

Modelling thin tissue compartiments using the immersed FEM (continuous Galerkin)

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Modelling thin tissue compartiments using the immersed FEM (continuous Galerkin)

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Biomag 2016

Motivations for Immersed FEM

Long term goal: Democratize EEG/MEG/sEEG FEM models. Reasons: Competitive, handles anisotropy, volumic, ...

Challenges:

- Make software widely available [e.g. SimBio, NeuroFEM since 2013].
- Interface it with standard toolboxes [see Johannes talk].
- Environment and tools to easily build FEM head models.
 - Head tissue segmentation [FSL, Free Surfer, BrainSuite, SPM, Slicer, BESA, ...]
 - Build the FEM model from segmentations.
 - Explore the solution.







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The EEG Forward Problem

Quasi-static electromagnetism



The potential V is related to the source currents $\mathbf{J}^{\mathbf{p}}$ by:

$$\nabla \cdot (\sigma \nabla V) = \nabla \cdot \mathbf{J}^{\mathsf{p}}$$
 in Ω

$$\sigma \nabla V \cdot \mathbf{n} = g \qquad \text{on} \quad \partial \Omega.$$

- Ω: head volume.
- σ : local conductivity.

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$$[V] = [\sigma \nabla V] = 0$$
 across interfaces.

The EEG Forward Problem

Quasi-static electromagnetism



Weak formulation:

$$C(V) = \frac{1}{2} \int_{\Omega} \sigma(\mathbf{r}) \|\nabla V(\mathbf{r})\|^2 d\mathbf{r} + \int_{\Omega} \nabla \cdot \mathbf{J}^{\mathsf{p}}(\mathbf{r}) V(\mathbf{r}) d\mathbf{r} - \int_{\partial \Omega} g V ds \; .$$

 \implies after FEM discretization: $[\Delta_{\sigma}]\mathbf{V} = [\nabla]^{T} \mathbf{J}^{p}$

The EEG Forward Problem

Quasi-static electromagnetism



MEG can be put in the same fraemwork (adjoint method).

EEG Forward Model (FEM) Head segmentation and mesh generation

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EEG Forward Model (FEM) Head segmentation and mesh generation



- Very thin structure:
 - Esp. for infants.
 - Very resoluted meshes.
 - Derefinement.
- Topology changes.
 - Self intersections.
 - Holes in the model.
 - Complicate head modelling.
- Lots of work.
- Huge meshes.
- Computational quality.

EEG Forward Model (FEM) Potentially much more complex geometries...



- Complex skull geometry.
- Blood vessels.

From surfacic meshes to matrices $[\Delta_{\sigma}]$

- Volumic meshing between surfaces.
- Merge volumic meshes.
- Matrix computation from base functions.



- For each subject.
- Size/accuracy tradeoff.
- Image data ignored during simplication.

Solution: Compute matrices directly from segmentations.

Direct Use of Segmentations Level sets representation of surfaces



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Direct Use of Segmentations Level sets representation of surfaces

Most curve (2D) or surface (3D) manipulations can be expressed as levelset manipulations. Segmentation.



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Direct Use of Segmentations Direct computation of $[\Delta_{\sigma}]$ from levelsets

The MRI image space is the computation reference space.

- Use the "image mesh". Trilinear interpolation method (Q1).
 ⇒ The solution is an image of potential.
- Interfaces defined by implicit functions (levelsets).
 ⇒ Computation of partial integrals wrt interfaces.



Implicit boundary:

 $f(x, y) = c_{11}xy + c_{10}x + c_{01}y + c_{00} = 0$

Sum green and red contributions

 \Rightarrow Closed form formula in 2D (integral of a rational fraction).

Direct Use of Segmentations 2D/3D algorithms

2D: • Apply transform (rotations, symmetries, sign inversion) to reduce to:



- Compute the explicit 2D formulae.
- Correct for the transform action.
- 3D: Numerical integration of 2D values:



2D topological changes.

Direct Use of Segmentations 2D/3D algorithms

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2D topological changes.



Multiple interfaces per voxel can be handled.

Non-differentiable Elements

- V and $\sigma \nabla V$ continuous over Ω . . .
- \ldots but σ is discontinuous across interface boundaries
- $\Longrightarrow \nabla V$ discontinuous: not representable with Q1 elements.
- \implies Use non-differential continuous elements (piecewise Q1):

$$\begin{cases} \phi(x, y, z) = p_1(x, y, z) & \text{in } C_1, \\ \phi(x, y, z) = p_2(x, y, z) & \text{in } C_2, \\ p_1 - p_2 = \lambda l, \quad \lambda \in \mathbb{R} & \int_{l=0} \sigma_1 \nabla p_1 \cdot \mathbf{n} - \sigma_2 \nabla p_2 \cdot \mathbf{n} = 0 \\ \\ I \\ C_1 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_1 = 10 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_1 = 10 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_1 = 10 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_1 = 10 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_2 = 1 & \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_1 = 1 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_1 = 1 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_1 = 1 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_1 = 1 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_1 = 1 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_1 = 1 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_1 = 1 \\ \sigma_2 = 1 & \sigma_1 = 10 \\ \sigma_2 = 1 & \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 & \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 & \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 \\ \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 \\ \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 \\ \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 \\ \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 \\ \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 \\ \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_1 = 1 \\ \sigma_2 = 1 \\ \sigma_2 = 1 \\ \sigma_1 = 1 \\ \sigma_2$$

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Results: Spherical Model



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Results: Spherical Model



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Results: Real Data

MRI image of size $256 \times 256 \times 160$

5 conductivity domains segmented using four levelsets.

Rescaled to $128\times128\times80$

 \Longrightarrow many thin structures and double interface cells are present !



Immersed FEM:

- Use MRI as the computational space.
- Direct connexion Q1 Levelsets \longleftrightarrow FEM matrices.
- Simple, fast and unsurprising: a real use case advantage.
- Thin structure easy to deal with.
- Accuracy similar to standard FEM.
- Easy visualization and exploration of the solution.
- Multiscale and parallelisation are easy.
- Control of boundary condition of solution more complex.

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Thank you !!

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Direct Use of Segmentations (2)

The MRI image space is the computation reference space.

- Use the "image mesh". Trilinear interpolation method (Q1).
- Interfaces defined by implicit functions (levelsets).
- \implies Computation of partial integrals wrt interfaces.

Implicit boundary: $f(x, y) = c_{11}xy + c_{10}x + c_{01}y + c_{00} = 0$. Green contribution:

$$\int \int_{Green} m(x, y) dy dx = \int_{0}^{x_{0}} \int_{0}^{-\frac{c_{10}x + c_{00}}{c_{11}x + c_{01}}} m(x, y) dy dx$$

$$= \int_{0}^{x_{0}} M\left(x, -\frac{c_{10}x + c_{00}}{c_{11}x + c_{01}}\right) dx .$$

- \implies Analytic formula (integral of a rational fraction).
 - Special cases (polynomial expressions): $c_{11}c_{00} c_{10}c_{01} = 0$, $c_{11} \simeq 0$ (taylor expansion).

Comparison with FD



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More Comparisons with P1 Implementation



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