

# Forward and Inverse EEG Source Modeling

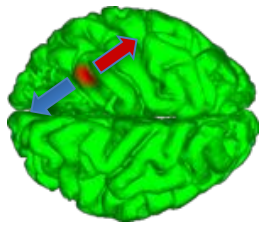


Scott Makeig

Institute for Neural Computation, UCSD

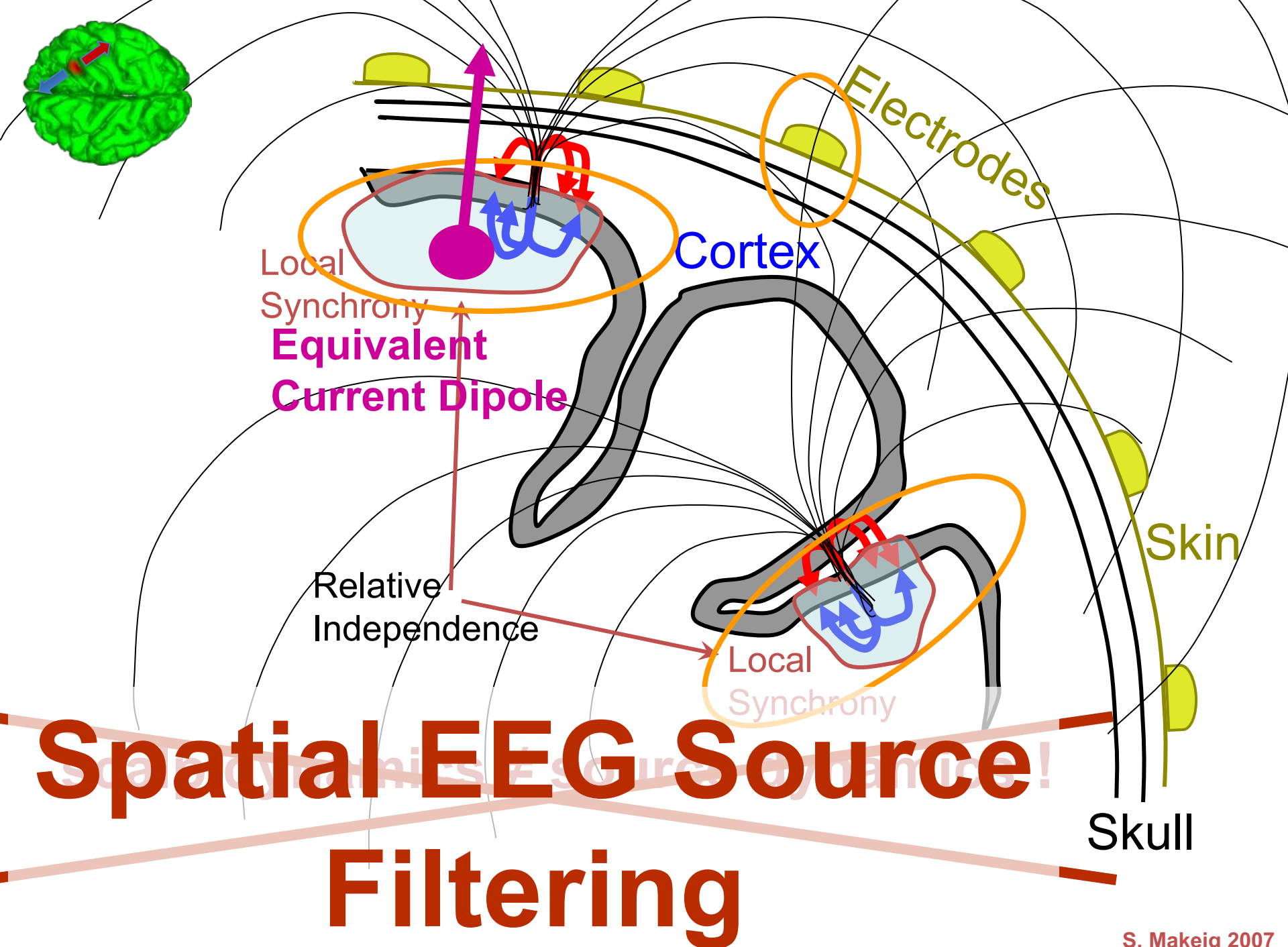
**EEGLAB Be'er Sheva, Israel**

**October, 2017**



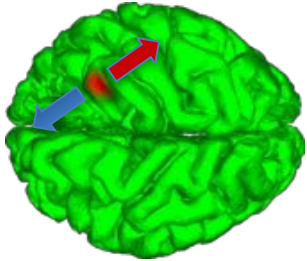
# Motivation

- 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



# EEG source modeling

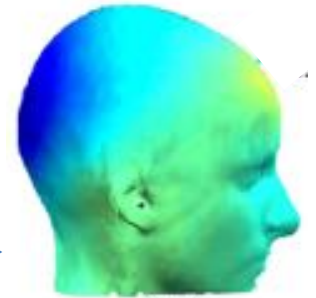
Source  
Space



Forward head model

**forward** problem

Sensor  
Space



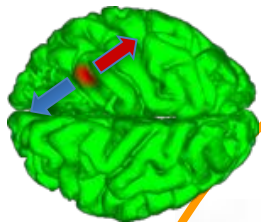
electrical  
currents

volume conduction  
through body tissues

recorded  
potentials

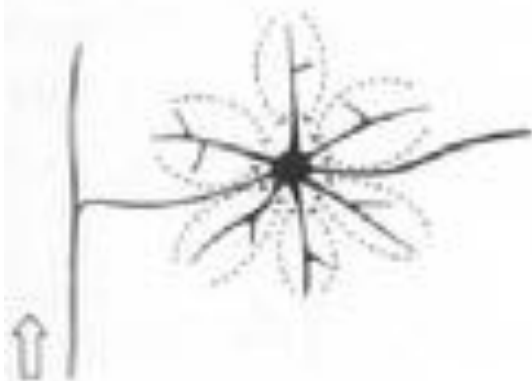
**inverse** problem

Inverse localization  
method



# Peri-neuronal currents

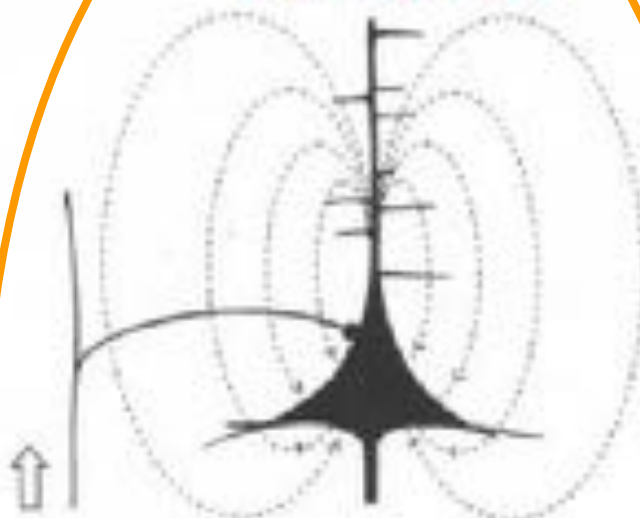
Stellate cell



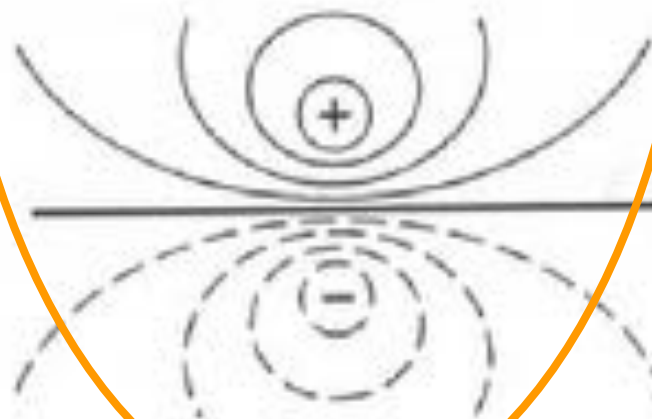
Closed field

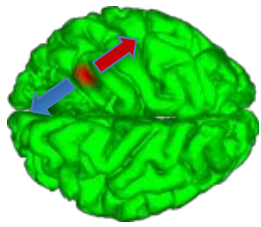


Pyramidal cell



