



Inverse parameter identification problems

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Contents

- ▶ Parameter identification problems
in diffuse optical tomography (DOT)
- ▶ Linearized reconstruction algorithms
in electrical impedance tomography (EIT)



Parameter identification problems

in diffuse optical tomography (DOT)



Diffuse optical tomography

Diffuse optical tomography (DOT):

- ▶ Transilluminate biological tissue with visible/near-infrared light
- ▶ **Goal:** Reconstruct spatial image of interior physical properties.

Relevant quantities (in diffusive regime):

- ▶ Scattering
- ▶ Absorption
- ▶ **Applications:**
 - ▶ Breast cancer detection
 - ▶ Bedside-imaging of neonatal brain function

Diffuse optical tomography



Figure 1 from Gibson, Hebden and Arridge (Phys. Med. Biol. 50, R1–R43, 2005)
Imaging Diagnostic System Inc.'s computed tomography laser breast imaging system.
See www.imds.com for more details.

Mathematical Model

- ▶ General Forward Model:

Photon transport models (Boltzmann transport equation)

cf., e.g., Bal, Inverse Problems 25, 053001 (48pp), 2009.

- ▶ For highly scattering media:

- ▶ DC diffusion approximation for photon density u :

$$-\nabla \cdot (a \nabla u) + cu = 0 \quad \text{in } B \subset \mathbb{R}^n,$$

$u : B \rightarrow \mathbb{R}$: photon density

$a : B \rightarrow \mathbb{R}$: diffusion/scattering coefficient

$c : B \rightarrow \mathbb{R}$: absorption coefficient

- ▶ Boundary measurements (idealized):

Neumann and Dirichlet data $u|_S, a\partial_\nu u|_S$ on $S \subseteq \partial B$.

Remaining boundary assumed to be insulated, $a\partial_\nu u|_{\partial B \setminus S} = 0$.

Forward & inverse problem

DC diffuse optical tomography:

$$-\nabla \cdot (a \nabla u) + cu = 0 \quad \text{in } B \subset \mathbb{R}^n, \quad n \geq 2,$$

B bounded with smooth bndry, $S \subseteq \partial B$ open part, $a, c \in L_+^\infty(B)$.

- ▶ $\forall g \in L^2(S) \exists!$ sol. $u \in H^1(B) : a \partial_\nu u|_{\partial B} = \begin{cases} g & \text{on } S, \\ 0 & \text{on } B \setminus \overline{S}. \end{cases}$
- ▶ (Local) Neumann-to-Dirichlet map

$$\Lambda_{a,c} : g \mapsto u|_S, \quad L^2(S) \rightarrow L^2(S)$$

is linear, compact and self-adjoint.

Inverse Problem: Can we reconstruct a and c from $\Lambda_{a,c}$?

Non-uniqueness

DC diffuse optical tomography:

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

Arridge/Lionheart (1998 Opt. Lett. 23 882–4):

- ▶ $v := \sqrt{a}u$ solves

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

- ▶ $a = 1$ around $S \rightsquigarrow (u|_S, a \partial_\nu u|_S) = (v|_S, \partial_\nu v|_S).$

$\rightsquigarrow \Lambda_{a,c}$ only depends on **effective absorption** $\eta = \eta(a, c).$

Absorption and scattering effects cannot be distinguished.

(Note: Argument requires smooth scattering coefficient a).



Experimental results

Theory: Absorption and scattering effects cannot be distinguished.

Practice:

Successful separate reconstructions of absorption and scattering
(from phantom experiment using dc diffusion model!)

Pei et al. (2001), Jiang et al. (2002), Schmitz et al. (2002), Xu et al. (2002)

~> Practice contradicts theory!

Pei et al. (2001):

"As a matter of established methodological principle (...) empirical facts have the right-of-way; if a theoretical derivation yields a conclusion that is at odds with experimental results, the reconciliatory burden falls on the theorist, not on the experimentalist."

