



Forward and inverse modelling and the EEGLAB dipfit tools

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Overview

Motivation and background

Forward modeling

- Source model

- Volume conductor model

Inverse modeling - general

- Single and multiple dipole fitting

- Distributed source models

- Beamforming methods

Inverse modeling - independent components

Summary

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Motivation 1

Strong points of EEG and MEG

Temporal resolution (~ 1 ms)

Characterize individual components of ERP

Oscillatory activity

Disentangle dynamics of cortical networks

Weak points of EEG and MEG

Measurement on outside of brain

Overlap of components

Low spatial resolution

Motivation 2

If you find a ERP/ERF component, you want to characterize it in physiological terms

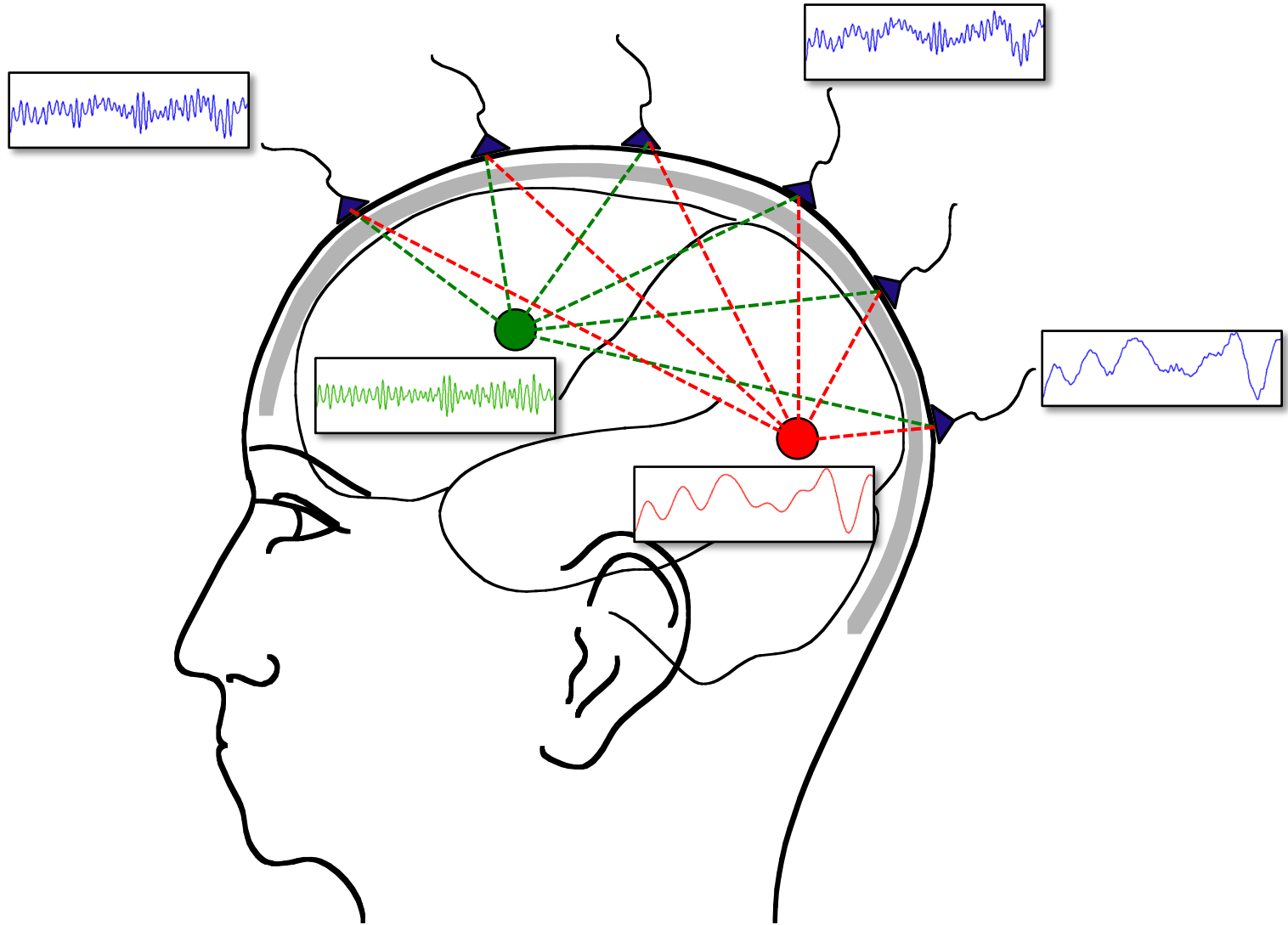
Time or frequency are the “natural” characteristics

