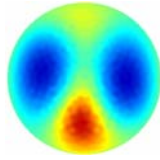


The d-bar reconstruction method for electrical impedance tomography

汁太念
砂夢理



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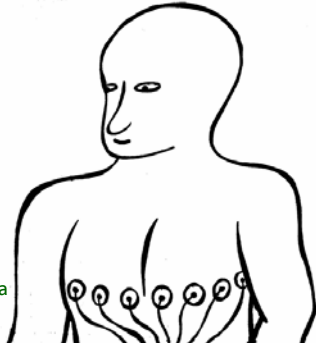
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Electrical impedance tomography (EIT) is an emerging medical imaging method

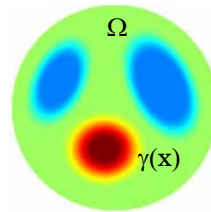
Feed electric currents through electrodes, measure voltages

Reconstruct the image of electric conductivity in a two-dimensional slice

Applications include:
monitoring heart and lungs of unconscious patients, detecting pulmonary edema (swollen lungs)



The inverse conductivity problem of Calderón is the mathematical model of EIT



$$\Lambda_\gamma f = \gamma \frac{\partial u}{\partial \nu} \Big|_{\partial \Omega},$$

$$\begin{aligned} \nabla \cdot \gamma \nabla u &= 0 \text{ in } \Omega, \\ u &= f \text{ on } \partial \Omega. \end{aligned}$$

Given the Dirichlet-to-Neumann map, how to reconstruct the conductivity?

The reconstruction problem is nonlinear and ill-posed.

Nonlinearity of Calderón's problem

The weak formulation of the DN map as an operator

$$\Lambda_\gamma : H^{1/2}(\partial \Omega) \rightarrow H^{-1/2}(\partial \Omega),$$

is given by

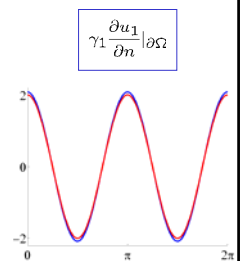
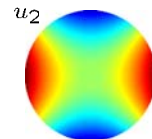
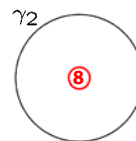
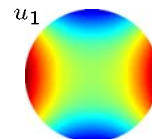
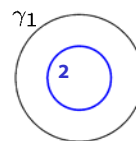
$$\langle \Lambda_\gamma f, g \rangle = \int_\Omega \gamma \nabla u \cdot \nabla v,$$

where v is any H^1 function with trace g , and u satisfies the Dirichlet problem

$$\begin{cases} \nabla \cdot \gamma \nabla u = 0 & \text{in } \Omega, \\ u = f & \text{on } \partial \Omega. \end{cases}$$

Now the map $\gamma \mapsto \Lambda_\gamma$ is nonlinear because u depends on γ .

Ill-posedness of Calderón's problem



$$\gamma_2 \frac{\partial u_2}{\partial n} \Big|_{\partial \Omega}$$

EIT reconstruction algorithms can be divided roughly into the following classes:

Linearization

Iterative output least-squares methods

Statistical inversion

The inverse scattering approach, or **d-bar method**

Theoretical development of the d-bar method can be found in these studies

1980 Calderón

1987 Sylvester and Uhlmann

1987 R G Novikov

1988 Nachman

1996 Nachman

1997 Liu

1997 Brown and Uhlmann

2000 Francini

2001 Barceló, Barceló and Ruiz

2003 Knudsen and Tamasan

2003 Mueller and S

2003 Astala and Päivärinta

Throughout this talk we use these definitions:

Let $\Omega \subset \mathbb{R}^2$ be the unit disc.

Let $\gamma \in C^4(\Omega)$ be strictly positive.

Assume that $\gamma \equiv 1$ near $\partial\Omega$.

Define $q = \frac{\Delta\gamma^{1/2}}{\gamma^{1/2}} \in C^2(\mathbb{R}^2)$ by zero extension.