New uniqueness result

Theorem (H., Inverse Problems 2009)

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

If $\Lambda_{a_1, c_1} = \Lambda_{a_2, c_2}$ then $a_1 = a_2$ and $c_1 = c_2$.

- ▶ Piecewise constantness seems fulfilled for phantom experiments.
- ↪ Result reconciles theory with practice.
- ▶ Measurements contain more than just the effective absorption!

Next slides: Idea of the proof using monotony and loc. potentials.

Monotony result

Lemma

Let $a_1, a_2, c_1, c_2 \in L_+^\infty(B)$. Then for all $g \in L^2(S)$,

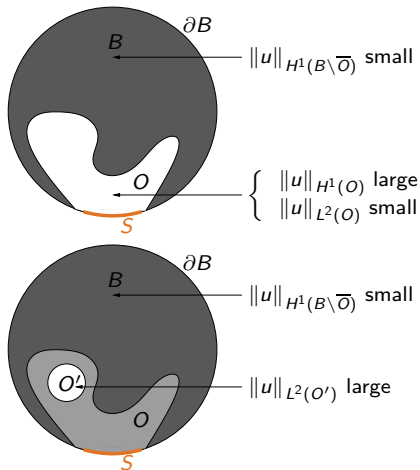
$$\begin{aligned} \int_B ((a_2 - a_1)|\nabla u_1|^2 + (c_2 - c_1)|u_1|^2) \, dx \\ \geq \langle (\Lambda_{a_1, c_1} - \Lambda_{a_2, c_2})g, g \rangle \\ \geq \int_B ((a_2 - a_1)|\nabla u_2|^2 + (c_2 - c_1)|u_2|^2) \, dx, \end{aligned}$$

$u_1, u_2 \in H^1(B)$: solutions for (a_1, c_1) , resp., (a_2, c_2) .

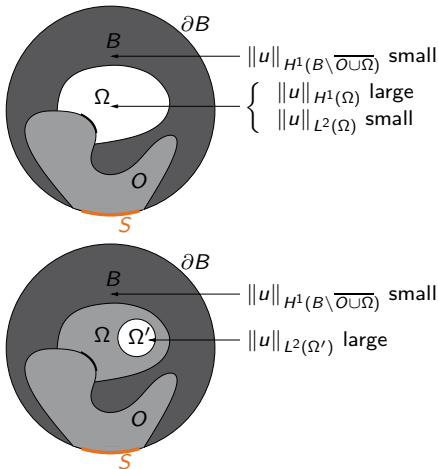
Can we control $|u_j|^2$ and $|\nabla u_j|^2$?

Localized potentials

Lemma There exist solutions u with



Lemma



Proof of uniqueness

Monotony

$$\begin{aligned} \int_B ((a_2 - a_1)|\nabla u_1|^2 + (c_2 - c_1)|u_1|^2) \, dx \\ \geq \langle (\Lambda_{a_1, c_1} - \Lambda_{a_2, c_2})g, g \rangle \\ \geq \int_B ((a_2 - a_1)|\nabla u_2|^2 + (c_2 - c_1)|u_2|^2) \, dx, \end{aligned}$$

Proof of the uniqueness result (*very sketchy ...*)

Start with region next to S

- ▶ Use loc. pot. with $|\nabla u|^2 \rightarrow \infty$ in that region $\rightsquigarrow a_1 = a_2$
- ▶ Then use loc. pot. with $|u|^2 \rightarrow \infty$ in that region $\rightsquigarrow c_1 = c_2$
- ▶ Repeat over all regions.

Uniqueness or not?

- ▶ Arridge/Lionheart (1998): Non-uniqueness for general smooth (a, c) .
- ▶ H. (2009): Uniqueness for piecew. constant a , piecew. analytic c .

What information about (a, c) does $\Lambda_{a,c}$ really contain?

Formally(!), $\Lambda_{a,c}$ can only determine $\eta = \frac{\Delta\sqrt{a}}{\sqrt{a}} + \frac{c}{a}$.

Jumps in a or $\nabla a \rightsquigarrow$ distributional singularities in $\Delta\sqrt{a}$.