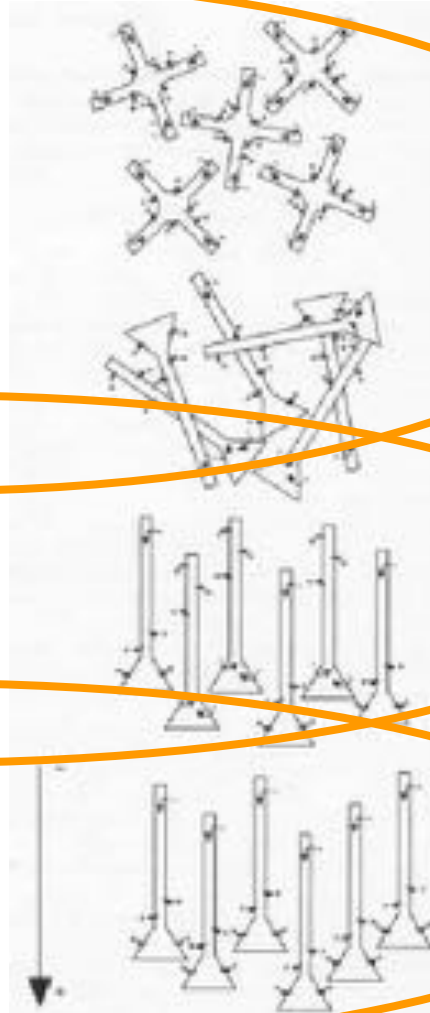
Open field





# Symmetry, orientation and activation

radially symmetric, i.e.  
randomly-oriented



Closed field

asynchronously activated

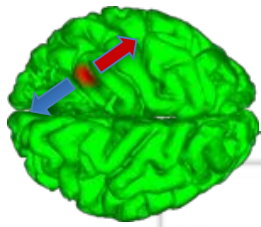
Phase  
cancellation

synchronously activated  
parallel-oriented

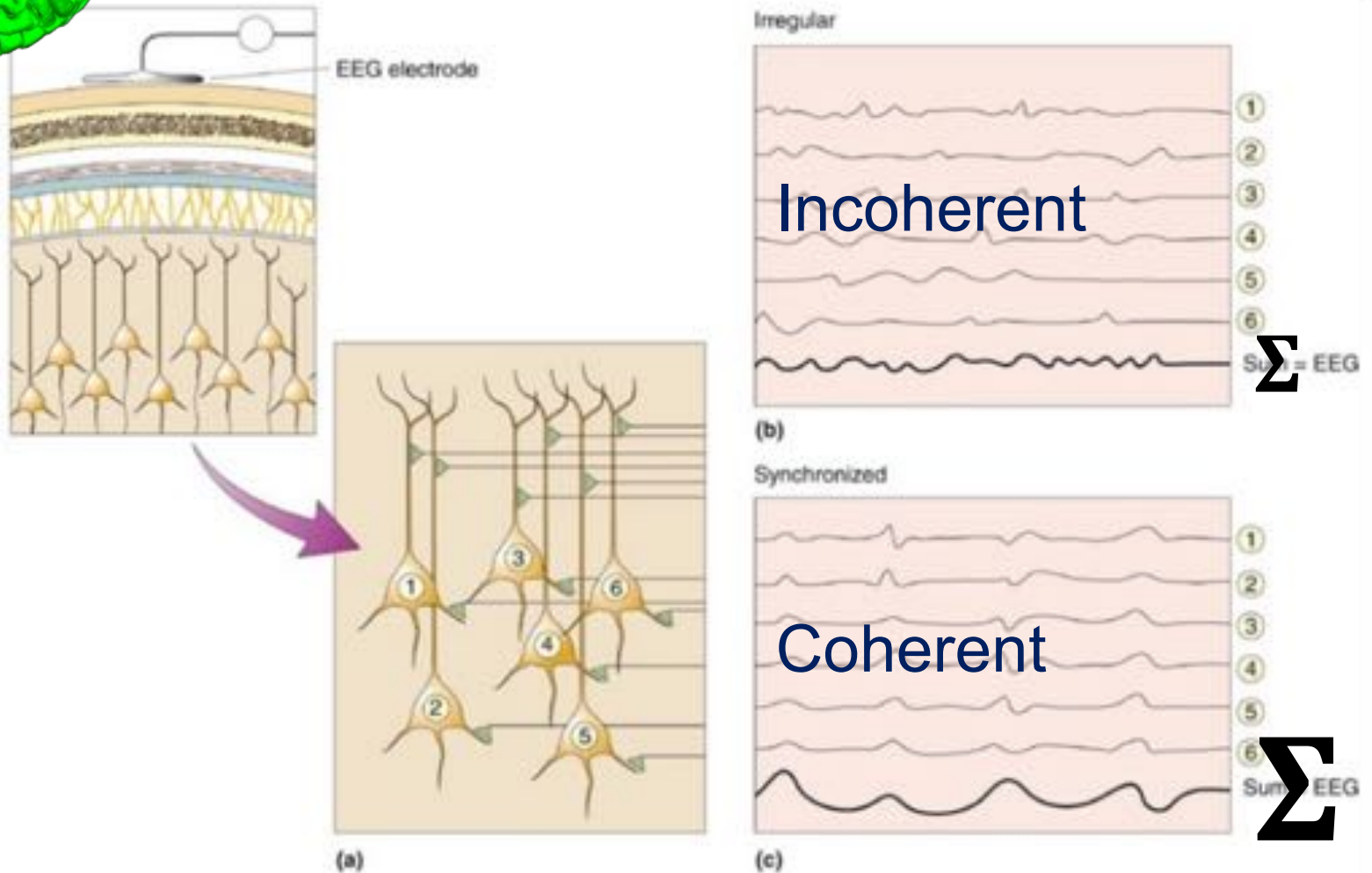
Open field

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

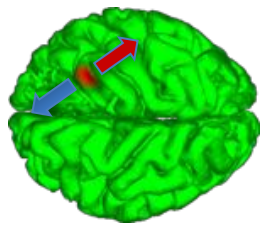




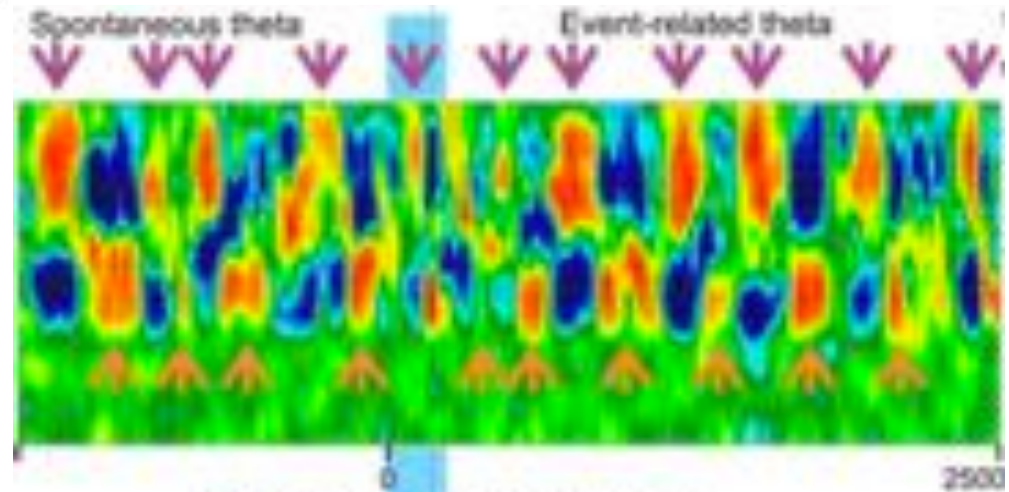
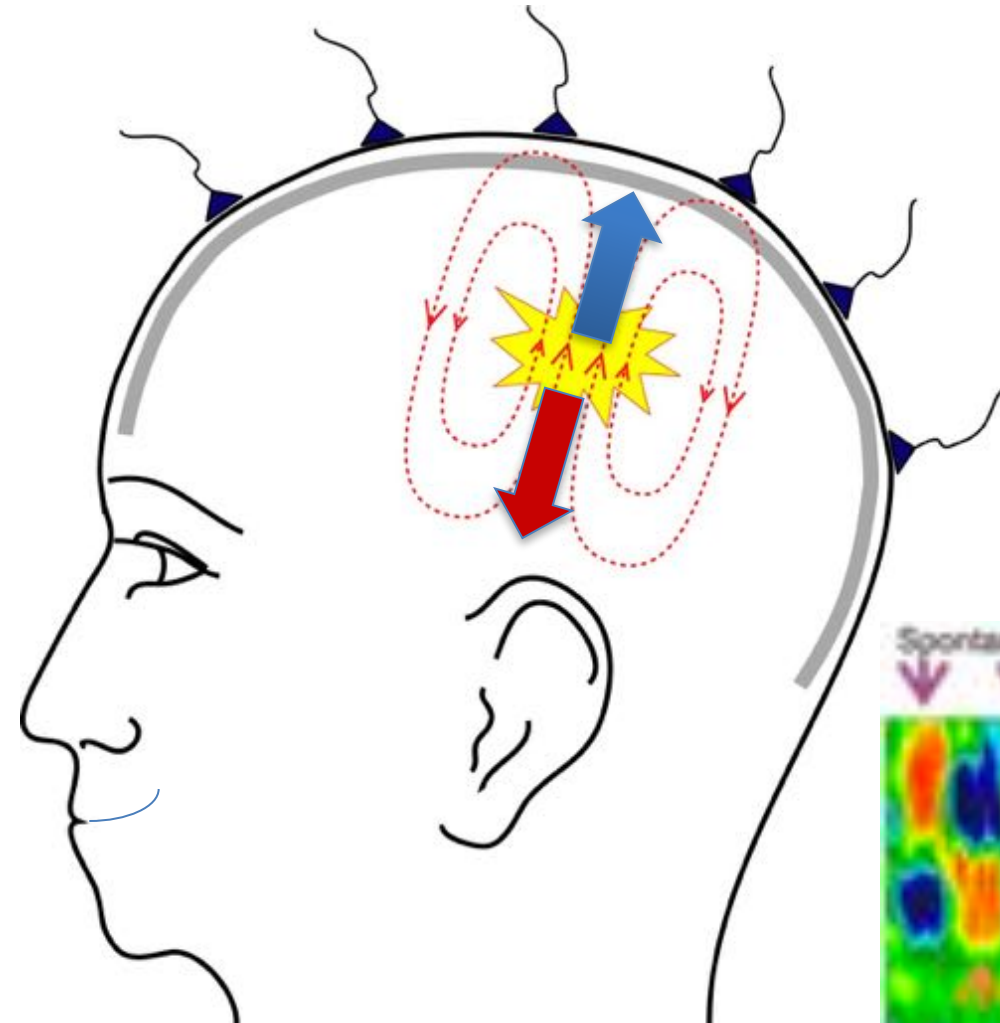
# EEG Effective Sources



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

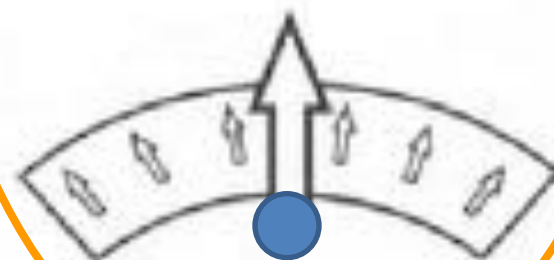
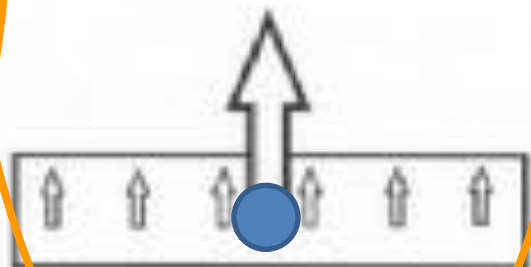
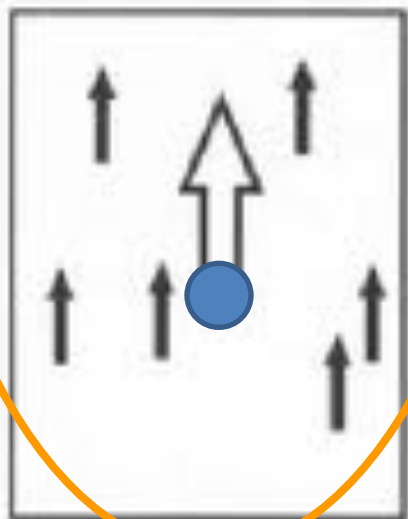
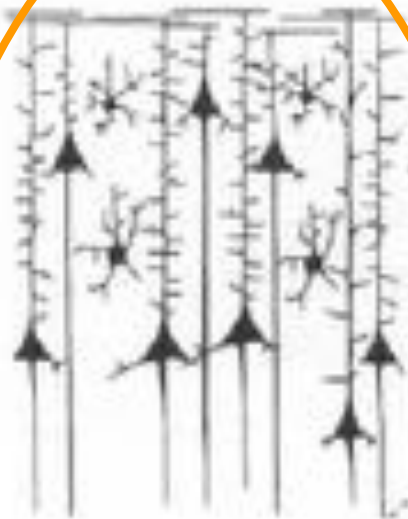
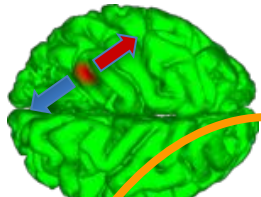


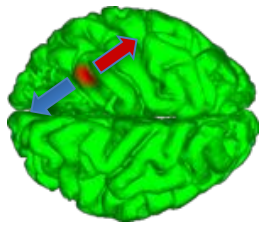
# EEG volume conduction of dipolar field patterns → effective sources



