

Bistatic Ultrasound Tomography

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This includes joint work with
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Ultrasound Reflection Tomography (URT)

Acoustic waves are emitted from a source and reflect off of inhomogeneities inside the body. Their echoes are measured by a receiver. Then, emitter and receiver are moved about the body.

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Acoustic waves are emitted from a source and reflect off of inhomogeneities inside the body. Their echoes are measured by a receiver. Then, emitter and receiver are moved about the body.

This measured data are then used to reconstruct the unknown ultrasonic reflectivity function, which is used to generate cross-sectional images of the body.

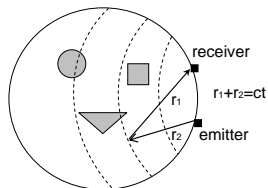
Advantage:

- URT is safe and cost effective.
- URT does a good job imaging soft tissue.

Bistatic Data Acquisition Method

The ultrasound emitter is focussed to send sound waves in a plane.

- The emitter and receiver are separated on the circle, a fixed distance apart.
- The emitter and receiver rotate around the body on the circle (c =constant speed of sound, t =time).

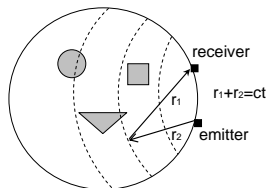


Separated emitted and receiver

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- The data can be modeled as integrals of the reflectivity over ellipses with foci the emitter and receiver.



Separated emitted and receiver

Figure: A sketch of ellipses of integration in bistatic URT

The Parameters

The emitter and receiver move along the unit circle and are 2α radians apart, $\alpha \in (0, \pi/2)$.

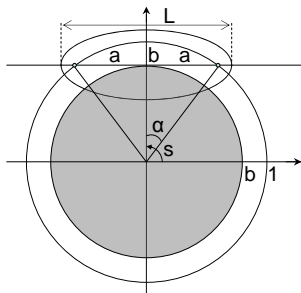


Figure: a , b , and D_b

$$a = \sin(\alpha),$$

$$b = \cos(\alpha)$$

$2a$ = distance between the foci

$$D_b = \{x \in \mathbb{R}^2 \mid \|x\| < b\}$$

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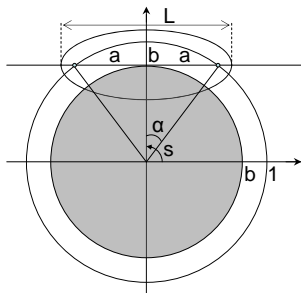


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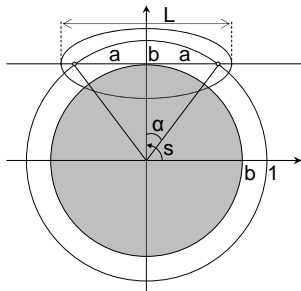
$$D_b = \{x \in \mathbb{R}^2 \mid \|x\| < b\}$$

For $s \in [0, 2\pi]$,

$$\gamma_E(s) = (\cos(s-\alpha), \sin(s-\alpha))$$

$$\gamma_R(s) = (\cos(s+\alpha), \sin(s+\alpha))$$

Ellipse: $E(s, L) = \{x \in \mathbb{R}^2 \mid \|x - \gamma_E(s)\| + \|x - \gamma_R(s)\| = L\}$



L is the major diameter—the sum of the distances from the foci to any point on the ellipse.

s = angle from the positive x axis to the center of the ellipse

Figure: An ellipse $E(s, L)$

The set of ellipses: $Y = \{(s, L) \mid s \in [0, 2\pi], L > 2a\}$

The Elliptical Radon Transform: $f \in \mathcal{E}'(D_b)$, $(s, L) \in Y$

$$\mathcal{R}f(s, L) = \int_{E(s, L)} f(x) dx$$

The Backprojection Transform: $g \in \mathcal{D}'(Y)$, $x \in D_b$

$$\mathcal{R}^*g(x) = \int_{s \in [0, 2\pi]} g(s, \|x - \gamma_E(s)\| + \|x - \gamma_R(s)\|) ds$$

ds is standard Lebesgue measure and \mathcal{R}^* is close to the L^2 adjoint of \mathcal{R} .

\mathcal{R}^* is an integral of g over all ellipses through x .

Our Reconstruction Algorithm

For $f \in \mathcal{E}'(D_b)$, $\mathcal{L}(f)(x) := \mathcal{R}^*(DRf)(x)$

where D is a second order differential operator to be chosen using microlocal analysis.