“Cortical location” requires interpretation of the scalp topography

Forward and inverse modeling helps to interpret the topography

Forward and inverse modeling helps to disentangle overlapping source timeseries

Superposition of source activity



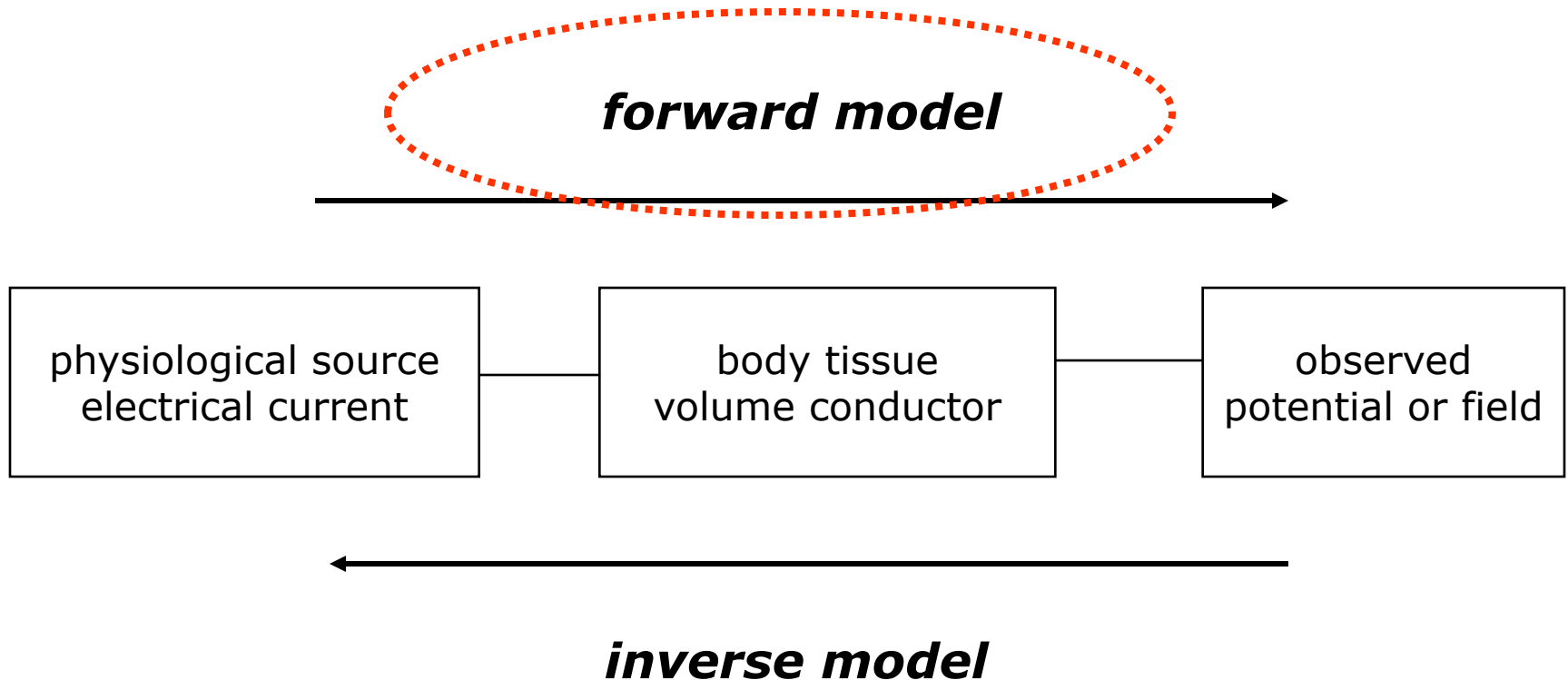
Superposition of source activity

Timecourse of each source contributes to each channel

The contribution of each source to each channel depends on its “visibility”

Activity on each channel is a **superposition** of all source activity

Biophysical source modelling: overview



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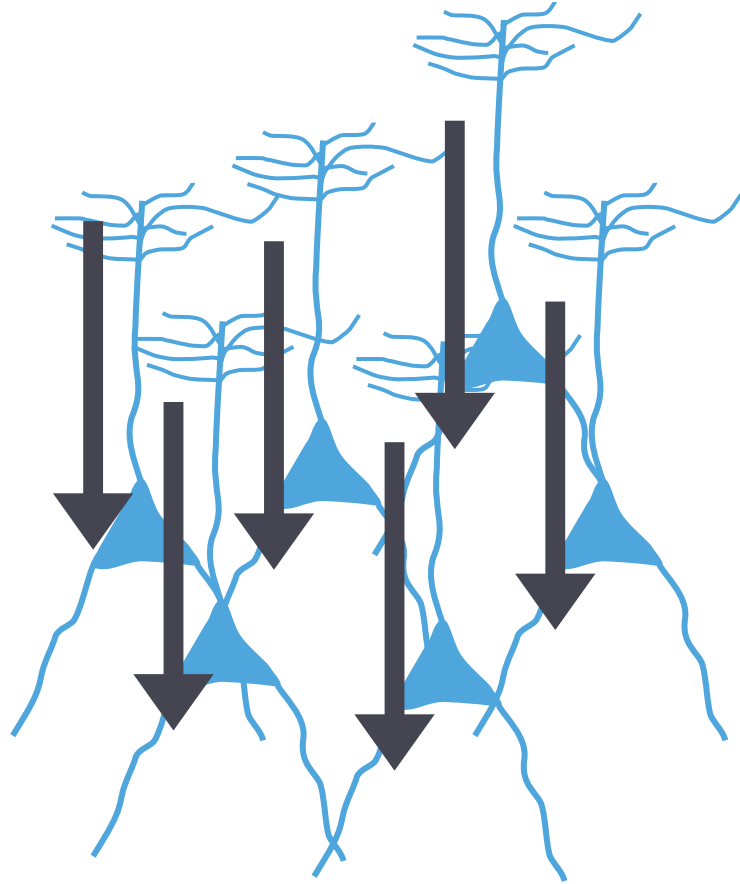
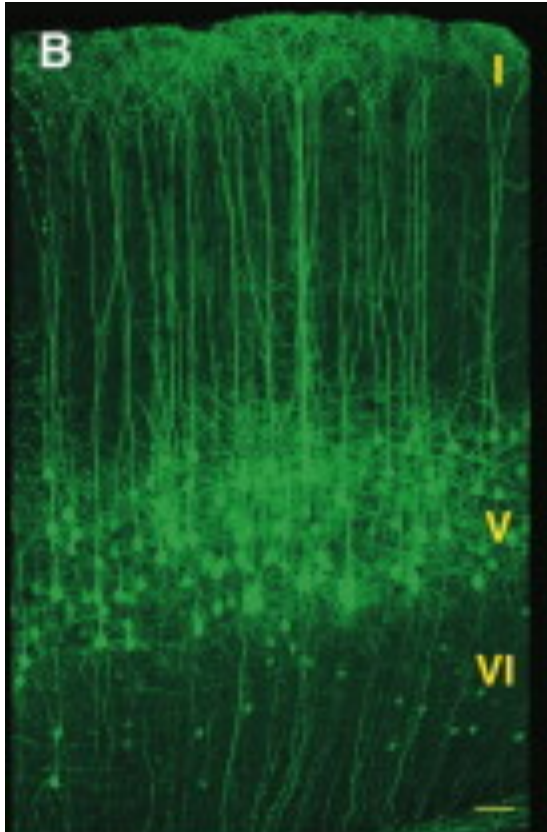
Distributed source models

