

# Recovering Coefficients in a System of Semilinear Helmholtz Equations from Internal Data

Nathan Soedjak



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Joint with Kui Ren (Columbia)

# The inverse problem

Let  $\Omega \subset \mathbb{R}^d$  ( $d \geq 2$ ) be a bounded domain with smooth boundary  $\partial\Omega$ .

- Model parameters:  $\sigma(\mathbf{x})$  (conductivity),  $\eta(\mathbf{x})$  (first-order susceptibility),  $\gamma(\mathbf{x})$  (second-order nonlinear susceptibility),  $\Gamma(\mathbf{x})$  (Grüneisen coefficient)
- Forward model (wave propagation with **second harmonic generation**):

$$(1) \quad \begin{aligned} \Delta u + k^2(1 + \eta)u + ik\sigma u &= -k^2\gamma u^*v, & \text{in } \Omega \\ \Delta v + (2k)^2(1 + \eta)v + i2k\sigma v &= -(2k)^2\gamma u^2, & \text{in } \Omega \\ u = g, \quad v = h, & & \text{on } \partial\Omega \end{aligned}$$

- Data (from **thermo-acoustic tomography**):

$$H(\mathbf{x}) = \Gamma(\mathbf{x})\sigma(\mathbf{x})(|u|^2 + |v|^2), \quad \mathbf{x} \in \Omega$$

## Inverse Problem

Determine the model parameters  $(\Gamma, \sigma, \eta, \gamma)$  from the map

$$\Lambda_{\Gamma, \sigma, \eta, \gamma} : (g, h) \mapsto H.$$

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# Well-posedness of the forward model

## Theorem

Let  $\alpha \in (0, 1)$ . There exist  $\varepsilon > 0$  and  $\delta > 0$  such that for all  $g, h \in \mathcal{C}^{2,\alpha}(\partial\Omega; \mathbb{C})$  with  $\|g\|_{\mathcal{C}^{2,\alpha}(\partial\Omega)} < \varepsilon$  and  $\|h\|_{\mathcal{C}^{2,\alpha}(\partial\Omega)} < \varepsilon$ , the forward model (1) has a unique solution

$$(u, v) \in \{f \in \mathcal{C}^{2,\alpha}(\overline{\Omega}; \mathbb{C}) : \|f\|_{\mathcal{C}^{2,\alpha}(\overline{\Omega})} \leq \delta\}^2.$$

Proof is a more-or-less straightforward application of the **Banach fixed point theorem**.

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- **Observation:** when  $g = h = 0$ , we know that the solution to the forward model is  $u = v = 0$ , *regardless of the unknown coefficients*  $(\eta, \sigma, \gamma)$ . This motivates **linearizing** around the boundary conditions  $g = h = 0$ .
- For a given small number  $\varepsilon > 0$ , let  $(u_\varepsilon, v_\varepsilon)$  be the solution to the system

$$\begin{aligned}\Delta u_\varepsilon + k^2(1 + \eta)u_\varepsilon + ik\sigma u_\varepsilon &= -k^2\gamma u_\varepsilon^* v_\varepsilon, & \text{in } \Omega \\ \Delta v_\varepsilon + (2k)^2(1 + \eta)v_\varepsilon + i2k\sigma v_\varepsilon &= -(2k)^2\gamma u_\varepsilon^2, & \text{in } \Omega\end{aligned}$$

with boundary conditions

$$(2) \quad u_\varepsilon = \varepsilon g, \quad v_\varepsilon = \varepsilon h, \quad \text{on } \partial\Omega.$$

Denote the corresponding data by  $H_\varepsilon := \Gamma\sigma(|u_\varepsilon|^2 + |v_\varepsilon|^2)$ .