This is a generalization of Lambda tomography for the line transform: $R^*(-d^2/dp^2)Rf = \sqrt{-\Delta}f$.

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We have developed derivative-backprojection algorithms

- SPECT [Bakhos, Chung, Q]
- Electron Microscopy [Q, Öktem]
- Common Offset Bistatic Synthetic Aperture Radar [Levinson, Q]
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Our Reconstruction Algorithm and Microlocal Analysis

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We have developed derivative-backprojection algorithms and analyzed their microlocal properties

- SPECT [Bakhos, Chung, Q] **Admissible Line Transform**
- Electron Microscopy [Q, Öktem] **Admissible Line Transform**
- Common Offset Bistatic Synthetic Aperture Radar [Levinson, Q] **Like an Elliptical Transform**
- Sonar [Q, Rieder, Schuster] **Spherical Transform**

In each case the forward operator is a FIO related to a Radon transform.

Incidence Relation: $Z = \{(s, L; x) \in Y \times D_b \mid x \in E(s, L)\}$.

Guillemin used the double fibration

$$\begin{array}{ccc} & Z & \\ \pi_L \swarrow & & \searrow \pi_R \\ Y & & D_b \end{array}$$

to prove microlocal properties of generalized Radon transforms.

[Ambartsoumian, Krishnan, Q] prove that \mathcal{R} fits into Guillemin's theory since both projections are fiber maps and π_R has compact fiber, S^1 .

Guillemin [G 1975, GS, G 1985] showed that generalized Radon transforms defined by such double fibrations are elliptic FIOs. The canonical relation is $\mathcal{C} = (N^*(Z) \setminus \mathbf{0})'$:

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The Main Theorem

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Theorem (Ambartsoumian, Krishnan, Q)

For functions supported in D_b , \mathcal{R} satisfies the Bolker Assumption: $\pi_L : \mathcal{C} \rightarrow T^(Y)$ is an injective immersion. Therefore, if D is elliptic, then our reconstruction operator $\mathcal{L} = \mathcal{R}^* D \mathcal{R}$ is an elliptic pseudodifferential operator from $\mathcal{E}'(D_b)$ to $\mathcal{D}'(D_b)$.*

- The composition of two FIOs does not have to be a FIO, and many FIO do not satisfy the Bolker assumption:
our transform in D_b is special.
- Our theorem implies that, for $f \in \mathcal{E}'(D_b)$, $\mathcal{L}(f)$ shows all singularities of f in D_b .

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- In the proof, we put coordinates on \mathcal{C} , (s, L, ϕ, η) where ϕ is a “polar angle” on the ellipse $E(s, L)$ and $\eta \in \mathbb{R} \setminus 0$ is a cotangent coordinate. We reduce to proving that, as a function of ϕ , Π_L is an injective immersion.

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My senior honors thesis student Howard Levinson implemented the algorithm and showed that $D = -d^2/dL^2$ gives better reconstructions than $-d^2/ds^2$ (for microlocal reasons!).

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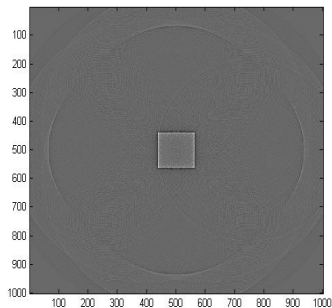


Figure: Reconstruction of a square

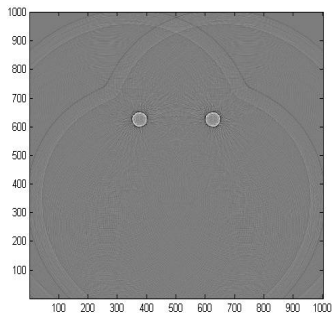


Figure: Reconstruction of two circles

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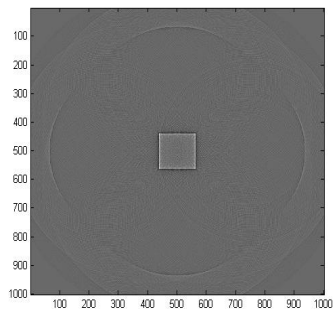


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Artifacts are outside of D_b !

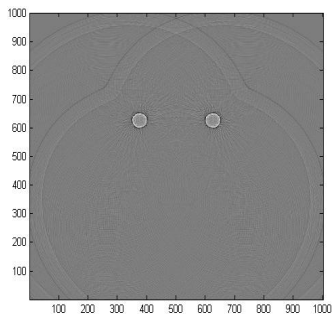


Figure: Reconstruction of two circles