

Forward and inverse models

Localizing sources using DIPFIT

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DIPFIT: localizing dipoles

- Motivation
- Ingredients
 - Source model
 - Volume conductor model
 - Analytical (spherical model)
 - Numerical (realistic model)
 - Considerations regarding EEG
- Inverse modeling
 - Single and multiple dipole fitting
 - Distributed source models

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Motivation

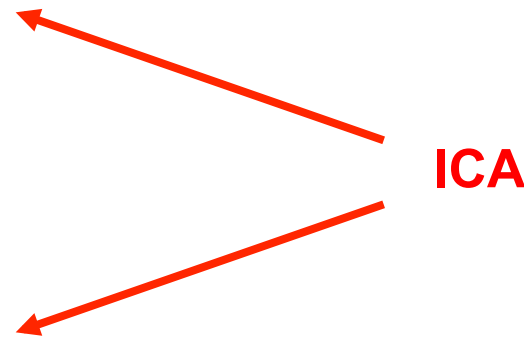
- Why fit dipoles?
- Why measure EEG?
- Why do ICA?
- Get extra information about brain processes
 - Time course of activity -----> **EEG**
 - Location of activity -----> **fMRI**

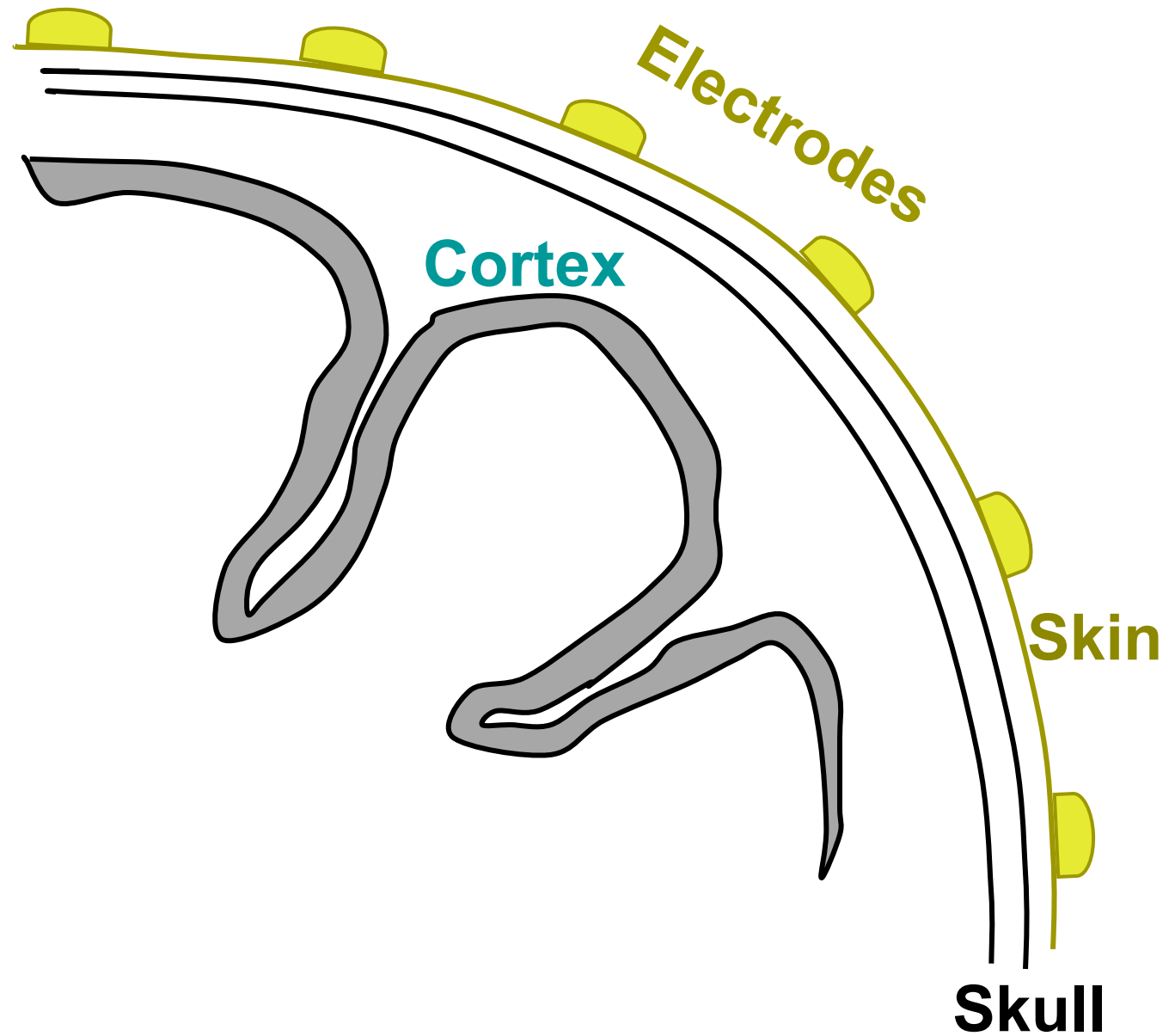
Difference between EEG and fMRI

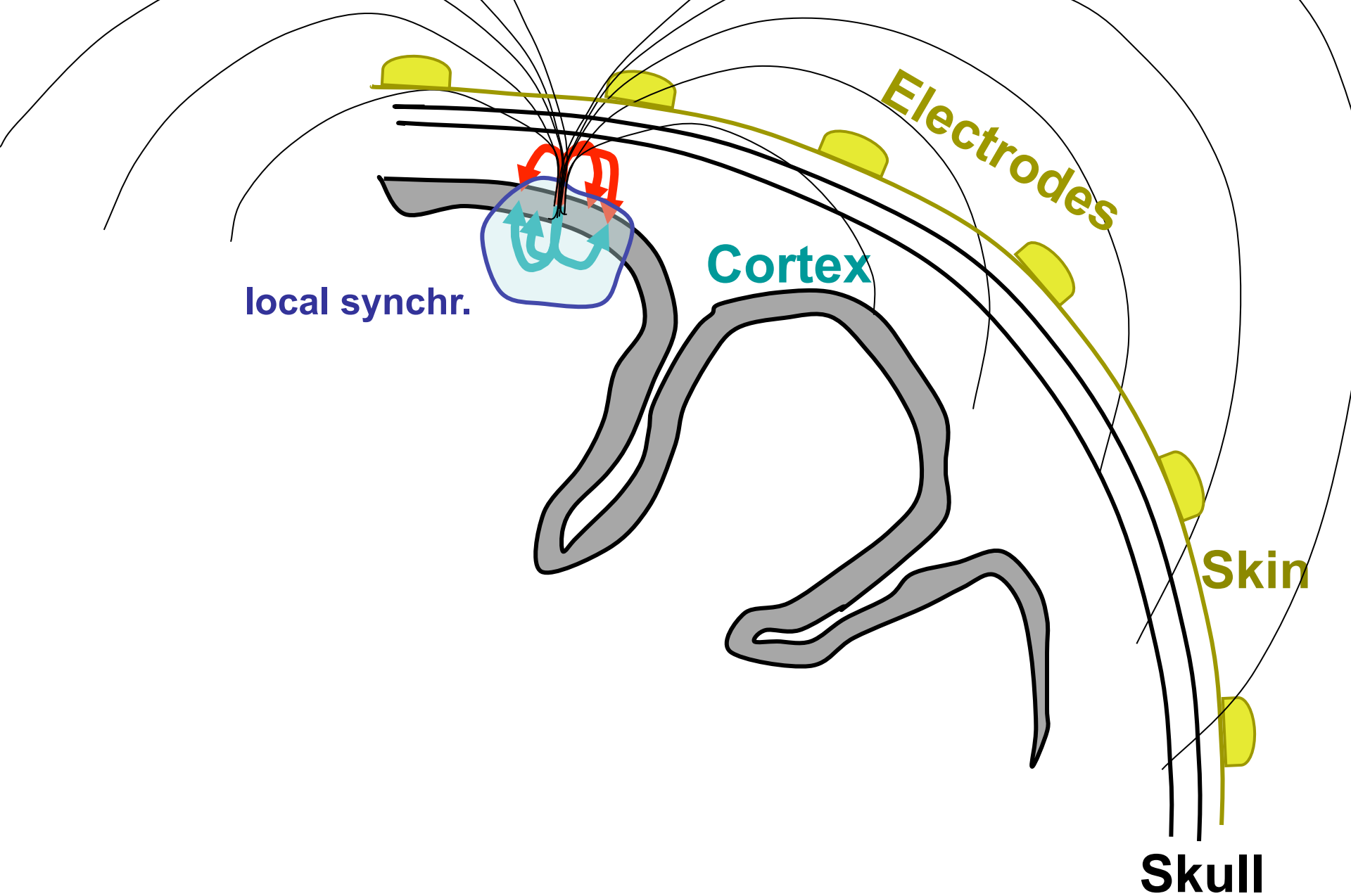
- EEG measures post-synaptic potentials
 - related to synchronized neuronal input
- fMRI measures BOLD
 - related to energy consumption
- Different characteristics in the time domain
- Different generators
- Timecourse *and* location