- When  $\varepsilon = 0$ , clearly  $u_0 = v_0 = H_0 = 0$ .
- Consider the derivatives  $u^{(n)} := \partial_\varepsilon^n u_\varepsilon|_{\varepsilon=0}$ ,  $v^{(n)} := \partial_\varepsilon^n v_\varepsilon|_{\varepsilon=0}$  and  $H^{(n)} := \partial_\varepsilon^n H_\varepsilon|_{\varepsilon=0}$ .

# First-order linearization

- Taking the derivative of the forward model with respect to  $\varepsilon$  at  $\varepsilon = 0$  yields

$$\begin{aligned}\Delta u^{(1)} + k^2(1 + \eta)u^{(1)} + ik\sigma u^{(1)} &= 0, & \text{in } \Omega \\ \Delta v^{(1)} + (2k)^2(1 + \eta)v^{(1)} + i2k\sigma v^{(1)} &= 0, & \text{in } \Omega \\ u^{(1)} = g, \quad v^{(1)} = h, & & \text{on } \partial\Omega.\end{aligned}$$

- Meanwhile,  $H^{(1)} = 0$  and

$$H^{(2)} = 2\Gamma\sigma(u^{(1)*}u^{(1)} + v^{(1)*}v^{(1)}).$$

## Theorem

*All of the above derivatives exist.*

# Reconstruction of $(\Gamma, \sigma, \eta)$

- By taking  $h = 0$ , the problem reduces to reconstructing  $(\Gamma, \sigma, \eta)$  from the data

$$H = \Gamma\sigma|u|^2,$$

with the model for  $u$  given by

$$\begin{aligned}\Delta u + k^2(1 + \eta)u + ik\sigma u &= 0, & \text{in } \Omega \\ u &= g, & \text{on } \partial\Omega.\end{aligned}$$

- When  $\Gamma$  and  $\eta$  are *known*, this problem was analyzed in Ammari et al. [1] (2012) and Bal et al. [4] (2011).

# Reconstruction of $(\Gamma, \sigma, \eta)$ (cont.)

- When  $\Gamma$  and  $\eta$  are *unknown*, use **polarization** identity. Let  $g_1$  and  $g_2$  be two incident sources. Measuring data corresponding to the illuminations  $g_1$ ,  $g_2$ ,  $g_1 + g_2$  and  $g_1 + ig_2$  gives us that the quantity

$$\Gamma \sigma u_1 u_2^* = \frac{1}{2} \Gamma \sigma (|u_1 + u_2|^2 + i|u_1 + iu_2|^2 - (1+i)|u_1|^2 - (1+i)|u_2|^2)$$

is known.

- So given illuminations  $\{g_j\}_{j=1}^2$ , we can reconstruct from the measured internal data the new data:

$$E_j = \Gamma \sigma u_j u_1^*,$$

for  $j = 1, 2$ . In particular, we know

$$\frac{u_2}{u_1} = \frac{E_2}{E_1}.$$

# Reconstruction of $(\Gamma, \sigma, \eta)$ (cont.)

## Theorem

Let  $\{g_j\}_{j=1}^2$  be a set of incident source functions, and suppose that the measured data  $\{E_j\}_{j=1}^2$  satisfy the following two conditions:

1  $E_1(\mathbf{x}) \geq \alpha_0 > 0$  for some  $\alpha_0$ , a.e.  $\mathbf{x} \in \Omega$ .

2 The vector field

$$\beta(\mathbf{x}) := \nabla \frac{E_2}{E_1}$$

is  $W^{1,\infty}$  field such that  $|\beta| \geq \beta_0 > 0$  for some  $\beta_0$ , a.e.  $\mathbf{x} \in \Omega$ .

Then  $\Gamma$ ,  $\eta$ , and  $\sigma$  are uniquely determined from the data  $\{E_j\}_{j=1}^2$ .

