

Forward and Inverse EEG Source Modeling

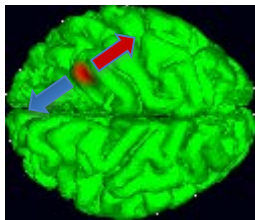


Scott Makeig

Institute for Neural Computation, UCSD

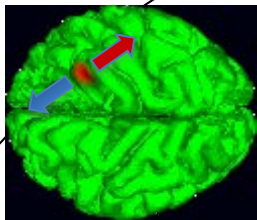
EEGLAB Tokyo

September, 2017



Motivation

- Why perform ICA?
- Why fit dipoles or distribution source models?
- Why measure EEG?!
- To obtain information about brain processes...
 - Time course of activities that produce the EEG signals
 - Locations of the activities that produce the EEG signals



Local
Synchrony
**Equivalent
Current Dipole**

Cortex

Electrodes

Skin

Relative
Independence

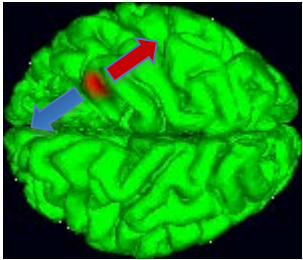
Local
Synchrony

Skull

~~Spatial EEG Source Filtering~~

EEG source modeling

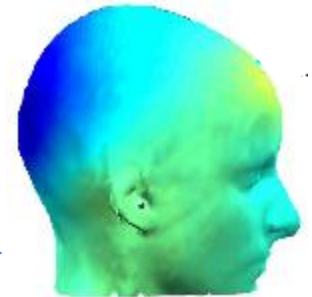
Source
Space



Forward head model

forward problem

Sensor
Space



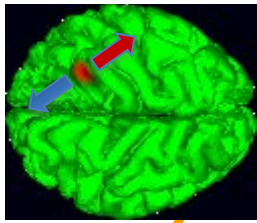
electrical
currents

volume conduction
through body tissues

recorded
potentials

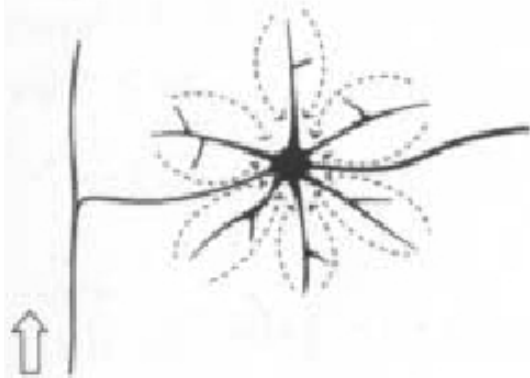
inverse problem

Inverse localization
method

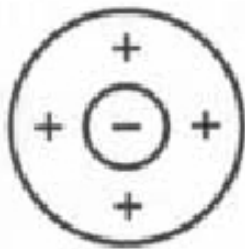


Peri-neuronal currents

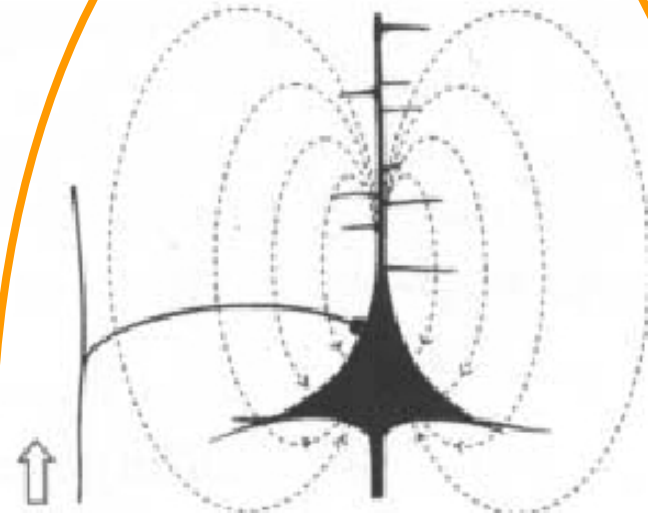
Stellate cell



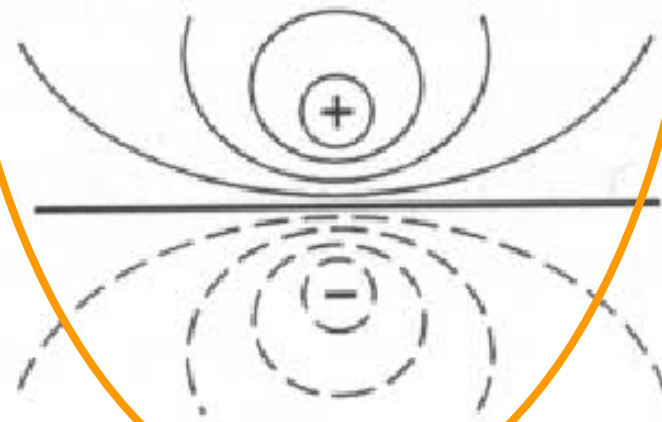
Closed field

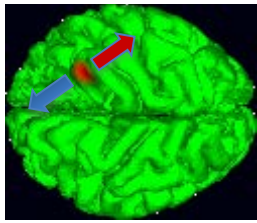


Pyramidal cell



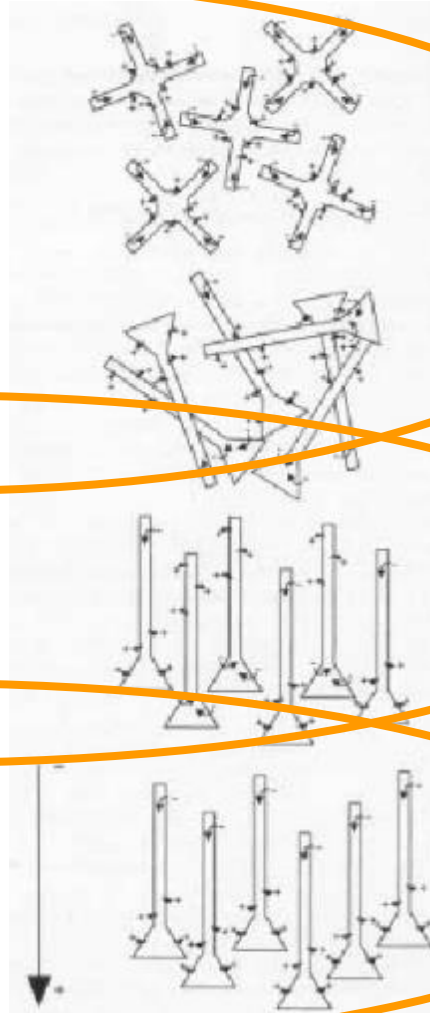
Open field





Symmetry, orientation and activation

radially symmetric, i.e.
randomly-oriented



Closed field

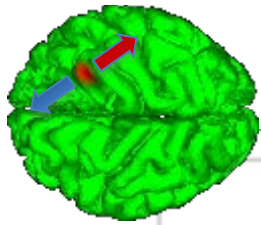
asynchronously activated

Phase
cancellation

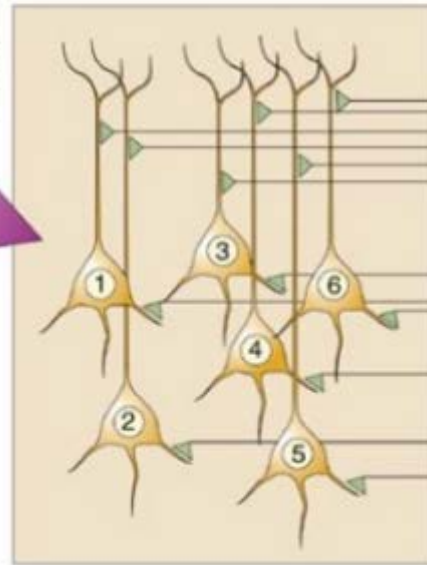
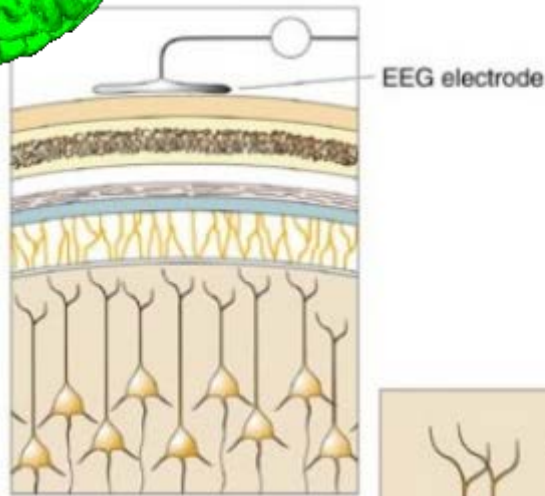
synchronously activated
parallel-oriented

Open field

When recorded at a distance, **dipolar** field components dominate

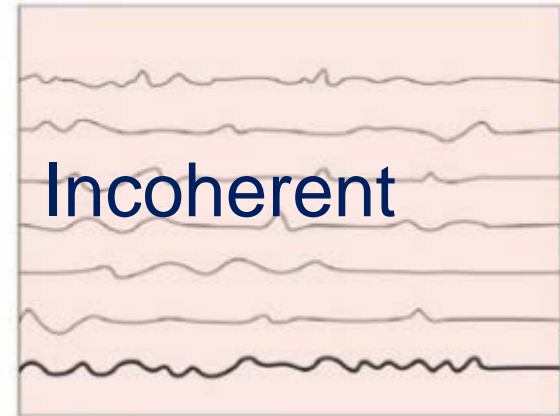


EEG Effective Sources



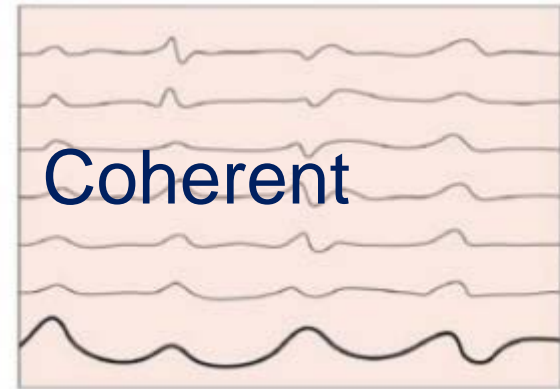
(a)

Irregular



(b)

Synchronized

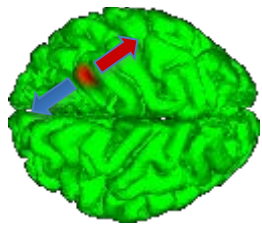


(c)

