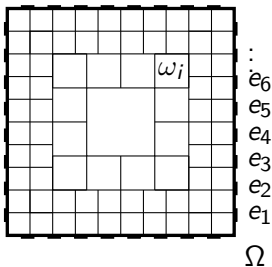
- ▶ Finite number of electrodes, CEM, noisy data  $\Lambda^\delta(\sigma)$
- ▶ Unknown background, e.g.,  $1 - \epsilon \leq \sigma_0(x) \leq 1 + \epsilon$
- ▶ Anomaly with some minimal contrast to background, e.g.,  

$$\sigma(x) = \sigma_0(x) + \kappa(x)\chi_D, \quad \kappa(x) \geq 1$$
- ▶ Can we **rigorously guarantee** to find inclusion  $D$ ?

### H./Ullrich: Monotonicity-based Rigorous Resolution Guarantee

- ▶ If  $D = \emptyset$ , methods return  $\emptyset$ .
- ▶ If  $D \supset \omega_j$  then it is detected.

(Here: 32 electrodes,  $\epsilon = 1\%$ ,  $\delta = 1.4\%$ )





## Conclusions

Using monotonicity and localized potentials we showed that

- ▶ Unique identifiability holds for *piecewise analytic* parameters.
- ▶ In shape reconstruction problems we can *avoid non-linearity*.
- ▶ *Resolution guarantees* for locating anomalies in unknown backgrounds with realistic finite precision data are possible.

Major limitations / open problems for our approach

- ▶ Piecewise analyticity required to prevent infinite oscillations.
- ▶ Voltage has to be measured on current-driven electrodes.