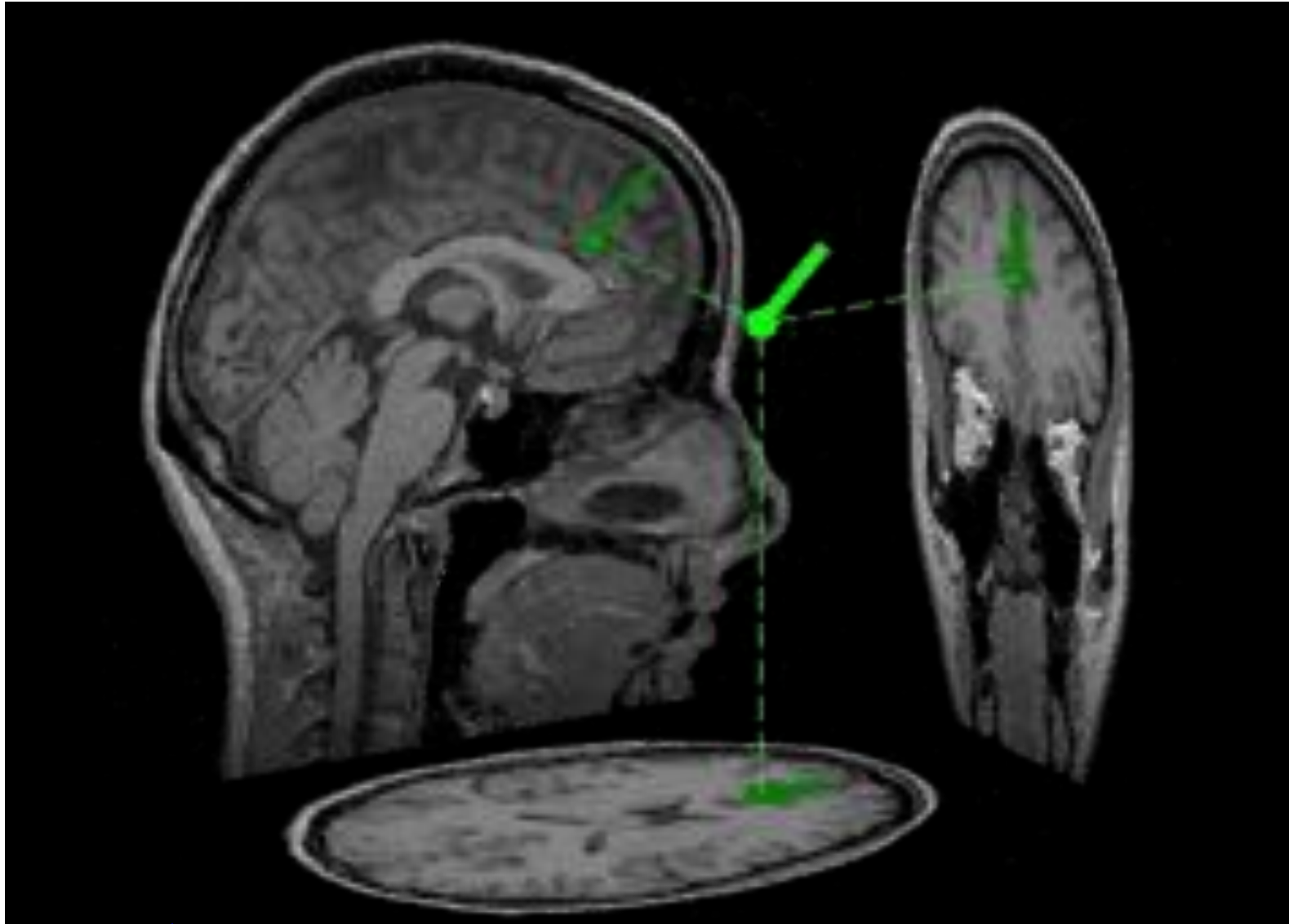


# The *equivalent* current dipole



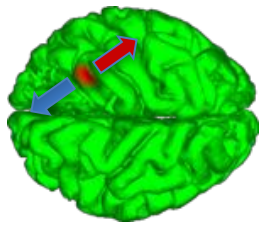


# Equivalent current dipole modeling

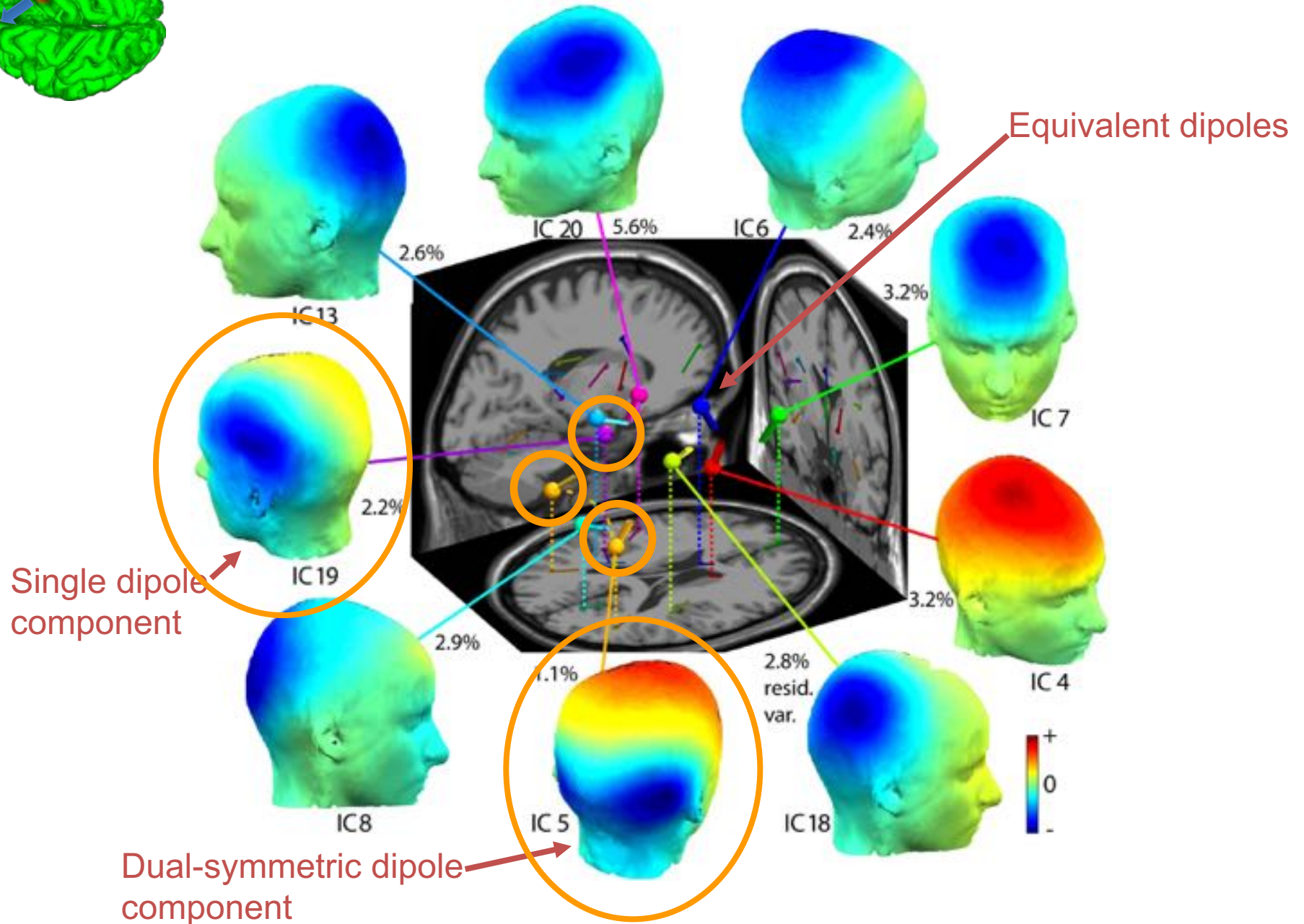


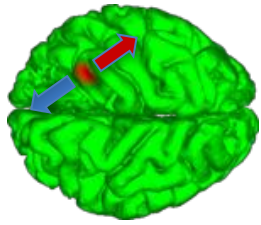
1<sup>st</sup> IC source fit in an individual head model via EEGLAB

