



# Inverse coefficient problems in elliptic partial differential equations

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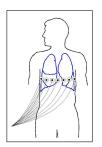
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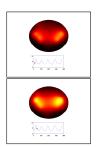




## Electrical impedance tomography (EIT)







- 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

### Electrical potential u(x) solves

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

 $\Omega \subset \mathbb{R}^n$ : 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

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

from all possible Dirichlet and Neumann boundary values

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

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

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

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



## Partial/local data

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

Recover  $\sigma$  from

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

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

$$\sigma \partial_{\nu} u|_{\Sigma} = \left\{ egin{array}{ll} g & ext{on } \Sigma, \\ 0 & ext{else.} \end{array} 
ight.$$





## Optimization and Inverse Problems

### Given measurements $\Lambda_{\text{measured}}$

- ▶ Inverse Problem: Solve  $\Lambda(\sigma) = \Lambda_{\text{measured}}$
- ▶ Optimization: Minimize  $\|\Lambda(\sigma) \Lambda_{\text{measured}}\|^2$  (+ regularization)

### Special challenges in inverse problems:

- Uniqueness is crucial.
- Local minima are usually useless.
- Convergence of iterates to true solution is crucial.
- ► Additional assumptions can often not be justified. (sufficient optimality conditions, constraint qualifications, source conditions, . . . )

Inverse coeff. problems pose major challenges even for simple PDEs.





## 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?

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





## Uniqueness



### 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)
- $ightharpoonup L^{\infty}$  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.  $-\Delta u + qu = 0$ ,  $q = \frac{\Delta\sqrt{\sigma}}{\sqrt{\sigma}}$ ).

Next: Uniqueness proof using monotonicity and loc. potentials.



### 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

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

for all solutions  $u_0$  of

$$abla \cdot (\sigma_0 
abla u_0) = 0, \quad \sigma_0 \partial_{\nu} u_0|_{\Sigma} = \left\{ egin{array}{ll} \mathsf{g} & \text{on } \Sigma, \\ 0 & \text{else.} \end{array} 
ight.$$

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

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





## Localized potentials

Theorem (H., 2008)

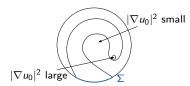
Let  $\sigma_0$  fulfill unique continuation principle (UCP),

$$\overline{D_1} \cap \overline{D_2} = \emptyset, \quad \text{ and } \quad \Omega \setminus \left(\overline{D}_1 \cup \overline{D}_2\right) \text{ be connected with } \Sigma.$$

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

$$\int_{D_1} \left| \nabla u_0^{(k)} \right|^2 \ \mathrm{d} x \to \infty \quad \text{ and } \quad \int_{D_2} \left| \nabla u_0^{(k)} \right|^2 \ \mathrm{d} x \to 0.$$





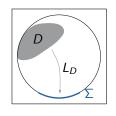


## Proof 1/3

### Virtual measurements:

$$L_D: H^1_{\diamond}(D)' \to L^2_{\diamond}(\Sigma), \quad f \mapsto u|_{\Sigma}, \text{ with }$$

$$\int_{\Omega} \sigma \nabla u \cdot \nabla v \, dx = \langle f, v|_{D} \rangle \quad \forall v \in H^{1}_{\diamond}(D).$$



By (UCP): If 
$$\overline{D_1} \cap \overline{D_2} = \emptyset$$
 and  $\Omega \setminus (\overline{D_1} \cup \overline{D_2})$  is connected with  $\Sigma$ , then  $\mathcal{R}(L_{D_1}) \cap \mathcal{R}(L_{D_2}) = 0$ .

Sources on different domains yield different virtual measurements.

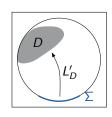


## Proof 2/3

### Dual operator:

$$L_D':\ L_\diamond^2(\Sigma) o H_\diamond^1(D), \quad g \mapsto u|_D,, \ \ \text{with}$$

$$abla \cdot (\sigma 
abla u) = 0, \quad \sigma \partial_{\nu} u|_{\Sigma} = \left\{ egin{array}{ll} g & ext{on } \Sigma, \\ 0 & ext{else.} \end{array} 
ight.$$



Evaluating solutions on D is dual operation to virtual measurements.



## Proof 3/3

### Functional analysis:

 $X, Y_1, Y_2$  reflexive Banach spaces,  $L_1 \in \mathcal{L}(Y_1, X)$ ,  $L_2 \in \mathcal{L}(Y_1, X)$ .

$$\mathcal{R}(L_1) \subseteq \mathcal{R}(L_2) \iff \|L_1'x\| \lesssim \|L_2'x\| \ \forall x \in X'.$$

Here: 
$$\mathcal{R}(L_{D_1}) \not\subseteq \mathcal{R}(L_{D_2}) \implies \|u_0|_{D_1}\|_{H_0^1} \not\lesssim \|u_0|_{D_2}\|_{H_0^1}.$$

If two sources do not generate the same data, then the respective evaluations are not bounded by each other.

Note:  $H^1_{\diamond}(D)'$ -source  $\longleftrightarrow$   $H^1_{\diamond}(D)$ -evaluation.