Nachman's 1996 proof consists of two steps:

$$\Lambda_\gamma \rightarrow \mathbf{t} \rightarrow \gamma$$

The intermediate object \mathbf{t} is a complex-valued function called *scattering transform* and defined as follows for complex k :

$$\mathbf{t}(k) := \int_{\mathbb{R}^2} e^{i\bar{k}\bar{x}} q(x) \psi(x, k) dx$$

$$q = \frac{\Delta\gamma^{1/2}}{\gamma^{1/2}}$$

$$(-\Delta + q)\psi(\cdot, k) = 0$$

$$\psi(x, k) \sim e^{ikx} = e^{i(k_1+ik_2)(x_1+ix_2)}$$

Step 1: from DN map to scattering transform

Solve traces of ψ from the boundary integral equation

$$\psi(\cdot, k)|_{\partial\Omega} = e^{ikx} - S_k(\Lambda_\gamma - \Lambda_1)\psi(\cdot, k),$$

where the single-layer operator has Faddeev Green's function as kernel.

Compute the scattering transform as

$$\mathbf{t}(k) = \int_{\partial\Omega} e^{i\bar{k}\bar{x}} (\Lambda_\gamma - \Lambda_1)\psi(x, k) d\sigma(x).$$

Step 2: from scattering transform to γ

Define $\mu(x, k) = e^{-ikx}\psi(x, k)$

Then the following d-bar equation holds:

$$\frac{\partial}{\partial \bar{k}} \mu(x, k) = \frac{\mathbf{t}(k)}{4\pi \bar{k}} e^{-i(kx + \bar{k}\bar{x})} \overline{\mu(x, k)}.$$

The d-bar equation has a unique solution for all x . The conductivity can be recovered from

$$\gamma^{1/2}(x) = \lim_{k \rightarrow 0} \mu(x, k).$$

To evaluate the scattering transform we need exponentially growing solutions

Recall the definition of the scattering transform:

$$t(k) = \int_{\mathbb{R}^2} e^{i\bar{k}\bar{x}} q(x) \psi(x, k) dx$$

Given a conductivity-type potential, we wish to evaluate numerically the solutions

$$(-\Delta + q)\psi(\cdot, k) = 0, \quad \psi(x, k) \sim e^{ikx}.$$

This can be done via the Lippmann-Schwinger type equation

$$\begin{aligned} \mu(x, k) &= 1 - \int_{\text{supp } q} g_k(x-y) q(y) \mu(y, k) dy, \\ \mu(x, k) &= e^{-ikx} \psi(x, k). \end{aligned}$$

In dimension two, the definition of Faddeev's Green function is as follows:

Let $k \in \mathbb{C} \setminus 0$. Define $g_k: \mathbb{R}^2 \rightarrow \mathbb{C}$ as

$$g_k(x) = \frac{1}{(2\pi)^2} \int_{\mathbb{R}^2} \frac{e^{ix \cdot \xi}}{|\xi|^2 + 2k(\xi_1 + i\xi_2)} d\xi_1 d\xi_2.$$

Then g_k satisfies

$$(-\Delta - 4ik\bar{\partial})g_k = \delta,$$

where

$$\bar{\partial} = \frac{1}{2} \left(\frac{\partial}{\partial x_1} + i \frac{\partial}{\partial x_2} \right).$$

Note the symmetry $g_k(x) = g_1(kx)$.

Equation for exponentially growing solutions can be rewritten in periodic setting

We wish to solve the equation

$$\mu(x, k) = 1 - \int_{\text{supp } q} g_k(x-y) q(y) \mu(y, k) dy.$$

We take $R > 0$ such that $\text{supp } q \subset B(0, R)$ and choose a square $S \supset B(0, 2R)$.

Then we consider the S -periodic equation

$$w = 1 - \tilde{g}_k * (qw),$$

where \tilde{g}_k is periodic extension of $g_k|_S$ and $*$ is convolution on the torus. It follows that

$$\mu|_{\text{supp } q} = w|_{\text{supp } q}.$$

We need three steps for numerical computation of the scattering transform

$$t(k) = \int_{\mathbb{R}^2} e^{i\bar{k}\bar{x}} q(x) \psi(x, k) dx$$

1. Evaluation of Faddeev's Green function

$$g_1(x) = \frac{1}{(2\pi)^2} \int_{\mathbb{R}^2} \frac{e^{ix \cdot \xi}}{|\xi|^2 + 2(\xi_1 + i\xi_2)} d\xi_1 d\xi_2.$$

2. Solution of the periodic equation

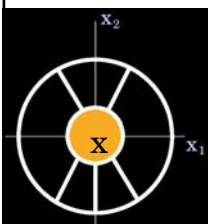
$$w = 1 - \tilde{g}_k * (qw)$$

3. Substituting the result to definition of $t(k)$

$$\psi(x, k)|_{\text{supp } q} = e^{ikx} w(x, k)|_{\text{supp } q}$$

For x in the unit disc we compute $g_1(x)$ with a formula by [Boiti et al 1987]