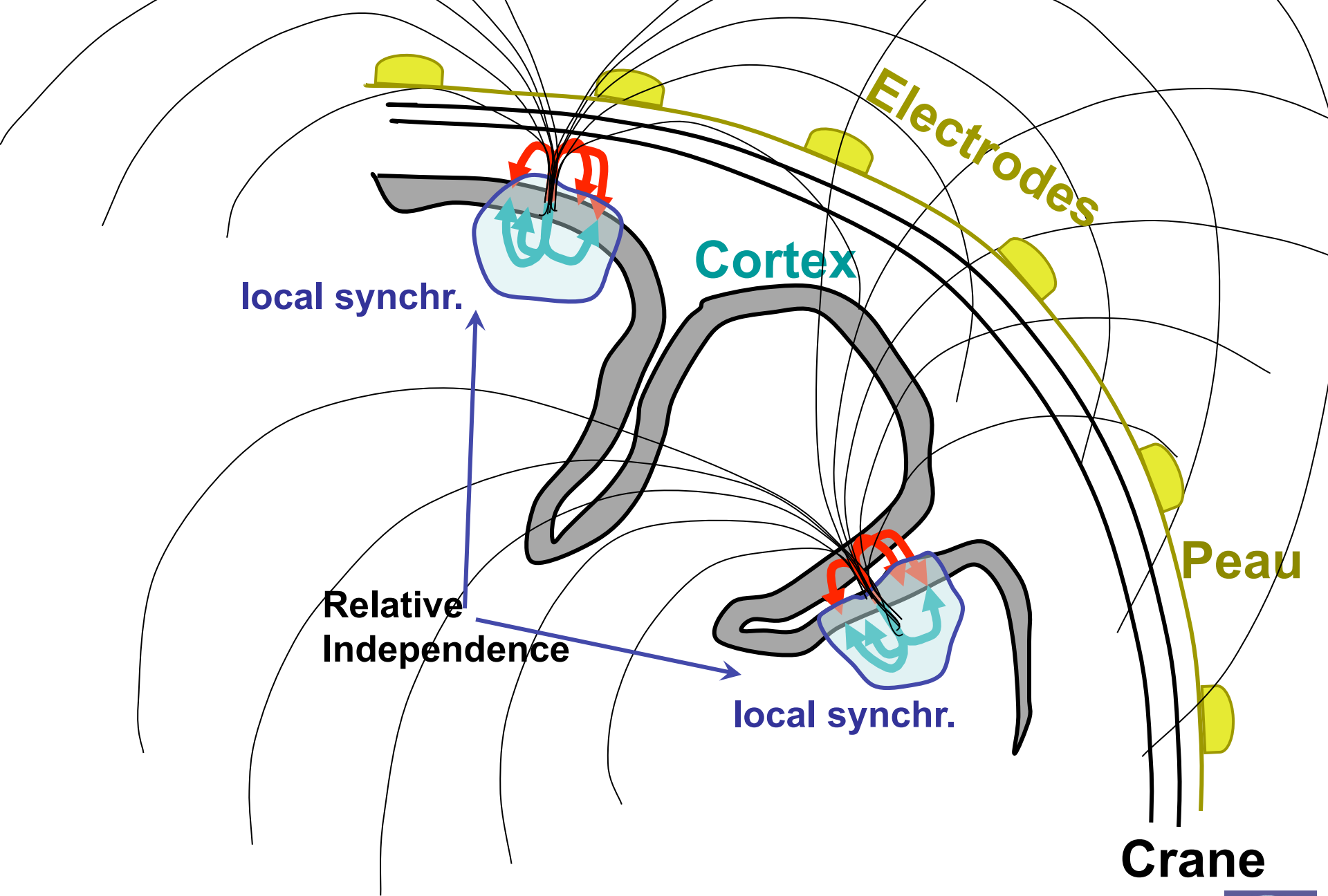
Why EEG: extra information

- Timecourse
 - ERSP
 - ERP
- Topography
 - Scalp distribution
 - Underlying generators





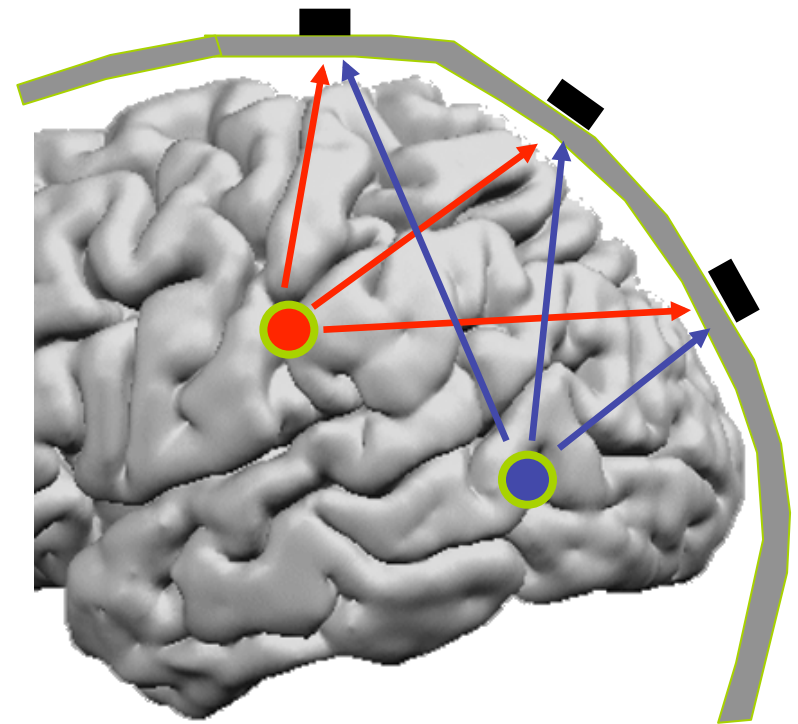
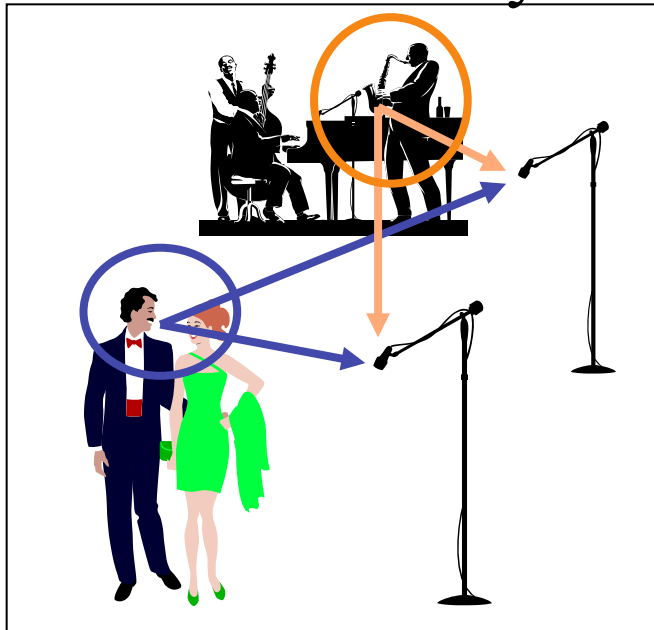