Beamforming methods

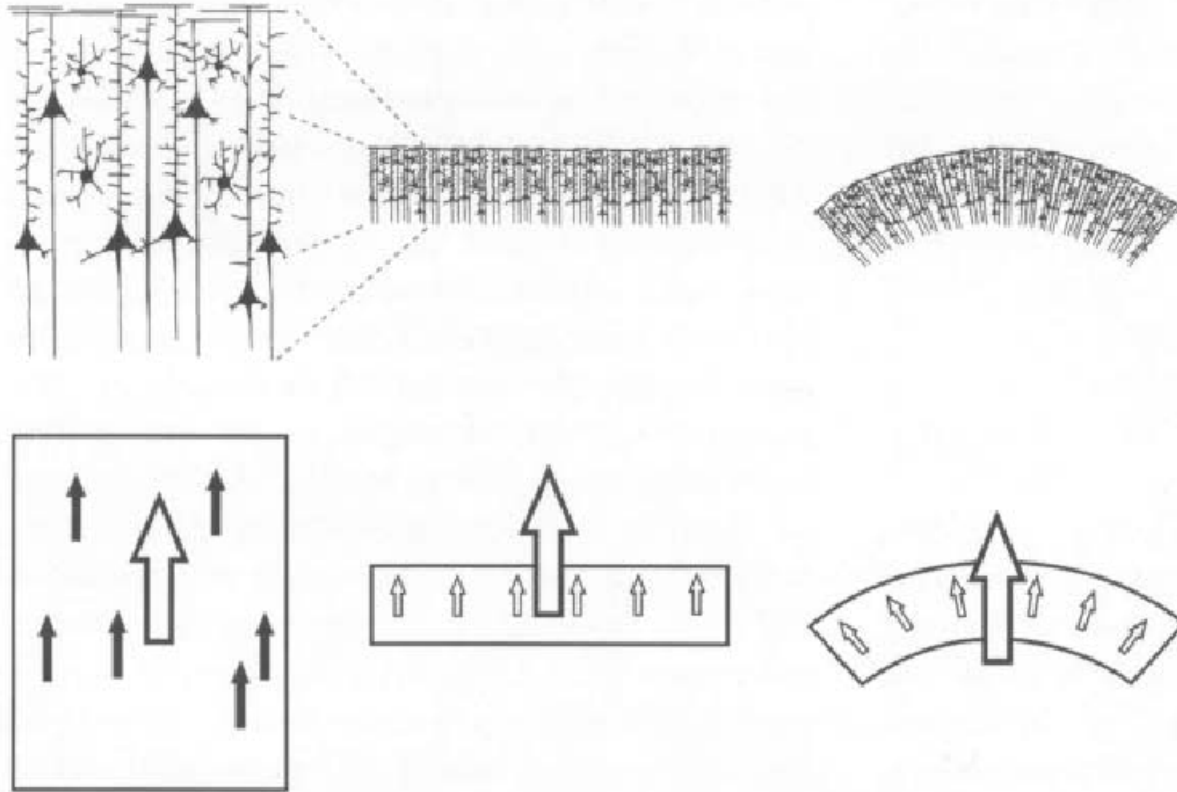
Inverse modeling - independent components

Summary

What produces the electric current



Equivalent current dipoles



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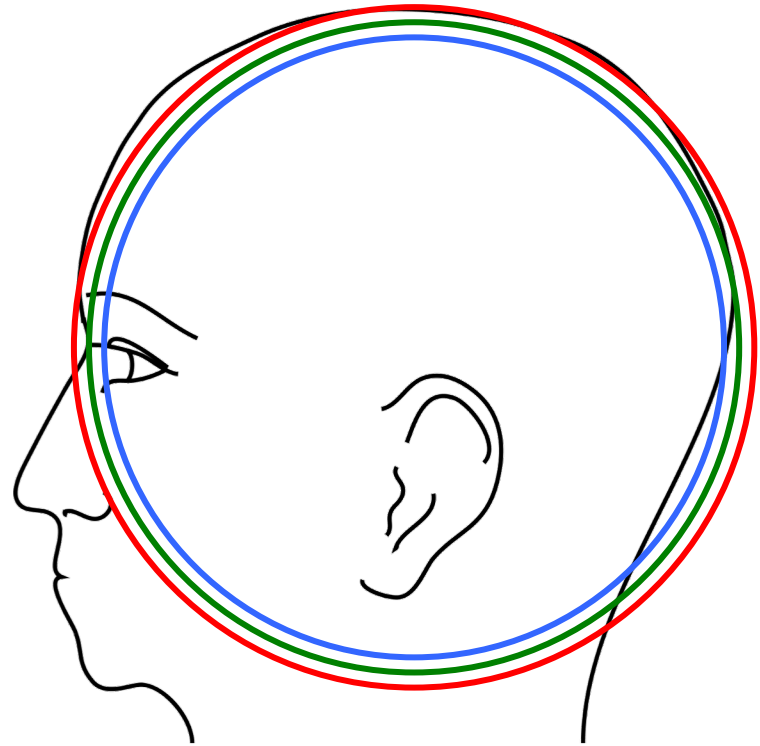
Volume conductor

described electrical properties of tissue

describes geometrical model of the head

describes **how** the currents flow, not where they originate from

same volume conductor for EEG as for MEG, but also tDCS, tACS, TMS, ...



Volume conductor

Computational methods for volume conduction problem that allow for realistic geometries

BEM *Boundary Element Method*

FEM *Finite Element Method*

FDM *Finite Difference Method*

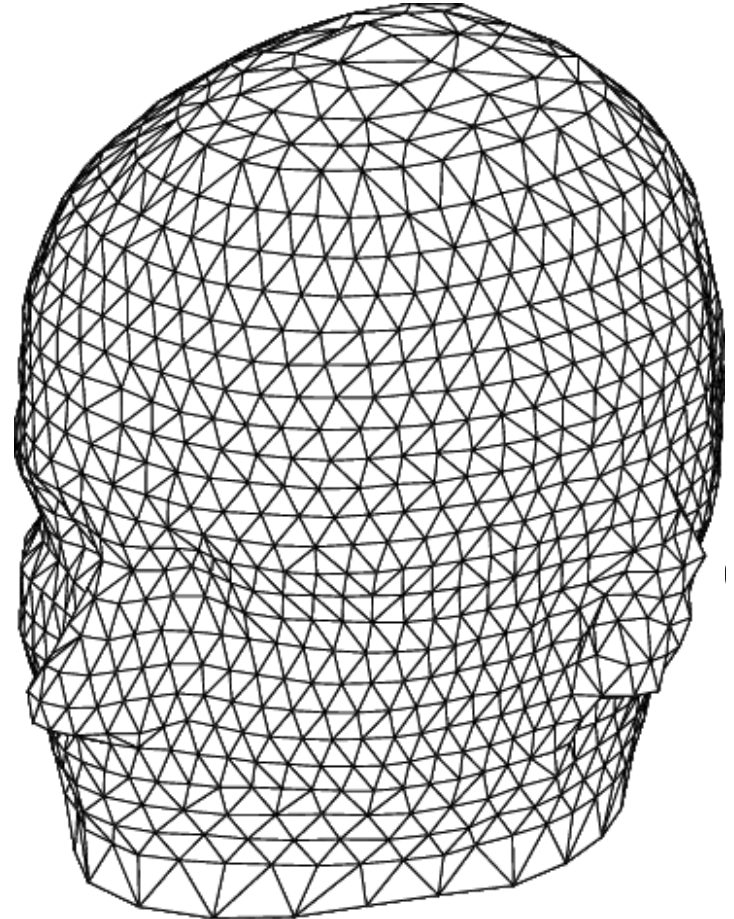
Volume conductor: Boundary Element Method

Each compartment is
homogenous
isotropic

Important tissues

skin
skull
brain
(CSF)

