

# Integral equation methods for Electrical Impedance Tomography

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# OUTLINE

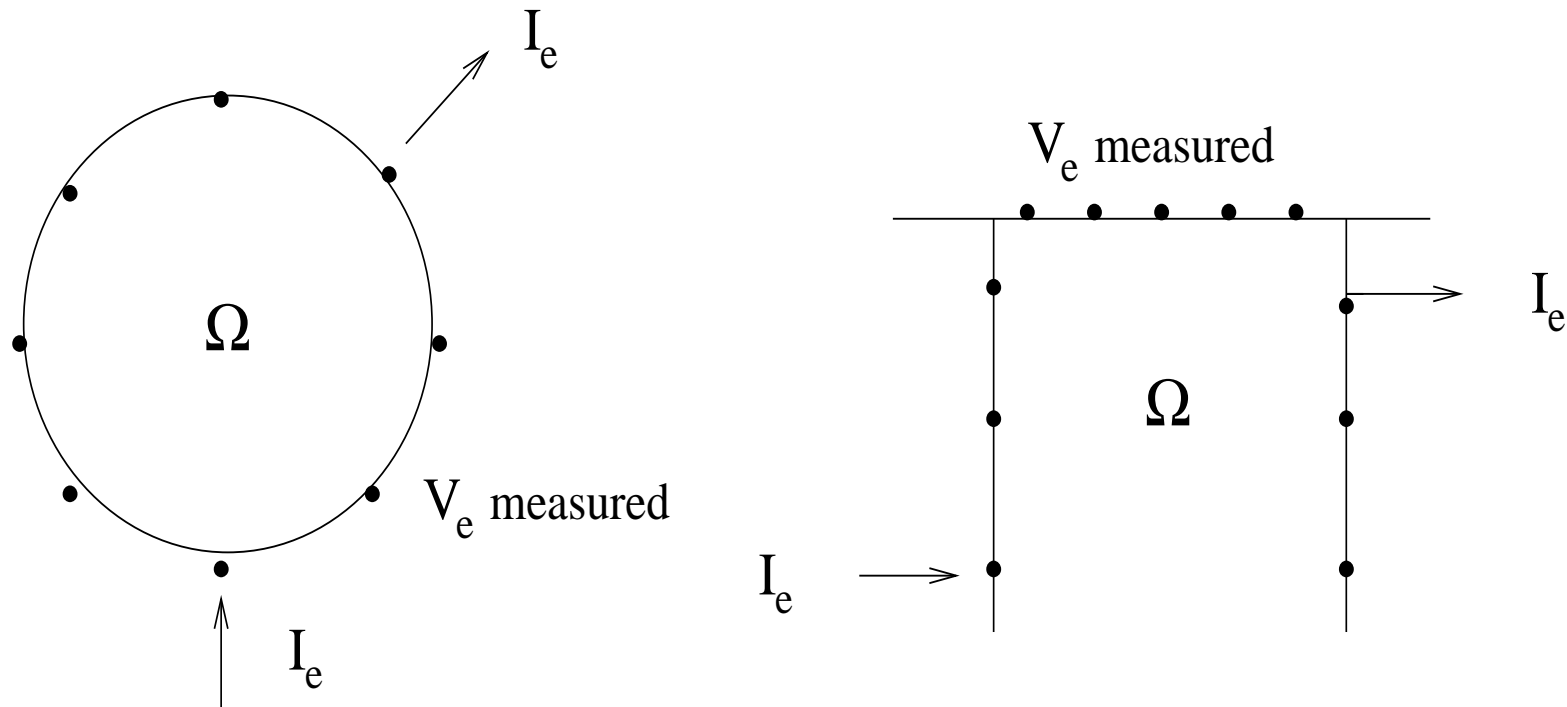
- **EIT and the inverse conductivity problem**
- **Integral equation formulation**
- **A Tikhonov regularization method**
- **A mollifier method**
- **Numerical examples**
- **Conclusions**

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- **EIT and the inverse conductivity problem**
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## EIT AND THE INVERSE CONDUCTIVITY PROBLEM

Electrical Impedance Tomography (EIT) is the inverse problem of determining the electrical conductivity in the interior of an object,  $\Omega$ , given simultaneous measurements of direct electric currents and voltages on the boundary of the object.