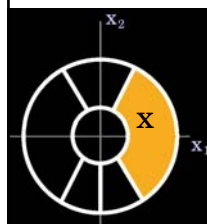
$$g_1(x) = -\frac{e^{-ix}}{4\pi} (2\gamma + \log|x|^2 + \sum_{n=1}^{\infty} \frac{(ix)^n + (-i\bar{x})^n}{nm!})$$



Here γ is Euler's constant

We write $g_1(x)$ as a formula containing a rapidly converging one-dimensional integral

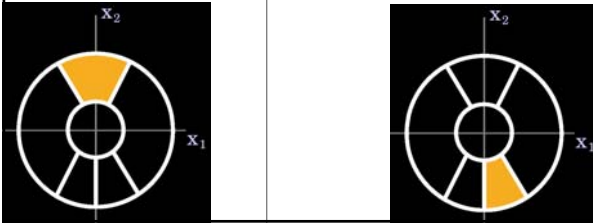
$$g_1(x) = \frac{e^{-ix_1}}{2\pi} \text{Re} \left[-e^{ix_1} \sum_{j=0}^N \frac{j!}{(ix)^{j+1}} + \frac{(N+1)! e^{ix_1}}{(-x)^{N+1}} \int_0^{\infty} \frac{e^{-t(x_1 + ix_2)}}{(t-i)^{N+2}} dt \right]$$



For other domains we use residue calculus and reflectional symmetry

Cases 3 and 4: The integral in case 2 is modified using the residue theorem

$$-i \int_0^\infty \frac{e^{-x_2 s + i x_1 s}}{(-i s - i)^{N+2}} ds \quad (1+i) \int_0^\infty \frac{e^{-i s(x_2 + x_1) + s(x_2 - x_1)}}{(s + i s - i)^{N+2}} ds$$



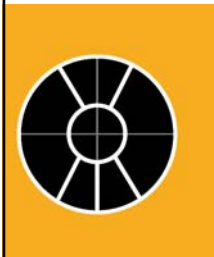
Cases 5 and 6 are reduced to cases 4 and 2 using the following symmetry:

$$g_1(-x_1, x_2) = \overline{g_1(x_1, x_2)}$$

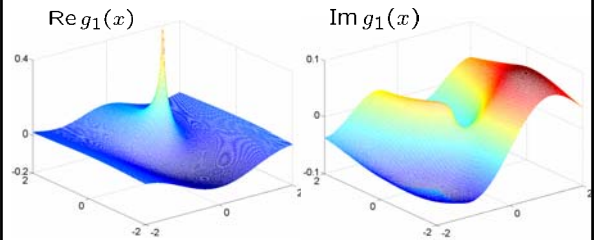


Case 7: for $|x| > 25$ we ignore the integral in case 2 and use the truncated sum

$$g_1(x) \approx \frac{e^{-i x_1}}{2\pi} \operatorname{Re} \left[-e^{i x_1} \sum_{j=0}^N \frac{j!}{(i x)^{j+1}} \right]$$



We now have a complete algorithm for the evaluation of Faddeev's Green function

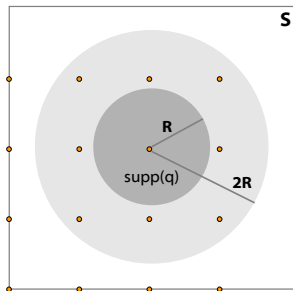


We solve the equation $w = 1 - \tilde{g}_k * (qw)$ following [Vainikko 2000]

We take a grid on S with $(2^m \times 2^m)$ points

Here $m=2$, in practice typically $m=8$.

This grid is suitable for the use of Fast Fourier Transform (FFT).



Vainikko's method is based on iterative solution of linear equations

We can solve the discretized periodic equation

$$[I + \tilde{g}_k * (q \cdot)] w = 1$$

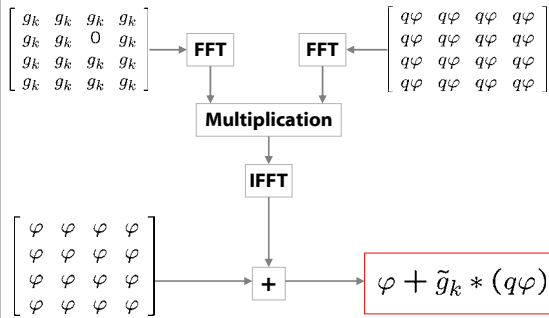
using the iterative GMRES method.