Triangulated surfaces
describe boundaries



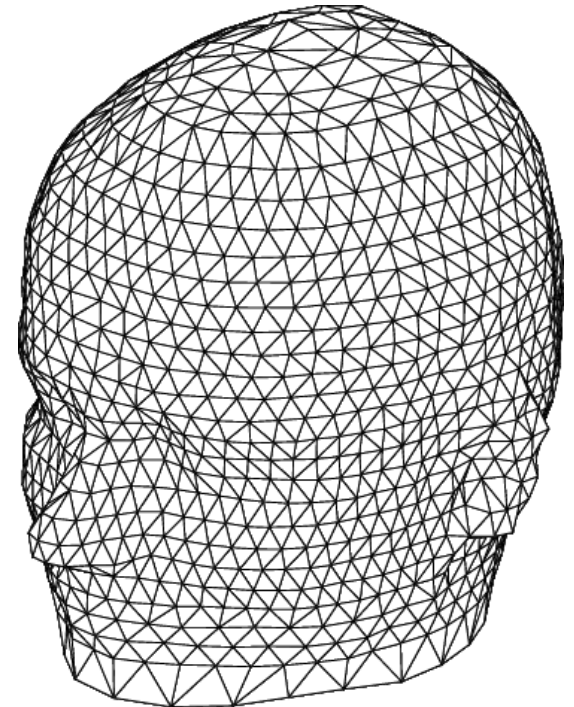
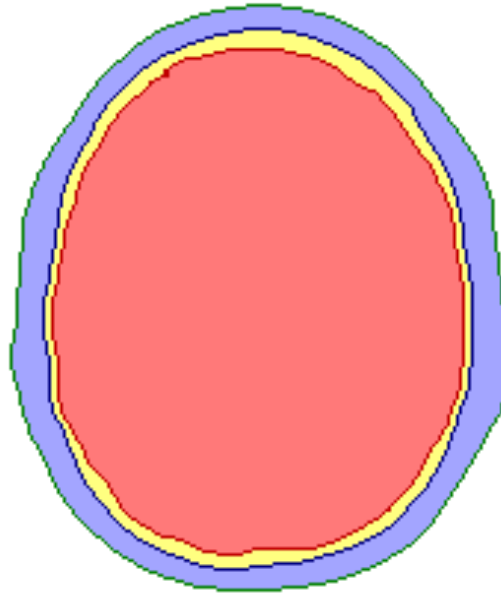
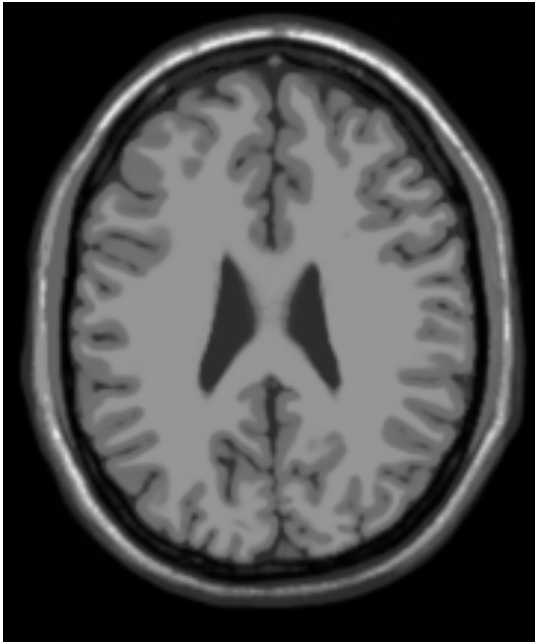
Volume conductor: Boundary Element Method

Construction of geometry

segmentation in different tissue types

extract surface description

downsample to reasonable number of triangles



Volume conductor: Boundary Element Method

Construction of geometry

- segmentation in different tissue types

- extract surface description

- downsample to reasonable number of triangles

Computation of model

- independent of source model

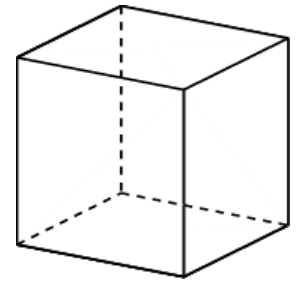
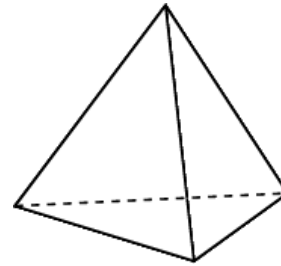
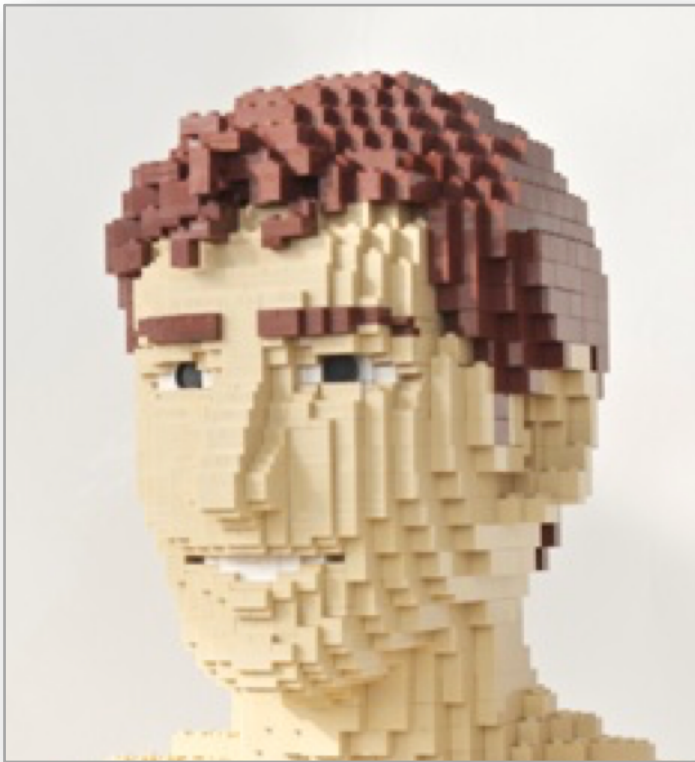
- only one lengthy computation