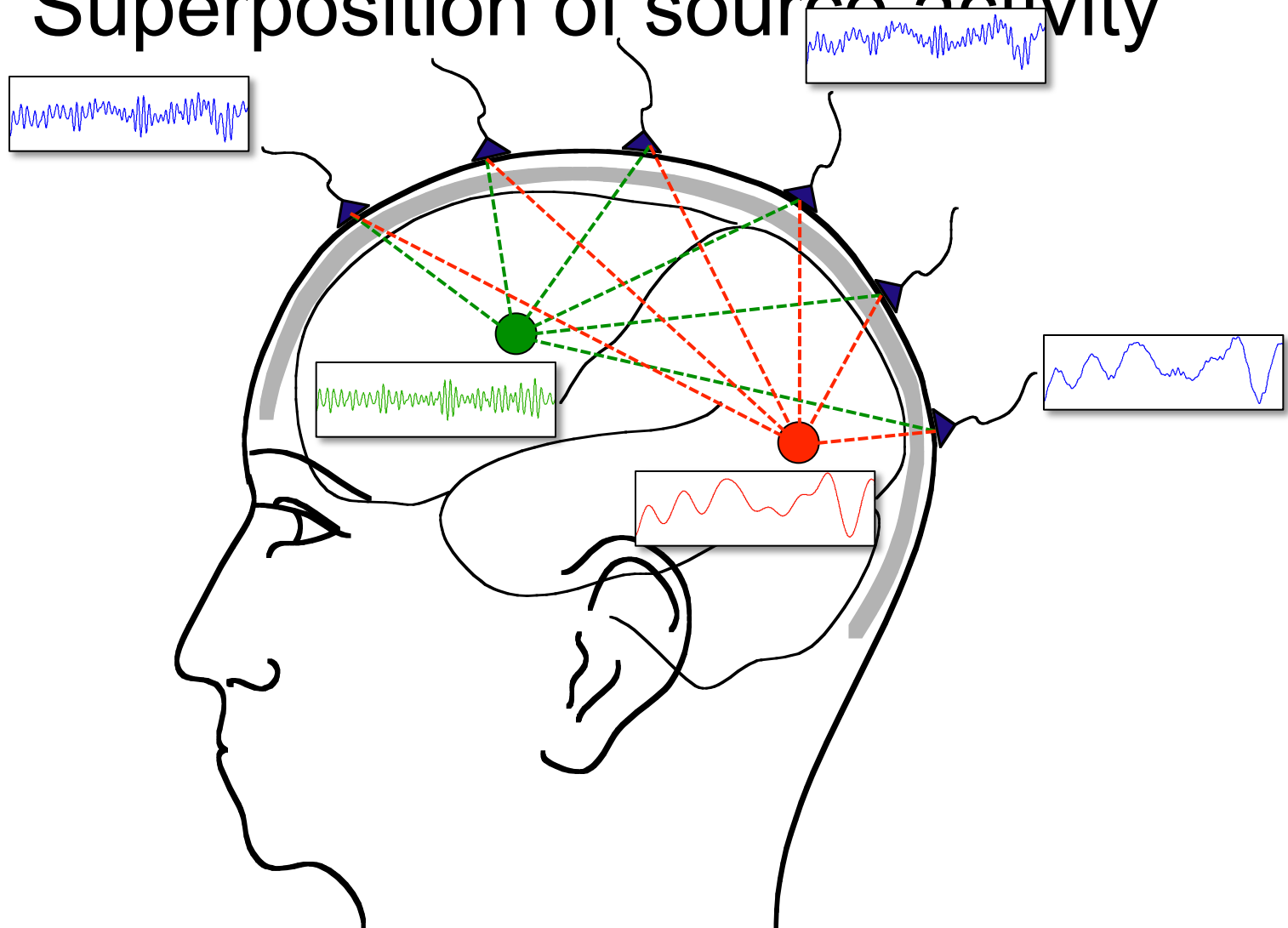
Independent component analysis

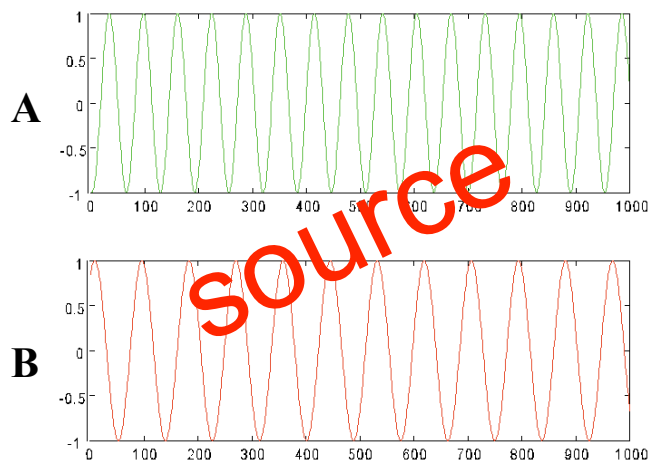
Mixture of Brain source activity

Cocktail Party

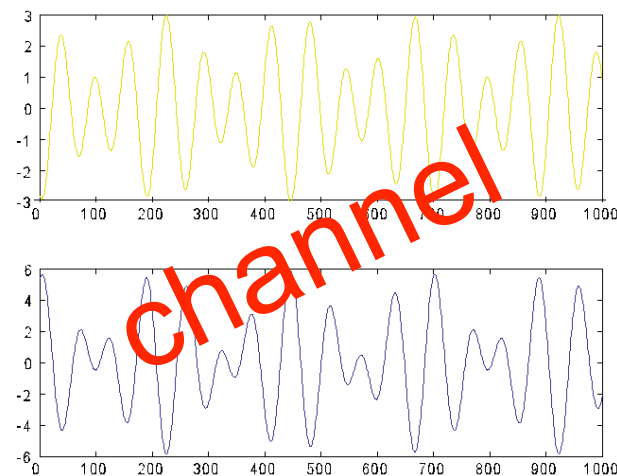
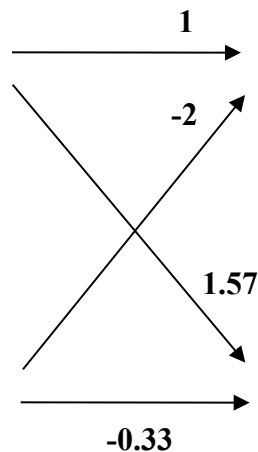