# EIT and the inverse conductivity problem

EIT applications:

- **medicine**

- detection of pulmonary emboli
- monitoring of heart functional blood flow
- breast cancer detection

- **geophysics**

- locating underground mineral deposits
- detection of leaks in underground storage tanks
- monitoring flows of injected fluids into the earth

- **nondestructive testing**

- detection of corrosion and of small defects in metals:  
cracks or voids

## EIT AND THE INVERSE CONDUCTIVITY PROBLEM

Let  $\Omega \subset \mathbb{R}^n$ ,  $n = 2, 3$  be a bounded domain and  $\sigma$  be an isotropic conductivity distribution ( $0 < c < \sigma < \infty$ ).

If an electric current  $j = \sigma \frac{\partial \Phi}{\partial n} \in H^{-\frac{1}{2}}(\partial\Omega)$  is applied then the induced electric potential  $\Phi \in H^1(\Omega)$  satisfies:

$$\nabla \cdot [\sigma(\mathbf{x}) \nabla \Phi(\mathbf{x})] = 0, \quad \mathbf{x} \in \Omega$$

**Definition.** *The ICP is to find  $\sigma(\mathbf{x})$ ,  $\mathbf{x} \in \Omega$ , given the NtD map*

$$\begin{aligned} (\Lambda_\sigma)^{-1} : \mathcal{I} &\rightarrow H^{1/2}(\partial\Omega) \\ j &\mapsto \Phi|_{\partial\Omega}, \end{aligned}$$

where  $\mathcal{I} = \{j(\mathbf{x}) \in H^{-\frac{1}{2}}(\partial\Omega) : \int_{\partial\Omega} j(\mathbf{x}) d\mathbf{x} = 0\}$ .

# EIT AND THE INVERSE CONDUCTIVITY PROBLEM

## Uniqueness of solutions

- the linearized problem

A.P. Calderón. *On an inverse boundary value problem*. Seminar on Numerical Analysis and its Applications to Continuum Physics. Soc. Brasileira de Mathématique, Rio de Janeiro, (1980).

- in  $n \geq 3$  dimensions for  $\gamma \in W^{3/2,\infty}(\Omega)$

L. Päivärinta, A. Panchenko and G. Uhlmann. *Complex geometric optics solutions for Lipschitz conductivities*. Rev. Mat. Iberoamericana **19**, (2003).

- in two dimensions  $\sigma \in L^\infty(\Omega)$

K. Astala and L. Päivärinta. *Calderón's inverse conductivity problem in the plane*. Ann. Math. **163**, (2006).

# EIT AND THE INVERSE CONDUCTIVITY PROBLEM

Reconstruction methods:

- back-projection methods
- iterative methods
- factorization approaches
- integral equation methods



# EIT AND THE INVERSE CONDUCTIVITY PROBLEM

Reconstruction methods:

- back-projection methods
- iterative methods
- factorization approaches
- integral equation methods

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- EIT and the inverse conductivity problem
- **Integral equation formulation**
- A Tikhonov regularization method
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- Conclusions

## INTEGRAL EQUATION FORMULATION

Assume  $\Omega \subset \mathbb{R}^n$ ,  $n = 2, 3$ , an open bounded Lipschitz domain with boundary  $\partial\Omega$  and  $\sigma \in C^{0,1}(\overline{\Omega})$ .