Bold guess: Maybe $\Lambda_{a,c}$ determines

- ▶ η where a and c are smooth,
- ▶ jumps in a and ∇a .

(However, note that $\Delta\sqrt{a}/\sqrt{a}$ is not well-defined for non-smooth $a \dots$)

Exact characterization

Theorem (H., Inverse Probl. Imaging 2012)

Let $a_1, a_2, c_1, c_2 \in L_+^\infty(B)$ piecewise analytic on joint partition

$$B = O_1 \cup \dots \cup O_J \cup \Gamma, \quad \partial O_1 \cup \dots \cup \partial O_J = \partial B \cup \Gamma.$$

Then, $\Lambda_{a_1, c_1} = \Lambda_{a_2, c_2}$ if and only if

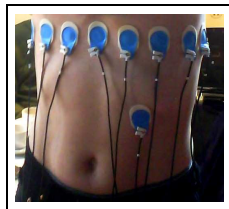
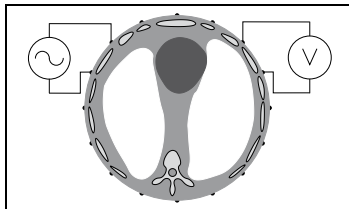
- (a) $a_1|_S = a_2|_S, \quad \text{and} \quad \partial_\nu a_1|_S = \partial_\nu a_2|_S \quad \text{on } S,$
- (b) $\frac{\partial_\nu a_1}{a_1}|_{\partial B \setminus \overline{S}} = \frac{\partial_\nu a_2}{a_2}|_{\partial B \setminus \overline{S}} \quad \text{on } \partial B \setminus \overline{S},$
- (c) $\eta_1 := \frac{\Delta \sqrt{a_1}}{\sqrt{a_1}} + \frac{c_1}{a_1} = \frac{\Delta \sqrt{a_2}}{\sqrt{a_2}} + \frac{c_2}{a_2} =: \eta_2 \quad \text{on } B \setminus \Gamma,$
- (d) $\frac{a_1^+|_\Gamma}{a_1^-|_\Gamma} = \frac{a_2^+|_\Gamma}{a_2^-|_\Gamma}, \quad \text{and} \quad \frac{[\partial_\nu a_2]_\Gamma}{a_2^-|_\Gamma} = \frac{[\partial_\nu a_1]_\Gamma}{a_1^-|_\Gamma} \quad \text{on } \Gamma.$



Linearized reconstruction algorithms

in electrical impedance tomography (EIT)

EIT



Electrical impedance tomography (EIT):

- ▶ Apply currents $\sigma \partial_\nu u|_{\partial B}$ (Neumann boundary data)
 $\rightsquigarrow$ Electric potential u solves

$$\nabla \cdot (\sigma \nabla u) = 0 \quad \text{in } B$$

- ▶ Measure voltages $u|_{\partial B}$ (Dirichlet boundary data)

Current-Voltage-Measurements $\rightsquigarrow$ Neumann-to-Dirichlet map $\Lambda(\sigma)$



Inverse problem

Non-linear forward operator of EIT

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

Inverse problem of EIT: $\Lambda(\sigma) \mapsto \sigma?$

Localized potentials $\rightsquigarrow$ Uniqueness for piecew. analytic conductivities
already known: Druskin (1982+85), Kohn/Vogelius (1984+85)

Linearization

Generic approach: **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.

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

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

Often: $\text{supp}(\sigma - \sigma_0) \subset\subset B$ compact. ("*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!



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.



Exact Linearization

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

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

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

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

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

Proof: Combination of monotony and localized potentials.

