

# Magnet Resonance Electrical Impedance Tomography (MREIT): convergence of the Harmonic $B_z$ Algorithm

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# MREIT: the Harmonic $B_z$ Algorithm<sup>1</sup>

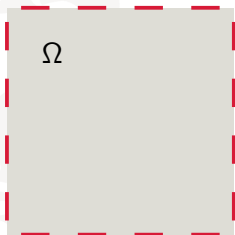
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<sup>1</sup>Based on Seo, Woo, et al. since 2003

D. Garmatter: MREIT: convergence of the  $\nabla^2 B_z$ -Algorithm

## Motivation: Electrical Impedance Tomography (EIT)

- ▶ **Setting:** Imaging object  $O \subset \mathbb{R}^3$  with **electrode pairs** attached
- ▶ **Aim:** reconstruct **cross-sectional** ( $\Omega = O \cap \{z = z_0\} \subset \mathbb{R}^2$ ) image of electrical conductivity inside  $\Omega$

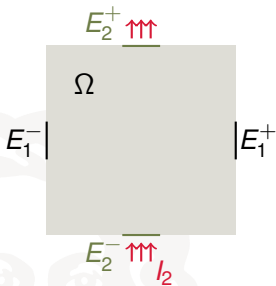


### In EIT

- ▶ **Data:** Current-Voltage measurements on boundary
- ▶ **Difficulty:** highly ill-posed  
 $\rightsquigarrow$  low spatial resolution

## MREIT: Setting

**Aim:** achieve higher resolution of conductivity  $\sigma$ .



- ▶ Object  $\Omega$  with two electrode pairs  $E_j^\pm$  attached
- ▶ Place object inside MRI-Scanner
- ▶ Apply **current** / between **electrode pair**
- ▶ Generates magnetic flux density  $B$  (z-comp. measurable with MRI scanner)

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↪ Full internal data set to overcome ill-posedness

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## Forward problem

Consider ( $j = 1, 2$  determines active electrode pair)

$$\nabla \cdot (\sigma \nabla u_j) = 0 \text{ in } \Omega, \quad I_j = \int_{E_j^+} \sigma \frac{\partial u_j}{\partial \mathbf{n}} ds = - \int_{E_j^-} \sigma \frac{\partial u_j}{\partial \mathbf{n}} ds \quad (1a)$$

$$\nabla u_j \times \mathbf{n} = 0, \text{ on } E_j^+ \cup E_j^-, \quad \sigma \frac{\partial u_j}{\partial \mathbf{n}} = 0 \text{ on } \partial\Omega \setminus \overline{(E_j^+ \cup E_j^-)} \quad (1b)$$

- ▶  $\mathcal{P} \equiv C_{\pm}^1(\bar{\Omega}) := \{\sigma \in C^1(\bar{\Omega}) \mid 0 < \underline{\sigma} \leq \sigma \leq \bar{\sigma} < \infty\}$
- ▶ Unique solution of (1) up to an additive constant

## Inverse Problem

**Aim:** Determine  $\sigma$  from  $B$ .

► Maxwell Equations

$$-\sigma \nabla u_j = \frac{1}{\mu_0} \nabla \times B^j \text{ (Ampère's law)} \quad \text{and} \quad \nabla \cdot B^j = 0$$

- $\nabla \times$  of Ampère's law  $\rightsquigarrow \nabla u_j \times \nabla \sigma = \frac{1}{\mu_0} \nabla^2 B^j$
- Only  $B_z^j$  measurable and two different electrode pairs/currents

### Core-relation (point-wise in $\Omega$ )

$$\begin{pmatrix} \frac{\partial \sigma}{\partial x} \\ \frac{\partial \sigma}{\partial y} \end{pmatrix} = \frac{1}{\mu_0} \mathbb{A}[\sigma]^{-1} \begin{pmatrix} \nabla^2 B_z^1 \\ \nabla^2 B_z^2 \end{pmatrix}, \quad \mathbb{A}[\sigma] := \begin{pmatrix} \frac{\partial u_1}{\partial y} & -\frac{\partial u_1}{\partial x} \\ \frac{\partial u_2}{\partial y} & -\frac{\partial u_2}{\partial x} \end{pmatrix} \quad \text{(CR)}$$