A. Delorme, ~2007



# Independent cortical components

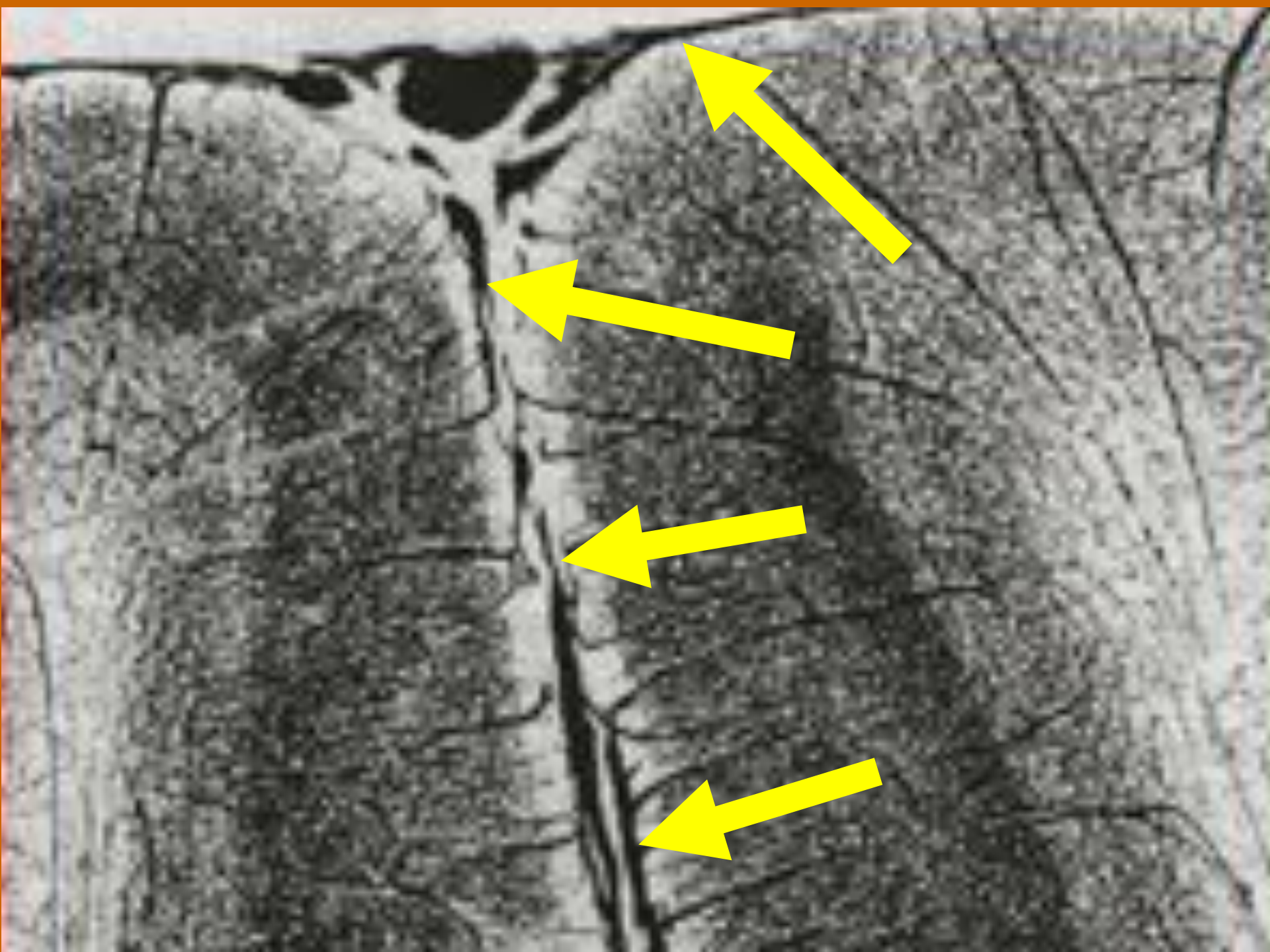


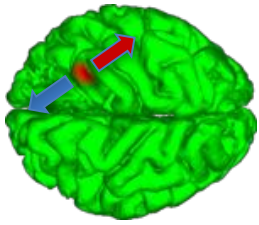


# Equivalent current dipole modeling

- **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



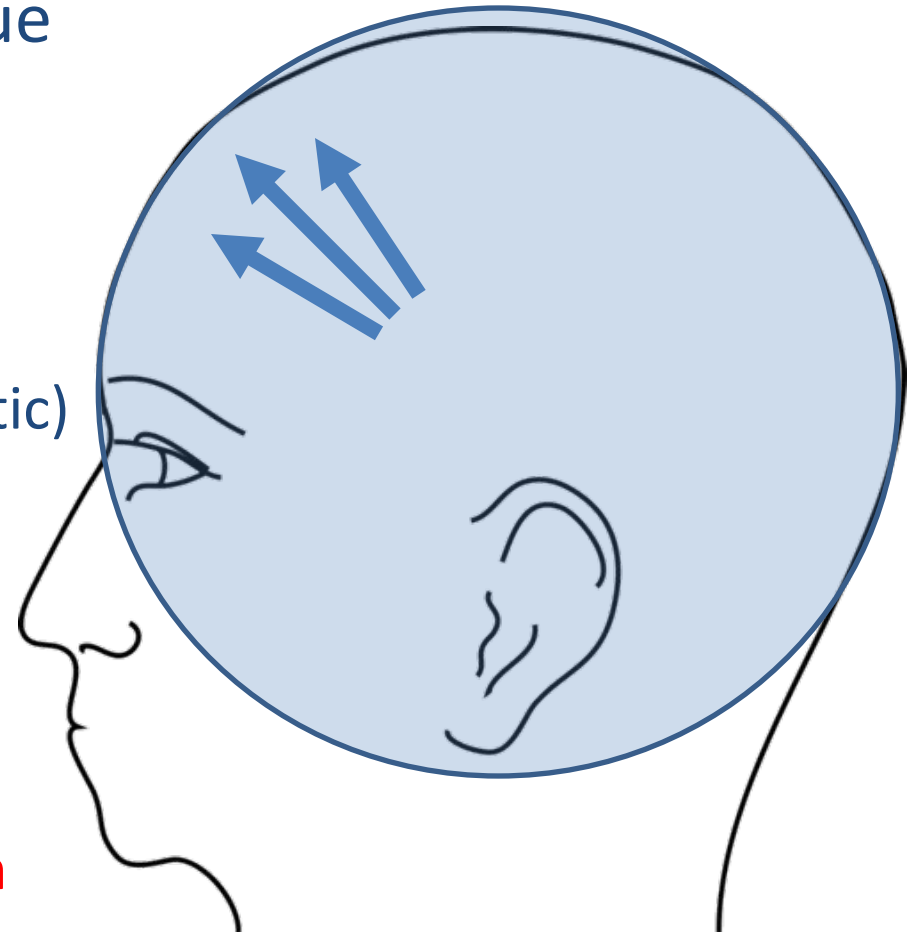




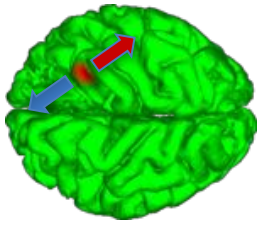
# Forward Head Models

- 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

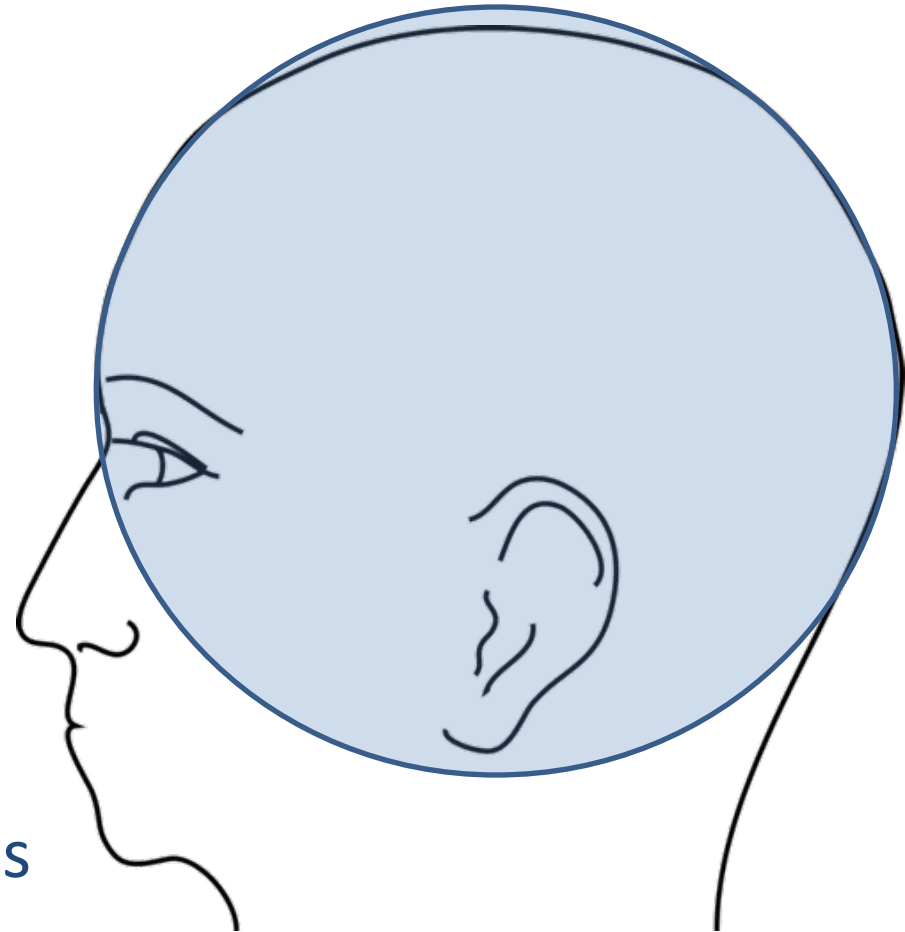


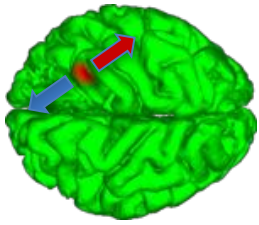




# Forward Head Models

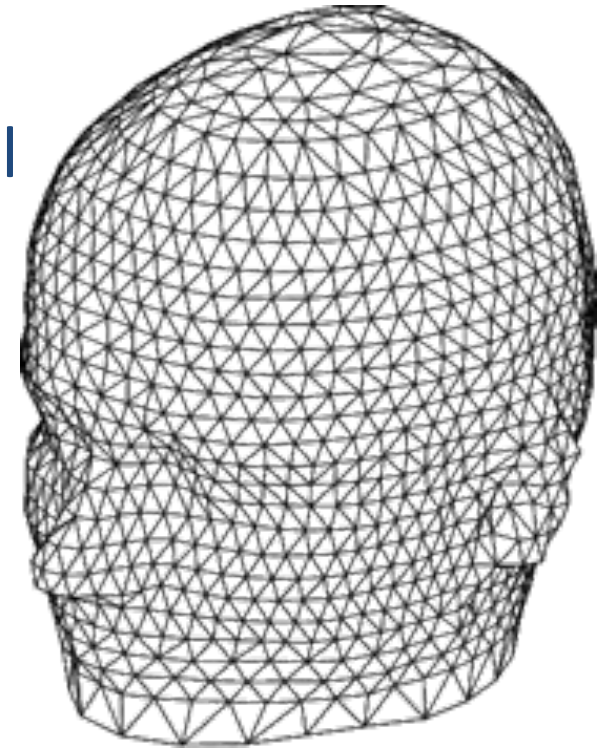
- Advantages of the **spherical** model
  - mathematically accurate
  - reasonably accurate
  - computationally fast
  - easy to use
- Disadvantages of the **spherical** model
  - inaccurate in some regions
  - difficult to align to head



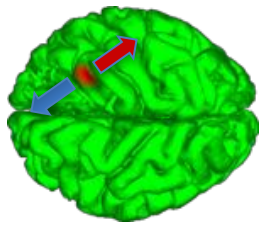


# Forward Head Models

- 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

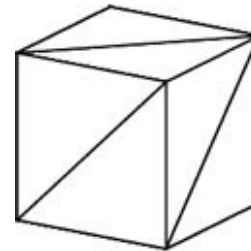
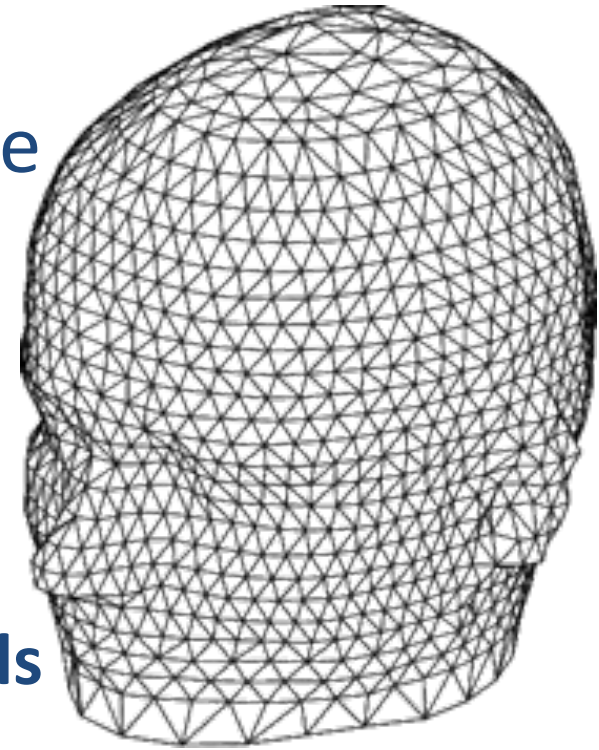


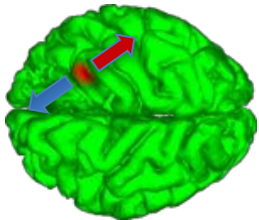
→ The pragmatic (easy, cheap) solution is to use  
a standard (mean) realistic head model (MNI).



# 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**



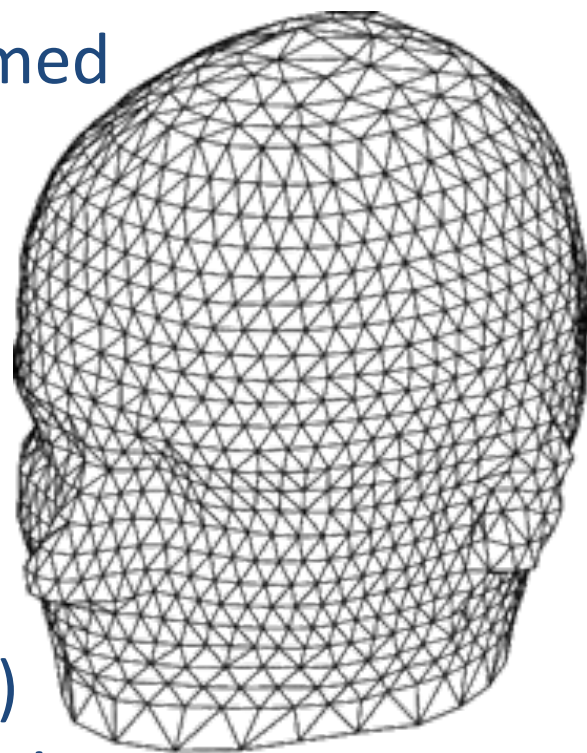


# Forward Head Models: BEM

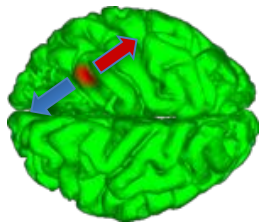
- **Boundary Element Method (BEM)** models
  - description of head geometry by tissue compartments
  - Tissue in each compartment is assumed
    - homogenous
    - isotropic