The inverse problem to solve is the following: find  $\sigma(\mathbf{x})$ ,  $\mathbf{x} \in \Omega$ , satisfying

$$\nabla \cdot [\sigma(\mathbf{x}) \nabla \Phi(\mathbf{x})] = 0, \quad \mathbf{x} \in \Omega, \quad \Phi \in H^1(\Omega),$$

$$\Phi(\mathbf{x}), \quad \mathbf{x} \in \partial\Omega,$$

$$j(\mathbf{x}) = \sigma(\mathbf{x}) \frac{\partial \Phi(\mathbf{x})}{\partial n}, \quad \mathbf{x} \in \partial\Omega,$$

$$\sigma(\mathbf{x}), \quad \mathbf{x} \in \partial\Omega.$$

# INTEGRAL EQUATION FORMULATION

$$\nabla \cdot [\sigma(\mathbf{x}) \nabla \Phi(\mathbf{x})] = 0, \quad \mathbf{x} \in \Omega$$

$$\partial\Omega : \sigma, \Phi, \frac{\partial\Phi}{\partial n} \Rightarrow \sigma(\mathbf{x}), \mathbf{x} \in \Omega$$

# INTEGRAL EQUATION FORMULATION

$$\nabla \cdot [\sigma(\mathbf{x}) \nabla \Phi(\mathbf{x})] = 0, \quad \mathbf{x} \in \Omega$$

$$\partial\Omega : \sigma, \Phi, \frac{\partial\Phi}{\partial n} \Rightarrow \sigma(\mathbf{x}), \mathbf{x} \in \Omega$$

$$\nabla \sigma(\mathbf{x}) \cdot \nabla \Phi(\mathbf{x}) + \sigma(\mathbf{x}) \Delta \Phi(\mathbf{x}) = 0$$

# INTEGRAL EQUATION FORMULATION

$$\nabla \cdot [\sigma(\mathbf{x}) \nabla \Phi(\mathbf{x})] = 0, \quad \mathbf{x} \in \Omega$$

$$\partial\Omega : \sigma, \Phi, \frac{\partial\Phi}{\partial n} \Rightarrow \sigma(\mathbf{x}), \mathbf{x} \in \Omega$$

$$\Delta\Phi(\mathbf{x}) = -\frac{\nabla\sigma(\mathbf{x})}{\sigma(\mathbf{x})} \cdot \nabla\Phi(\mathbf{x})$$

# INTEGRAL EQUATION FORMULATION

$$\nabla \cdot [\sigma(\mathbf{x}) \nabla \Phi(\mathbf{x})] = 0, \quad \mathbf{x} \in \Omega$$

$$\partial\Omega : \sigma, \Phi, \frac{\partial\Phi}{\partial n} \Rightarrow \sigma(\mathbf{x}), \mathbf{x} \in \Omega$$

$$\Delta\Phi(\mathbf{x}) = -\nabla\tilde{\sigma}(\mathbf{x}) \cdot \nabla\Phi(\mathbf{x})$$

$$\tilde{\sigma}(\mathbf{x}) = \ln \sigma(\mathbf{x})$$

# INTEGRAL EQUATION FORMULATION

$$\nabla \cdot [\sigma(\mathbf{x}) \nabla \Phi(\mathbf{x})] = 0, \quad \mathbf{x} \in \Omega$$

$$\partial\Omega : \sigma, \Phi, \frac{\partial\Phi}{\partial n} \Rightarrow \sigma(\mathbf{x}), \mathbf{x} \in \Omega$$

$$\Delta\Phi(\mathbf{x}) = -Y(\mathbf{x})$$

$$\tilde{\sigma}(\mathbf{x}) = \ln \sigma(\mathbf{x}), \quad Y(\mathbf{x}) = \nabla \tilde{\sigma}(\mathbf{x}) \cdot \nabla \Phi(\mathbf{x})$$



# INTEGRAL EQUATION FORMULATION

$$\nabla \cdot [\sigma(\mathbf{x}) \nabla \Phi(\mathbf{x})] = 0, \quad \mathbf{x} \in \Omega$$

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$$\Delta\Phi(\mathbf{x}) = -Y(\mathbf{x})$$

$$\partial\Omega : \tilde{\sigma}, \Phi, \frac{\partial\Phi}{\partial n} \Rightarrow \tilde{\sigma}(\mathbf{x}), \quad \mathbf{x} \in \Omega$$