## Recovery of $\sigma$ from $\nabla\sigma$

**So far:**  $\nabla\sigma$  obtainable in  $\Omega$  from  $\nabla^2 B_z^j$  via **(CR)**. How to get  $\sigma$ ?

- **Fundamental solution:** for  $\mathbf{r} = (x, y), \mathbf{r}' = (x', y') \in \Omega \subset \mathbb{R}^2$

$$\Phi(\mathbf{r} - \mathbf{r}') := \frac{1}{2\pi} \ln |\mathbf{r} - \mathbf{r}'| \quad \text{fulfilling} \quad \nabla^2 \Phi(\mathbf{r} - \mathbf{r}') = \delta(\mathbf{r} - \mathbf{r}')$$

- For  $\sigma \in C^1(\bar{\Omega})$  one can show

$$\begin{aligned} \sigma(\mathbf{r}) = & - \int_{\Omega} \nabla_{\mathbf{r}'} \Phi(\mathbf{r} - \mathbf{r}') \cdot \nabla \sigma(\mathbf{r}') d\mathbf{r}' \\ & + \int_{\partial\Omega} \nu(\mathbf{r}') \cdot \nabla_{\mathbf{r}'} \Phi(\mathbf{r} - \mathbf{r}') \sigma(\mathbf{r}') d\ell_{\mathbf{r}'} \end{aligned} \quad (2)$$

## Harmonic $B_z$ Algorithm<sup>2</sup>

- ▶ **Assume:**  $\sigma^* \in \mathcal{P}$  with  $\sigma^* |_{\Omega \setminus \tilde{\Omega}} = \sigma_b$ ,  $\tilde{\Omega} \subset \Omega$ ,  $\sigma_b > 0$  known
- ▶ **Idea:** iteration for (2) + **(CR)** + In-formulation for positivity

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### Algorithm 1 BZ( $\sigma_b, \mu_0, tol$ )

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- 1:  $n := 0, \sigma^0 := \sigma_b$
  - 2: Calculate  $BI(\mathbf{r}) := \int_{\partial\Omega} \nu(\mathbf{r}') \cdot \nabla_{\mathbf{r}'} \Phi(\mathbf{r} - \mathbf{r}') \ln \sigma^*(\mathbf{r}') dI_{\mathbf{r}'}$
  - 3: **repeat**
  - 4:  $F^{n+1}(\mathbf{r}) := \frac{1}{\mu_0} \frac{1}{\sigma^n} \mathbb{A}[\sigma^n]^{-1} \begin{pmatrix} \nabla^2 B_{z,*}^1 \\ \nabla^2 B_{z,*}^2 \end{pmatrix}(\mathbf{r})$
  - 5:  $\ln \sigma^{n+1}(\mathbf{r}) := BI(\mathbf{r}) - \int_{\Omega} \nabla_{\mathbf{r}'} \Phi(\mathbf{r} - \mathbf{r}') \cdot F^{n+1}(\mathbf{r}') d\mathbf{r}'$
  - 6:  $n := n + 1$
  - 7: **until**  $\| \ln \sigma^n - \ln \sigma^{n-1} \|_{\mathcal{P}} \leq tol$
  - 8: **return**  $\sigma_{BZ} := \sigma^n$
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<sup>2</sup>Seo, Woo 2003



## Harmonic $B_z$ Algorithm: convergence

- Define  $\Xi(\sigma_b, \epsilon_0) := \{\sigma \in \mathcal{P} \mid \sigma|_{\Omega \setminus \tilde{\Omega}} = \sigma_b, \|\nabla \ln \sigma\|_{C(\Omega)} < \epsilon_0\}$ .

### Existing result<sup>3</sup>

There exists  $0 < \epsilon < \epsilon_0$ , such that for each  $\sigma^* \in \Xi(\sigma_b, \epsilon_0)$  with  $\|\nabla \ln \sigma^*\|_{C(\Omega)} \leq \epsilon$  the sequence  $\{\sigma^n\}$  generated by the Harmonic  $B_z$  Algorithm with initial guess  $\sigma_b$  satisfies

$$\sigma^n \equiv \sigma_b \text{ in } \Omega \setminus \tilde{\Omega}, \quad \|\ln \sigma^n - \ln \sigma^*\|_{C^1(\tilde{\Omega})} \leq K \left(\frac{1}{2}\right)^n \epsilon,$$

with  $K := \text{diam}(\Omega) + 1$ .