We just need to implement the linear operator

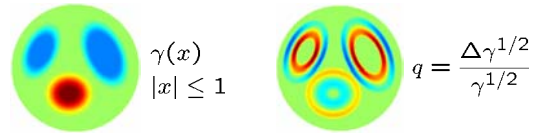
$$\varphi \mapsto \varphi + \tilde{g}_k * (q\varphi)$$

for a function φ given on the grid points.

The linear operator is implemented using Fast Fourier Transform (FFT)

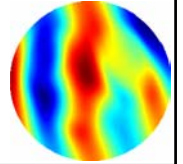
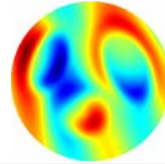


Given a conductivity, we compute the potential and Faddeev solutions μ

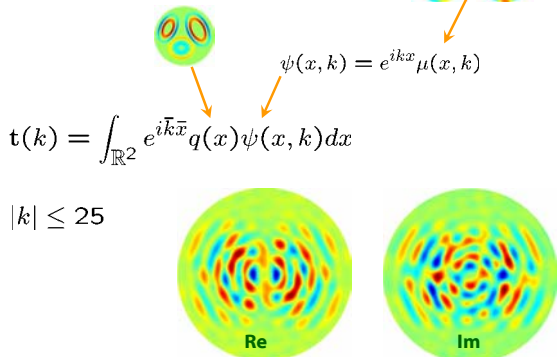


$\text{Re}(\mu(x, 3 + i) - 1)$

$\text{Im}(\mu(x, 3 + i))$



We can now use μ to compute the scattering transform



Practical step 1: compute stable approximation to scattering transform from noisy data

With noisy data, we cannot solve equation

$$\psi(\cdot, k)|_{\partial\Omega} = e^{ikx} - S_k(\Lambda_\gamma - \Lambda_1)\psi(\cdot, k),$$

so we introduce the approximate scattering transform:

$$\mathbf{t}^{\text{exp}}(k) = \int_{\partial\Omega} e^{i\bar{k}\bar{x}} (\Lambda_\gamma - \Lambda_1) e^{ikx} d\sigma(x)$$

Further, we regularize the computation by truncation:

$$\mathbf{t}_R^{\text{exp}}(k) := \begin{cases} \mathbf{t}^{\text{exp}}(k), & |k| < R, \\ 0, & |k| \geq R. \end{cases}$$

Practical step 2: solve the d-bar equation with approximate kernel

Write the approximate dbar equation

$$\frac{\partial}{\partial \bar{k}} \mu_R(x, k) = \frac{\mathbf{t}_R^{\text{exp}}(k)}{4\pi \bar{k}} e^{-i(kx + \bar{k}\bar{x})} \overline{\mu_R(x, k)}$$

In integral form:

$$\mu_R(x, k) = 1 + \frac{1}{\pi k} * \left(\frac{\mathbf{t}_R^{\text{exp}}(k)}{4\pi \bar{k}} e^{-i(kx + \bar{k}\bar{x})} \overline{\mu_R(x, k)} \right)$$

This Lippmann-Schwinger-type equation can be solved numerically with modified Vainikko's algorithm. Then

$$\gamma_R^{1/2}(x) = \mu_R(x, 0).$$

The d-bar equation can be solved in a bounded domain using periodization

Instead of the d-bar equation

$$\mu_R(x, k) = 1 + \frac{1}{\pi k} * \left(\frac{\mathbf{t}_R^{\text{exp}}(k)}{4\pi \bar{k}} e^{-i(kx + \bar{k}\bar{x})} \overline{\mu_R(x, k)} \right)$$

valid in the k -plane, we solve the S -periodic equation

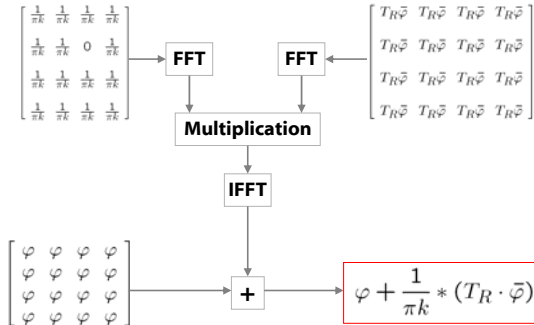
$$\left[I + \frac{1}{\pi k} * (T_R \cdot) \right] w = 1$$

$$T_R(k) = -\frac{\mathbf{t}_R^{\text{exp}}(k)}{4\pi \bar{k}} e^{-i(kx + \bar{k}\bar{x})}$$

The d-bar equation is also solved since it can be shown that

$$\mu_R(x, \cdot)|_{B(0, R)} = w|_{B(0, R)}$$

The convolution is effectively implemented using Fast Fourier Transform (FFT)



Note: real and imaginary parts must be kept separate!