# INTEGRAL EQUATION FORMULATION

Let  $H : \Omega \times \Omega \rightarrow \mathbb{R}$  satisfy the Helmholtz equation

$$\Delta_{\mathbf{y}} H(\mathbf{x}, \mathbf{y}) + \lambda H(\mathbf{x}, \mathbf{y}) = 0, \quad \mathbf{x}, \mathbf{y} \in \Omega, \quad \lambda > 0$$

Let  $\mathcal{F}$  be the set of functions  $f : \Omega \rightarrow \mathbb{R}$  which satisfy the Helmholtz equation

$$\Delta_{\mathbf{y}} f(\mathbf{y}) + \lambda f(\mathbf{y}) = 0, \quad \mathbf{y} \in \Omega, \quad \lambda > 0$$

Let  $\mathcal{G}_0(\mathbf{x}, \mathbf{y})$  be the free space Green's function

$$\Delta_{\mathbf{y}} \mathcal{G}_0(\mathbf{x}, \mathbf{y}) + \lambda \mathcal{G}_0(\mathbf{x}, \mathbf{y}) = -\delta(\mathbf{x} - \mathbf{y}), \quad \mathbf{x}, \mathbf{y} \in \Omega$$

Green's second identity

$$\int_{\Omega} (u \Delta v - v \Delta u) = \int_{\partial \Omega} \left( u \frac{\partial v}{\partial n} - v \frac{\partial u}{\partial n} \right)$$

## INTEGRAL EQUATION FORMULATION

$$0 = \zeta(\mathbf{x}) - \int_{\Omega} d\mathbf{y} H(\mathbf{x}, \mathbf{y})(Y(\mathbf{y}) - \lambda\Phi(\mathbf{y}))$$

$$\zeta(\mathbf{x}) = \int_{\partial\Omega} d\mathbf{y} \left( \Phi(\mathbf{y}) \frac{\partial H}{\partial n}(\mathbf{x}, \mathbf{y}) - H(\mathbf{x}, \mathbf{y}) \frac{\partial \Phi(\mathbf{y})}{\partial n} \right)$$

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$$\Phi(\mathbf{x}) = \zeta_0(\mathbf{x}) - \int_{\Omega} d\mathbf{y} \mathcal{G}_0(\mathbf{x}, \mathbf{y})(Y(\mathbf{y}) - \lambda\Phi(\mathbf{y}))$$

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$$X(\mathbf{y}) = Y(\mathbf{y}) - \lambda\Phi(\mathbf{y})$$

## INTEGRAL EQUATION FORMULATION

$$0 = \zeta(\mathbf{x}) - \int_{\Omega} d\mathbf{y} H(\mathbf{x}, \mathbf{y}) X(\mathbf{y})$$

$$\zeta(\mathbf{x}) = \int_{\partial\Omega} d\mathbf{y} \left( \Phi(\mathbf{y}) \frac{\partial H}{\partial n}(\mathbf{x}, \mathbf{y}) - H(\mathbf{x}, \mathbf{y}) \frac{\partial \Phi(\mathbf{y})}{\partial n} \right)$$

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$$X(\mathbf{y}) = Y(\mathbf{y}) - \lambda \Phi(\mathbf{y})$$

# INTEGRAL EQUATION FORMULATION

- Solve the linear problem  $\int_{\Omega} d\mathbf{y} H(\mathbf{x}, \mathbf{y}) X(\mathbf{y}) = \zeta(\mathbf{x})$

# INTEGRAL EQUATION FORMULATION

- Solve the linear problem  $AX = \zeta$



# INTEGRAL EQUATION FORMULATION

- Solve the linear problem  $AX = \zeta$

$$\begin{aligned}
 A : L^2(\Omega) &\longrightarrow L^2(\Omega) \\
 X &\longmapsto \int_{\Omega} dy H(\mathbf{x}, \mathbf{y}) X(\mathbf{y}) .
 \end{aligned}$$