<sup>3</sup>Seo, Woo, Liu, 2010

## Harmonic $B_z$ Algorithm: our approach

- ▶ **Idea:** Replace  $u_j^{\sigma^n}$  (see  $\mathbb{A}[\sigma^n]$ ) in Algo. 1 by approximations  $u_{N,j}^{\sigma^n}$
- ▶ **Assumptions** (for all  $n = 0, 1, 2, \dots$ )
  - ▶  $u_{N,j}^{\sigma^n} \in C^1(\tilde{\Omega})$
  - ▶  $\|\nabla u_{N,j}^{\sigma^n} - \nabla u_j^{\sigma^n}\|_{C(\tilde{\Omega})} \leq \epsilon^{n+1} C, \quad \epsilon$  from Theorem and  $C > 0$

Assumptions fulfilled  $\Rightarrow$  Existing result is replicated

### Applications

- ▶ Fineness of FEM-mesh (actual numerical convergence)
- ▶ **Our scope:** approximations  $u_{N,j}^{\sigma^n}$  via **Reduced Basis Methods**  
 $\rightsquigarrow$  also speed-up the method

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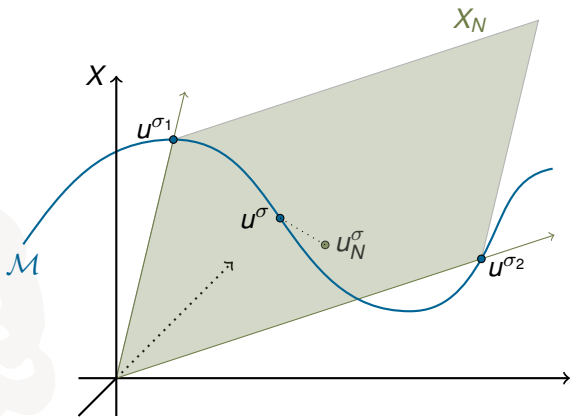
# Reduced Basis Methods & MREIT<sup>4</sup>

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<sup>4</sup>Based on **G.**/Haasdonk/Harrach, A Reduced Basis Landweber method for nonlinear inverse problems, 2016 Inverse Problems 32 (3) (doi).

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## Reduced Basis Methods (RBM): Idea



- ▶ Solution manifold  $\mathcal{M} := \{u^\sigma \mid \sigma \in \mathcal{P}\}$
- ▶ Construction of  $X_N$  via *carefully chosen snapshots*  $u^{\sigma_i}$

## RBM: problem formulation & properties (for $j=1,2$ )

Assume:  $X_{N,j} := \text{span}\{u_j^{\sigma_1}, \dots, u_j^{\sigma_N}\}$  is given.

### Reduced forward problem (Galerkin projection)

For  $\sigma \in \mathcal{P}$ , find  $u_{N,j}^\sigma \in X_{N,j} \subset X$ , the **reduced solution**, of

$$b(u_{N,j}^\sigma, v; \sigma) = f(v), \quad \forall v \in X_{N,j}.$$

- ▶ **Reproduction of solutions:**  $u_j^\sigma \in X_{N,j} \Rightarrow u_{N,j}^\sigma = u_j^\sigma$
- ▶ **Offline/Online-decomposition:** rapid computation of  $u_{N,j}^\sigma$
- ▶ **Certification:**  $\|u_j^\sigma - u_{N,j}^\sigma\|_X \leq \Delta_{N,j}(\sigma) := \frac{\|v_{r,j}\|_X}{\alpha(\sigma)}$

## Reduced Basis Harmonic $B_z$ Algorithm

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### Algorithm 2 RBZ( $\sigma_b, \mu_0, tol_1, tol_2, X_{N,1}, X_{N,2}$ )