## Important tissue types

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

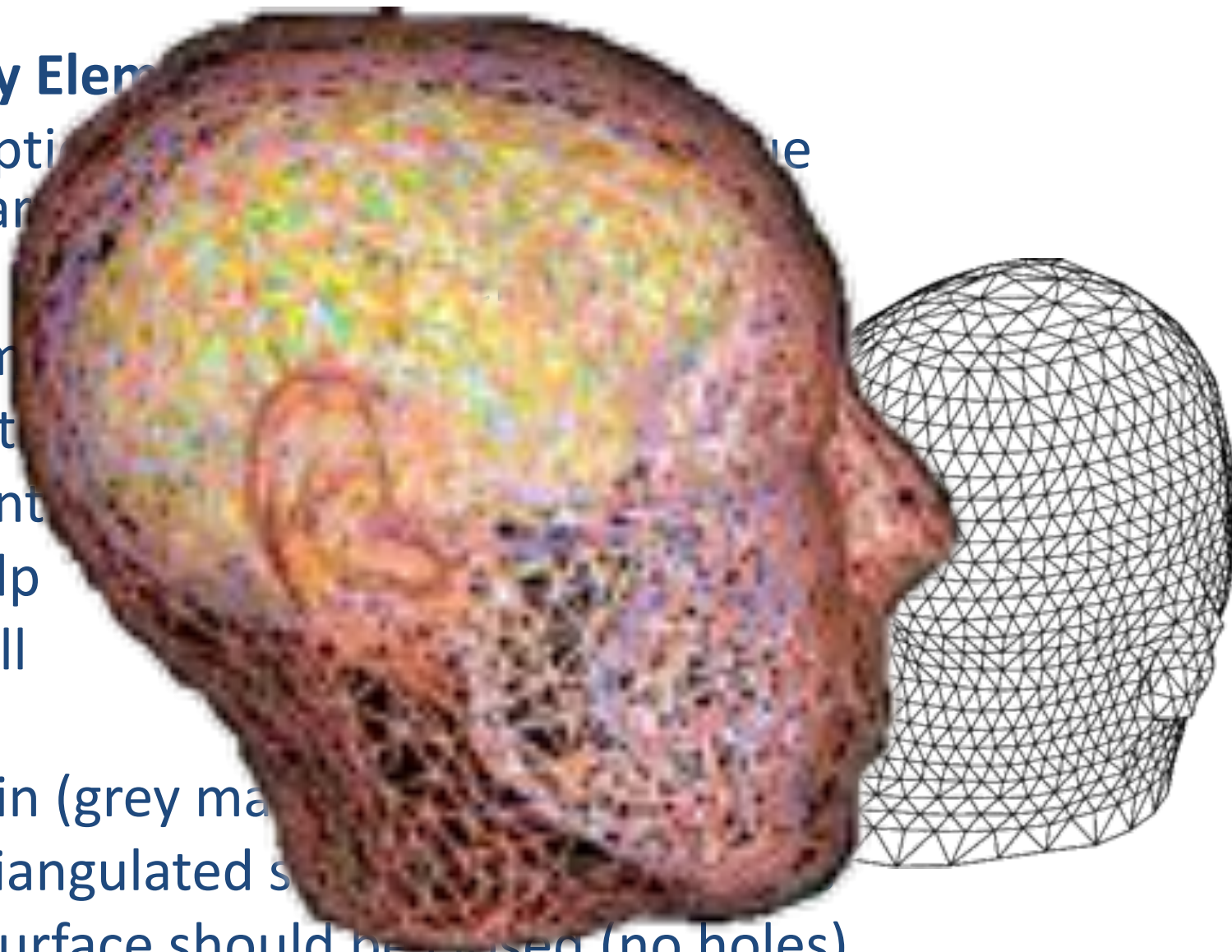


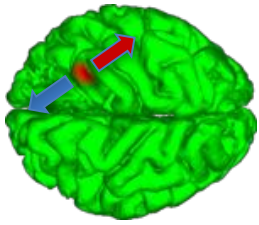




# Forward Head Models: BEM

- **Boundary Element Method**
  - description of the volume by the boundary
  - 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)

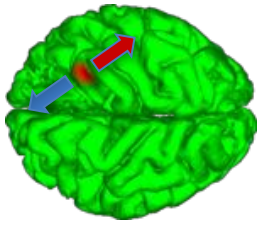




# Forward head models: Modeling the skull

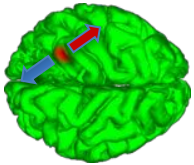
- **Potential differences between electrodes** measures summed current flowing through scalp
  - However, only a tiny fraction of *brain source currents* pass through the skull
  - Therefore a forward head model should describe *brain, skull, and scalp tissues* as accurately as possible.





# Forward head models: Modeling the skull

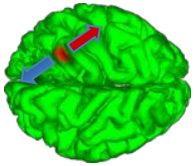
- Problems with skull modeling
  - Poorly visible in the anatomic MRI (T2) image
  - Thickness varies regionally
  - Conductivity is not homogeneous
  - Complex geometry at front and base of skull
- **Skull conductivity** variable & unknown



# 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

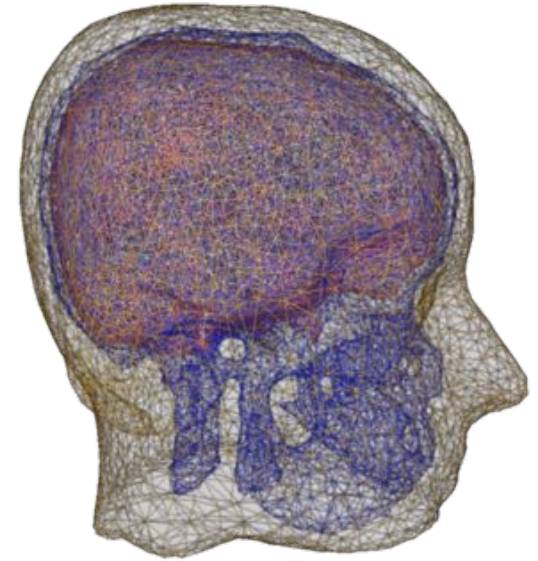
Head Geometry Errors

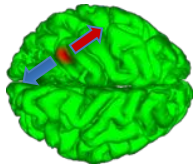
EXCLUSION of white matter

Too Few electrodes

Poor distribution of electrodes

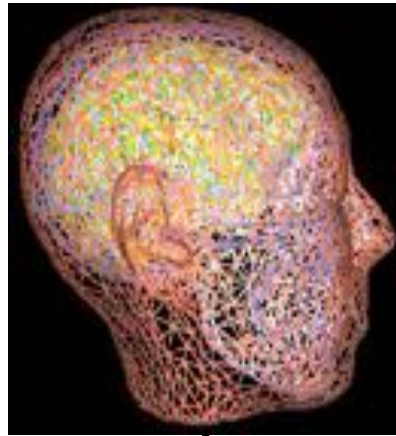
→ mis-estimation of skull conductivity





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

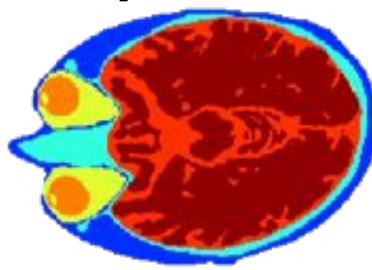
Solve the forward problem using realistic head models (BEM)



Mesh generation

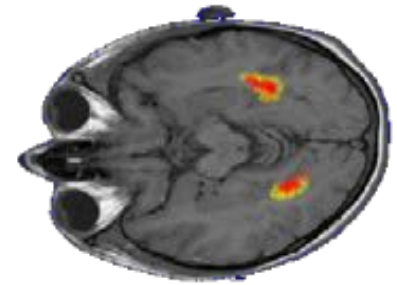


MRI



Segmentation

Simple Map



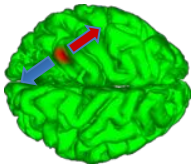
Source Image

Sensor Localization

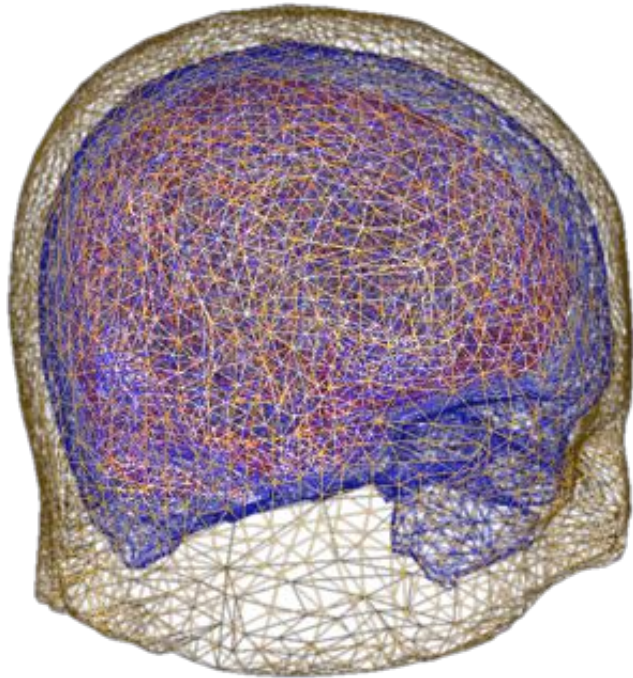
Signal Processing



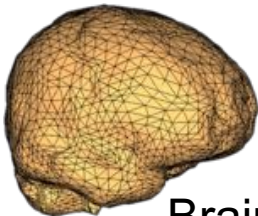
EEG/MEG



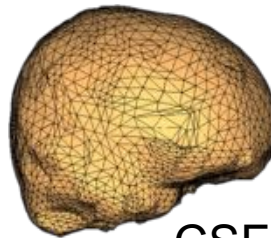
# The MNI Head Model



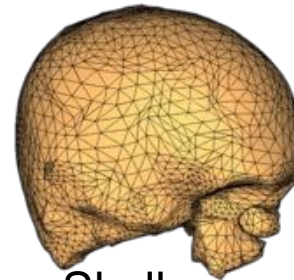
- 4-layer
  - 16856 nodes
  - 33696 elements
- 3-layer
  - 12730 nodes
  - 25448 elements



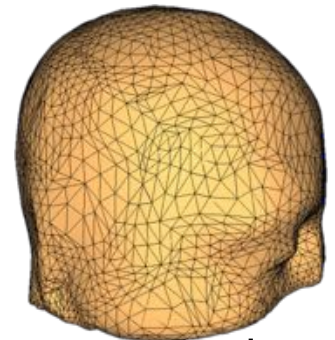
Brain



CSF

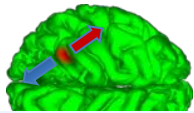


Skull



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

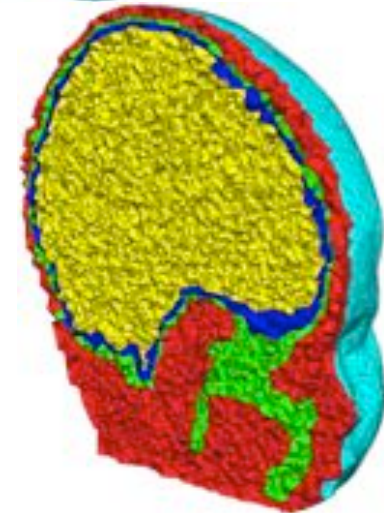
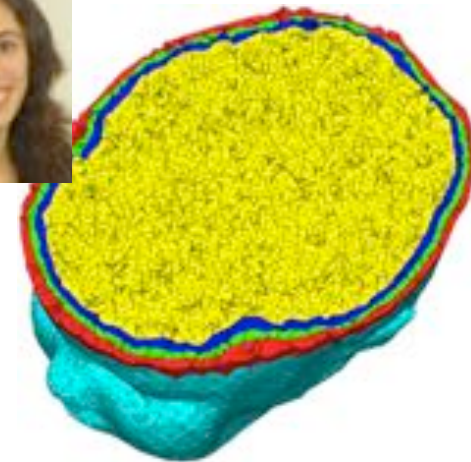
Electrode Co-Registrati...

