

Uniqueness and global convergence for inverse coefficient problems with finitely many measurements

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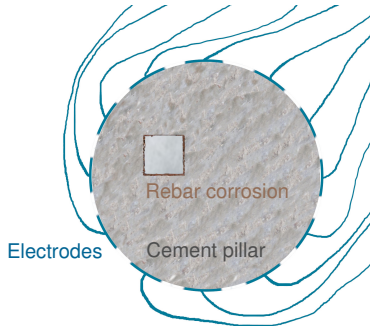
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Electrical impedance tomography for corrosion detection



Non-destructive EIT-based corrosion detection:

- ▶ Apply electric currents on outer boundary $\partial\Omega$
 - ▶ Measure necessary voltages
 - ≈ Detect corrosion on inner boundary $\Gamma = \partial D$
-

Idealized mathematical model: Robin PDE

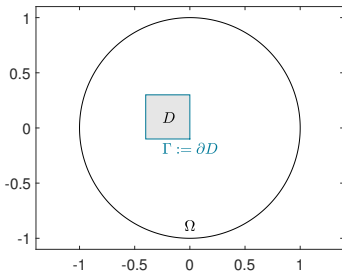
Electric potential $u: \Omega \rightarrow \mathbb{R}$ solves

$$(1) \quad \Delta u = 0 \quad \text{in } \Omega \setminus \Gamma,$$

$$(2) \quad \partial_\nu u|_{\partial\Omega} = g \quad \text{on } \partial\Omega,$$

$$(3) \quad [[u]]_\Gamma = 0 \quad \text{on } \Gamma,$$

$$(4) \quad [[\partial_\nu u]]_\Gamma = \sigma u \quad \text{on } \Gamma$$



Inverse Problem: Recover σ from Neumann-to-Dirichlet-Operator

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

where u solves Robin PDE (1)–(4).

(Similar to but simpler than famous Calderón Problem)

Finitely many measurements and unknowns

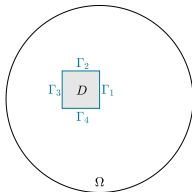
- ▶ Finitely many measurements:

$$\int_{\partial\Omega} g_j \Lambda(\sigma) g_k \, ds \quad \text{for finitely many } g_1, \dots, g_m$$

- ▶ Finite desired resolution:

$$\sigma = \sum_{j=1}^n \sigma_j \chi_{\Gamma_j} \quad \text{with } \sigma_j \in \mathbb{R}, j = 1, \dots, n$$

with partition $\Gamma = \bigcup_{j=1}^n \Gamma_j$



- ▶ A-priori bounds: $\sigma := (\sigma_1, \dots, \sigma_n)^T \in [a, b]^n$, $b > a > 0$ known

Finite-dimensional non-linear inverse problem (IP): Determine

$$\sigma = (\sigma_j)_{j=1}^n \in [a, b]^n \quad \text{from} \quad F(\sigma) := \left(\int_{\partial\Omega} g_j \Lambda(\sigma) g_k \, ds \right)_{j,k=1}^m \in \mathbb{R}^{m \times m}$$

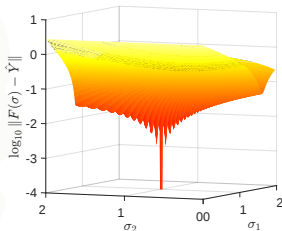
Data fitting approaches

(IP) Determine $\hat{\sigma} \in [a, b]^n \subset \mathbb{R}^n$ from $\hat{Y} := F(\hat{\sigma}) \in \mathbb{R}^{m \times m}$

- Natural approach: Least-squares data fitting

$$\text{minimize } \|F(\sigma) - \hat{Y}\|_F^2 \quad (+ \text{Regularization})$$

↪ High-dim., non-convex minimization. Problem of local minima.



(Example from Calderón problem)

Next slides: Equivalent convex reformulation of (IP)

Main result 1/3

Theorem. (H., Optim. Lett. 2021)

If sufficiently many measurements are taken, then

- ▶ $\hat{Y} := F(\hat{\sigma}) \in \mathbb{R}^{m \times m}$ uniquely determines $\hat{\sigma} \in [a, b]^n$.
- ▶ $\hat{\sigma}$ is the unique solution of

$$\text{minimize } \|\sigma\|_1 = \sum_{j=1}^n \sigma_j \quad \text{s.t.} \quad \sigma \in [a, b]^n, F(\sigma) \leq \hat{Y}.$$

- ▶ The constraint set $\sigma \in [a, b]^n, F(\sigma) \leq \hat{Y}$ is convex.