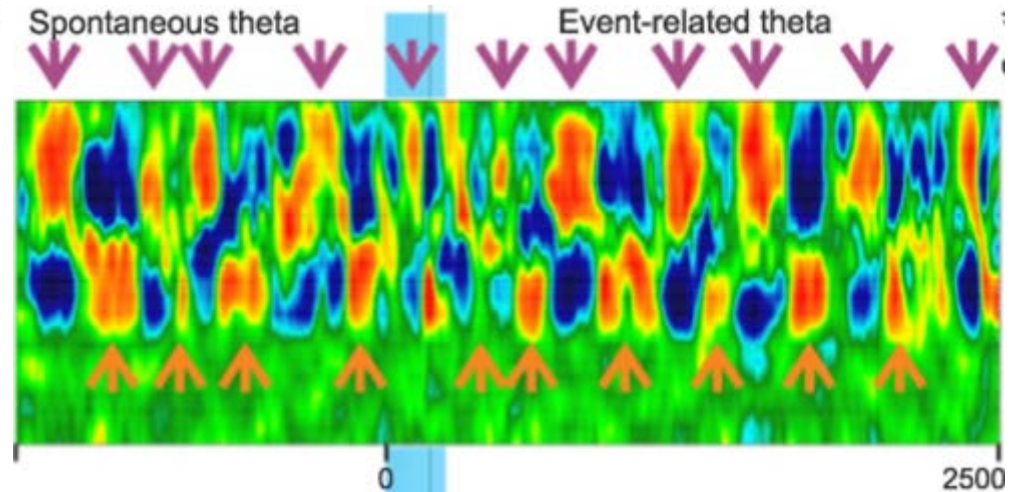
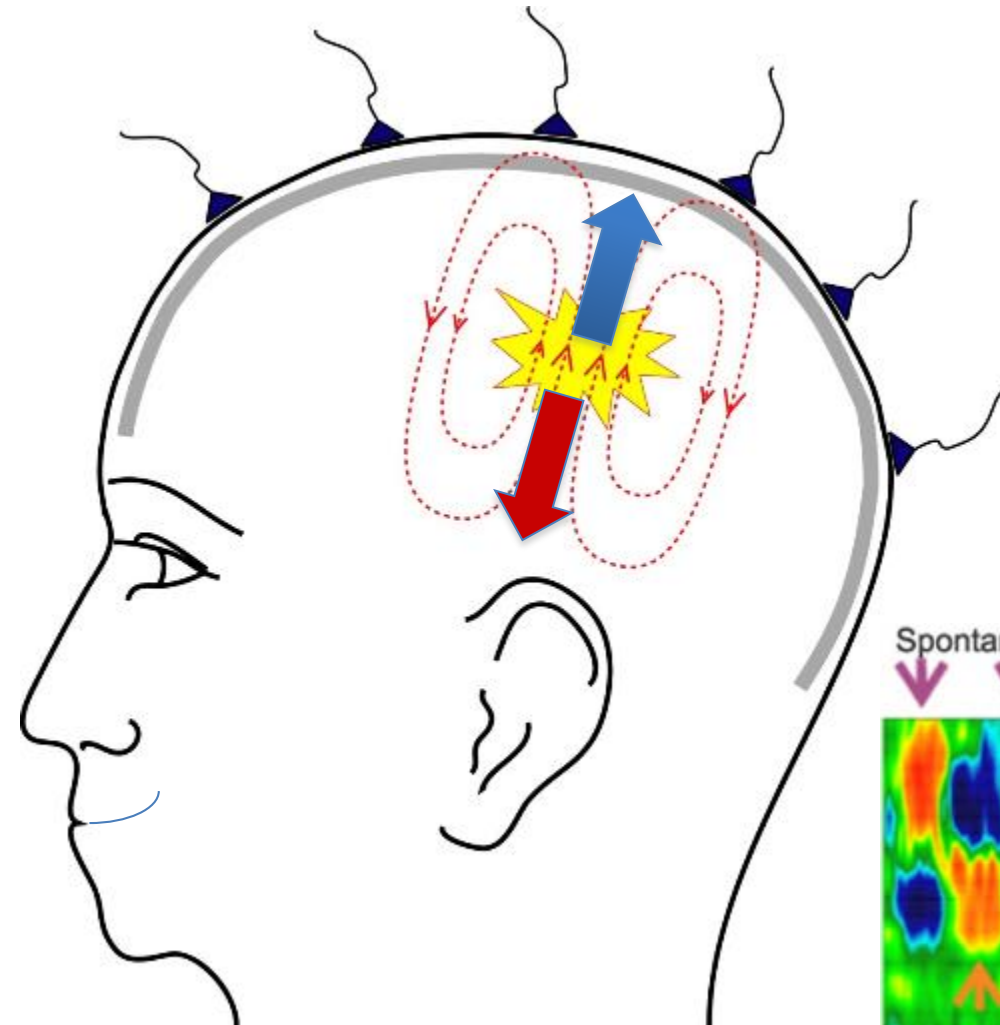
Incoherent

Coherent

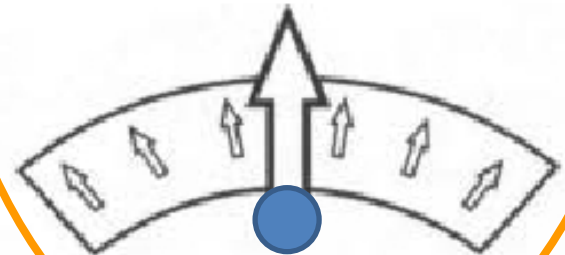
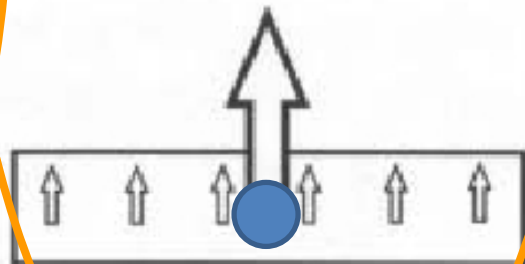
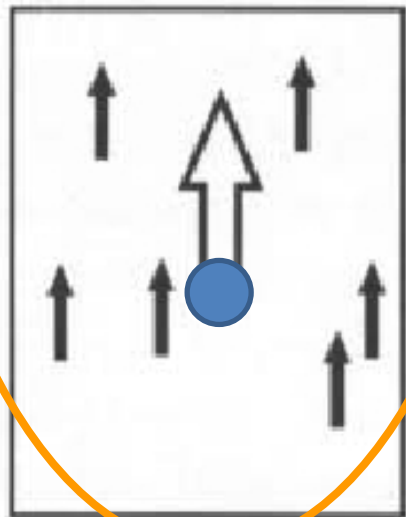
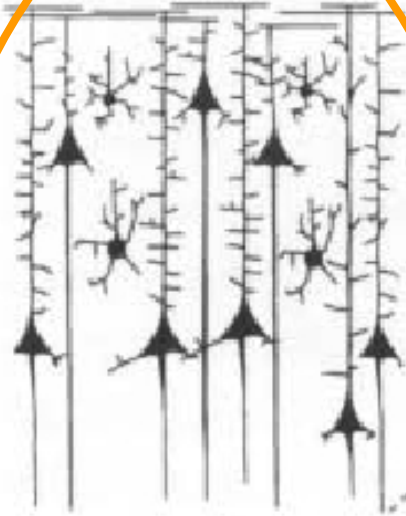
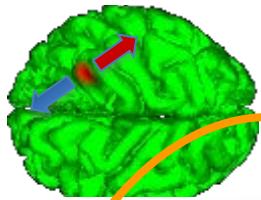
Many neurons need to sum their local field activities to be detectable at EEG electrodes. Synchronized neural activity produces large far field signals.