# INTEGRAL EQUATION FORMULATION

- Solve the linear problem  $AX = \zeta$

$$A : L^2(\Omega) \longrightarrow L^2(\Omega)$$

$$X \longmapsto \int_{\Omega} dy H(\mathbf{x}, \mathbf{y}) X(\mathbf{y}).$$

$$X(\mathbf{x}) \approx Y(\mathbf{x}) - \lambda \Phi(\mathbf{x}),$$

# INTEGRAL EQUATION FORMULATION

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- $\Phi(\mathbf{x}) = \zeta_0(\mathbf{x}) - \int_{\Omega} dy \mathcal{G}_0(\mathbf{x}, \mathbf{y}) X(\mathbf{y}), \mathbf{x} \in \Omega$

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- Solve the linear problem  $AX = \zeta$

$$\begin{aligned} A : L^2(\Omega) &\longrightarrow L^2(\Omega) \\ X &\longmapsto \int_{\Omega} dy H(\mathbf{x}, \mathbf{y}) X(\mathbf{y}). \end{aligned}$$

- $\Phi(\mathbf{x}) = \zeta_0(\mathbf{x}) - \int_{\Omega} dy \mathcal{G}_0(\mathbf{x}, \mathbf{y}) X(\mathbf{y}), \quad \mathbf{x} \in \Omega$
- Compute  $\tilde{\sigma}(\mathbf{x}) = \ln \sigma(\mathbf{x}), \quad \mathbf{x} \in \Omega$ , by solving

$$\nabla \tilde{\sigma}(\mathbf{x}) \cdot \nabla \Phi = -Y(\mathbf{x}), \quad \text{subject to } \sigma(\tilde{\mathbf{x}}), \quad \mathbf{x} \in \partial\Omega.$$

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- EIT and the inverse conductivity problem
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- **A Tikhonov regularization method**
- A mollifier method
- Numerical examples
- Conclusions and future work

# A TIKHONOV REGULARIZATION METHOD

$$AX = \zeta$$

$$\begin{array}{lcl}
 A : L^2(\Omega) & \longrightarrow & L^2(\Omega) \\
 X & \longmapsto & \int_{\Omega} d\mathbf{y} H(\mathbf{x}, \mathbf{y}) X(\mathbf{y}) .
 \end{array}$$



# A TIKHONOV REGULARIZATION METHOD

$$AX = \zeta$$

$$A : L^2(\Omega) \longrightarrow L^2(\Omega)$$

$$X \longmapsto \int_{\Omega} dy H(\mathbf{x}, \mathbf{y}) X(\mathbf{y}) .$$

$$\|\zeta(\mathbf{x}) - AX_{reg}(\mathbf{x})\|_{L^2} \rightarrow \min \text{ subject to } \|X_{reg}(\mathbf{x}) - X_{mod}(\mathbf{x})\|_{L^2} \leq \delta$$

# A TIKHONOV REGULARIZATION METHOD

$$AX = \zeta$$

$$A : L^2(\Omega) \longrightarrow L^2(\Omega)$$

$$X \longmapsto \int_{\Omega} dy H(\mathbf{x}, \mathbf{y}) X(\mathbf{y}) .$$

$$\|\zeta(\mathbf{x}) - AX_{reg}(\mathbf{x})\|_{L^2}^2 + \mu \left( \|X_{reg}(\mathbf{x}) - X_{mod}(\mathbf{x})\|_{L^2}^2 - \delta^2 \right) \rightarrow \min$$

## A TIKHONOV REGULARIZATION METHOD

$$AX = \zeta$$

$$\begin{aligned} A : L^2(\Omega) &\longrightarrow L^2(\Omega) \\ X &\longmapsto \int_{\Omega} dy H(\mathbf{x}, \mathbf{y}) X(\mathbf{y}). \end{aligned}$$

$$\|\zeta(\mathbf{x}) - AX_{reg}(\mathbf{x})\|_{L^2}^2 + \mu \left( \|X_{reg}(\mathbf{x}) - X_{mod}(\mathbf{x})\|_{L^2}^2 - \delta^2 \right) \rightarrow \min$$