- $\hat{\sigma}$ is the lower left corner of the convex constraint set
- Problem can be solved by convex semidefinite programming

Global convergence is feasible.

(H., Numer. Math. 2020: Global Newton convergence for this Robin problem)

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Main result 2/3

Theorem. (H., Optim. Lett. 2021)

- ▶ Suff. many measurements are taken if $\lambda_{\max}(F'(z_{j,k})d_j) > 0$ for $z_{j,k} := \frac{a}{2}e'_j + \left(a + k\frac{a}{4(n-1)}\right)e_j \in \mathbb{R}^n$, $d_j := \frac{2b-a}{a}(n-1)e'_j - \frac{1}{2}e_j \in \mathbb{R}^n$, with $j = 1, \dots, n$, $k = 1, \dots, \lceil \frac{4(n-1)b}{a} \rceil - 4n + 5$.
- ▶ This **criterion** is fulfilled if $(g_j)_{j=1}^{\infty}$ has dense span in $L^2(\partial\Omega)$, and sufficiently many g_j are used.

($e_j \in \mathbb{R}^n$: j -th unit vector, $e'_j := 1 - e_j \in \mathbb{R}^n$: negated j -th unit vector)

- ↪ Explicit, easy-to-check criterion whether a desired resolution can be achieved with a certain number of measurements

Achievable resolution can be characterized.

Main result 3/3

Theorem. (H., Optim. Lett. 2021)

- ▶ Let the **criterion** hold with lower bound $\lambda > 0$.
- ▶ Let $\delta > 0$, and $Y^\delta \in \mathbb{R}^{m \times m}$ be symmetric with $\|\hat{Y} - Y^\delta\|_2 \leq \delta$.

Then there exist solutions of

$$\text{minimize } \|\sigma\|_1 = \sum_{j=1}^n \sigma_j \quad \text{s.t.} \quad \sigma \in [a, b]^n, F(\sigma) \leq Y^\delta + \delta I.$$

and every such minimum σ^δ fulfills

$$\|\hat{\sigma} - \sigma^\delta\|_\infty \leq \frac{2\delta(n-1)}{\lambda}$$

Explicit error estimates, convergence for $\delta \rightarrow 0$.

Proof ingredients & possible generalizations

- **Monotonicity & Convexity:** $F : \mathbb{R}_+^n \rightarrow \mathbb{S}_m \subset \mathbb{R}^{m \times m}$ fulfills

$$\begin{aligned} F'(\sigma)d &\leq 0 && \text{for all } \sigma \in \mathbb{R}_+^n, 0 \leq d \in \mathbb{R}^n \\ F(\tau) - F(\sigma) &\geq F'(\sigma)(\tau - \sigma) && \text{for all } \sigma, \tau \in \mathbb{R}_+^n \end{aligned}$$

↪ holds for general elliptic PDEs (H., Jahresber. DMV, 2021)

- **Localized potentials:** For any $C > 0$, there exist currents g s.t.

$$g^T (F'(\sigma)(e_j - Ce'_j))g = \int_{\Gamma_j} |\nabla u|^2 \, dx - C \int_{\Gamma \setminus \Gamma_j} |\nabla u|^2 > 0$$

$$\implies \lambda_{\max}(F'(z)(e_j - Ce'_j)) > 0 \text{ for suff. many measurement.}$$

↪ holds for many elliptic problems, but in more complicated form

Conclusions

For elliptic coefficient inverse problems

- ▶ least-squares residuum functionals may be highly non-convex
- ▶ local minima are usually useless

Possible remedy

- ▶ utilize monotonicity & convexity with respect to Loewner order
- ▶ utilize localized potentials to control directional derivatives

For an inverse Robin coefficient problem we can obtain

- ▶ equivalent reformulation as convex semidefinite program
- ▶ globally convergent solution algorithms
- ▶ explicit characterizations of achievable resolution
- ▶ explicit error estimates for noisy data