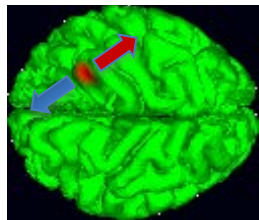


EEG volume conduction of dipolar field patterns → effective sources

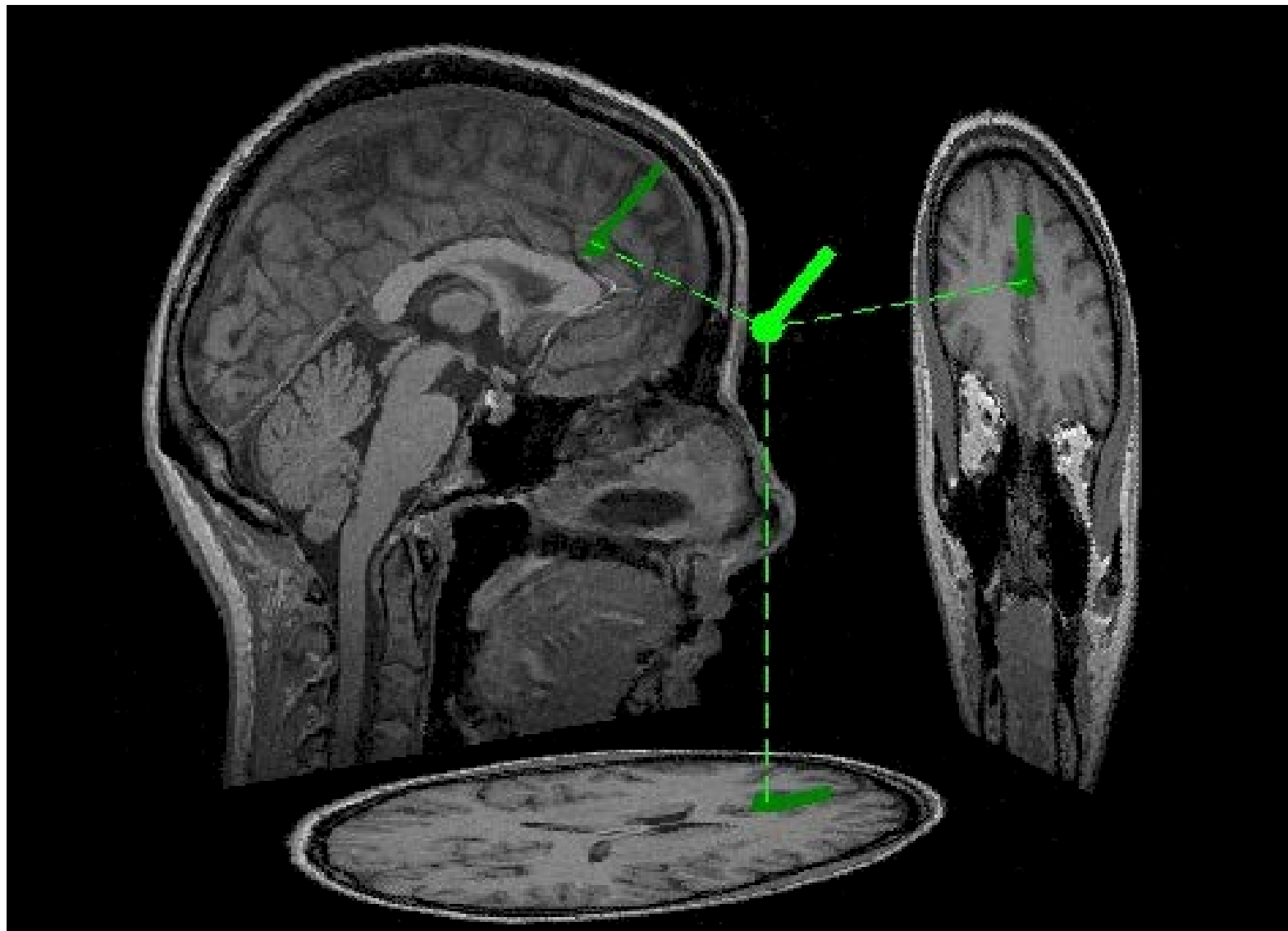


The *equivalent* current dipole



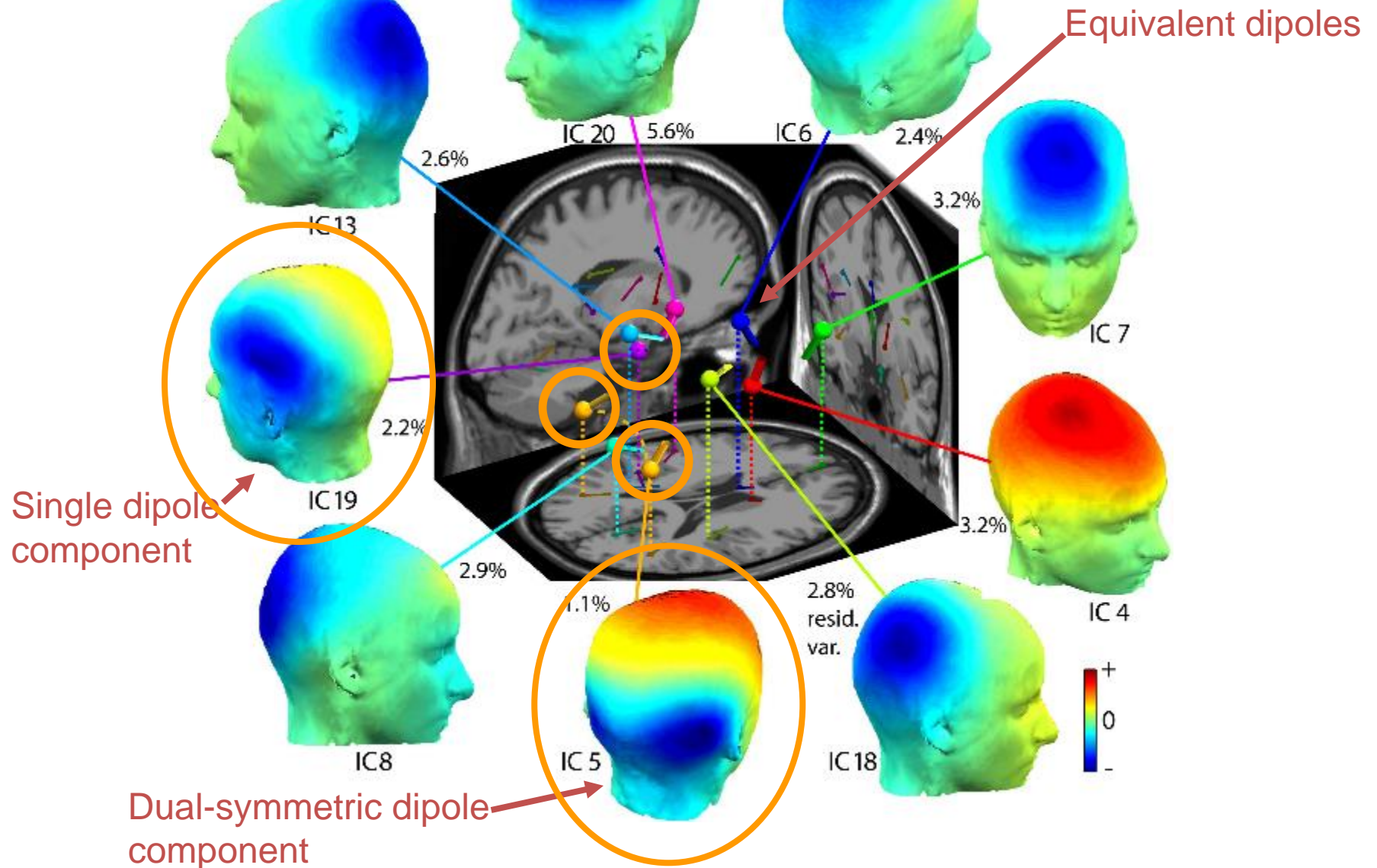


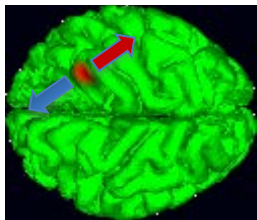
Equivalent current dipole modeling



1st IC source fit in an individual head model via EEGLAB

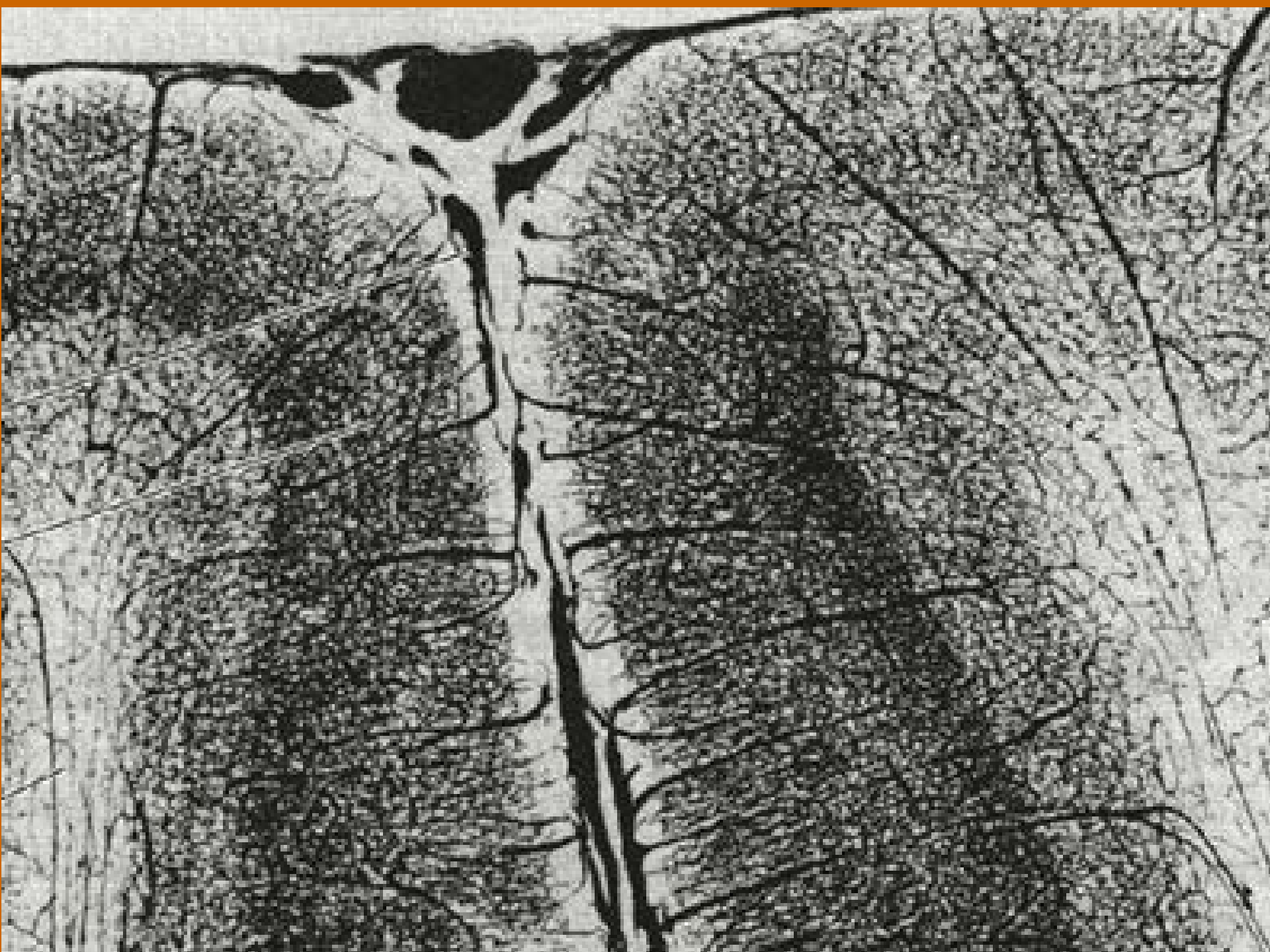
A. Delorme, ~2007

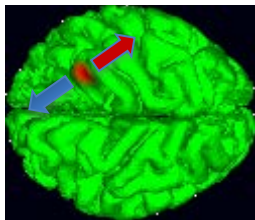




Equivalent current dipole modeling

- **Physical/mathematical motivation**
 - Any current distribution can be written as a multipole expansion
 - First term: monopole (must be 0)
 - Second term: dipole
 - Higher order terms: quadrupole, octopole, ...
 - **In far-field recordings, the dipolar term dominates.**
- For convenience + accuracy, therefore
 - **Dipoles** can be used as building blocks in distributed EEG effective source models

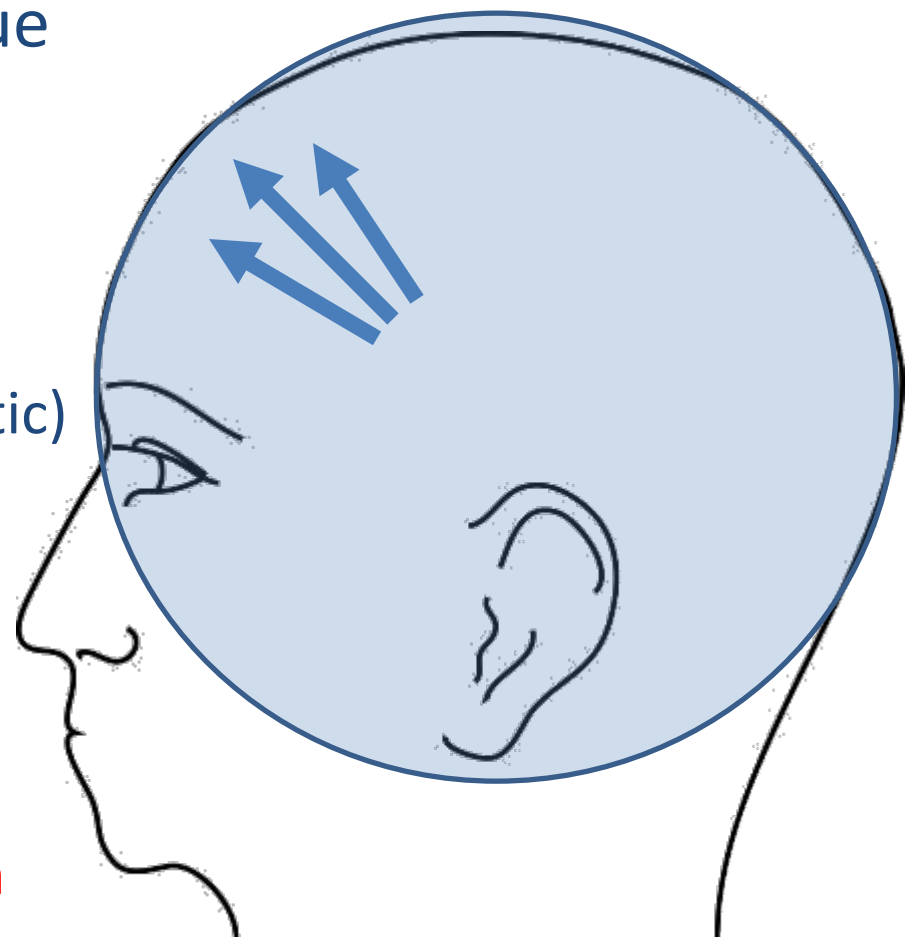


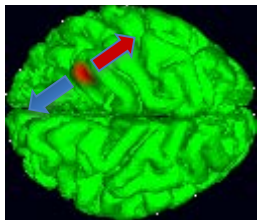


Forward Head Models

- Electrical properties of tissue
 - Conductivity
 - Anisotropy
- Geometrical description
 - Spherical model? (less realistic)
 - Realistically shaped model

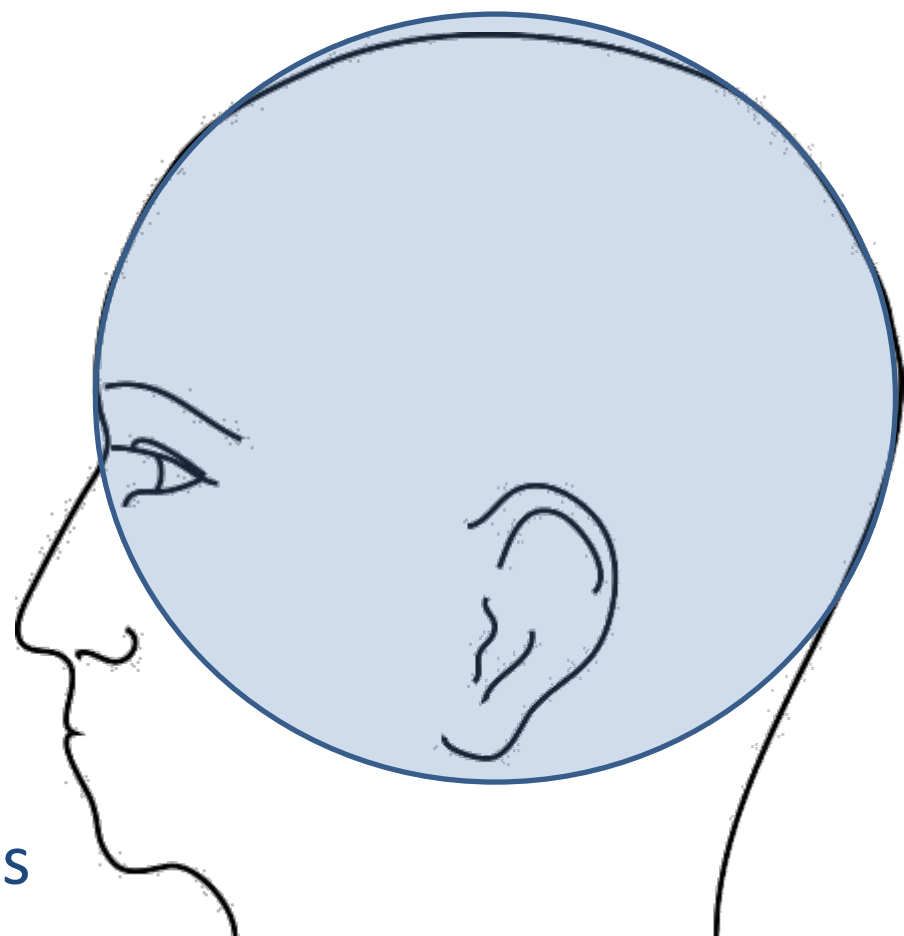
→ A **forward model** describes
how the currents flow
from all possible points of origin