$$X_{reg}(\mathbf{x}) = X_{mod}(\mathbf{x}) + \frac{1}{\mu} \int_{\Omega} dy H(\mathbf{x}, \mathbf{y}) \zeta(\mathbf{y}) - \frac{1}{\mu} \int_{\Omega} dy H_2(\mathbf{x}, \mathbf{y}) X_{reg}(\mathbf{y})$$

where  $H_2(\mathbf{x}, \mathbf{y}) = \int_{\Omega} dz H(\mathbf{x}, \mathbf{z}) H(\mathbf{z}, \mathbf{y})$ .

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# A MOLLIFIER METHOD

$$AX = \zeta$$

$$\begin{array}{lcl}
 A : L^2(\Omega) & \longrightarrow & L^2(\Omega) \\
 X & \longmapsto & \int_{\Omega} d\mathbf{y} H(\mathbf{x}, \mathbf{y}) X(\mathbf{y}) .
 \end{array}$$

## A MOLLIFIER METHOD

$$AX = \zeta$$

$$A : L^2(\Omega) \longrightarrow L^2(\Omega)$$

$$X \longmapsto \int_{\Omega} dy H(\mathbf{x}, \mathbf{y}) X(\mathbf{y}) .$$

### Advantages of mollifier methods:

- Locally adapted resolution can be easily incorporated.
- Solve an operator equation for every reconstruction point  $\mathbf{x}$ .
- All the pointwise reconstruction vectors can be precomputed.
- The choice of the function  $H$  can be optimized.

## A MOLLIFIER METHOD

$$\boxed{AX = \zeta}$$

$$\int_{\Omega} d\mathbf{y} X(\mathbf{y}) e_{\gamma}(\tilde{\mathbf{y}}, \mathbf{y}) \rightarrow X(\tilde{\mathbf{y}}), \text{ as } \gamma \rightarrow 0$$

$$e_{\gamma}(\tilde{\mathbf{y}}, \mathbf{y}) = \frac{1}{|B(\tilde{\mathbf{y}}, \gamma)|} \cdot \begin{cases} 1 & : \mathbf{y} \in B(\tilde{\mathbf{y}}, \gamma) \\ 0 & : \text{otherwise} \end{cases},$$

$$B(\tilde{\mathbf{y}}, \gamma) = \{\mathbf{y} : \|\tilde{\mathbf{y}} - \mathbf{y}\| \leq \gamma\}$$

▪

$$X_{\gamma}(\tilde{\mathbf{y}}) = \langle e_{\gamma}(\tilde{\mathbf{y}}, \cdot), X \rangle_{L^2(\Omega)},$$

## A MOLLIFIER METHOD

$$AX = \zeta$$

$$\int_{\Omega} d\mathbf{y} X(\mathbf{y}) e_{\gamma}(\tilde{\mathbf{y}}, \mathbf{y}) \rightarrow X(\tilde{\mathbf{y}}), \text{ as } \gamma \rightarrow 0$$

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▪

$$X_{\gamma}(\tilde{\mathbf{y}}) = \langle \tilde{e}_{\gamma}(\tilde{\mathbf{y}}, \cdot), X \rangle_{L^2(\Omega)},$$

where  $\tilde{e}_{\gamma} \in R(A^*) \subset \mathcal{F}$ .



## A MOLLIFIER METHOD

$$AX = \zeta$$

**Theorem.** *The mollified approximation to the solution  $X$  of the above equation at  $\tilde{\mathbf{y}}$  is given by*