Template Warping

FP Solution with BEM

FP Solution with FEM

Dipole Fitting

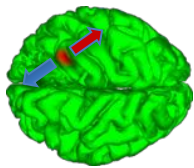


FEM models

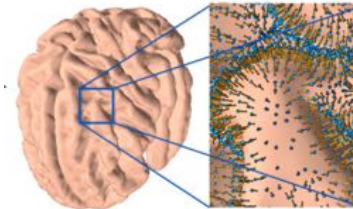


BEM models



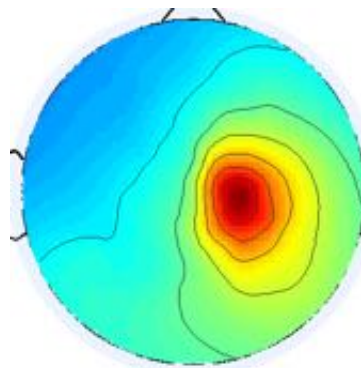


# NIST

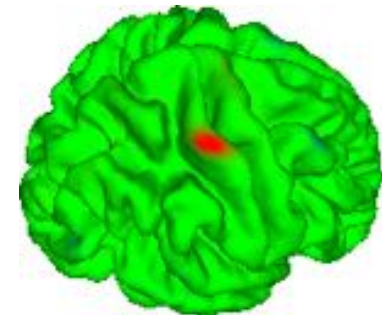


Source space

Scalp map

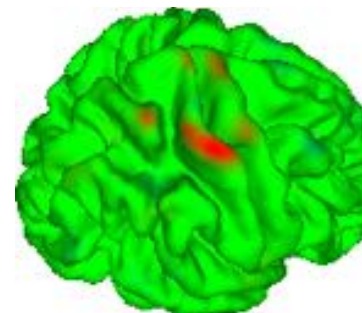


SCS

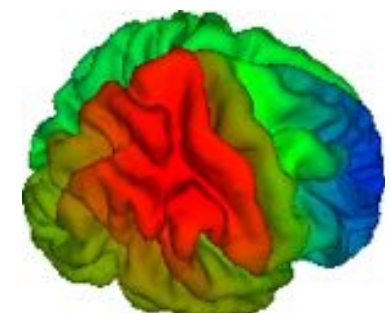


Cheng Cao, 2012

Patch-based SBL



sLORETA



Distributed\_Source\_Localization

Load MRI /data/cta/zeynep/NFTdene/13test001.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

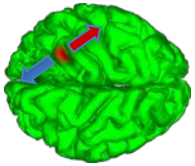
FP Solution with FEM

Component indices

Select Source Localization Method

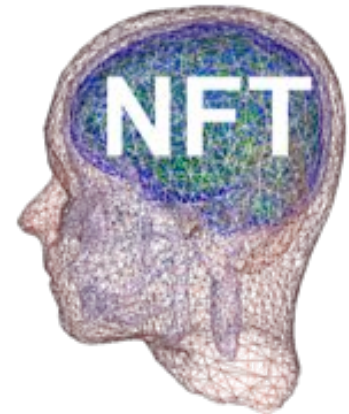
Start Source Localization

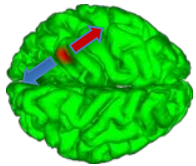
Visualization



# Head Model Generation Summary

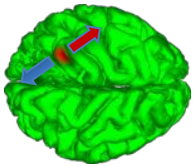
- **Subject-specific Head Model (NFT)**
  - From whole head T1 weighted MR of the subject
  - 4-layer realistic BEM model
- **MNI Template Head model (DIPFIT)**
  - From the MNI head
  - 3-layer and 4-layer template BEM model
- **Warped MNI Template Head Model (NFT)**
  - Warp MNI template to EEG sensors
- **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.



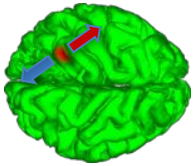
# Inverse methods

## 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!**

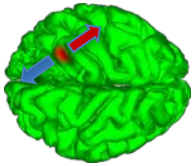
→ ICA solves ‘the first half’ of the inverse problem ‘What?’

→ 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

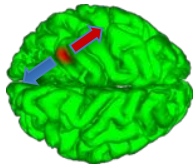


# DIPFIT: Dipole fitting 1. Grid search

## 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

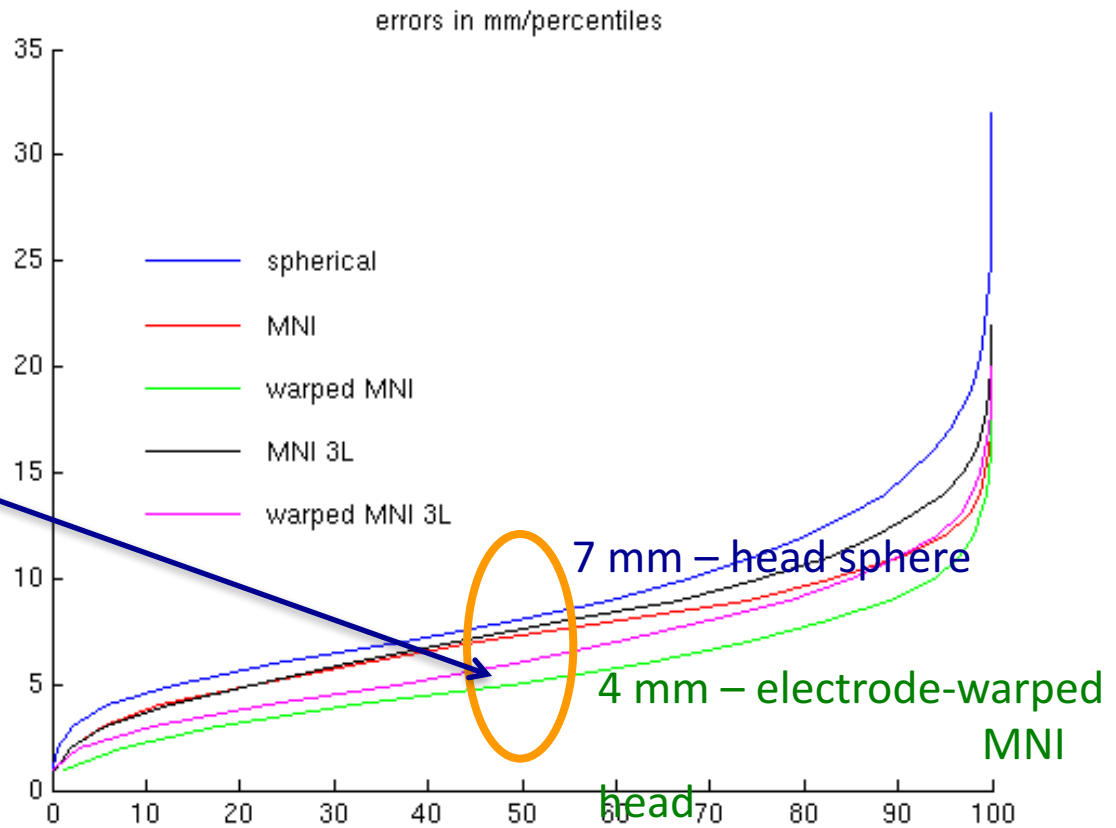
Start with the initial guess from coarse fitting

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

Number of iterative steps required =  $\sim 100$

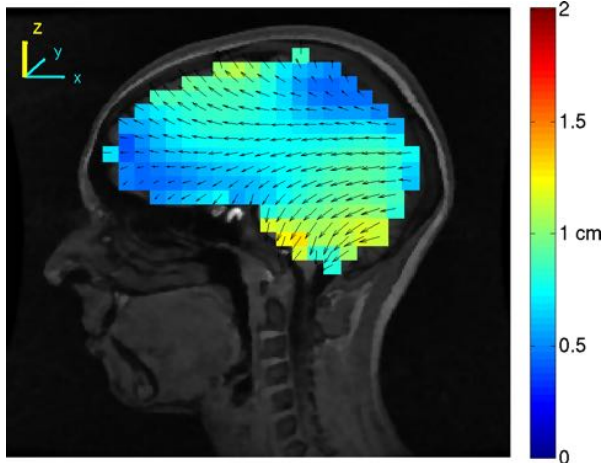
# Effect of Template Head Model Choice On Estimated Dipole Locations

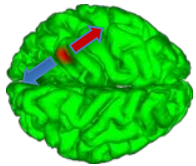
**By Simulation:** The median geometric error in dipole localization using the MNI template head model warped to measured electrode positions is only 4 mm.



## **BUT Additional dipole error contributors:**

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



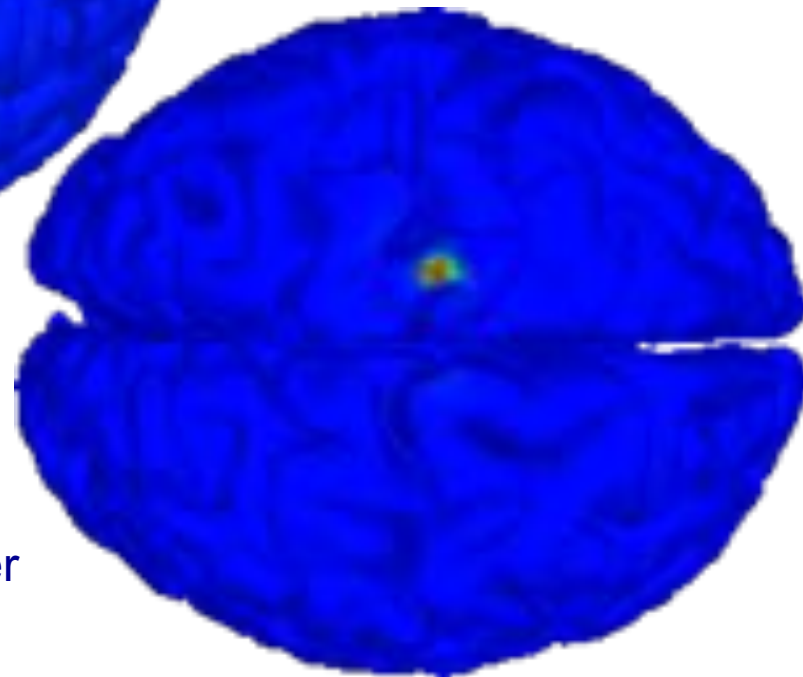
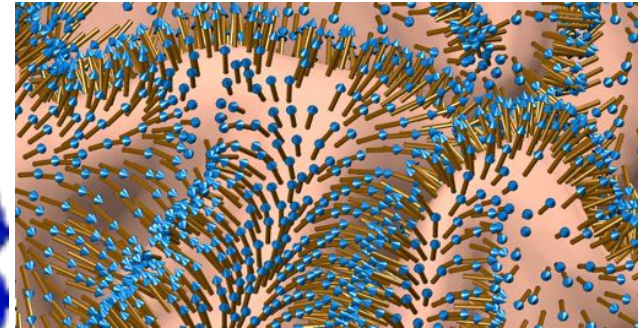
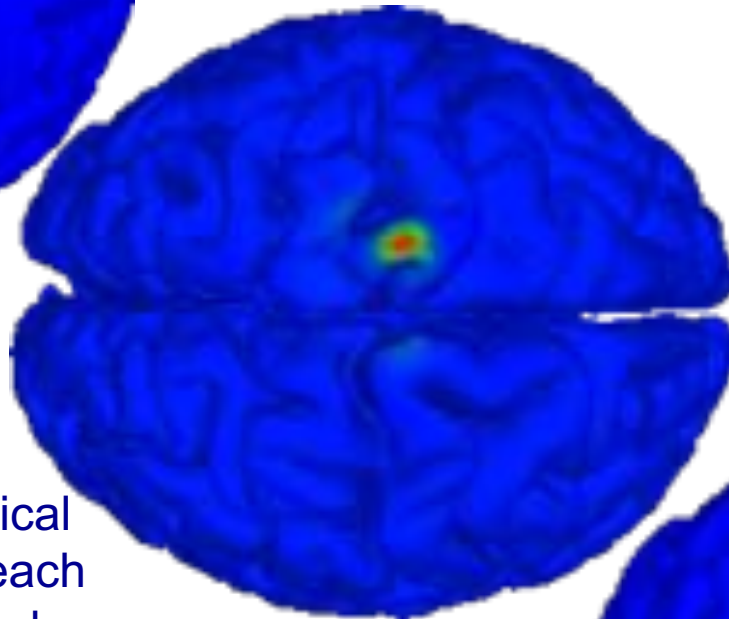
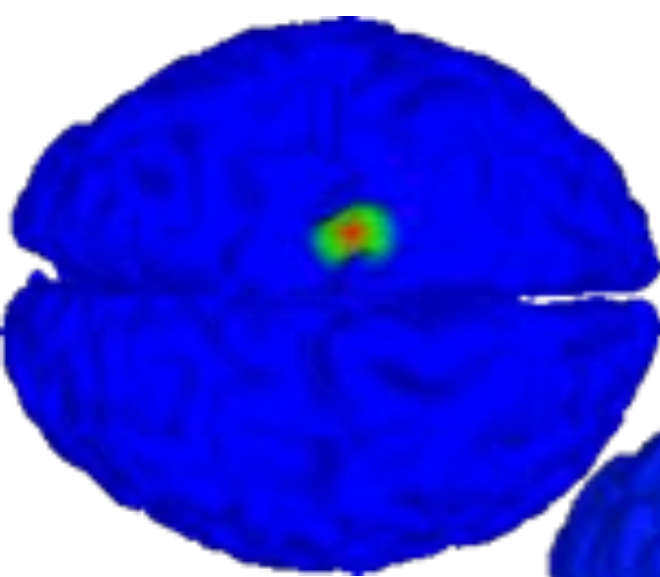