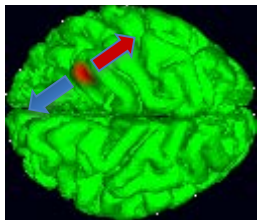




Forward Head Models

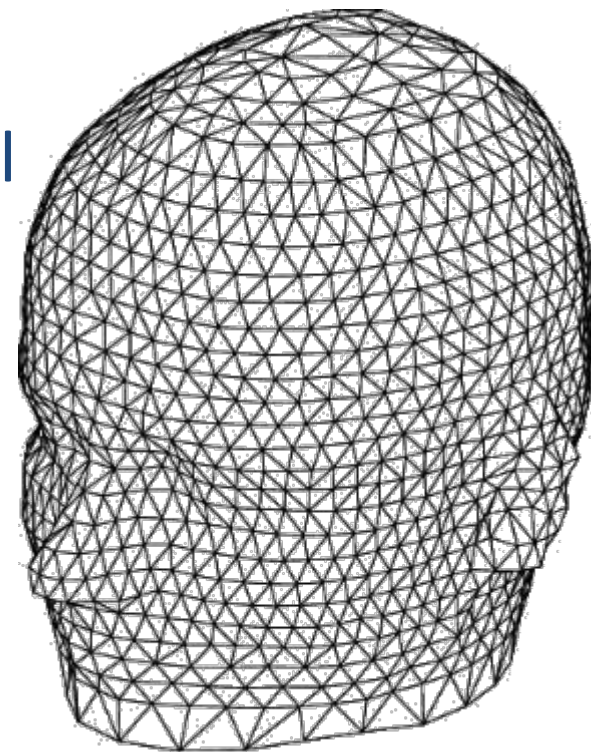
- Advantages of the **spherical** model
 - mathematically accurate
 - reasonably accurate
 - computationally fast
 - easy to use
- Disadvantages of the **spherical** model
 - inaccurate in some regions
 - difficult to align to head



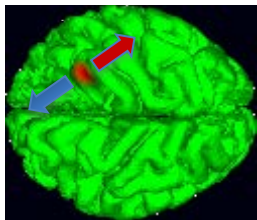


Forward Head Models

- Advantages of a **realistic** head model
 - accurate solution for EEG
- Disadvantages of a **realistic** model
 - more work
 - computationally slower
 - numerically instable?
 - Difficult for inter-individual comparisons

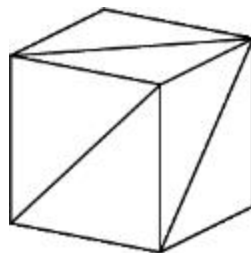
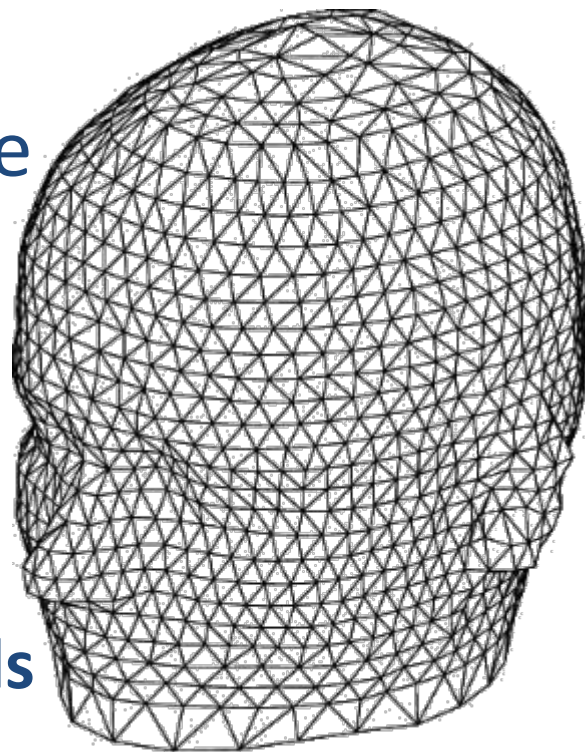


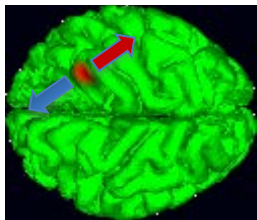
→ The pragmatic (easy, cheap) solution is to use a standard (mean) realistic head model (MNI).



Forward Head Models

- Computational methods for volume conduction problem that allow realistic geometries
 - **Boundary Element Method (BEM) models**
 - **Finite Element Method (FEM) models**
- Geometrical description
 - **Triangles (2-D) \rightarrow BEM**
 - **Tetrahedra (3-D) \rightarrow FEM**



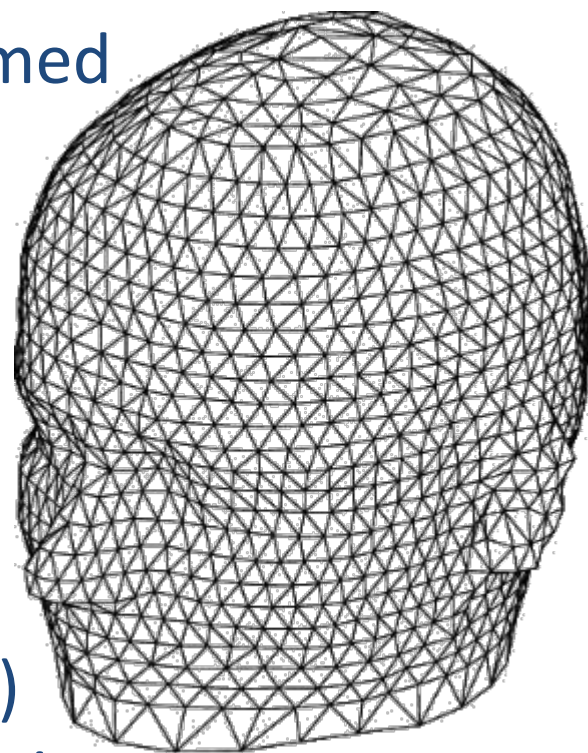


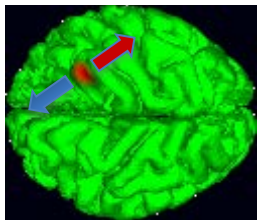
Forward Head Models: BEM

- **Boundary Element Method (BEM)** models
 - description of head geometry by tissue compartments
 - Tissue in each compartment is assumed
 - homogenous
 - isotropic

Important tissue types

- Scalp
- Skull
- CSF
- Brain (grey matter / white matter)
- Use triangulated surfaces as boundaries
- Each surface should be closed (no holes)





Forward Head Models: BEM

- **Boundary Elements**

- descriptive of the geometry of the head
- comparison of the forward and inverse problem

- Tissue properties

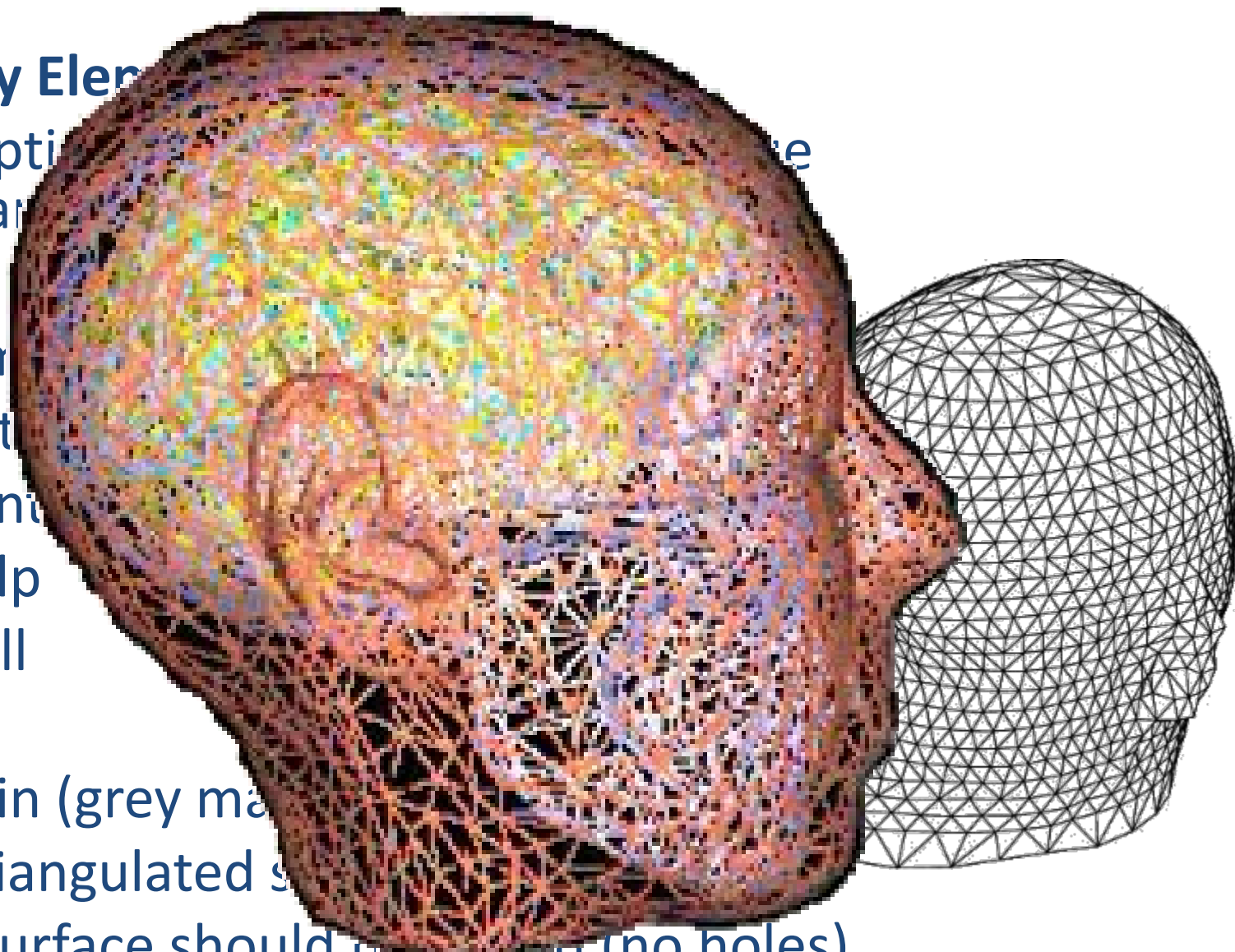
- homogeneous
- isotropic

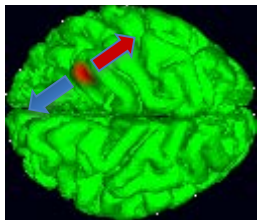
- Important layers

- Scalp
- Skull
- CSF
- Brain (grey matter)

- Use triangulated surfaces

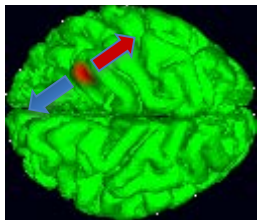
- Each surface should be closed (no holes)





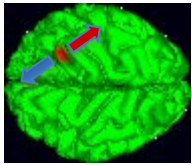
Forward head models: Modeling the skull

- **Potential differences between electrodes** measures summed current flowing through scalp
 - However, only a tiny fraction of *brain source currents* pass through the skull
 - Therefore a forward head model should describe *brain, skull, and scalp tissues* as accurately as possible.



Forward head models: Modeling the skull

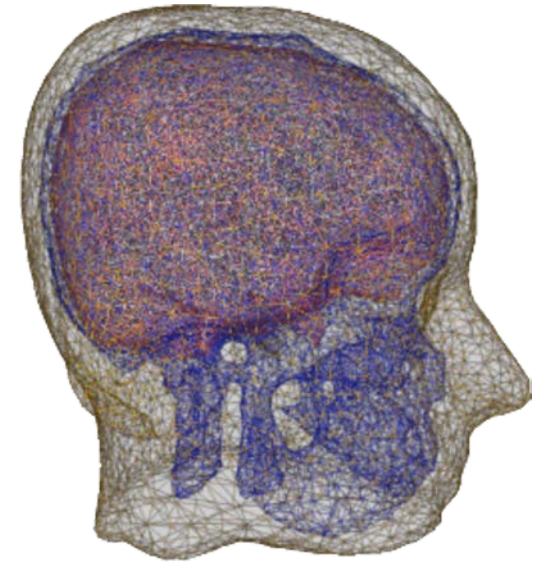
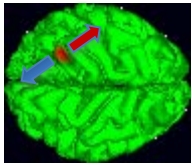
- Problems with skull modeling
 - Poorly visible in the anatomic MRI (T2) image
 - Thickness varies regionally
 - Conductivity is not homogeneous
 - Complex geometry at front and base of skull
- **Skull conductivity** variable & unknown



Volume conductor: FEM

To make a Finite Element Method (FEM) head model:

- **Tessellate the 3-D volume into solid tetrahedra**
 - Contains a large number of 3-D elements
 - Each tetrahedron can have its own conductivity
 - Each tetrahedron can have its own *anisotropy*
(direction-dependent conductivity differences)
- **FEM is the more complete numerical method (> BEM)**
 - But is computationally expensive
 - Note: Accurate conductivities are not known, particularly for skull (and scalp?).