Superposition of source activity





source



channel

$$Y=[A;B]$$

Linear Combination

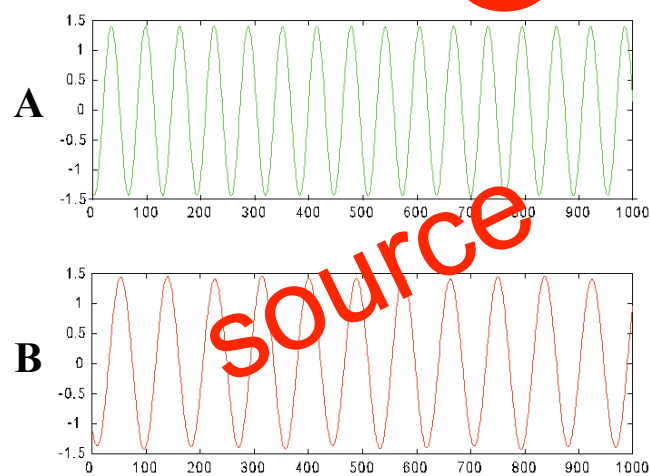
$$X=WY$$

ICA

ICA

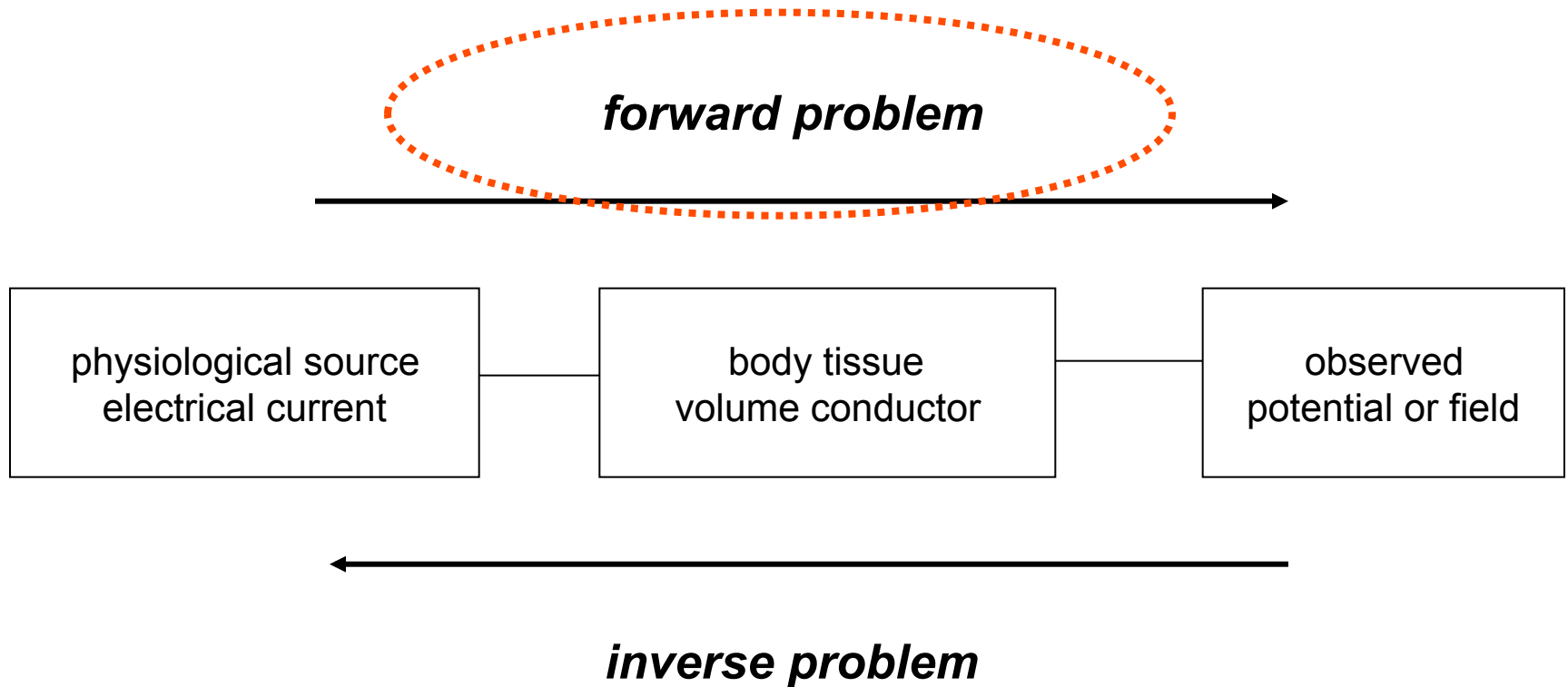


$$\tilde{Y}=W^{-1}\tilde{X}$$



source

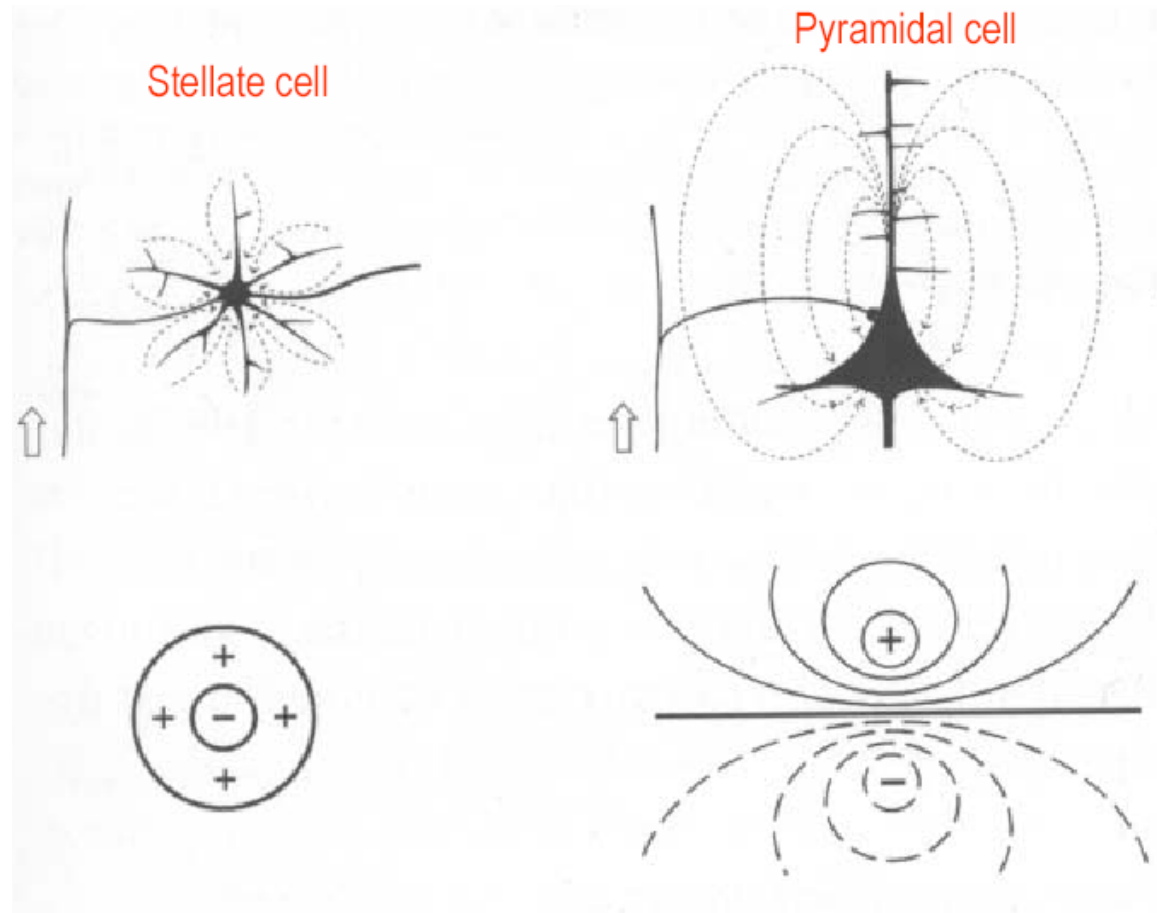
Source modelling



Overview

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Neuronal currents



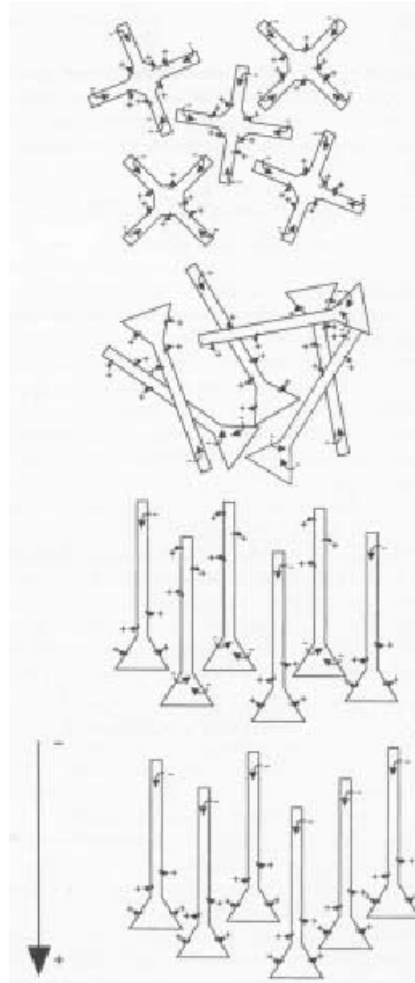
Symmetry, orientation and activation

radial symmetric

random oriented

asynchronously activated

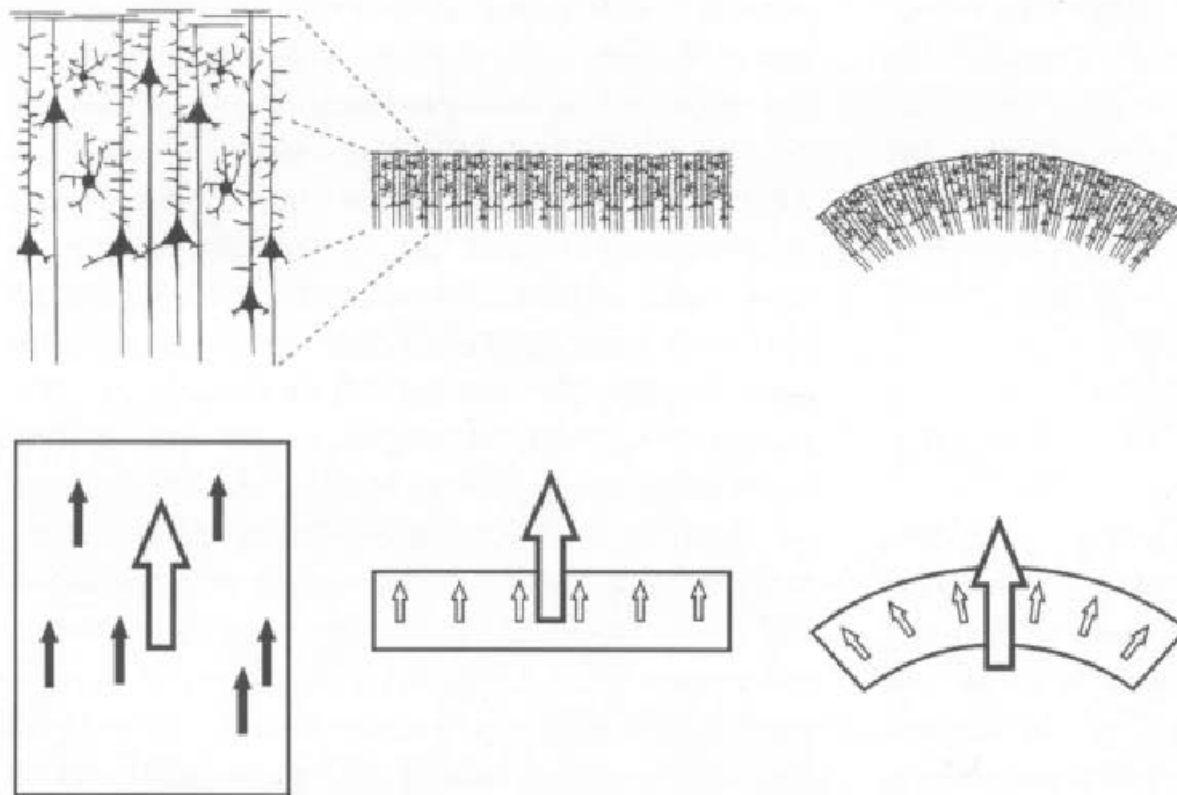
synchronously activated
parallel oriented



Motivation for current dipoles

- Neurophysiological motivation

Equivalent current dipoles



Motivation for current dipoles

- Neurophysiological motivation
- Physical/mathematical motivation
 - Any current distribution can be written as a multipole expansion
 - First term: monopole (must be zero)
 - Second term: dipole
 - Higher order terms: quadrupole, octupole

Motivation for current dipoles

- Neurophysiological motivation
- Physical/mathematical motivation
 - Any current distribution can be written as a multipole expansion
 - First term: monopole (must be zero)
 - Second term: dipole
 - Higher order terms: quadrupole, octupole
- Convenience