- fast during application to real data

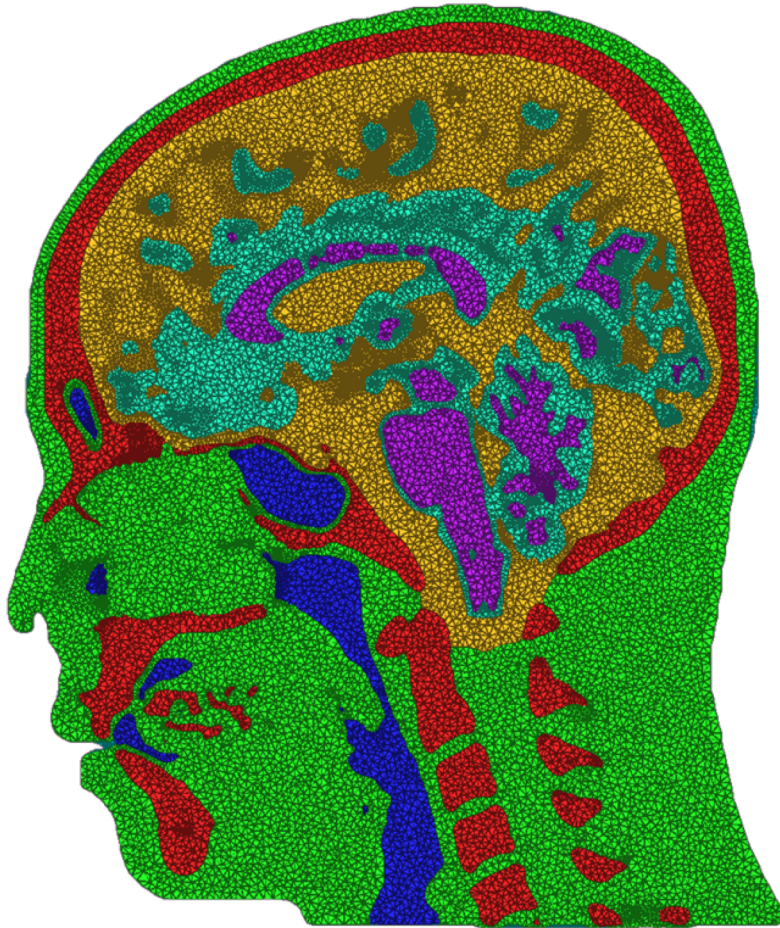
Can (almost) be arbitrary complex

Volume conductor: Finite Element Method

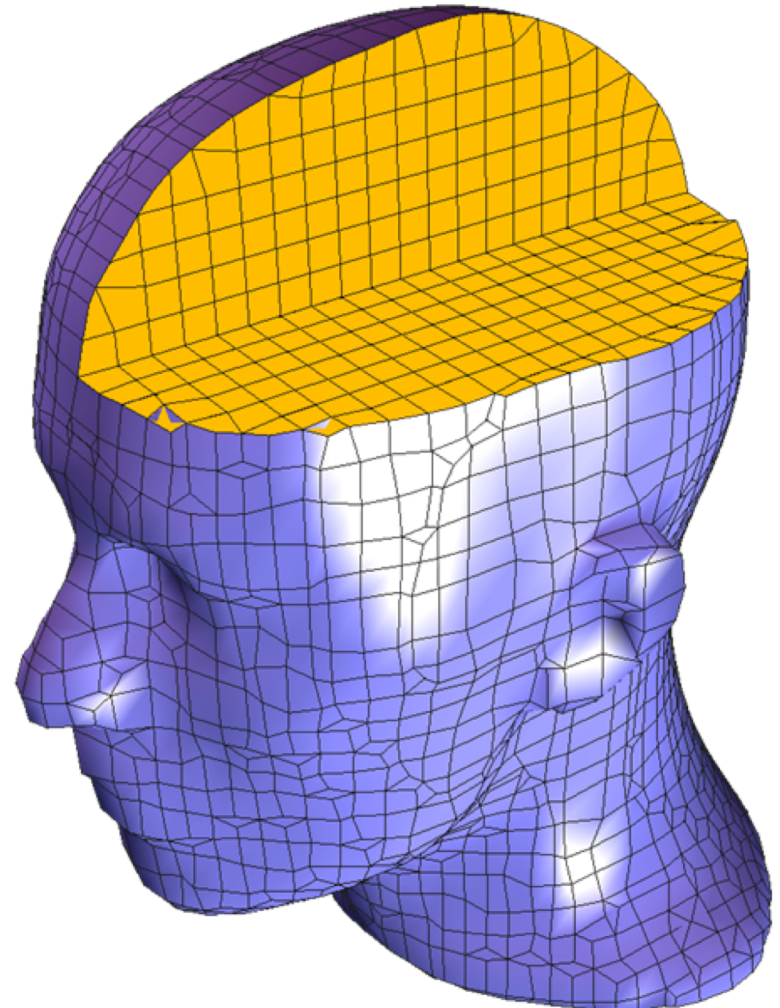
Tesselation of 3D volume in tetraeders or hexaheders