Truncation of scattering transform gives asymptotically correct reconstruction

Theorem [Mueller & S 2003].

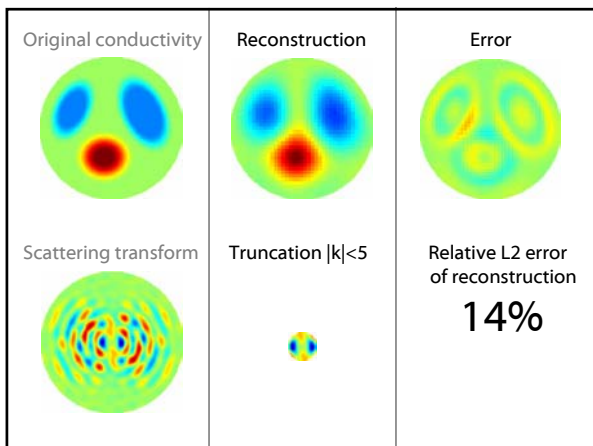
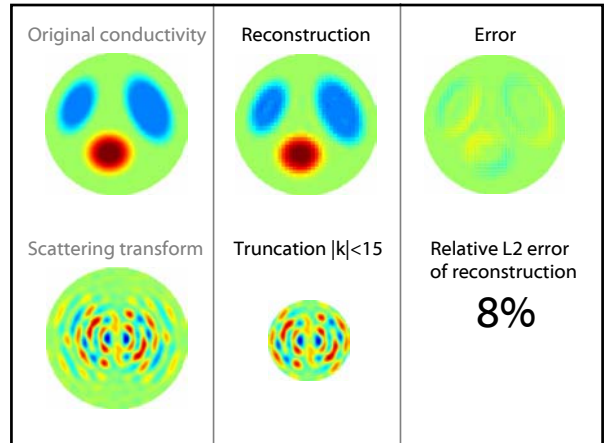
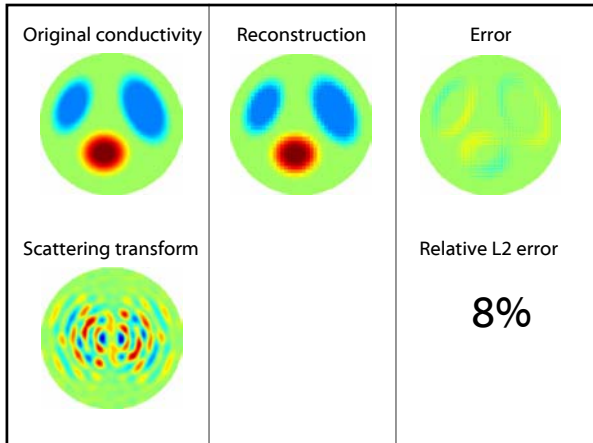
Let μ_R be the solution of the $\bar{\partial}$ equation

$$\frac{\partial}{\partial \bar{k}} \mu_R(x, k) = \frac{t_R(k)}{4\pi k} e^{-i(kx + \bar{k}x)} \overline{\mu_R(x, k)}.$$

Then the following estimate holds for large R :

$$\|\sqrt{\gamma} - \mu_R(\cdot, 0)\|_{L^\infty(\Omega)} \leq CR^{-1}.$$

Next we demonstrate this theorem numerically.



Current-to-voltage measurements are described with the ND map

The Neumann-to-Dirichlet (ND) map is defined by

$$R_\gamma f = u|_{\partial\Omega} - \frac{1}{|\partial\Omega|} \int_{\partial\Omega} u,$$

where u satisfies the Neumann problem

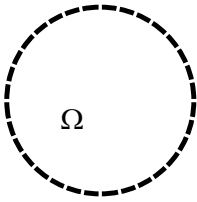
$$\begin{cases} \nabla \cdot \gamma \nabla u = 0 & \text{in } \Omega, \\ \gamma \frac{\partial u}{\partial \nu} = f & \text{on } \partial\Omega. \end{cases}$$

In practice, currents are applied and voltages measured.

This is because

the ND operator is smoothing and suppresses noise, while the DN operator is roughing and amplifies noise.