Proof

- ▶ Exact linearization: $\Lambda'(\sigma_0)\kappa = \Lambda(\sigma) - \Lambda(\sigma_0)$
- ▶ Monotony: For all "reference solutions" u_0 :

$$\begin{aligned} \int_B (\sigma - \sigma_0) |\nabla u_0|^2 \, dx \\ \geq \underbrace{\langle g, (\Lambda(\sigma) - \Lambda(\sigma_0)) g \rangle}_{= \int_B \kappa |\nabla u_0|^2 \, dx} \geq \int_B \frac{\sigma_0}{\sigma} (\sigma - \sigma_0) |\nabla u_0|^2 \, dx. \end{aligned}$$

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

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



Interpretation

σ_0 : reference conductivity

σ : true conductivity

- ▶ Current paths depend on unknown conductivity σ
(Non-linearity of EIT)
- ▶ Linearizing EIT around ref. conductivity σ_0 corresponds to assuming that currents take the same paths as in a ref. body.
- ▶ Paths may differ considerably when $\sigma - \sigma_0$ is large.
- ▶ However, taking the (wrong) reference paths for reconstruction still yields the correct shape information!

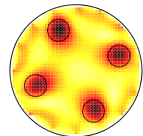
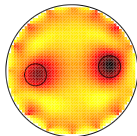
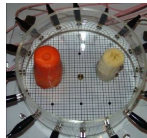
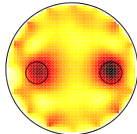
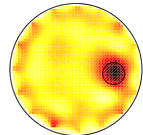
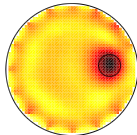
Practical issues: Theorem requires exact sol. of linearized equation

- ▶ Does an approximate solution of the linearized equation show approximately correct shapes?

Experimental result (frequency difference data)

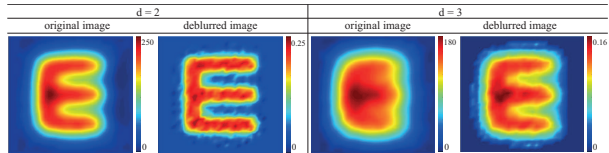
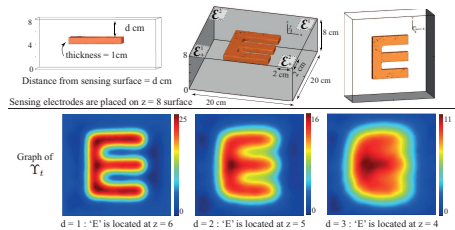
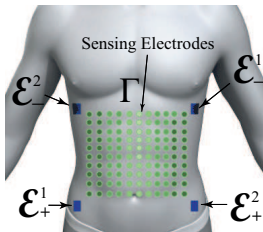
H./Seo/Woo (*IEEE Trans. Med. Imaging* 2010):

Heuristic combination of linearization and localized potentials for fdEIT.



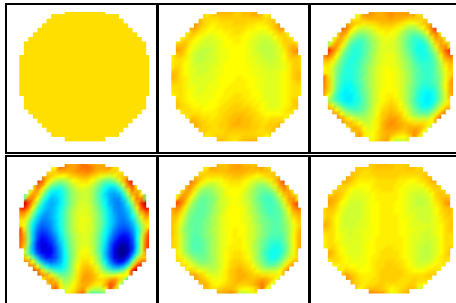
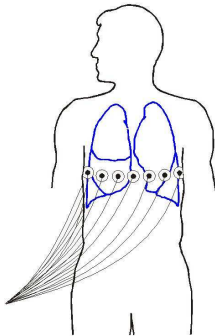
Planar EIT

Ts/Lee/Seo/H./Kim (submitted): Partial inversion plus linearization



Lung monitoring - real human data

BMBF-project on regularizing EIT in the medical and geophysical sciences.





Conclusion

- ▶ Novel tomography techniques lead to mathematical parameter identification problems.
- ▶ Uniqueness questions may have non-trivial answers.
In diffuse optical tomography:
 - ▶ Effective absorption plus jumps in diffusion coefficient (and its derivative) can be reconstructed simultaneously.
 - ↪ Uniqueness for piecewise constant coefficients, but not for general smooth coefficients
- ▶ Non-linearity is main challenge for stable/convergent algos.
 - ▶ Shape information is not affected by linearization errors.
 - ▶ Stable shape reconstruction is possible.
- ▶ Close interplay between uniqueness arguments and convergent reconstruction algorithms.