 - dipoles can be used as building block in distributed source models

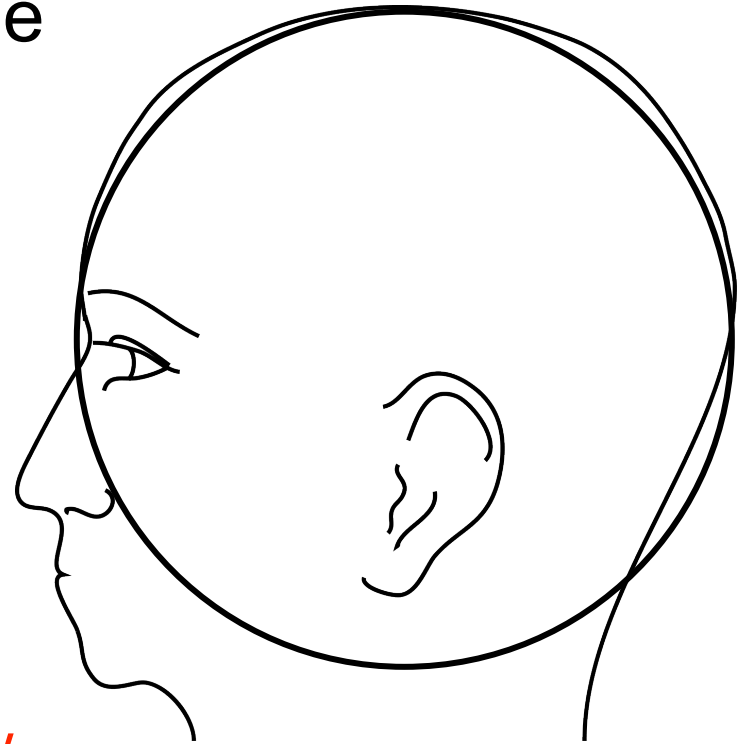
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Volume conductor

- electrical properties of tissue
- geometrical description
- spherical model
- realistic shaped model

→ Describes how the currents flow,
not where they originate from



Volume conductor

- Advantages spherical model
 - mathematically accurate
 - reasonably accurate
 - computationally fast
 - easy to use
- Disadvantages spherical model
 - inaccurate, esp. in some regions
 - difficult alignment with anatomy

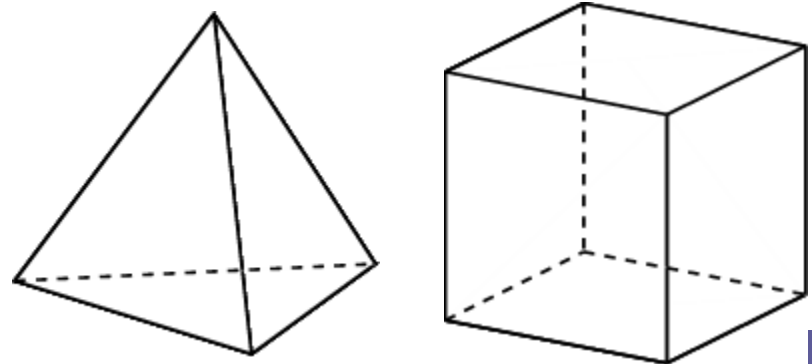
Volume conductor

- Advantages realistic model
 - accurate solution for EEG
- Disadvantages realistic model
 - more work
 - individual anatomical MRI required
 - computationally slow(er)
 - numerically instable
 - difficult in interindividual comparison

→ The pragmatic solution is to use a standard realistic headmodel for EEG

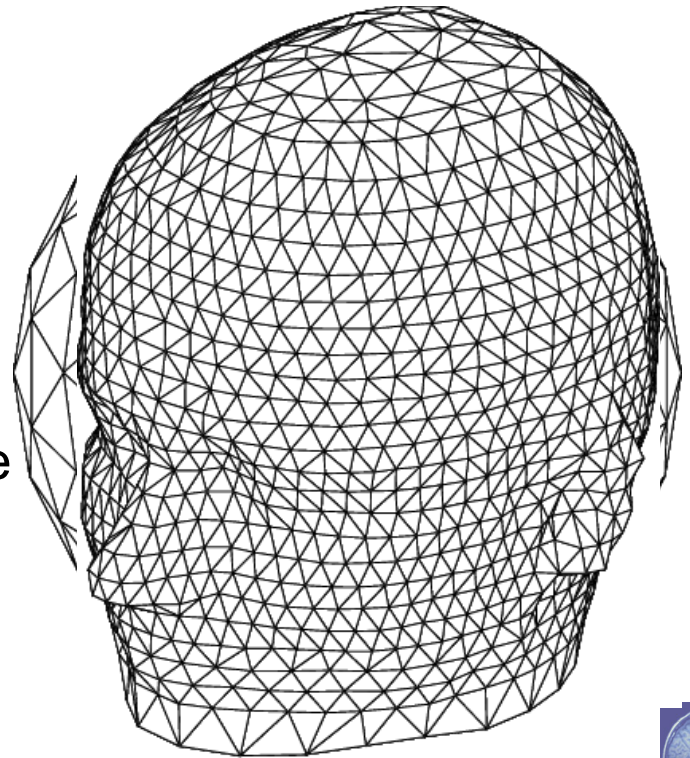
Realistic volume conductor

- Computational methods for volume conduction problem that allow realistic geometries
 - Boundary Element Method (BEM)
 - Finite Element Method (FEM)
- Geometrical description
 - triangles
 - tetraeders/voxels