Head Modeling Errors

Electrode & MRI Co-registration errors

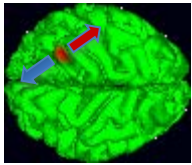
Head Geometry Errors

EXCLUSION of white matter

Too Few electrodes

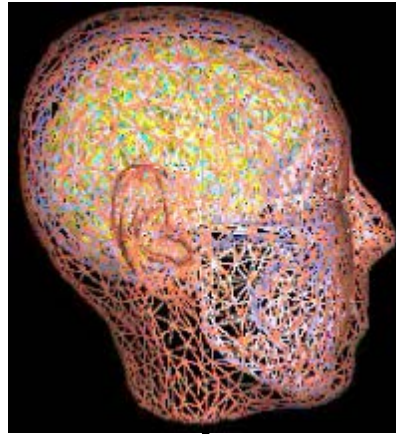
Poor distribution of electrodes

→ mis-estimation of skull conductivity

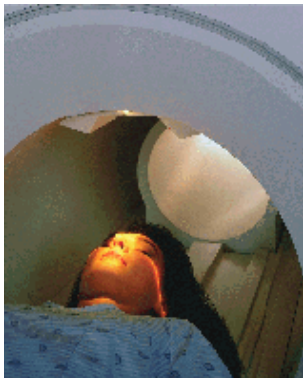


Electromagnetic source localization using realistic head models (Dipfit, NFT)

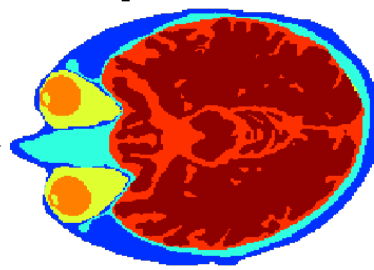
Solve the forward problem using realistic head models (BEM)



Mesh generation

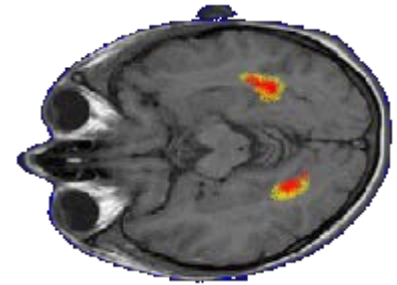


MRI



Segmentation

Simple Map



Source Image

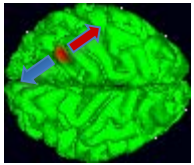
Sensor Localization

Signal Processing

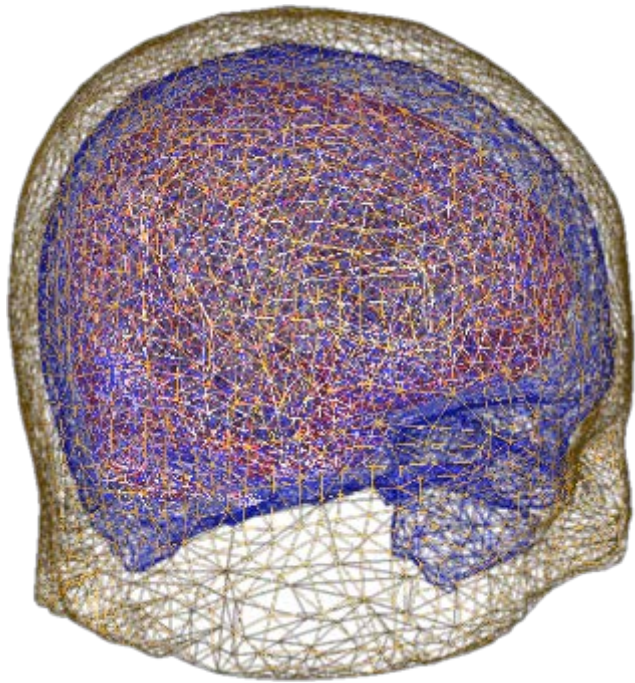


EEG/MEG

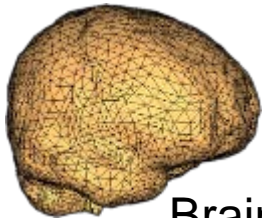




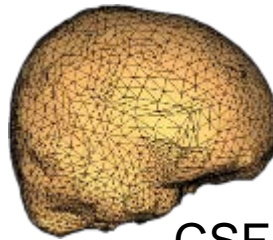
The MNI Head Model



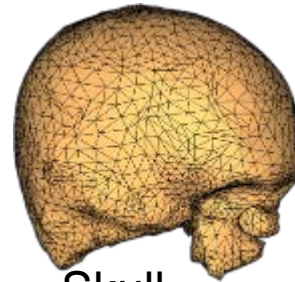
- 4-layer
 - 16856 nodes
 - 33696 elements
- 3-layer
 - 12730 nodes
 - 25448 elements



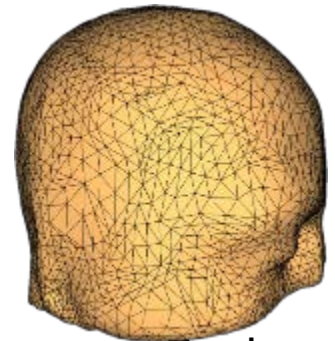
Brain



CSF

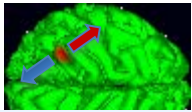


Skull



Scalp

NFT



Subject Folder

Subject Name

Session Name

Head Modeling

From a magnetic
Resonance Image

From electrode
Position Data

Image Segmentation

Mesh Generation

Source Space Generation

Electrode Co-Registrati...

Template Warping

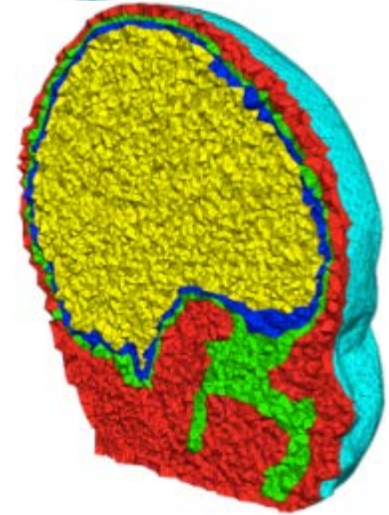
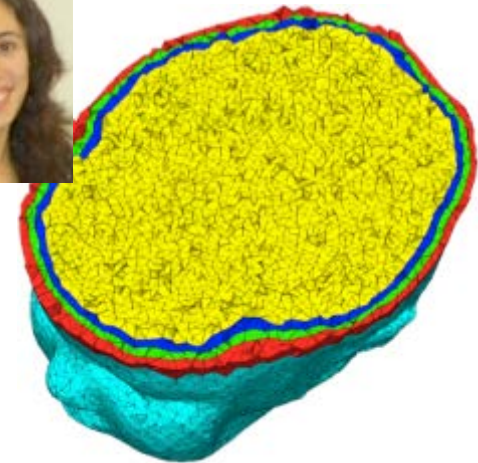
FP Solution with BEM

FP Solution with FEM

Dipole Fitting

adult

6-month old

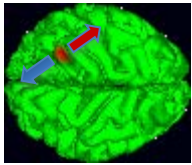


FEM models

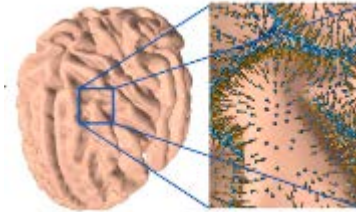
BEM
models



Akalin Acar & Makeig, 2010

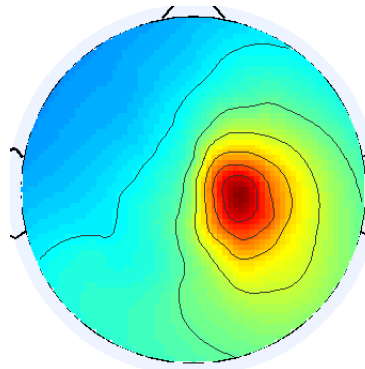


NIST

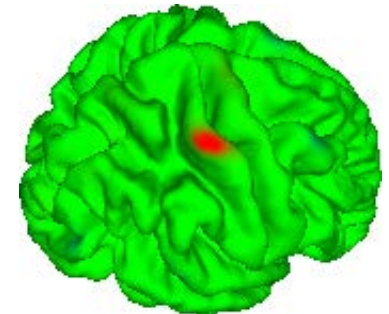


Source space

Scalp map

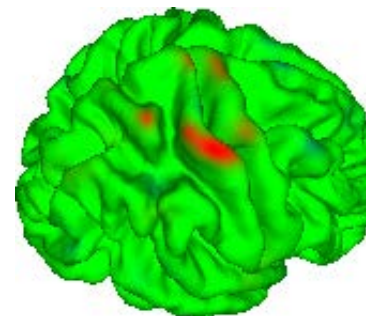


SCS

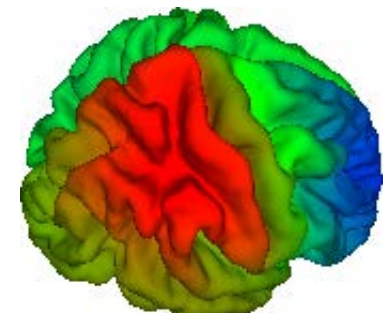


Cheng Cao, 2012

Patch-based SBL



sLORETA



Distributed_Source_Localization

Load MRI

/data/cta/zeynep/NFTdene/t3test001.img

Start Freesurfer

Running Freesurfer for cortical segmentation...

Cortical source space

80000

of dipoles in source space

10 mm

Generate patches

FP Solution with BEM

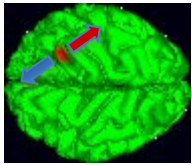
FP Solution with FEM

Component indices

Select Source Localization Method

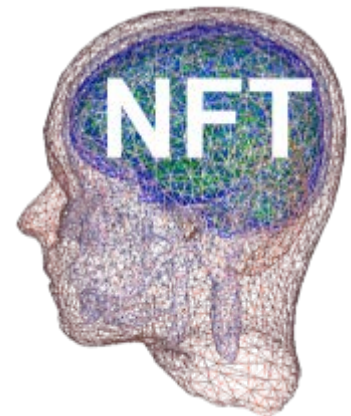
Start Source Localization

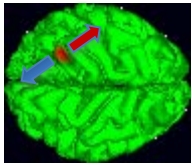
Visualization



Head Model Generation Summary

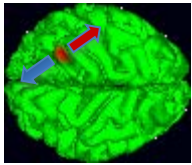
- **Subject-specific Head Model (NFT)**
 - From whole head T1 weighted MR of the subject
 - 4-layer realistic BEM model
- **MNI Template Head model (DIPFIT)**
 - From the MNI head
 - 3-layer and 4-layer template BEM model
- **Warped MNI Template Head Model (NFT)**
 - Warp MNI template to EEG sensors
- **Spherical Head model (no longer in use)**
 - 3-layer concentric spheres
 - Fitted to EEG sensor locations
 - Not accurate





Inverse source localization

- **Single and multiple dipole models**
 - Minimize error between the model and the measured potential/field
- **Distributed dipole models**
 - Seek perfect fit to the measured potential or field
 - Must minimize **some additional source constraint**
 - LORETA assumes a smooth source current distribution
 - Minimum Norm (L2), min. total cortical $|\text{current}|^2$
 - Minimum Current (L1) min. total cortical $|\text{current}|$
 - Note: L2/L1 need some weighting scheme to keep source models from being too broad & superficial.



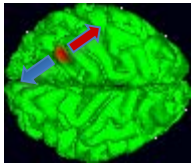
Inverse methods

Spatial filtering approaches

- **Scan whole brain** with single dipole and compute the filter output at every location (using sensor covariance)
 - MUSIC
 - *Beamforming* (e.g. LCMV, SAM, DICS)
- **Perform ICA decomposition** (higher-order statistics) on the continuous data.
 - ICA gives the projections of the sources to the scalp surface → **‘simple’ maps!**

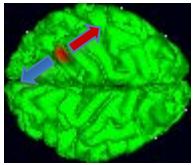
→ ICA solves ‘the first half’ of the inverse problem ‘What?’