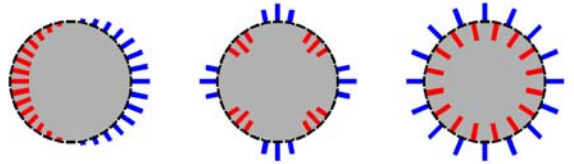
This is a typical configuration for electrode measurements in EIT



Here we have $N=32$ electrodes.
The machine is in Rensselaer Polytechnic Institute, USA.

There is a finite number of linearly independent current patterns

Here are three examples with $N=32$:



Altogether, there are $N-1$ linearly independent current patterns due to conservation of charge.

We represent a finite set of EIT measurements with a complex 31×31 matrix

Define trigonometric current patterns as

$$T_\ell^k = \begin{cases} M \cos(k\theta_\ell), & k = 1, \dots, \frac{N}{2} - 1, \\ M \cos(\pi\ell), & k = N/2, \\ M \sin((k - N/2)\theta_\ell), & k = \frac{N}{2} + 1, \dots, N - 1, \end{cases}$$

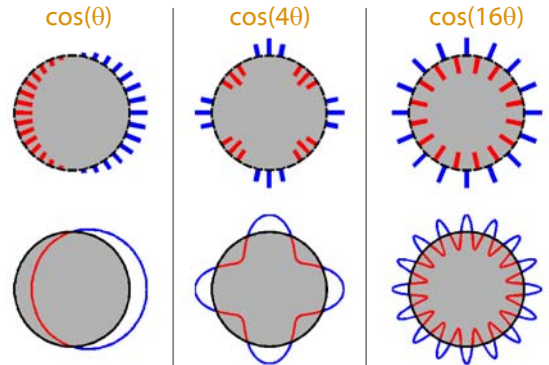
where M is current amplitude. Let V_ℓ^k denote the voltage measured on the ℓ th electrode corresponding to the k th current pattern with

$$\sum_{\ell=1}^N V_\ell^k = 0.$$

Up to normalization, the ND matrix is given by

$$R_\gamma[m, n] := \sum_{\ell} T_\ell^m V_\ell^n.$$

We use trigonometric current patterns in both discrete and continuous form



The approximate scattering transform can now be written in terms of measured data

Expand the exponential function as series:

$$e^{ikz} = \sum_{n=-\infty}^{\infty} a_n(k) e^{in\theta} \quad \text{with} \quad a_n(k) = \begin{cases} \frac{(ik)^n}{n!}, & n \geq 0, \\ 0, & n < 0. \end{cases}$$

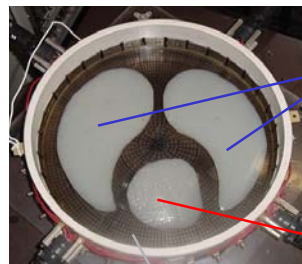
Then we can write

$$t^{\text{exp}}(k) = \sum_{m=0}^{\infty} \sum_{n=0}^{\infty} a_m(\bar{k}) a_n(k) \int_{\partial\Omega} e^{im\theta} (\Lambda_\gamma - \Lambda_1) e^{in\theta} d\sigma(\theta).$$

The relation between DN map the measurements is, roughly,

$$\Lambda_\gamma = R_\gamma^{-1}.$$

At the RPI lab, we construct a chest phantom consisting of saline and agar

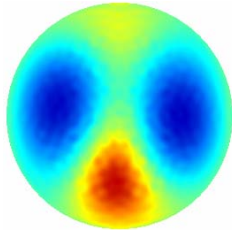
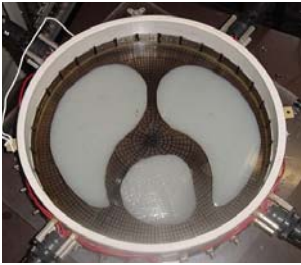


"Lungs" with lower conductivity than background (240 mS/m)

"Heart" with higher conductivity than background (750 mS/m)

Background of salt water, conductivity 424 mS/m.
Diameter of the tank is 30cm.

Reconstruction from measured data



Relative error 23% (lung) and 12% (heart).
Dynamical range is 94% of the true range.

Conclusion

We have developed a new kind of EIT algorithm
Large contrast details recovered remarkably well
The algorithm has rigorous mathematical background

Future challenges

Understanding reconstruction of discontinuities
Removing the requirement $\gamma=1$ near boundary
Taking the shape of domain into account
Developing a 3D algorithm

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