# Reconstruction of $(\Gamma, \sigma, \eta)$ (cont.)

## Proof.

Following Bal and Ren [3] (2011), we multiply the equation for  $u_1$  by  $u_2$  and multiply the equation for  $u_2$  by  $u_1$ , and subtract to get

$$u_1 \Delta u_2 - u_2 \Delta u_1 = 0.$$

We can then rewrite this into

$$\nabla \cdot u_1^2 \nabla \frac{u_2}{u_1} = \nabla \cdot u_1^2 \nabla \frac{E_2}{E_1} = \nabla \cdot (u_1^2 \beta) = 0.$$

This is a **transport equation** for  $u_1$ . Together with the boundary condition  $u_1 = g_1$  on  $\partial\Omega$  and assumption 2, classical results show that there exists a unique weak solution  $u_1$ . Hence we can reconstruct

$$k^2(1 + \eta) + ik\sigma = -\frac{\Delta u_1}{u_1}.$$

This gives us  $\eta$  and  $\sigma$ . The last step is to reconstruct  $\Gamma$  as  $\Gamma = \frac{H_1}{\sigma|u_1|^2}$ . □

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# Higher-order linearization

- Recall the original system:

$$\begin{aligned}\Delta u_\varepsilon + k^2(1 + \eta)u_\varepsilon + ik\sigma u_\varepsilon &= -k^2\gamma u_\varepsilon^* v_\varepsilon, & \text{in } \Omega \\ \Delta v_\varepsilon + (2k)^2(1 + \eta)v_\varepsilon + i2k\sigma v_\varepsilon &= -(2k)^2\gamma u_\varepsilon^2, & \text{in } \Omega \\ u_\varepsilon = \varepsilon g, & v_\varepsilon = \varepsilon h, & \text{on } \partial\Omega.\end{aligned}$$

- Taking the 2nd derivative of the forward model with respect to  $\varepsilon$  at  $\varepsilon = 0$  yields

$$\begin{aligned}\Delta u^{(2)} + k^2(1 + \eta)u^{(2)} + ik\sigma u^{(2)} &= -2k^2\gamma u^{(1)*}v^{(1)}, & \text{in } \Omega \\ \Delta v^{(2)} + (2k)^2(1 + \eta)v^{(2)} + i2k\sigma v^{(2)} &= -2(2k)^2\gamma(u^{(1)})^2, & \text{in } \Omega \\ u^{(2)} = 0, & v^{(2)} = 0, & \text{on } \partial\Omega.\end{aligned}$$

Note that  $u^{(1)}$  and  $v^{(1)}$  are known.

- Meanwhile, the 3rd-order derivative of the data is

$$H^{(3)} = 3\Gamma\sigma \left( u^{(1)*}u^{(2)} + u^{(1)}u^{(2)*} + v^{(1)*}v^{(2)} + v^{(1)}v^{(2)*} \right).$$

- If not for the conjugated terms  $u^{(2)*}$  and  $v^{(2)*}$  in  $H^{(3)}$ , these 3 equations would constitute a **system of linear PDEs** for  $u^{(2)}$ ,  $v^{(2)}$ ,  $\gamma$ .

# Reconstruction of $\gamma$

- **Trick:** Take the complex conjugate of the equations for  $u^{(2)}$  and  $v^{(2)}$  to get the following **system of linear PDEs** for  $(u^{(2)}, v^{(2)})$ ,  $(u^{(2)*}, v^{(2)*})$ , and  $\gamma$ :

$$\begin{aligned}(\Delta + q_1)u^{(2)} + 2k^2 u^{(1)*} v^{(1)} \gamma &= 0, & \text{in } \Omega \\(\Delta + q_1^*)u^{(2)*} + 2k^2 u^{(1)} v^{(1)*} \gamma &= 0, & \text{in } \Omega \\(\Delta + q_2)v^{(2)} + 2(2k)^2 (u^{(1)})^2 \gamma &= 0, & \text{in } \Omega \\(\Delta + q_2^*)v^{(2)*} + 2(2k)^2 (u^{(1)*})^2 \gamma &= 0, & \text{in } \Omega \\u^{(1)*} u^{(2)} + u^{(1)} u^{(2)*} + v^{(1)*} v^{(2)} + v^{(1)} v^{(2)*} &= \frac{H^{(3)}}{3\Gamma\sigma}, & \text{in } \Omega \\(u^{(2)}, u^{(2)*}, v^{(2)}, v^{(2)*}) &= (0, 0, 0, 0), & \text{on } \partial\Omega,\end{aligned}$$