→ ICA gives ‘simple’ source maps, helping to locate ‘Where?’



Single or multiple dipole models

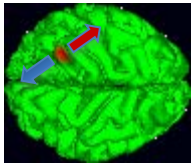
- Manipulate source parameters to **minimize error** between measured and model data
 - The **position** of each source
 - The **orientation** of each source
 - The **strength (magnitude)** of each source
- **Dipole orientation** and **strength** together correspond to the “**dipole moment**,” estimated linearly
- **Dipole position** is estimated non-linearly by source parameter estimation



DIPFIT: Dipole fitting 1. Grid search

1. Coarse fit step

- Define a grid with possible dipole locations
- Compute optimal dipole moment at each location
- Compute value of goal-function (fit to given map)
- Plot value of goal-function on the grid → find best fit.
- Number of evaluations:
 - single dipole, 1 cm grid: ~4,000
 - single dipole, ½ cm grid: ~32,000
 - BUT two dipoles, 1 cm grid: ~16,000,000



DIPFIT: Dipole fitting 2. Nonlinear search

2. Fine fit step

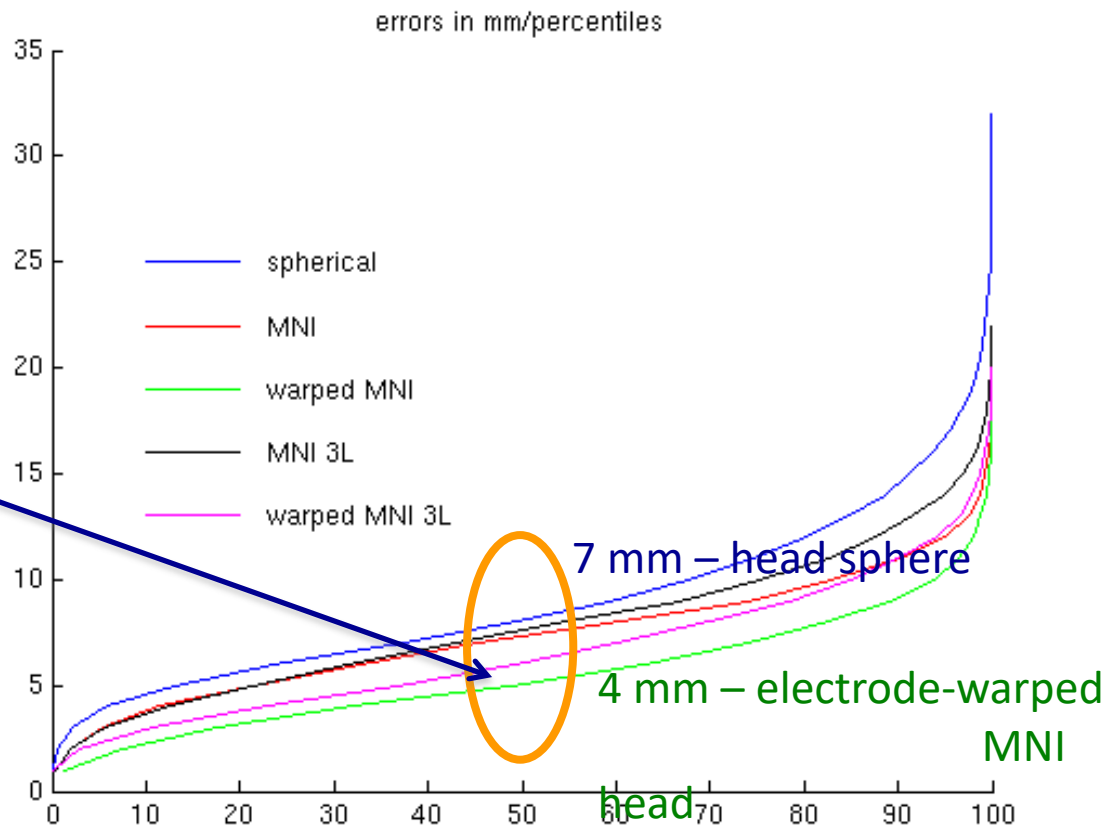
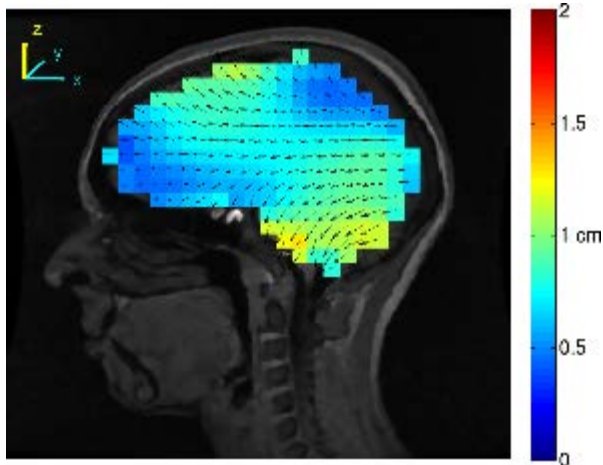
Start with the initial guess from coarse fitting

- Evaluate the local derivative of the goal (fit) function
- Then “walk down hill” to the most optimal solution

Number of iterative steps required = ~ 100

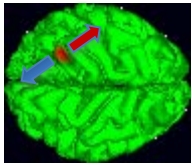
Effect of Template Head Model Choice On Estimated Dipole Locations

By Simulation: The median geometric error in dipole localization using the MNI template head model warped to measured electrode positions is only 4 mm.



BUT Additional dipole error contributors:

- Electrode co-registration error
- ICA numerical error (not enough data?)
- Source model geometry error
- **Conductance value error (skull) !**

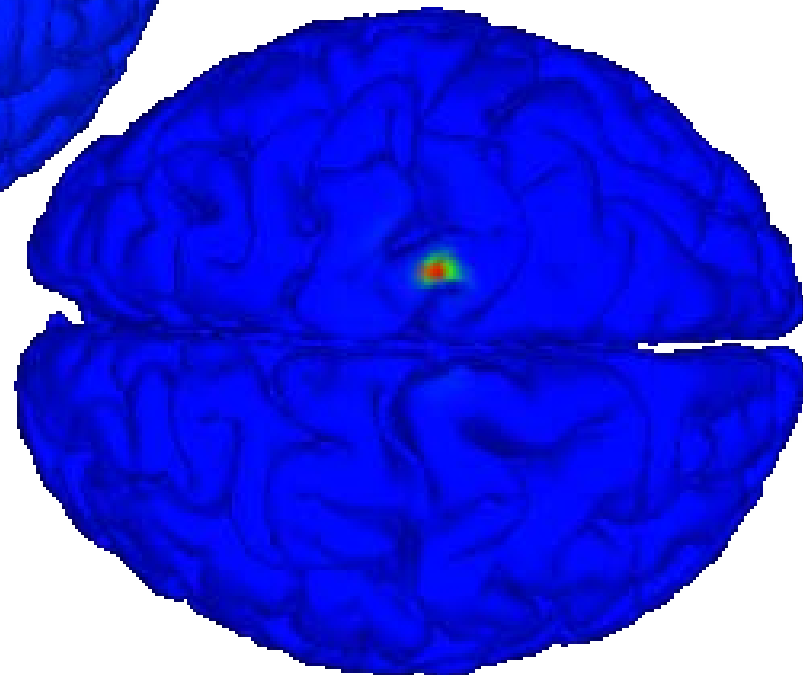
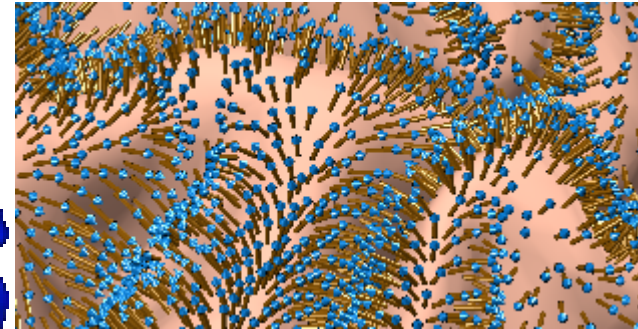
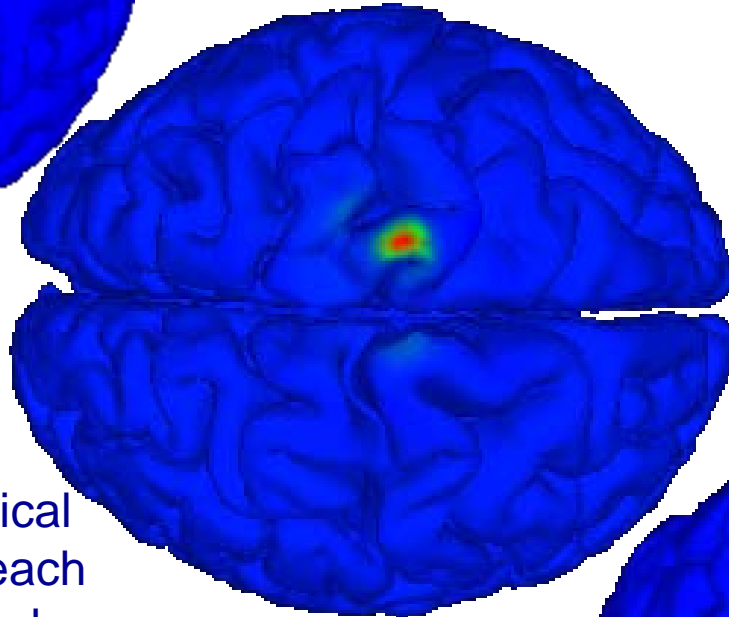
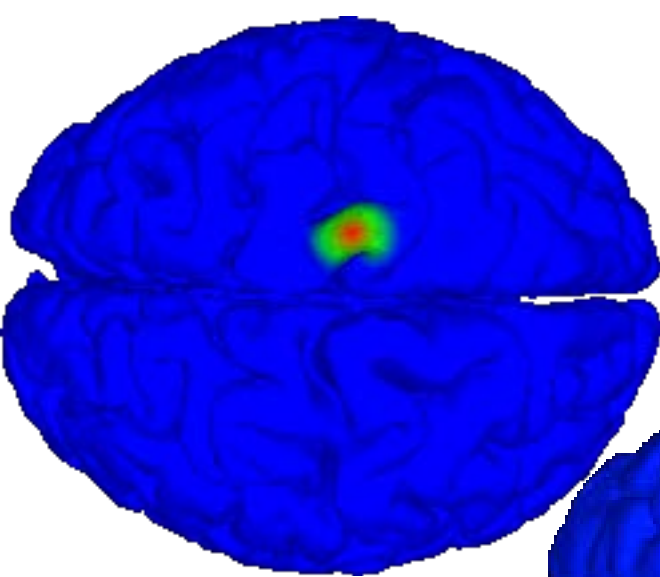


Distributed source models

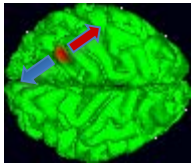
- The position of the source is not estimated as a whole
- Instead, On a pre-defined *source space* grid (3-D volume or cortical 2-D sheet)
 - Dipole strength is estimated *at each grid element*
 - In principle, a linear problem, easy to solve, BUT...
 - More “unknowns” (parameters) than “knowns” (channels, measurements), so ...
 - An infinite number of solutions can explain the data perfectly (not necessarily physiologically plausible!)
 - **Therefore**, additional source constraints are required ...

High-Resolution Distributed Source Localization

using a multiscale patch basis

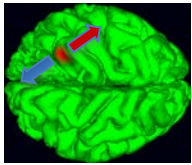


0. Build a high-res. cortical surface mesh; give each voxel an oriented dipole.
1. Compute a 'dictionary' of Gaussian patches conforming to the cortical surface centered at each cortical mesh voxel.
2. Use a 'sparsifying' approach to find the sum of the *fewest* of these patches that together produce the given source scalp or grid map.



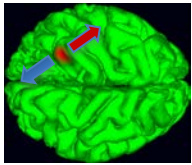
Summary

- An electromagnetic **forward head model** is required to interpret the sources of scalp maps
- Interpretation of scalp maps in terms of brain source distributions is “**inverse** source estimation”
- Mathematical techniques are available to aid in interpreting scalp maps as arising from particular brain sources
- These require an **inverse source model**, i.e. assumptions about the possible locations and nature of the sources (i.e., what attributes make them ***physiologically plausible***).
- Then search for the ***most plausible*** source model.