### Consequences

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

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

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

Higher conductivity reachable by the bndry cannot be balanced out.

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

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



## Diffuse optical tomography

Same strategy shows uniqueness for two coefficients (H. 2009+12):

$$-\nabla \cdot (a\nabla u) + cu = 0$$
 in  $\Omega$ .

▶ Let a, c pcw. constant, then

$$\Lambda(a_1,c_1)=\Lambda(a_2,c_2) \iff a_1=a_2 \text{ and } c_1=c_2.$$

▶ Let a, c pcw. anal., then  $\Lambda(a_1, c_1) = \Lambda(a_2, c_2)$ 

$$\iff \left\{\begin{array}{l} \frac{\Delta\sqrt{a_1}}{\sqrt{a_1}} + \frac{c_1}{a_1} = \frac{\Delta\sqrt{a_2}}{\sqrt{a_2}} + \frac{c_2}{a_2} \text{ on smooth parts} \\ \frac{a_1^+|_\Gamma}{a_1^-|_\Gamma} = \frac{a_2^+|_\Gamma}{a_2^-|_\Gamma}, \text{ and } \frac{[\partial_\nu a_2]_\Gamma}{a_2^-|_\Gamma} = \frac{[\partial_\nu a_1]_\Gamma}{a_1^-|_\Gamma} \text{ on discontinuity set} \end{array}\right.$$

Proof: Monotonicity + localized potentials





## Non-linearity



## Non-linearity

Back to the non-linear forward operator of EIT

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

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 / . . .

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

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

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

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





### Linearization

#### 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  $\kappa \approx \sigma - \sigma_0$ .

- ▶ 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

#### 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  $\kappa \approx \sigma - \sigma_0$ .

- ▶ 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

Theorem (H./Seo 2010)

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

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

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

- 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*<sub>0</sub>:

$$\begin{split} \int_{\Omega} (\sigma - \sigma_0) |\nabla u_0|^2 \, \mathrm{d}x \\ & \geq \underbrace{\int_{\Sigma} g \left( \Lambda(\sigma_0) - \Lambda(\sigma) \right) g}_{\sum} \geq \int_{\Omega} \frac{\sigma_0}{\sigma} (\sigma - \sigma_0) |\nabla u_0|^2 \, \mathrm{d}x. \\ & = \int_{\Sigma} g \left( \Lambda'(\sigma_0) \kappa \right) g = \int_{\Omega} \kappa |\nabla u_0|^2 \, \mathrm{d}x \end{split}$$

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

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

In shape reconstruction problems we can avoid non-linearity.





## Reconstruction from realistic data



## 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  $\Lambda(1 + \chi_B) \ge \Lambda(\sigma)$
- Result: upper bound of D.

Only an upper bound? Converse monotonicity relation?



## Converse monotonicity relation

Theorem (H./Ullrich, 2013)

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

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

→ Monotonicity method detects exact shape.

For faster implementation:

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

### Proof: Monotonicity + localized potentials



### General case

Theorem (H./*Ullrich*, 2013). 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  $\sigma - 1$ .

Result also holds with linearized condition

$$\exists \alpha > 1: \ \Lambda(1) + \alpha \Lambda'(1) \chi_{\mathcal{C}} \leq \Lambda(\sigma) \leq \Lambda(1) - \alpha \Lambda'(1) \chi_{\mathcal{C}}.$$

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

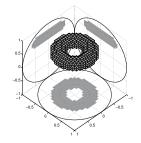




## Monotonicity based shape reconstruction

### 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



### Approach is closely related to (and heavily inspired by)

- ► Factorization Method of Kirsch and Hanke (in EIT: Brühl, Hakula, H., Hyvönen, Lechleiter, Nachman, Päivärinta, Pursiainen, Schappel, Schmitt, Seo, Teirilä, Woo, . . .)
- ► Ikehata's Enclosure Method and probing with Sylvester-Uhlmann-CGOs (*Ide, Isozaki, Nakata, Siltanen, Wang, ...*)
- ► Classic inclusion detection results (Friedmann, Isakov, ...)



### Realistic data & Uncertainties

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

$$\sigma(x) = \sigma_0(x) + \kappa(x)\chi_D, \quad \kappa(x) \ge 1$$

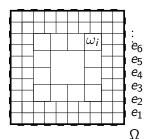
► Can we rigorously guarantee to find inclusion *D*?

## Monotonicity-based Rigorous Resolution Guarantee

(**H.**/Ullrich, to appear):

- ▶ If  $D = \emptyset$ , method returns  $\emptyset$ .
- ▶ If  $D \supset \omega_i$  then it is detected.

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







### Conclusions

Using monotonicity and localized potentials we can show

- uniqueness results for piecewiese anal. coefficients.
- ▶ invariance of shape information under linearization.
- resolution guarantees for locating anomalies in unknown backgrounds with realistic finite precision data.

Major limitations / open problems for our approach

- ▶ Piecewise analyticity required to prevent infinite oscillations.
- ► Approach requires operator/matrix-structure of measurements.

  ( Voltage has to be measured on current-driven electrodes.)