where we have used the notation

$$q_1 := k^2(1 + \eta) + ik\sigma, \quad q_2 := (2k)^2(1 + \eta) + i2k\sigma.$$

- The rest of the proof consists of analyzing the **uniqueness** of the solution to this linear system. The analysis is based on the uniqueness theory for redundant elliptic systems reviewed in Bal [2] (2013).

# Reconstruction of $\gamma$ (cont.)

To guarantee uniqueness, we found it convenient to make the following assumptions:

- 1 In addition to the internal data, we also have access to the **Dirichlet-to-Neumann map**

$$\Pi_\gamma : (g, h) \mapsto \left( \frac{\partial u}{\partial \nu} \Big|_{\partial\Omega}, \frac{\partial v}{\partial \nu} \Big|_{\partial\Omega} \right).$$

- 2 The domain  $\Omega$  is **sufficiently small**.

We believe that these assumptions can be removed without breaking the uniqueness and stability results.

# Reconstruction of $\gamma$ (cont.)

## Theorem

When the domain  $\Omega$  is sufficiently small, there exist illumination directions  $g$  and  $h$  in (2) such that

$$(H^{(3)}, J_u^{(2)}, J_v^{(2)}) = (\tilde{H}^{(3)}, \tilde{J}_u^{(2)}, \tilde{J}_v^{(2)}) \implies \gamma = \tilde{\gamma},$$

and  $\forall p > 1$ ,

$$\|\gamma - \tilde{\gamma}\|_{L^p(\Omega)} \leq C \left( \left\| H^{(3)} - \tilde{H}^{(3)} \right\|_{W^{2,p}(\Omega)} + \left\| J_u^{(2)} - \tilde{J}_u^{(2)} \right\|_{W^{1-1/p,p}(\partial\Omega)} + \left\| J_v^{(2)} - \tilde{J}_v^{(2)} \right\|_{W^{1-1/p,p}(\partial\Omega)} \right),$$

for some constant  $C = C(\Omega, \Gamma, \eta, \sigma) > 0$ .

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# Numerical experiments

- We perform numerical reconstructions with a standard **least-squares** optimization approach in which we find the coefficients  $\eta, \sigma, \gamma, \Gamma$  by minimizing the functional

$$\Psi(\eta, \sigma, \gamma) := \frac{1}{2} \sum_{j=2}^N \left\| \frac{|u_j|^2 + |v_j|^2}{|u_1|^2 + |v_1|^2} - \frac{H_j}{H_1} \right\|_{L^2(\Omega)}^2 \\ + \beta_1 \|\nabla \eta\|_{L^1(\Omega)} + \beta_2 \|\nabla \sigma\|_{L^1(\Omega)} + \beta_3 \|\nabla \gamma\|_{L^2(\Omega)},$$

where we assumed that we collected data from  $N$  different boundary conditions. The regularization parameters  $\beta_1, \beta_2, \beta_3$  are selected by a trial and error approach.

- The gradient of the objective function is computed using the standard **adjoint state method**.