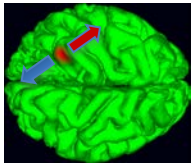
Summary

- **Inverse modeling**
 - Model assumption for volume conductor
 - Model assumption for source (i.e. dipole)
 - Additional assumptions on source
- **Single point-like sources**
- **Multiple point-like sources**
- **Distributed sources**
 - Different mathematical solutions
 - Dipole fitting (linear and nonlinear)
 - Linear estimation (regularized)



Summary

- **If we have MRI of the subject ...**
 - Subject specific head model
 - Distributed source localization
- **If we don't have the MRI head image ...**
 - Warped 4-layer MNI model (NFT)
 - Dipole source localization
- **Skull conductivity estimation** is *as important* as the head model used (SCALE)
- **White matter** modeling does not have a huge effect on source localization.

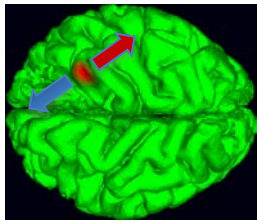


Acknowledgments

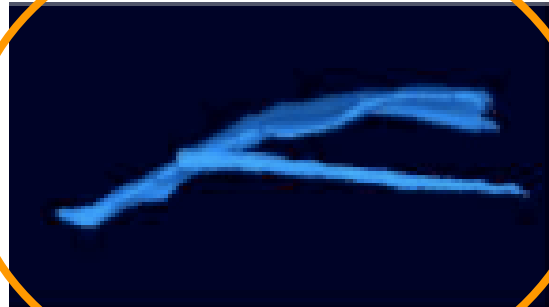
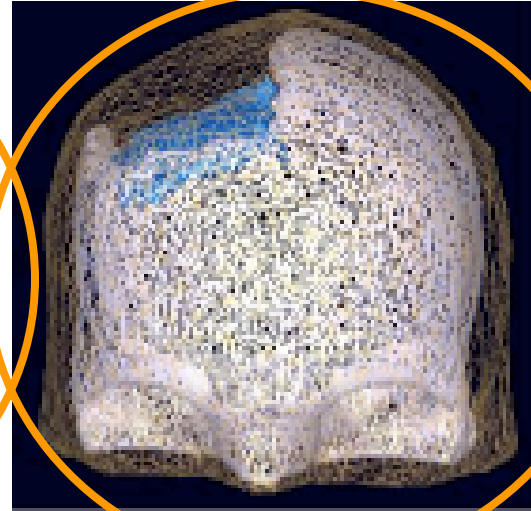
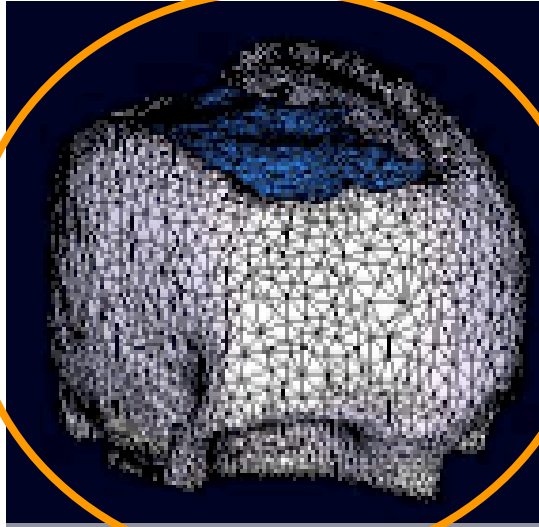


This presentation by Scott Makeig from materials by

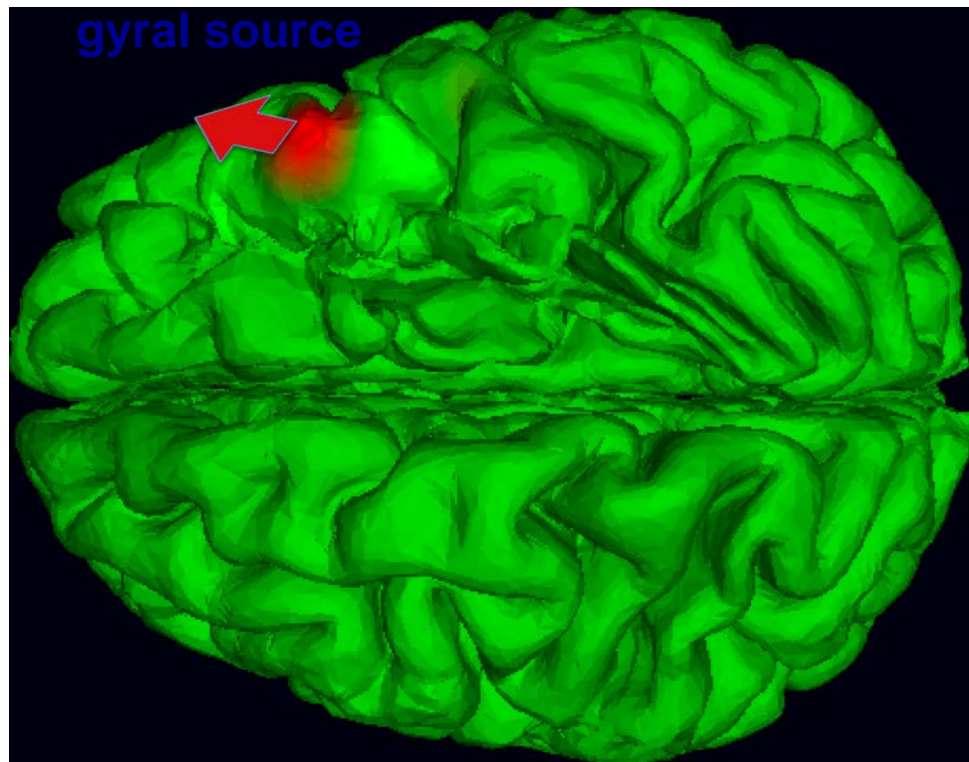
- **Robert Oostenveld** (Donders Institute, Nijmegen)
- **Zeynep Akalin Acar** (SCCN, UCSD)
- **Arnaud Delorme** (SCCN)



Mapping sources of intracranial data recorded to plan brain surgery

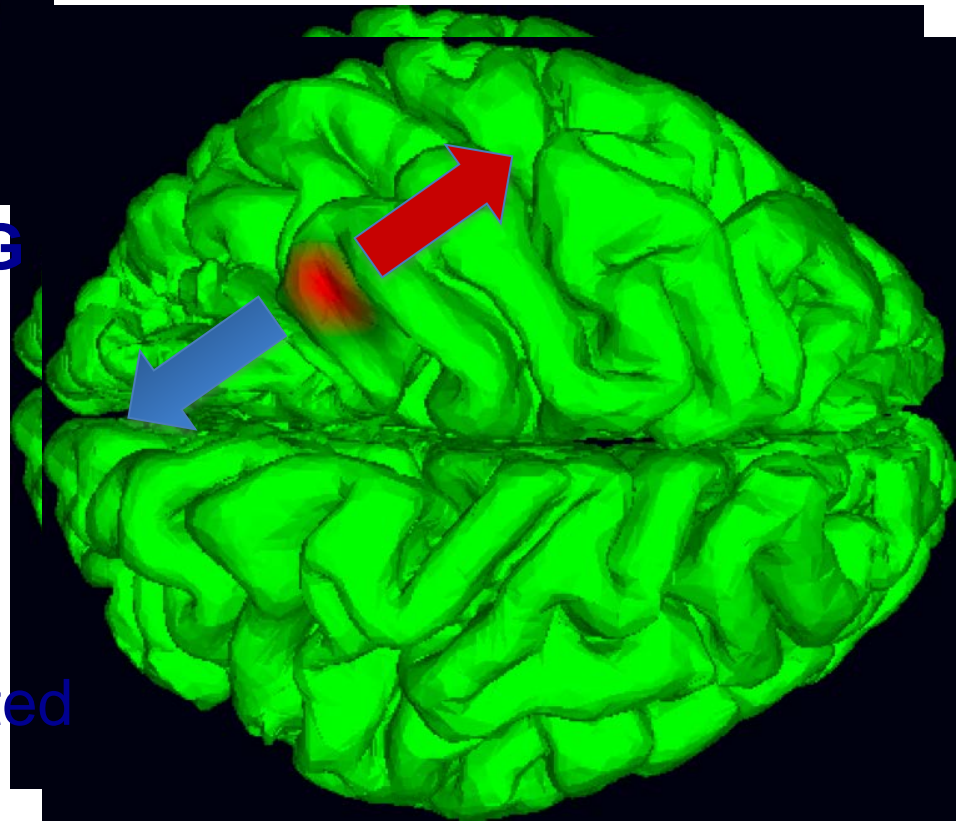


Non-conductive 'plastic
layer' (the ECoG
electrode sheet and strip)



ECoG Data Source Decomposition by ICA

sulcal source



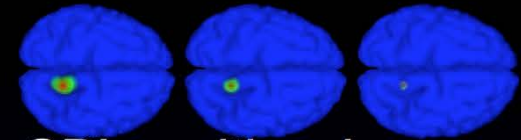
Given ECoG data from an ECoG grid:

0. Apply ICA to ECoG data.
1. Can find radially oriented 'gyral' ECoG sources (left)
2. Can also find tangentially oriented 'sulcal' ECoG sources (right).

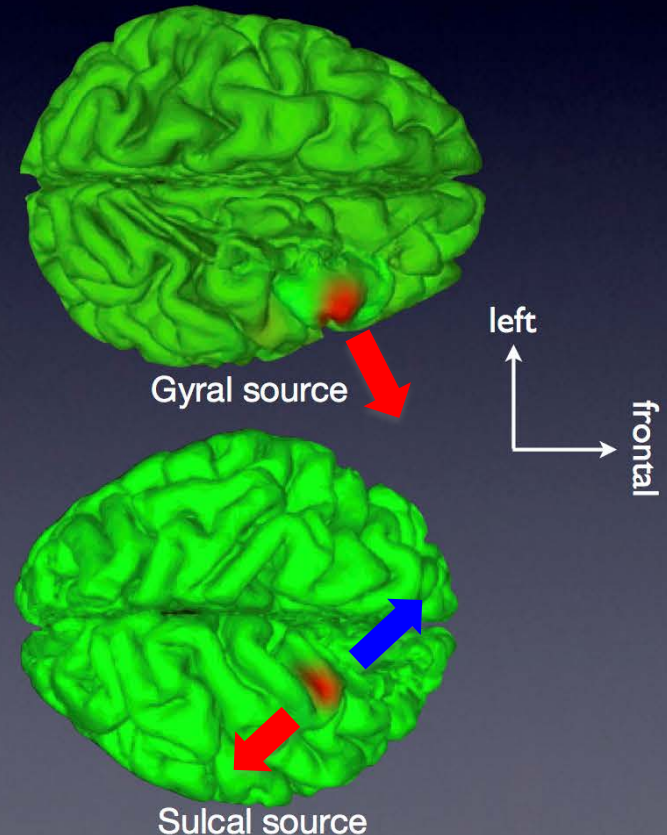
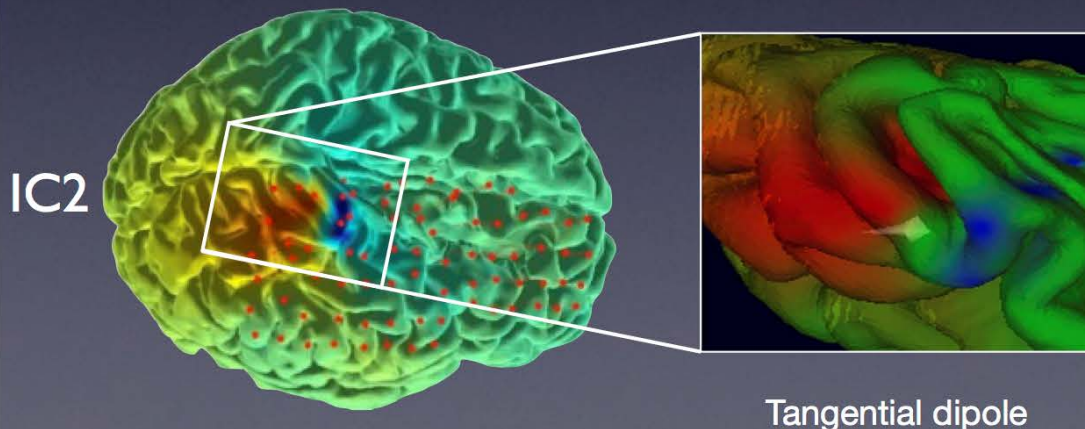
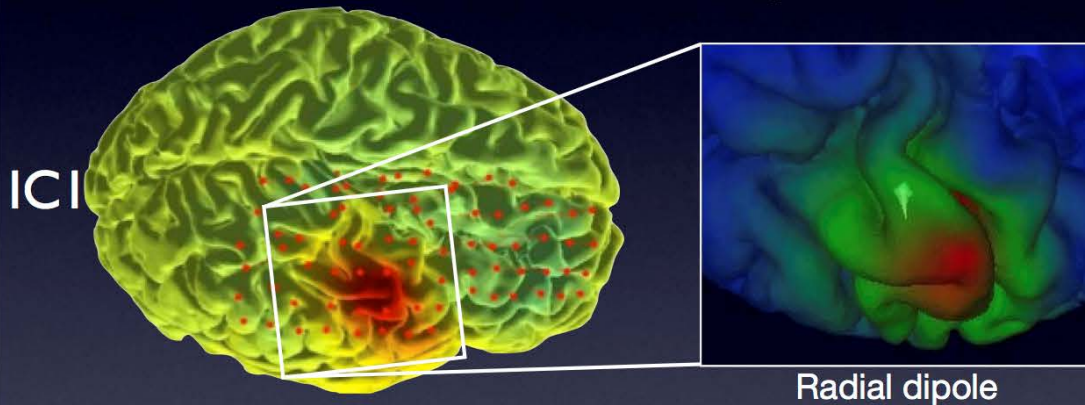
SBL Localization of Epileptogenic IC Sources

