



Inverse coefficient problems and shape reconstruction

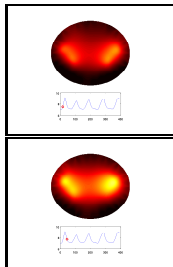
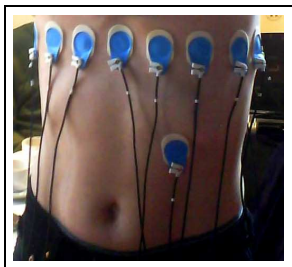
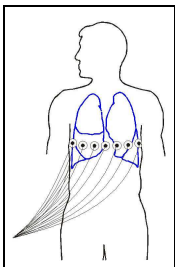
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Darmstadt, November 19, 2013.

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

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

from all possible Dirichlet and Neumann boundary values

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

Equivalent: Recover σ 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 $\sigma \partial_\nu u|_{\partial\Omega} = g$.

Partial/local data

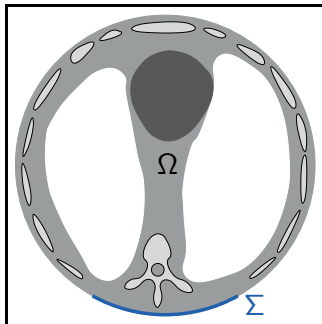
Measurements on open part of boundary $\Sigma \subset \partial\Omega$:
 ($\partial\Omega \setminus \Sigma$ is kept insulated.)

Recover σ 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 σ uniquely determined from the NtD $\Lambda(\sigma)$?
- ▶ Non-linearity and ill-posedness
 - ▶ Reconstruction algorithms to determine σ 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)
- ▶ L^∞ 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)

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

Localized potentials

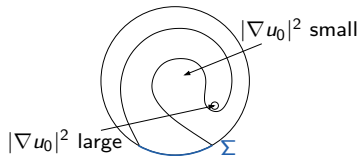
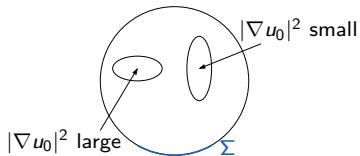
Theorem (H., 2008)

Let σ_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

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

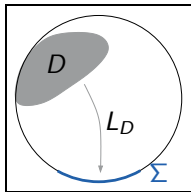


Proof 1/3

Virtual measurements:

$L_D : H_{\diamond}^1(D)' \rightarrow L_{\diamond}^2(\Sigma)$, $f \mapsto u|_{\Sigma}$, with

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



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

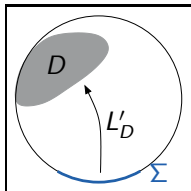
Sources on different domains yield different virtual measurements.

Proof 2/3

Dual operator:

$L'_D : L^2_\diamond(\Sigma) \rightarrow H^1_\diamond(D)$, $g \mapsto u|_D$, with

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



Evaluating solutions on D is dual operation to virtual measurements.

Proof 3/3

Functional analysis:

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

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

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

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

Note: $H_\diamond^1(D)'$ -source $\longleftrightarrow H_\diamond^1(D)$ -evaluation.

Consequences

- ▶ Back to Calderón: Let $\Lambda(\sigma_0) = \Lambda(\sigma_1)$, σ_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 Σ where $\sigma_1 \geq \sigma_0$ but $\sigma_1 \neq \sigma_0$
- ↪ 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.

Two coefficients

Can we recover two coefficients $a(x), c(x) \in L^{\infty}_+(\Omega)$ in

$$-\nabla \cdot (a\nabla u) + cu = 0 \quad \text{in } \Omega \quad (1)$$

from the NtD (with partial data)

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

where u solves (1) with

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

Application: Diffuse optical tomography (DOT).

Quasilinear case $a(u), c(x)$: *Egger, Pietschmann, Schlottbom 2013*

Monotonicity

$$\begin{aligned}
 & \int_{\Omega} ((a_2 - a_1)|\nabla u_1|^2 + (c_2 - c_1)|u_1|^2) \, dx \\
 & \geq \int_{\Sigma} g (\Lambda(a_1, c_1) - \Lambda(a_2, c_2)) g \, ds \\
 & \geq \int_{\Omega} ((a_2 - a_1)|\nabla u_2|^2 + (c_2 - c_1)|u_2|^2) \, dx,
 \end{aligned}$$

Method of localized potentials:

- ▶ Again, sources on different regions produce different data.
- ▶ $(H^1)'$ -sources produce different data than L^2 -sources

$$\implies \|u\|_{H^1(D)} \not\approx \|u\|_{L^2(D)}.$$

We can control $|\nabla u_1|^2$ and $|u_1|^2$ separately.