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- 1:  $n := 0, \sigma^0 := \sigma_b$
- 2: Calculate  $BI(\mathbf{r}) := \int_{\partial\Omega} \nu(\mathbf{r}') \cdot \nabla_{\mathbf{r}'} \Phi(\mathbf{r} - \mathbf{r}') \ln \sigma^*(\mathbf{r}') dI_{r'}$
- 3: **repeat**
- 4:   enrich  $X_{N,1}, X_{N,2}$  using  $u_1^{\sigma^n}, u_2^{\sigma^n}$
- 5:    $i := 1, \sigma^i := \sigma^n$
- 6:   **repeat**
- 7:      $F_N^{n+1}(\mathbf{r}) := \frac{1}{\mu_0} \frac{1}{\sigma^i} \mathbb{A}_N[\sigma^i]^{-1} \begin{pmatrix} \nabla^2 B_{z,*}^1 \\ \nabla^2 B_{z,*}^2 \end{pmatrix}(\mathbf{r})$
- 8:      $\ln \sigma^{i+1}(\mathbf{r}) := BI(\mathbf{r}) - \int_{\Omega} \nabla_{\mathbf{r}'} \Phi(\mathbf{r} - \mathbf{r}') \cdot F_N^{n+1}(\mathbf{r}') d\mathbf{r}'$
- 9:      $i := i + 1$
- 10:   **until**  $\|\sigma^i - \sigma^{i-1}\|_{\mathcal{P}} \leq tol_1$  **or**  $\Delta_{N,1}(\sigma_i) > tol_2$  **or**  $\Delta_{N,2}(\sigma_i) > tol_2$
- 11:    $\sigma^{n+1} := \sigma^i, n := n + 1$
- 12: **until**  $\|\sigma^i - \sigma^{i-1}\|_{\mathcal{P}} \leq tol_1$
- 13: **return**  $\sigma_{RBZ} := \sigma^n$

- ▶  $\Omega := [-1, 1] \times [-2, 2]$
- ▶  $E_1^\pm := \{(\pm 1, y) \mid |y| < 0.1\}$ ,  $E_2^\pm := \{(x, \pm 2) \mid |x| < 0.1\}$
- ▶ For  $r = \sqrt{x^2 + y^2}$ ,  $x, y \in \Omega$

$$\sigma^* \approx \sigma(r) := \begin{cases} 10 \left( \cos(r) - \frac{\sqrt{3}}{2} \right) + 2, & 0 \leq r \leq \pi/6 \\ 2, & \text{otherwise} \end{cases}$$

using  $40 \times 80$  rectangles (piecewise constant approximation)

- ▶  $\sigma_b = 2$ ,  $\mu_0 = 1$ ,  $tol_1 = 10^{-5}$ ,  $tol_2 = \frac{1}{100}$ , no noise

## Numerics - Comparison

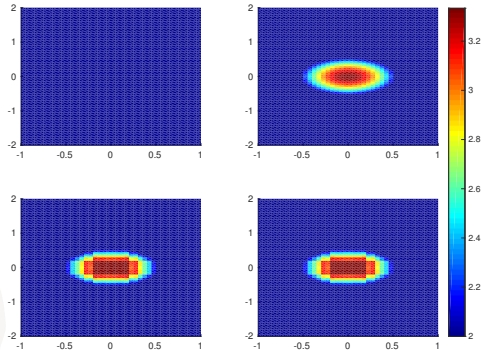


Figure:  $\sigma_b$  (top left),  $\sigma^*$  (top right),  $\sigma_{BZ}$  (bottom left),  $\sigma_{RBZ}$  (bottom right).

- ▶ **BZ:** 3.45s and 16 PDE solves | **RBZ:** 2.21s and 6 PDE solves
- ▶  $\|\sigma^* - \sigma_{BZ}\|_{\mathcal{P}} \approx 0.05$ ,  $\|\sigma_{BZ} - \sigma_{RBZ}\|_{\mathcal{P}} \approx 6.760 \cdot 10^{-6}$



- ▶ Summary:
  - ▶ Presented Harmonic  $B_z$  Algorithm to solve MREIT imaging problem
  - ▶ Extended existing convergence result to numerical applicability
  - ▶ Applied adaptive RB approach to also speed-up the algorithm
- ▶ Ongoing work:
  - ▶ Improve theoretical and numerical results
  - ▶ Publish paper

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Thank you for your attention!

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