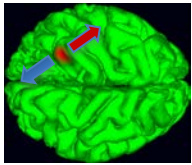
IC maps interpolated on cortical surface mesh

Equivalent Current Dipole solution



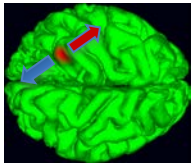
SBL multiscale patch solution





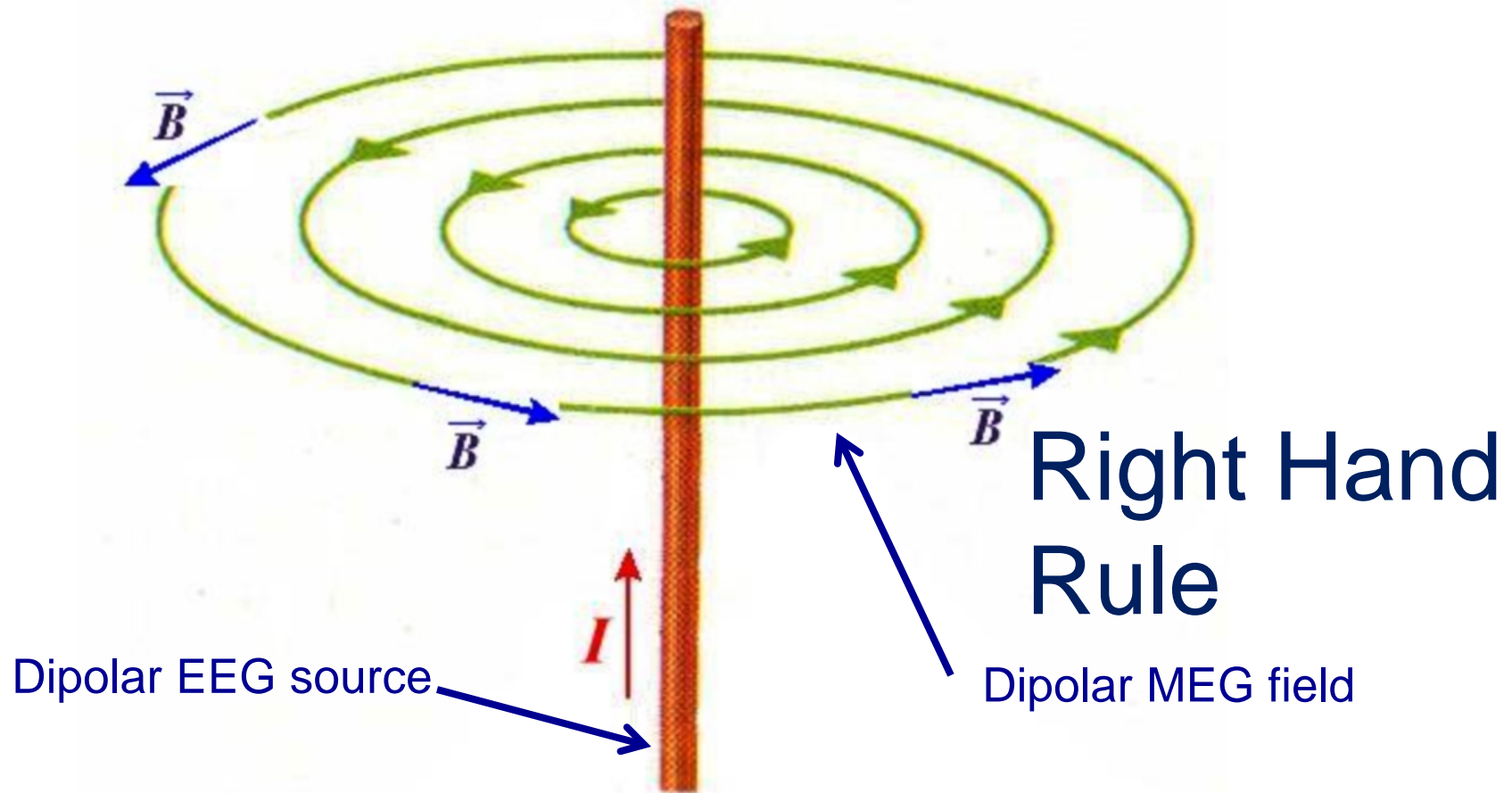
EEG vs. fMRI

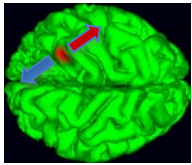
- EEG measures extra-cellular potentials
 - Indexing **synchronous neuronal activity** (*phase*)
- fMRI measures BOLD (blood oxygenation level diff.)
 - Indexing **local energy consumption** (*amplitude*)
- Different characteristics
 - Generators different (neural currents, blood flow)
 - Time course differences (***BOLD is slow***)



EEG versus MEG

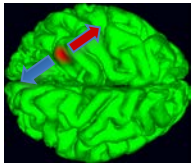
Electric current \leftrightarrow magnetic field





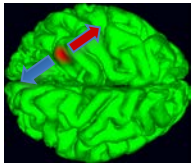
MEG volume conduction

- Measures sum of fields associated with
 - *Primary* currents
 - but also *secondary currents* at current distortions !
- However, only a fraction of the currents pass through the poorly conductive skull.
 - Therefore skull and scalp may be ~neglected in the MEG forward model (**simpler source imaging**).
- Local conductivity assumption near the dipole are important, so
 - **Both the geometries AND the conductivity of head model tissues are important in MEG forward models !**



Differences between EEG and MEG

- EEG scalp projection more blurred by volume conduction
- **BUT** MEG is insensitive to radial sources!!
- → **Therefore – EEG sees more sources!**
- EEG may be more noisy (electrode-skin impedance ~)
- but MEG is more sensitive to environmental fields!
- MEG requires no electrode gel
- but MEG requires the head to remain fixed !
- MEG is **MUCH** more **expen\$\$\$ive** than EEG!

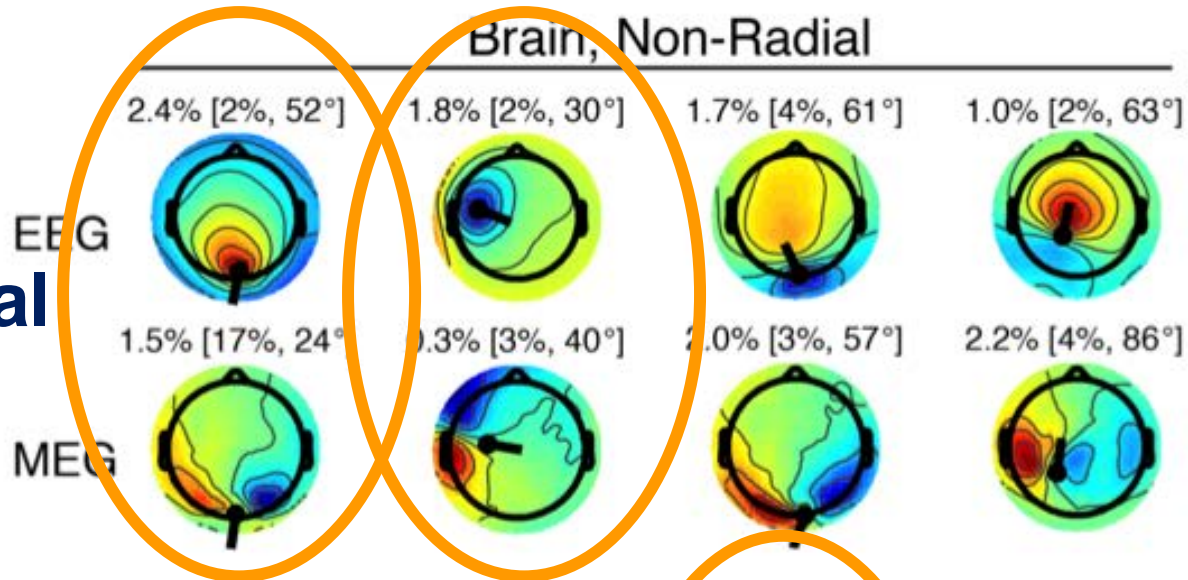


Differences between EEG and MEG

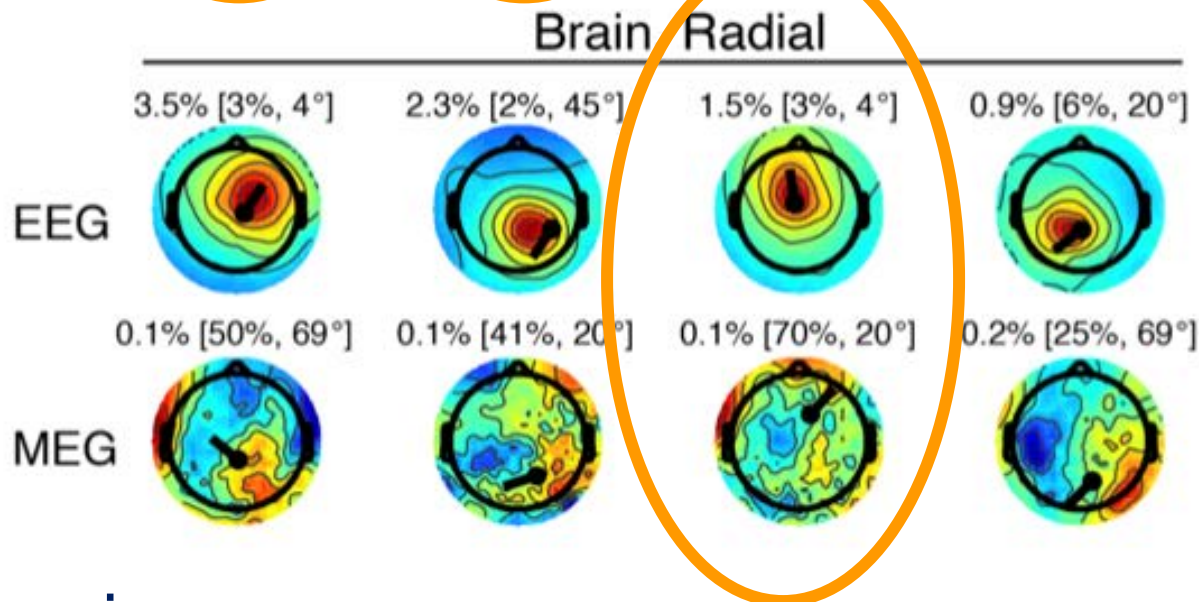
- EEG sees potential differences, requires choice of reference electrode – but **ICA makes the data reference-free.**
- MEG sensor measures have no common reference
- MEG can use a simpler but still somewhat accurate forward model
 - E.g., a multiple non-concentric sphere model,
Here, each sensor has its own local sphere fitted to the head position of brain relative to MEG sensors
 - But may vary when/if the head moves
 - Must be individually fit for each session

EEG+MEG Joint ICA Decomposition

**Tangential
Sources**



**Radial
Sources**



MEEG plug-in