# Distributed source models

- 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 ...

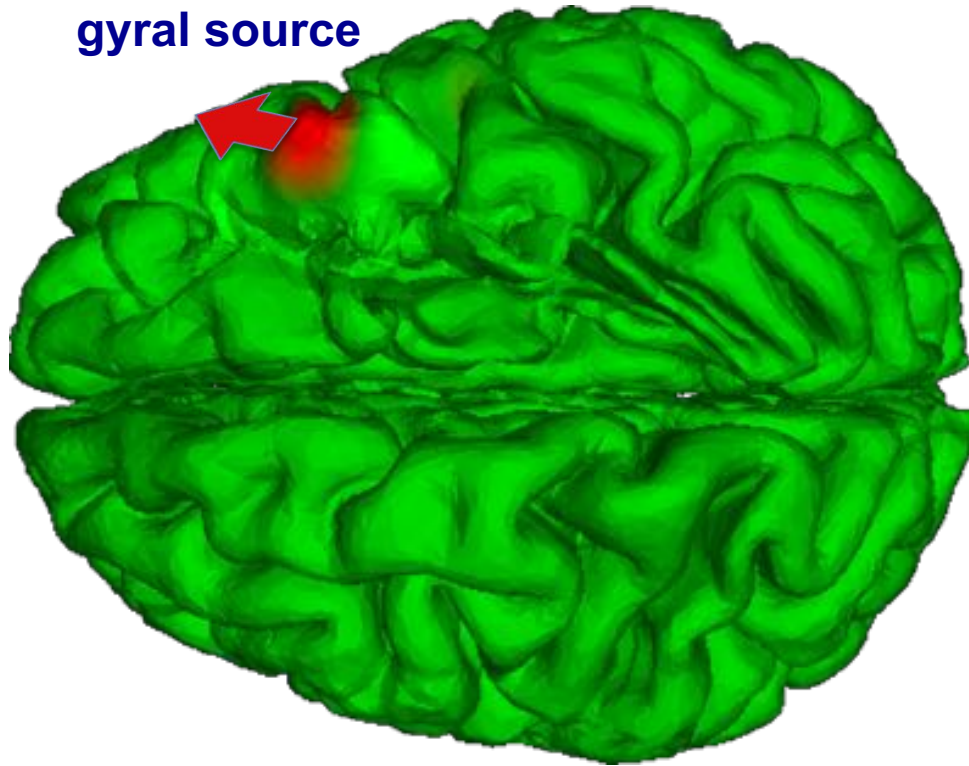
# High-Resolution Distributed Source Localization

using a multiscale patch basis



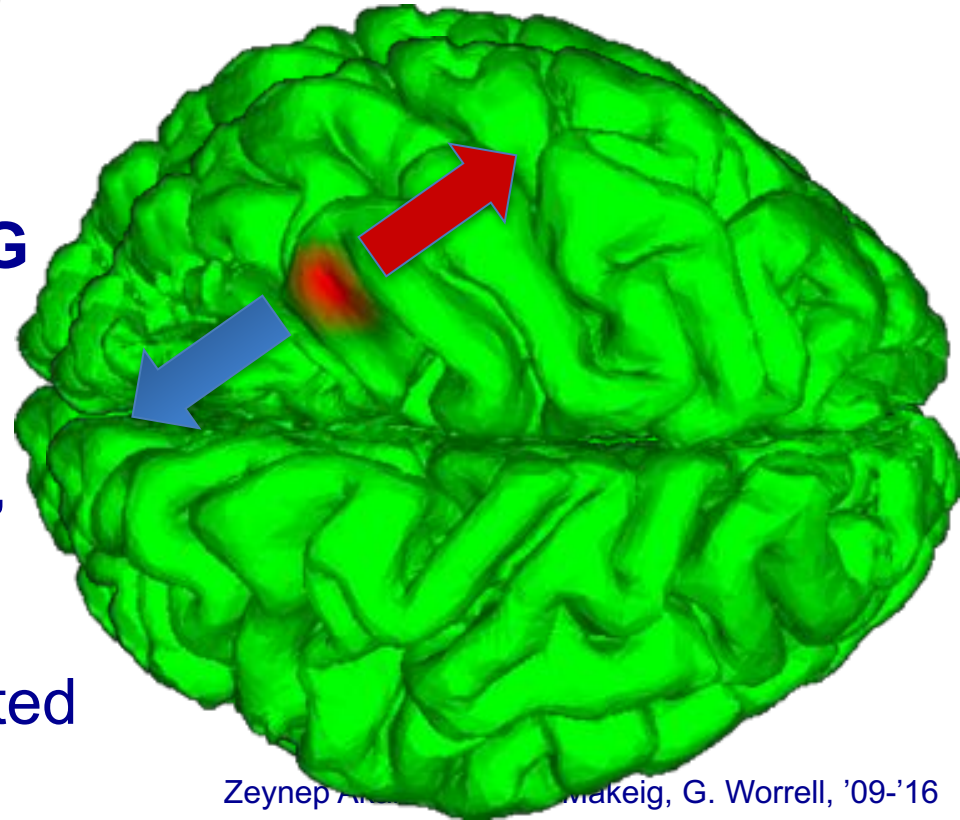
0. Build a high-res. cortical surface mesh; give each voxel an oriented dipole.
1. Compute a 'dictionary' of Gaussian patches conforming to the cortical surface centered at each cortical mesh voxel.
2. Use a 'sparsifying' approach (SCS) to find the sum of the *fewest* of these patches that together produce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

gyral source



## ECoG Data Source Decomposition by ICA

sulcal source



**Given ECoG data from an ECoG grid:**

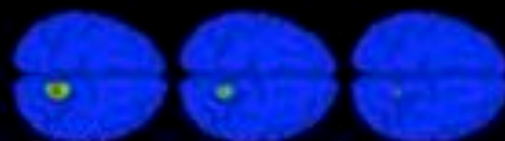
0. Apply ICA to ECoG data.
1. Can find radially oriented 'gyral' ECoG sources (left)
2. Can also find tangentially oriented 'sulcal' ECoG sources (right).



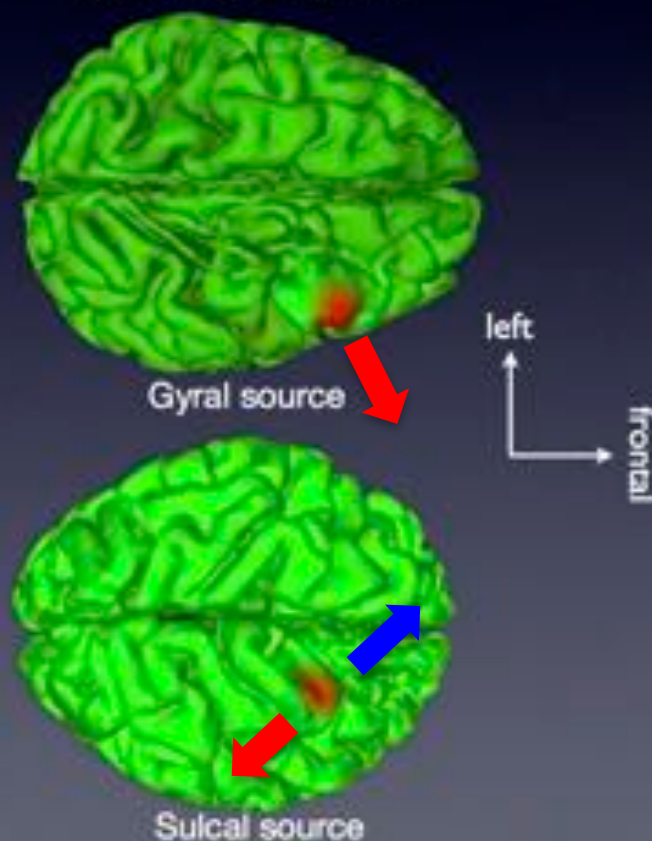
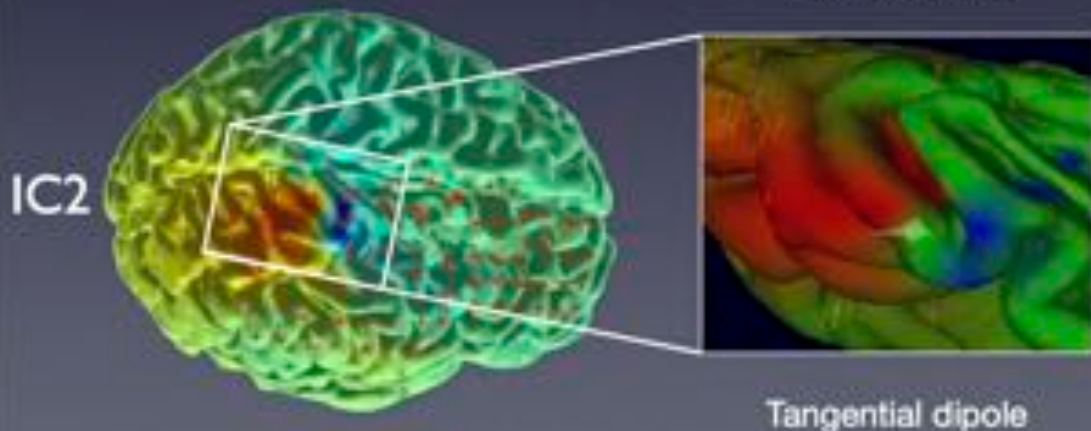
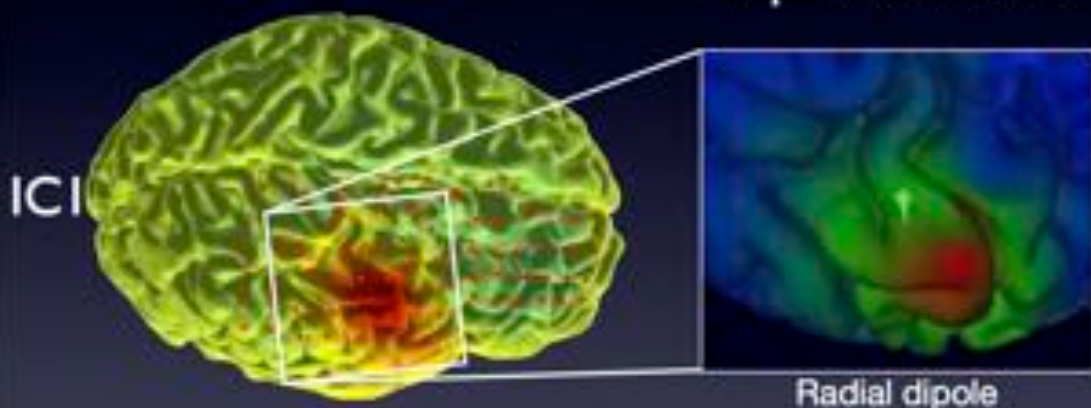
# SBL Localization of Epileptogenic IC Sources