Uniqueness

Theorem (H., 2009)

Let

- ▶ $a_1, a_2 \in L_+^\infty(\Omega)$ piecewise constant,
- ▶ $c_1, c_2 \in L_+^\infty(\Omega)$ piecewise analytic.

Then

$$\Lambda(a_1, c_1) = \Lambda(a_2, c_2) \iff a_1 = a_2, \quad c_1 = c_2.$$

Note that $v := \sqrt{a}u$ transforms $-\nabla \cdot (a\nabla u) + cu = 0$ into

$$-\Delta v + \eta v = 0, \quad \eta := \frac{\Delta\sqrt{a}}{\sqrt{a}} + \frac{c}{a}$$

(when the coefficients are smooth).

Uniqueness

Theorem (H., 2012)

Let $a_1, a_2, c_1, c_2 \in L^{\infty}_+(\Omega)$ be piecew. analytic. Then $\Lambda(a_1, c_1) = \Lambda(a_2, c_2)$ if and only if

$$(a) \quad a_1|_{\Sigma} = a_2|_{\Sigma}, \quad \partial_{\nu} a_1|_{\Sigma} = \partial_{\nu} a_2|_{\Sigma} \quad \text{on } \Sigma,$$

$$(b) \quad \frac{\partial_{\nu} a_1}{a_1}|_{\partial B \setminus \bar{S}} = \frac{\partial_{\nu} a_2}{a_2}|_{\partial B \setminus \bar{S}} \quad \text{on } \partial\Omega \setminus \Sigma,$$

$$(c) \quad \eta_1 = \eta_2 \quad \text{in smooth regions,}$$

$$(d) \quad \frac{a_1^+|_{\Gamma}}{a_1^-|_{\Gamma}} = \frac{a_2^+|_{\Gamma}}{a_2^-|_{\Gamma}}, \quad \frac{[\partial_{\nu} a_2]_{\Gamma}}{a_2^-|_{\Gamma}} = \frac{[\partial_{\nu} a_1]_{\Gamma}}{a_1^-|_{\Gamma}} \quad \text{on inner boundaries } \Gamma.$$

NtD $\Lambda_{(a,c)}$ determines $\eta = \frac{\Delta\sqrt{a}}{\sqrt{a}} + \frac{c}{a}$ and the jumps of a and ∇a .



Non-linearity



Non-linearity

Back to the non-linear forward operator of EIT

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

Generic approach for inverting Λ : **Linearization**

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

σ_0 : known reference conductivity / initial guess / ...

$\Lambda'(\sigma_0)$: 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

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 κ , σ , σ_0 piecewise analytic and $\Lambda'(\sigma_0)\kappa = \Lambda(\sigma) - \Lambda(\sigma_0)$. Then

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

supp_{Σ} : 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_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 $|\nabla u_0|^2$

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

□

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. τ 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. $\sigma = 1 + \chi_D$.

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

\rightsquigarrow Monotonicity method detects exact shape.

For faster implementation:

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

\rightsquigarrow Linearized monotonicity method detects exact shape.

Proof: Monotonicity + localized potentials

General case

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

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

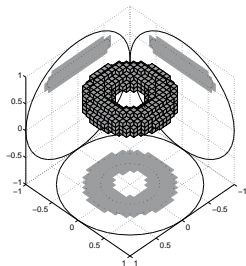
$$\exists \alpha > 1 : \Lambda(1) + \alpha\Lambda'(1)\chi_C \leq \Lambda(\sigma) \leq \Lambda(1) - \alpha\Lambda'(1)\chi_C.$$

- ▶ 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



Approach is closely related to (and heavily inspired by)

- ▶ Factorization Method of Kirsch and Hanke
(in EIT: Brühl, Hakula, H., Hyvönen, Lechleiter, Nachman, Päiväranta, 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

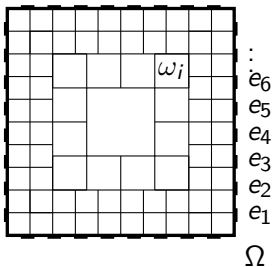
- ▶ Finite number of electrodes, CEM, noisy data $\Lambda^\delta(\sigma)$
- ▶ Unknown background, e.g., $1 - \epsilon \leq \sigma_0(x) \leq 1 + \epsilon$
- ▶ Anomaly with some minimal contrast to background, e.g.,

$$\sigma(x) = \sigma_0(x) + \kappa(x)\chi_D, \quad \kappa(x) \geq 1$$
- ▶ 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

- ▶ Uniqueness results for *piecewise smooth* parameters may significantly differ from that for *globally smooth* ones.
- ▶ 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.