# Numerical experiments (cont.)

## ■ Reconstruction of $\gamma$

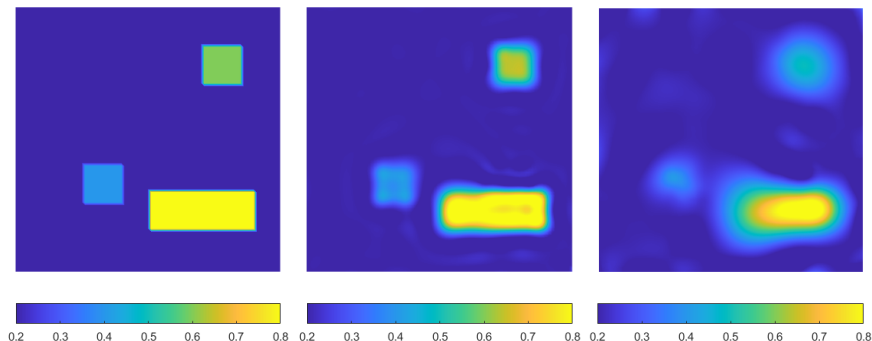


Figure 1: True  $\gamma$  (left),  $\gamma$  reconstructed from noise-free data (middle), and  $\gamma$  reconstructed from data containing 1% random noise (right).

# Numerical experiments (cont.)

## ■ Simultaneous reconstruction of $\eta, \sigma, \gamma, \Gamma$

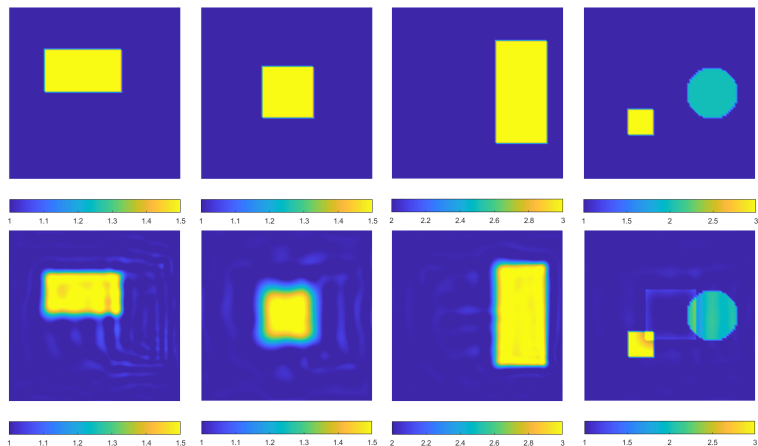


Figure 2: True (top) and reconstructed (bottom) ( $\eta, \sigma, \gamma, \Gamma$ ). From left to right are  $\eta, \sigma, \gamma$ , and  $\Gamma$ .

# Future directions

- Well-posedness of the forward model for **large** boundary data?
- We have assumed Dirichlet boundary condition for the generated second harmonic wave  $v$  in the forward model. This should certainly be replaced with **homogeneous Robin-type** boundary conditions that are more physical.
- Try to remove the two assumptions in the reconstruction theorem for the second-order susceptibility parameter  $\gamma$ .
- The linearization method requires access to a one-parameter family of datasets generated from  $\varepsilon$ -dependent boundary source. This is a large amount of data. It would be of great interest to see if our uniqueness and stability results can be reproduced for a **finite** number of measurements.

- [1] H. AMMARI, J. GARNIER, W. JING, AND L. NGUYEN, *Quantitative thermo-acoustic imaging: An exact reconstruction formula*, Submitted to Archive for Rational Mechanics and Analysis, (2011).
- [2] G. BAL, *Hybrid inverse problems and redundant systems of partial differential equations*, in Inverse Problems and Applications, P. Stefanov, A. Vasy, and M. Zworski, eds., vol. 615 of Contemporary Mathematics, American Mathematical Society, 2013, pp. 15–48.
- [3] G. BAL AND K. REN, *Multi-source quantitative PAT in diffusive regime*, Inverse Problems, 27 (2011). 075003.
- [4] G. BAL, K. REN, G. UHLMANN, AND T. ZHOU, *Quantitative thermo-acoustics and related problems*, Inverse Problems, 27 (2011). 055007.
- [5] K. REN AND N. SOEDJAK, *Recovering coefficients in a system of semilinear Helmholtz equations from internal data*, arXiv:2307.01385, (2023).