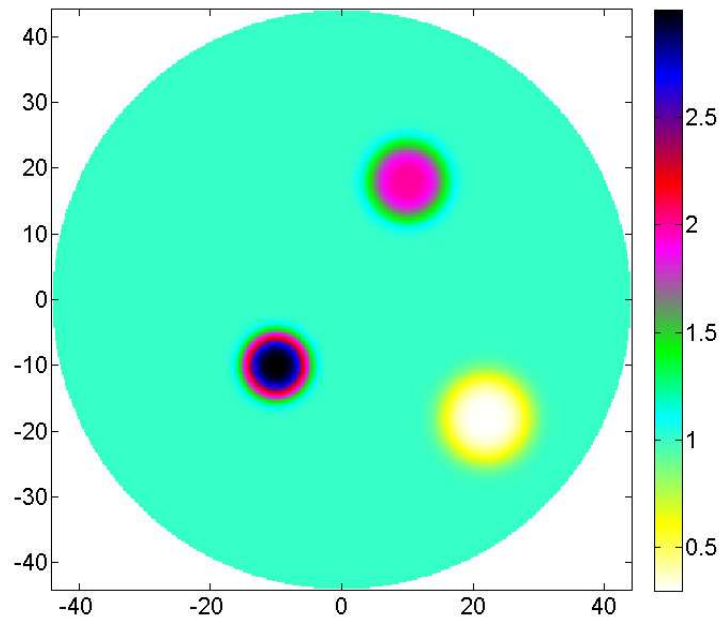
$$\begin{aligned} X_\gamma(\tilde{\mathbf{y}}) &= \langle \tilde{e}_\gamma(\tilde{\mathbf{y}}, \cdot), X \rangle_{L^2(\Omega)} \\ &= \int_{\partial\Omega} \left( \frac{\partial \tilde{e}_\gamma}{\partial n}(\tilde{\mathbf{y}}, \mathbf{x}) \Phi(\mathbf{x}) - \tilde{e}_\gamma(\tilde{\mathbf{y}}, \mathbf{x}) \frac{\partial \Phi}{\partial n}(\mathbf{x}) \right) d\mathbf{x} . \end{aligned}$$

# OUTLINE

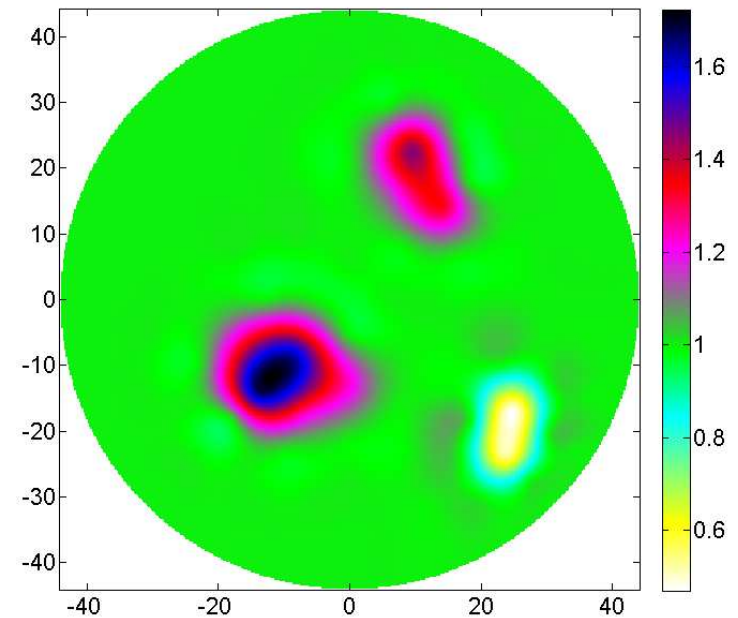
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## NUMERICAL EXAMPLES

**Example.** *The conductivity distribution consists of two off-centered high conductivity regions and a low conductivity region within a constant background. We reconstructed conductivity for data with 2% random errors, regularization parameter  $\gamma = 0.1$  and  $\lambda = 1$ .*



(a) The exact conductivity distribution.



(b) The reconstructed conductivity

# OUTLINE

- The inverse conductivity problem
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# CONCLUSIONS

- Integral Equation formulation for reconstructing  $\sigma \in C^{0,1}(\Omega)$
- The kernel of the Fredholm integral equation of the first kind satisfies the Helmholtz equation
- The approaches are not geometrically constrained
- The algorithms are quite stable with respect to the noise level in the data