Volume conductor: BEM

- Boundary Element Method
 - description of geometry by compartments
 - each compartment is
 - homogenous
 - isotropic
 - important tissues
 - skin
 - skull
 - brain
 - (CSF)
 - triangulated surfaces as boundaries
 - surfaces should be closed

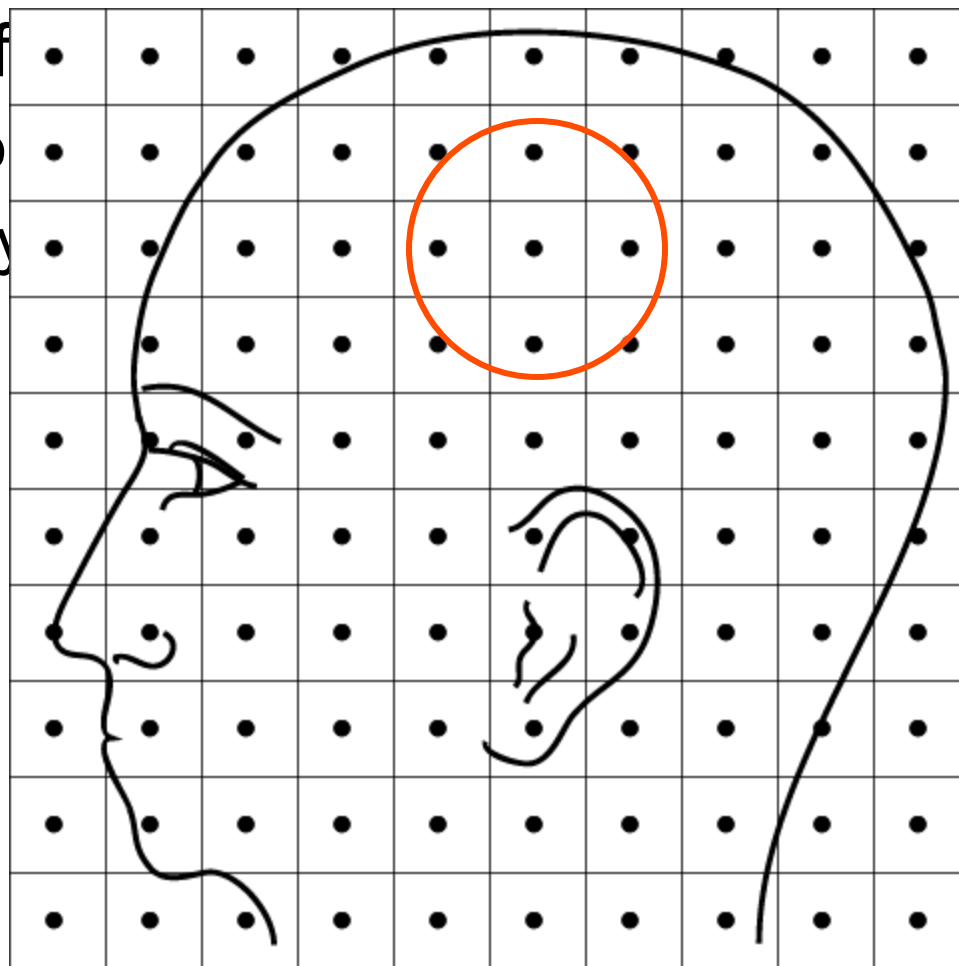


Volume conductor: FEM

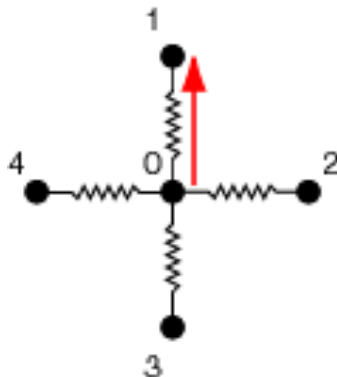
- Tesselation of 3D volume in tetraeders
- Large number of elements
- Each tetraeder can have its own conductivity
- FEM is most accurate numerical method
- Computationally expensive
- Accurate conductivities are not (well) known

Volume conductor: FDM

- Finite Dif
 - easy to
 - not very



Volume conductor: FDM



$$\left. \begin{array}{l} I_1 + I_2 + I_3 + I_4 = 0 \\ V = I * R \end{array} \right\} \Rightarrow$$

$$\Delta V_1 / R_1 + \Delta V_2 / R_2 + \Delta V_3 / R_3 + \Delta V_4 / R_4 = 0 \Rightarrow$$