IC maps interpolated on cortical surface mesh

Equivalent Current Dipole solution



SBL multiscale patch solution



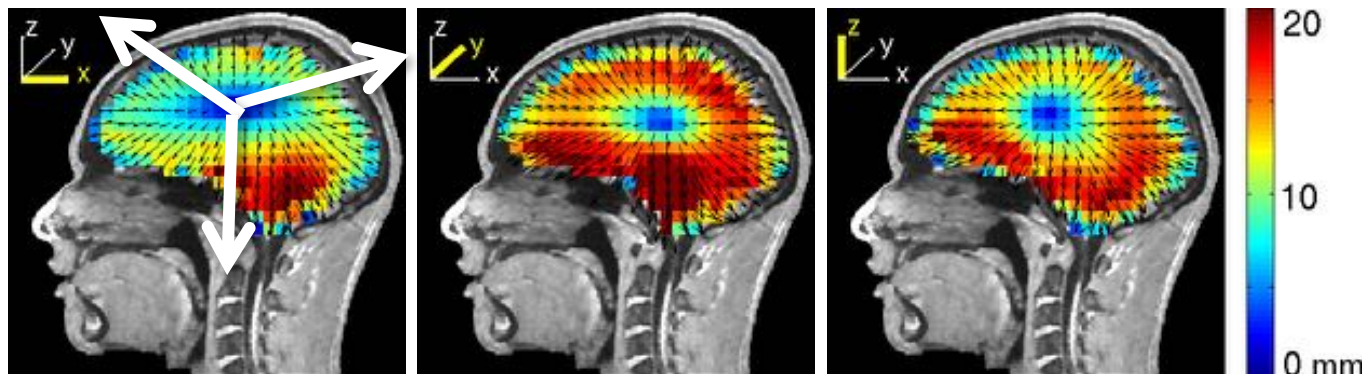
**BSCR**

**Simulate 25**

↑ RLS<sub>25-4</sub>  
↓ RLS<sub>80-4</sub>

**Assume 80**

(same  
head model)

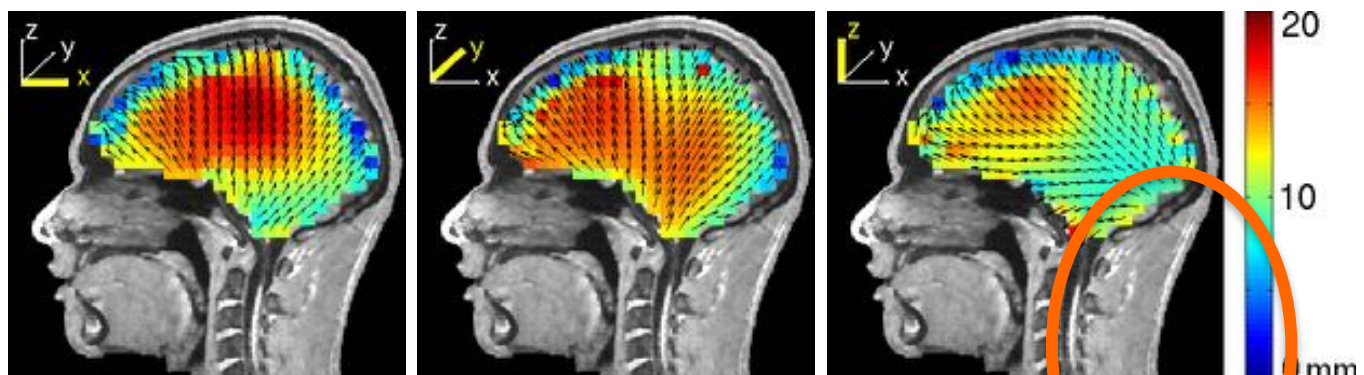


**Simulate 25**

↑ RLS<sub>25-4</sub>  
↓ wMNI<sub>80-4</sub>

**Assume 80**

(template  
head model)

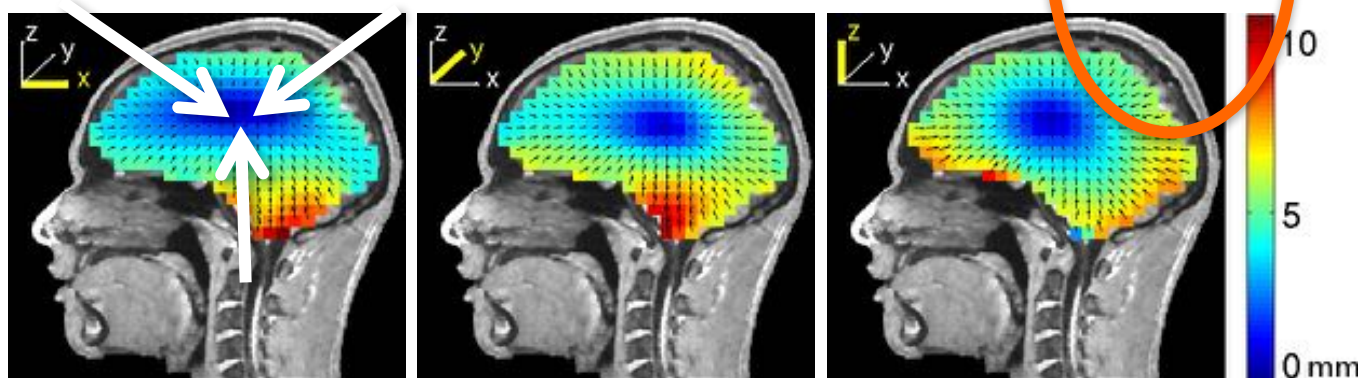


**Simulate 25**

↑ RLS<sub>25-4</sub>  
↓ RLS<sub>15-4</sub>

**Assume 15**

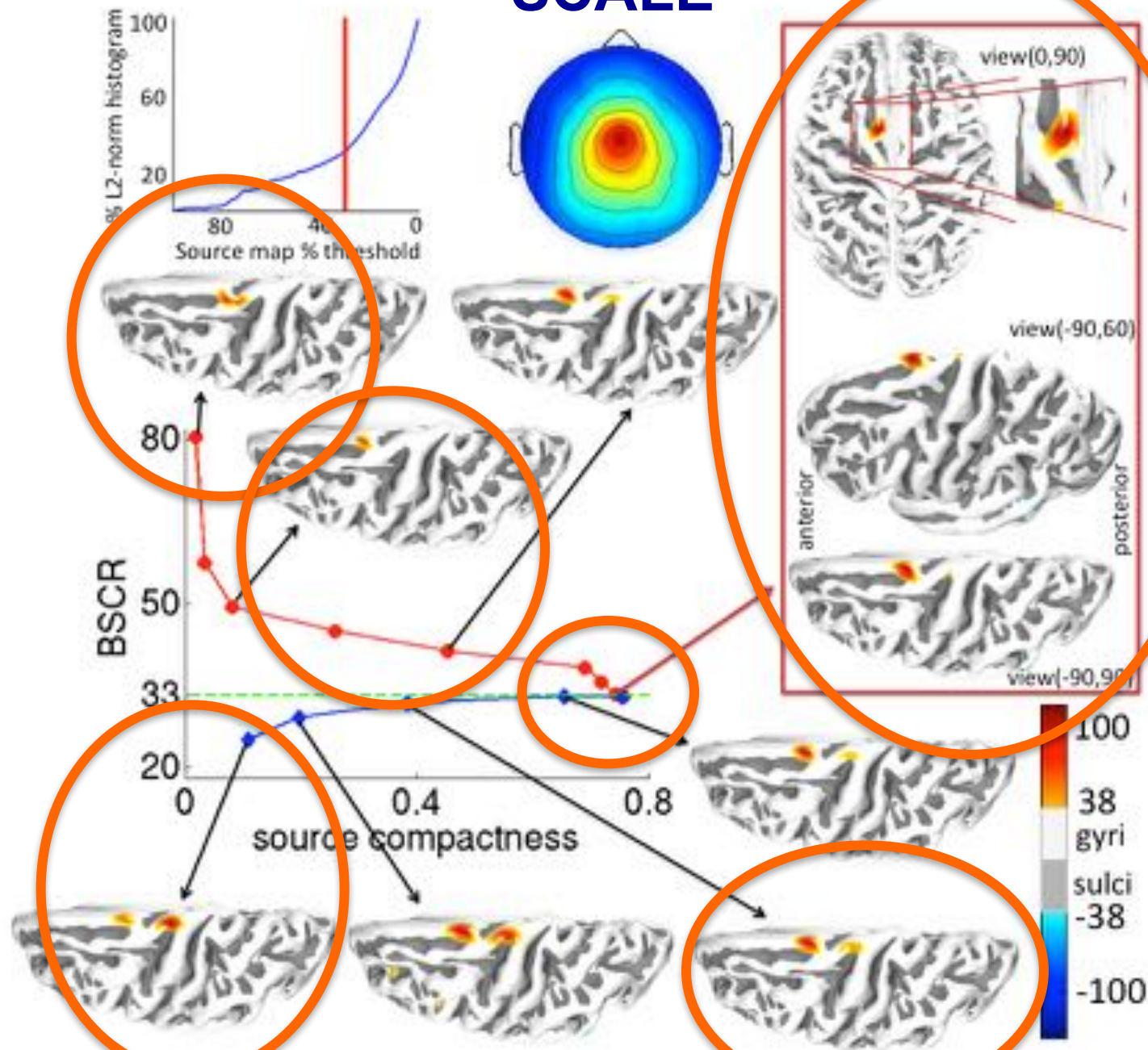
(template  
head model)

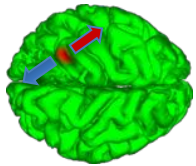


**Effects of Mis-Estimating Skull Conductivity**



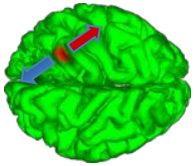
# SCALE





# Summary-1

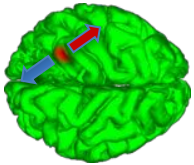
- An electromagnetic **forward head model** is required to interpret the sources of scalp maps
- Interpretation of scalp maps in terms of brain source distributions is “**inverse** source estimation”
- Mathematical techniques are available to aid in interpreting scalp maps as arising from particular 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-2

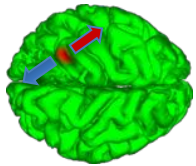
- **Inverse modeling**
  - Model assumption for volume conductor
  - Model assumption for source (i.e. dipole)
  - Additional assumptions on source
- **Single point-like sources**
- **Multiple point-like sources**
- **Distributed sources**
  - Different mathematical solutions
    - Dipole fitting (linear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





## Summary-3

- **If we have MRI of the subject**
  - Subject specific head model
  - Distributed source localization
- **If we don't have the MRI**
  - 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



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