Volume conductor: Finite Element Method



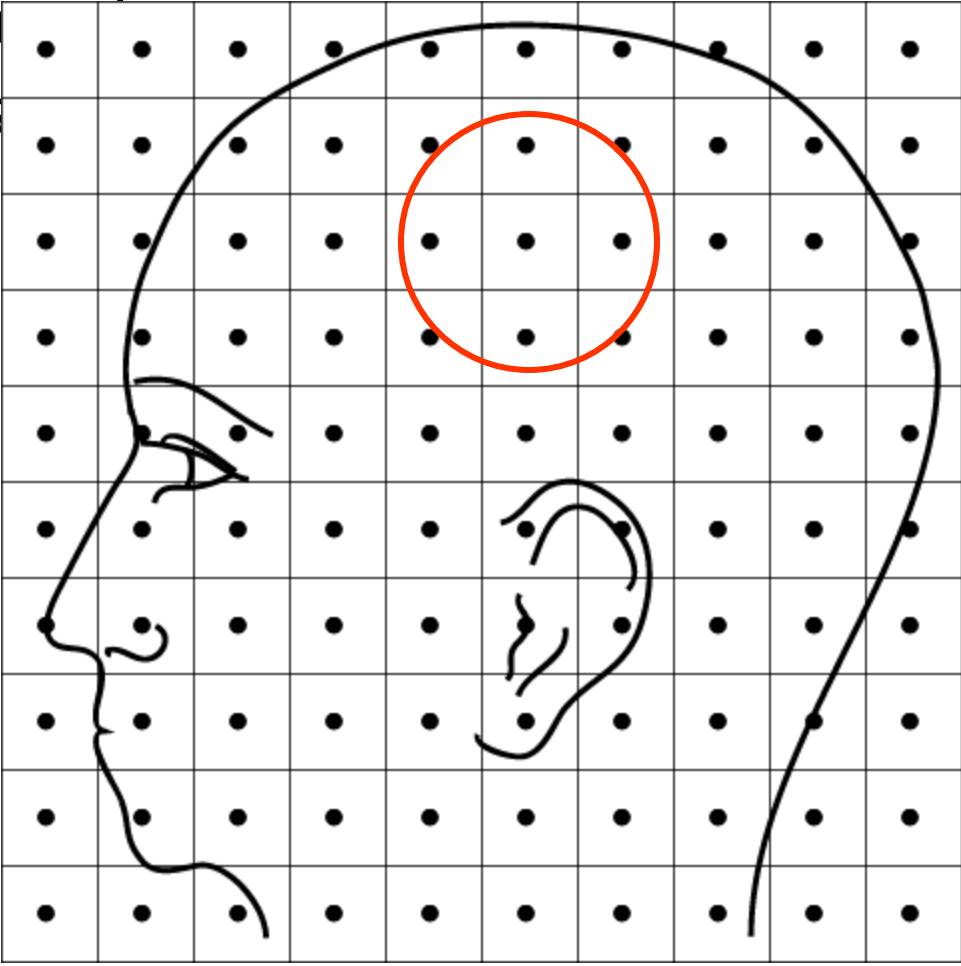
tetraeders



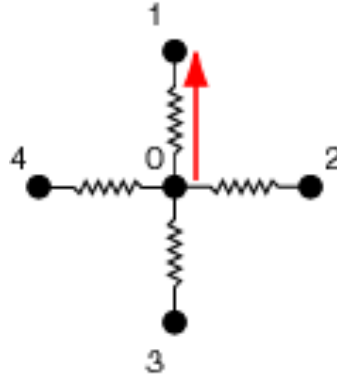
hexaheders

Volume conductor: Finite Difference Method

Easy to con
Not very us



Volume conductor: Finite Difference Method



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

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

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

Volume conductor: Finite Difference Method

Unknown potential V_i at each node

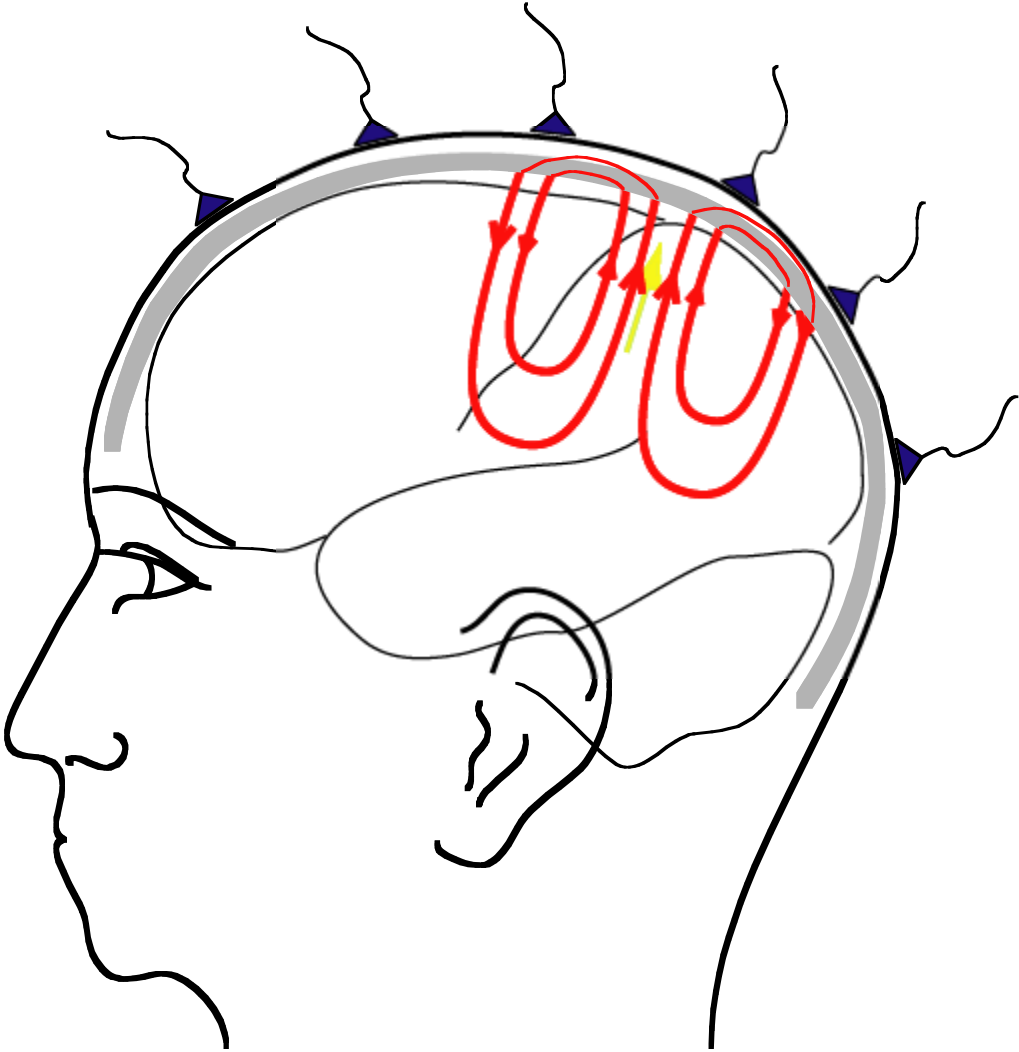
Linear equation for each node

approx. $100 \times 100 \times 100 = 1.000.000$ linear equations
just as many unknown potentials