$$(V_1 - V_0) / R_1 + (V_2 - V_0) / R_2 + (V_3 - V_0) / R_3 + (V_4 - V_0) / R_4 = 0$$

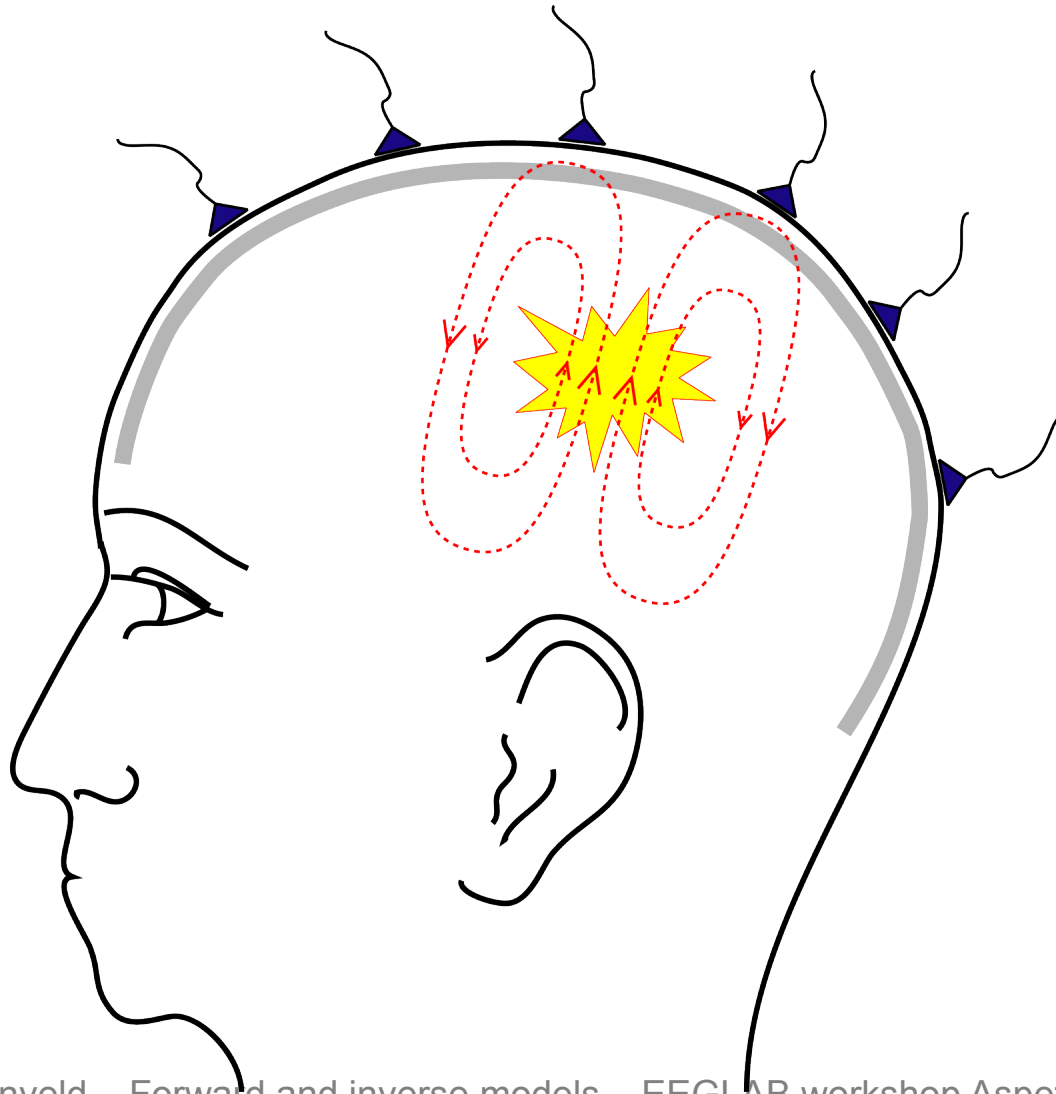
Volume conductor: FDM

- unknown potential V_i at each node
- linear equation for each node
 - approx. $100 \times 100 \times 100 = 1.000.000$ linear equations
 - just as many unknown potentials
- add a source/sink
 - sum of currents is zero for all nodes, except
 - sum of current is I_+ for a certain node
 - sum of current is I_- for another node
- solve system of linear equations for the potential

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EEG volume conduction



EEG volume conduction

- Potential difference between electrodes corresponds to current flowing through skin
- Only tiny fraction of current passes through skull
- Therefore the model should describe skull and skin as accurately as possible
- Challenges with skull
 - Not visible in anatomical MRI
 - Thickness varies
 - Conductivity is not homogeneous
 - Complex geometry at base of skull

MEG volume conduction

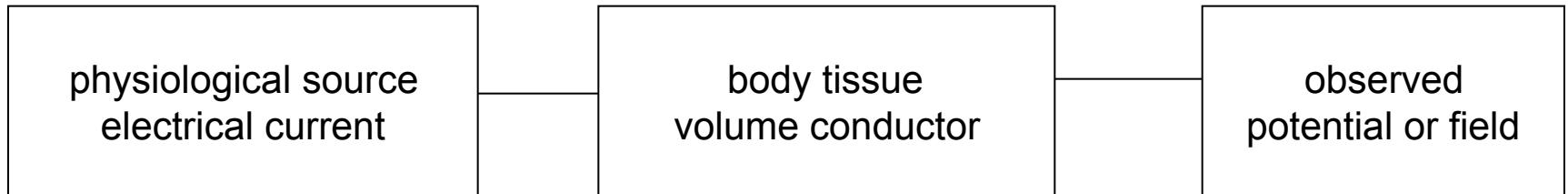
- Only tiny fraction of current passes through the poorly conductive skull
- Therefore skull and skin **can be neglected** in the MEG model
- However
 - difficulties with head movements
 - not all sources are equally visible
 - expensive and not widely available

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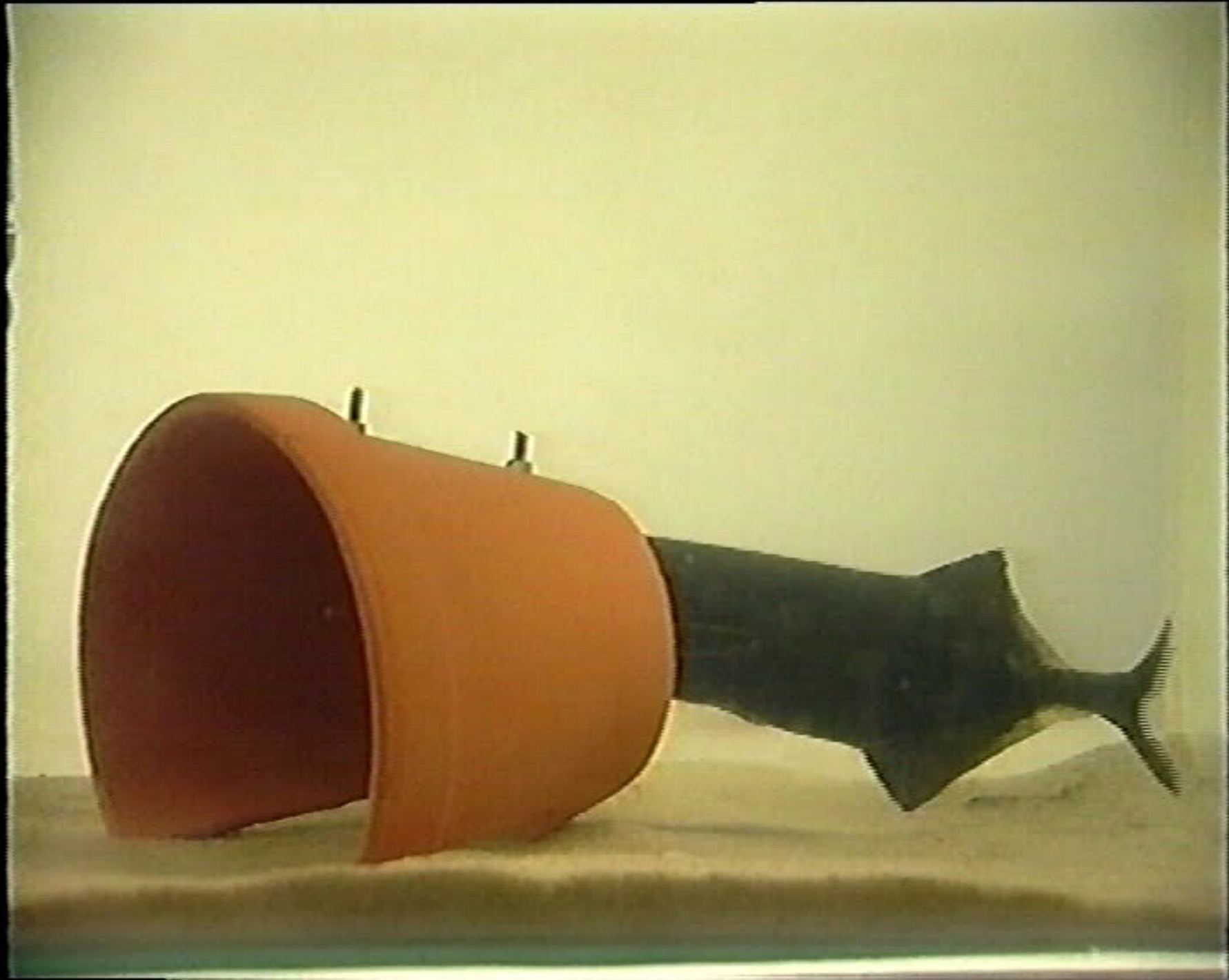
Source modelling

forward problem



inverse problem





Inverse methods

- Single and multiple dipole models
 - Minimize error between model and measured potential/field
- Distributed dipole models
 - Perfect fit of model to the measured potential/field
 - Minimize additional constraint on sources
 - LORETA (smoothness)
 - Minimum Norm (L2)
 - Minimum Current (L1)
- Spatial filtering
 - Scan whole brain with single dipole and compute the filter output at every location
 - MUSIC
 - Beamforming (e.g. LCMV, SAM, DICS)

