Multi-Cluster, Mixed-Mode Computational Modeling of Human Head Conductivity

 Adnan Salman¹, Sergei Turovets¹, Allen Malony¹, and Vasily Volkov
 ¹ NeuroInformatics Center, University of Oregon
 ²Institute of Mathematics, Minsk, Belarus

Collaboration

- NeuroInformatics Center, University of Oregon:
- Robert Frank
- Electrical Geodesic, Inc :
- Peter Lovely, Colin Davey, Pieter Poolman, Jeff Eriksen, and Don Tucker

Motivation

• Goal: To estimate the electrical conductivities of human head based on realistic segmented MRI or CT scans

Necessary for ...

- Source Localization: find the electrical source generator for the potential that can be measured at the scalp
- Detecting abnormalities: cracks, holes, ... etc

Building Computational Head Models

To relate the neural activity in the head to the EEG measurements on the scalp

- Three parts in constructing a human head model
 - 1. Geometry: Geometrical Model of the head with its tissue types
 - Sphere models: 4-sphere model, 3-sphere model
 - MRI or CT: determines the boundaries of the major head tissues
 - 2. Electrical Conductivity model: Assign a conductivity value for each tissue type
 - ➔ Homogenous: Assign an average value for the entire MRI segment
 - ➔ Known: For each tissue type it varies considerably
 - 3. Forward problem: Evolution of the potential within each tissue.

Given the conductivities of the head tissue and the current sources, find the potential at each point in the head.













Mesh **Solution**

MRI

The governing equation is:

• The Poisson equation

 $\nabla \bullet (\sigma \nabla \phi) = \nabla \bullet J^s$, in Ω

• With the boundary condition $\sigma(\nabla\phi) \bullet n = 0$, on Γ_{Ω} .

Where, $\sigma = \sigma_{ij}(x,y,z)$ is a tensor of the head tissues conductivity, J^s , current source.

Multi-component ADI Method:

- unconditionally stable in 3D
- accurate to $O(\tau + \Delta x^2 + \Delta y^2 + \Delta z^2)$

$$\frac{\phi_i^{n+1} - \overline{\phi}^n}{\tau} + \delta_x \phi_i^{n+1} + \delta_y \phi_j^n + \delta_z \phi_k^n = S$$

$$\frac{\phi_j^{n+1} - \overline{\phi}^n}{\tau} + \delta_x \phi_i^{n+1} + \delta_y \phi_j^{n+1} + \delta_z \phi_k^n = S$$

$$\frac{\phi_k^{n+1} - \overline{\phi}^n}{\tau} + \delta_x \phi_i^{n+1} + \delta_y \phi_j^{n+1} + \delta_z \phi_k^{n+1} = S$$

Here : $\phi^n = (\phi_i^n + \phi_j^n + \phi_k^n)/3$ $\delta_{x,y,z}$ is notation for an 1D second order spatial difference operator

Reference: Abrashin et al, Differential Equations 37 (2001) 867

Multi-component ADI algorithm:

- Each time step is split into 3 substeps
- In each substep we solve a 1D tridiagonal systems



Computational Head Models: Forward problem: solution



External Current Injection (Electrical Impedance Tomography) Intracranial Dipole Source Field (Epileptic Source Localization)

Computational Head Models: Forward problem: Validation



Electrode Number

Computational Head Models: Forward problem: Parallelization

- The computation to solve the system of equations in each substep is independent of each other
- Example: in the x direction we can solve the $N_y N_z$ equations concurrently on different processors
- The Parallel program structure is:
 - For each time step
 - Solve $N_y N_z$ tridiagonal equations
 - Solve $N_x N_y$ tridiagonal equations
 - Solve $N_y N_z$ tridiagonal equations

End

• We used openMP to implement the parallel code in a shared memory clusters



Computational Head Models: Forward problem: Parallelization speedup

Forward Solution Speedup on IBM-P690



Computational Head Models: Inverse Problem

Given the measured electric potential at the scalp V_i, the current sources and the head tissue geometry
 Estimate the conductivities of the head tissues

The procedure to estimate the tissue conductivities is:

- Small currents are injected between electrode pairs
- Resulting potential measured at remaining electrodes
- Find the conductivities that produce the best fit to measurements by minimizing the cost function:

$$E = \left[\frac{1}{N}\sum_{i=1}^{N} (\phi_{i}^{p} - V_{i})^{2}\right]^{1/2}$$

• Computationally intensive



Schematic view of the parallel computational system



Performance Statistics

Dynamics of Inverse Search



Performance Statistics

Dynamics of Inverse Search



optix

Inverse Problem: Simplex Algorithm simulated data (real MRI)

1/2

 $\Delta\sigma(\Omega^{-1}m^{-1})$

0.0099

0.0311

0.00017

0.00024

 $\sigma(\Omega^{-1}m^{-1})$

0.2491

1.7933

0.0180

0.4400



Exact Values

Inverse Problem: Simplex Algorithm simulated data (real MRI)



Summary

- Finite Difference ADI algorithm based 3D solvers for the forward electrical have been developed and tested for variety of geometries;
- The electrical forward solver has been optimized and parallelized within OpenMP protocol of multi-threaded, shared memory parallelism to run on different clusters;
- The successful demonstrations of solving the nonlinear inverse problem with use of HPC for search and estimation of the unknown head tissues conductivity have been made for 4tissues segmentation on the realistic MRI based geometry (128^3 resolution) of the human head;
- > The work with experimental human data is in progress

Thank you

Questions