Inject some current $+I$ and $-I$ at two of the nodes

Solve for unknown potential

EEG volume conduction



EEG volume conduction

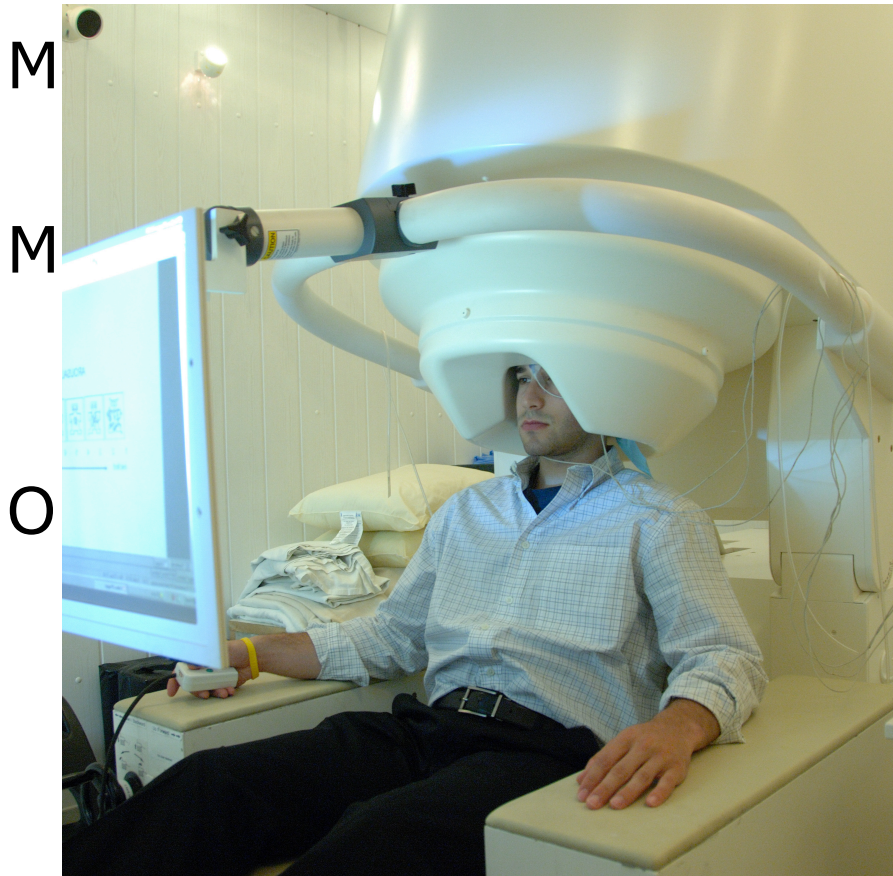
Potential difference between electrodes
corresponds to current flowing through skin

Only tiny fraction of current passes through skull

Therefore the model should describe the skull and
skin **as accurately as possible**

MEG volume conduction

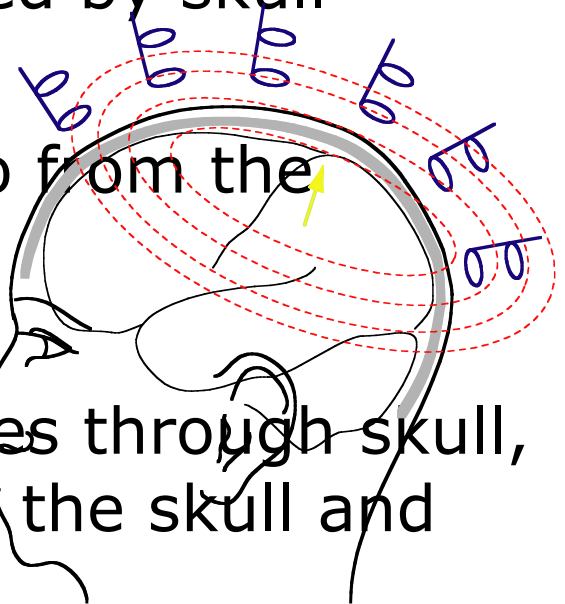
MEG measures magnetic field over the scalp