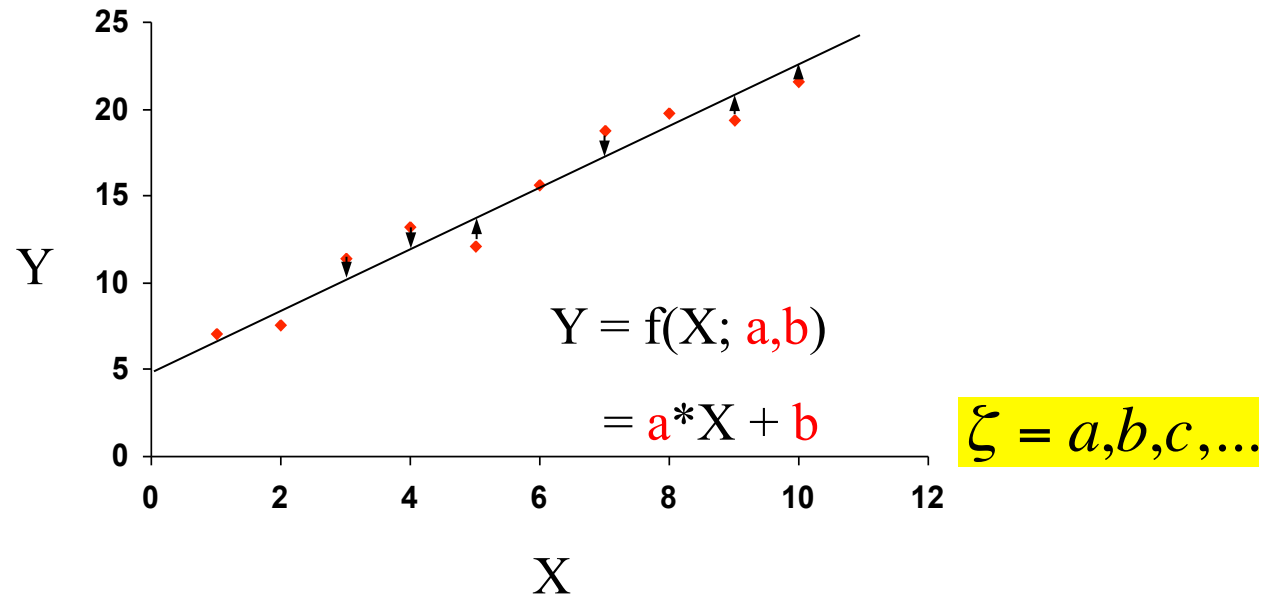
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Single or multiple dipole models

- Manipulate source parameters to minimize error between measured and model data
 - Location of each source
 - Orientation of each source
 - Strength of each source
- Orientation and strength together correspond to the “dipole moment” and can be estimated linearly
- Position is estimated non-linearly
- Source **parameter estimation**

Parameter estimation



Parameter estimation: model

measured potential

$$V_i = V(\vec{r}_i) + \text{noise}$$

forward model for the data

$$Y_i = Y(r_i; \xi) + \text{noise}$$

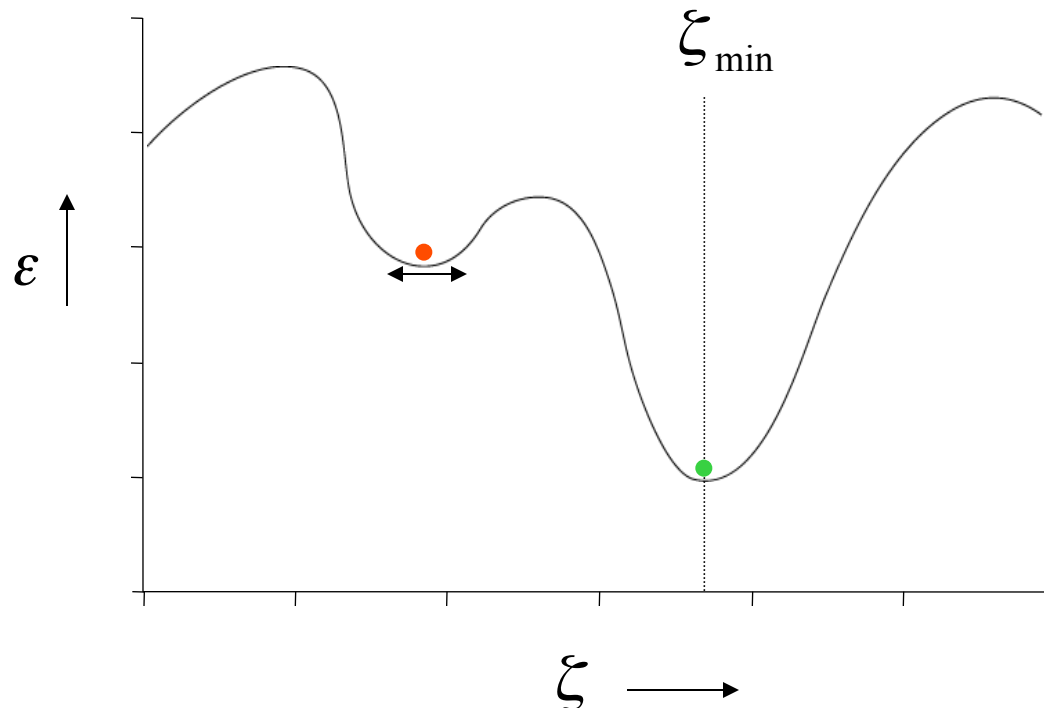
select “optimal” model

$$\min_{\xi} \left\{ \sum_{i=1}^N (Y(r_i; \xi) - V(r_i))^2 \right\}$$

Select optimal model

$$error(\zeta) = \sum_{i=1}^N (Y_i(\zeta) - V_i)^2 \Rightarrow \min_{\zeta}(error(\zeta))$$

$$\zeta = a, b, c, \dots$$



Dipole *scanning*: grid search

- define grid with allowed dipole locations
 - compute optimal dipole moment for each location
 - compute value of goal-function
 - plot value of goal-function on grid
-
- number of evaluations:
 - single dipole, 1 cm grid: ~4 000
 - single dipole, $\frac{1}{2}$ cm grid: ~32 000
 - two dipoles, 1cm grid: ~16 000 000

Dipole *fitting*: nonlinear search

- start with an initial guess
 - evaluate the local derivative of goal-function
 - “walk down hill” to the most optimal solution
-
- number of evaluations: ~100

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Distributed source model

- Position of the source is **not estimated** as such
 - Pre-defined grid (3D volume or on cortical sheet)
- Strength is estimated
 - In principle easy to solve, however...
 - More “unknowns” (parameters) than “knowns” (measurements)
 - Infinite number of solutions can explain the data perfectly
 - Additional constraints required
 - Linear estimation problem

Distributed source model

- Linear estimation

$$\vec{\Psi} = q_1 \vec{\Psi}_1 + q_2 \vec{\Psi}_2 + \dots = \begin{bmatrix} \Psi_{1,1} & \Psi_{2,1} & \dots \\ \Psi_{1,2} & \Psi_{2,2} & \dots \\ \vdots & \vdots & \ddots \\ \Psi_{1,N} & \Psi_{2,N} & \dots \end{bmatrix} \cdot \begin{bmatrix} q_1 \\ q_2 \\ \vdots \end{bmatrix} = L \cdot \vec{q}$$

$$\vec{q} = L^{-1} \cdot \vec{\Psi}$$

Distributed source model

$$V = L \cdot q + \text{Noise}$$

$$\min_q \{ \|V - L \cdot q\|^2 \} = 0 \quad !!$$

- Regularized linear estimation:

$$\min_q \{ \|V - L \cdot q\|^2 + \lambda^2 \cdot \|D \cdot q\|^2 \}$$

- Constrained linear estimation:

$$\min_q \{ q^T \cdot W \cdot q \} \quad \text{while} \quad \|V - L \cdot q\|^2 = 0$$

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Spatial filtering

- position of the source as such is not estimated
- scanning the whole brain
 - single dipole as source
 - estimate activity at each grid location
 - that explains a part of the data
 - that suppresses other activity
- various methods
 - multiple signal classification (MUSIC)
 - beamforming
 - LCMV, SAM, DICS, ...
- not a distributed source model, but a distributed representation of the single dipole estimate

Spatial filtering

- unmixing of data into “signal” and “noise” sources
- requires assumptions on
 - temporal relation between sources and
 - biophysical model
- with ICA we have already separated the timeseries and we only have spatial topographies that need to be explained
- hence spatial filtering (beamforming etc.) are incompatible with ICA

Summary 1

- Forward modelling
 - Required for the interpretation of scalp topographies
 - Interpretation of scalp topography *is* “source estimation”
 - Mathematical techniques are available that aid in interpreting scalp topographies -> inverse modeling

Summary 2

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