Magnetic field is distorted by skull

but also from the

currents that pass through skull, ignore the skull and



Practical differences between EEG and MEG

fixed sensor positions in MEG

flexible cap in EEG

MEG requires head size to be known in analysis

using individual anatomical MRI

position of sensors is accurately known

EEG requires the electrode positions to be known
in analysis

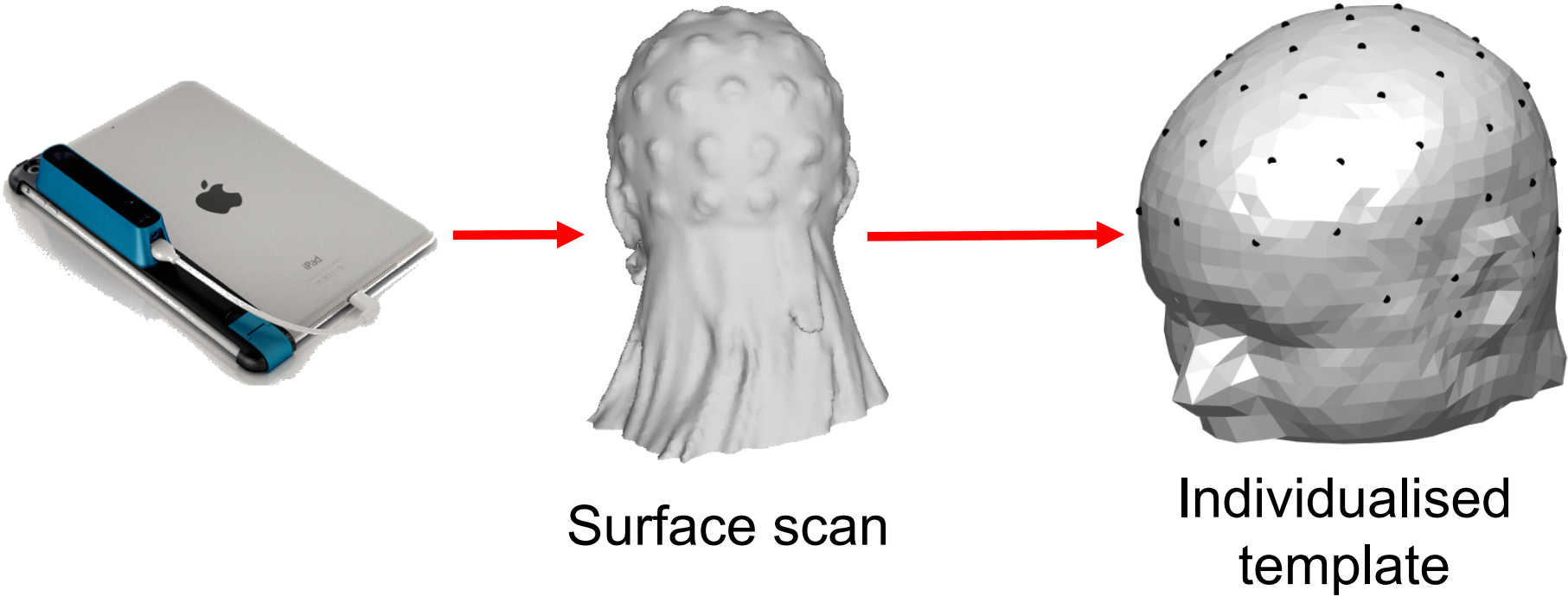
Obtaining geometrical data



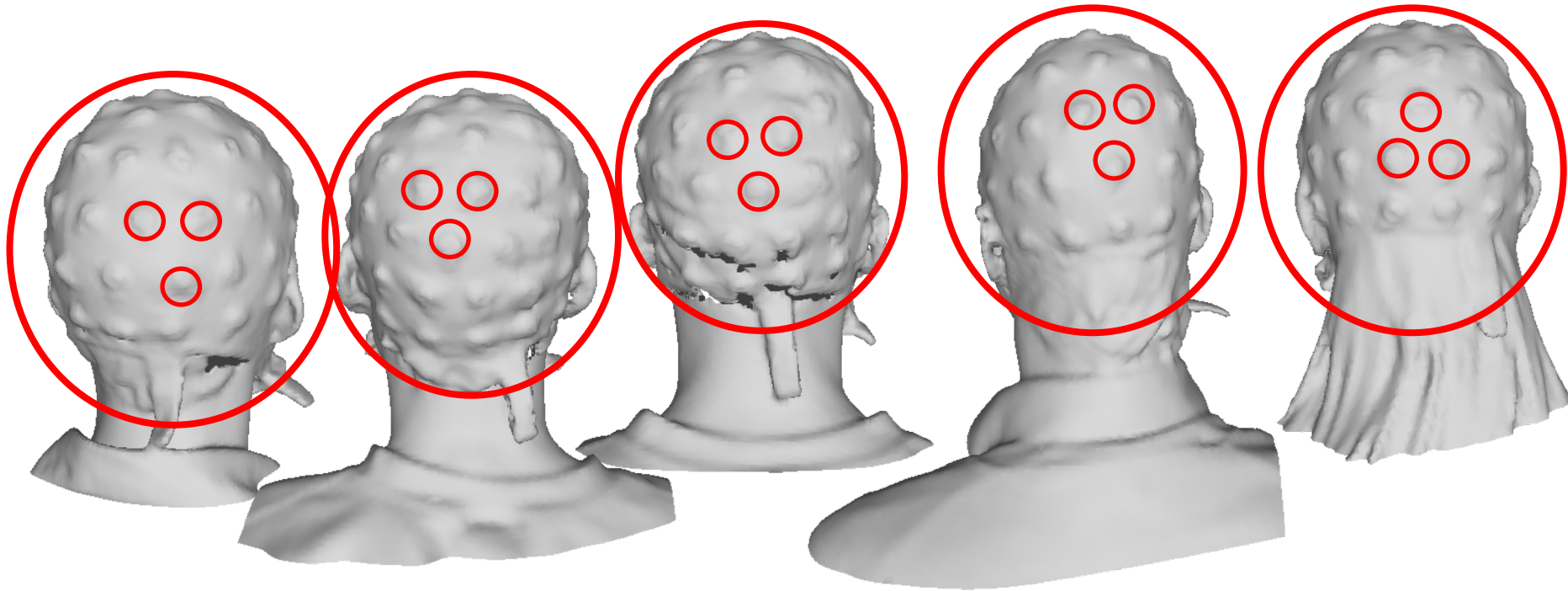
3D scanning instead of MRI



3D scanning - pipeline for EEG modelling

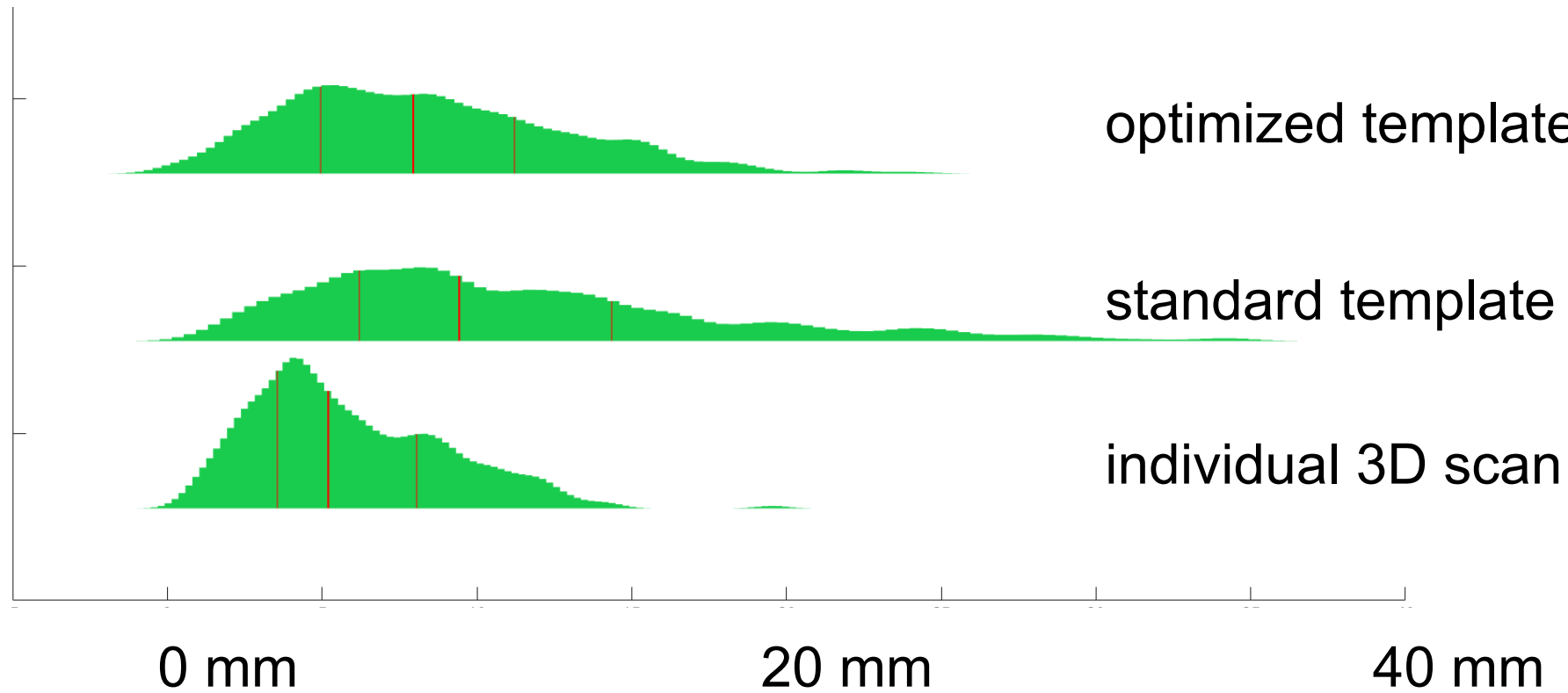


3D scanning



Homölle & Oostenveld
in preparation

3D scanning - Electrode position accuracy



Forward modeling – practical considerations

Most accurate source estimate using individual headmodels and electrode positions

Decent accurate source estimate with template headmodel and individual electrode positions

Reasonably accurate source estimate with template BEM headmodel and template electrodes

Least accurate source estimate with spherical model and template electrodes

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Inverse modeling - general

- Single and multiple dipole fitting

- Distributed source models

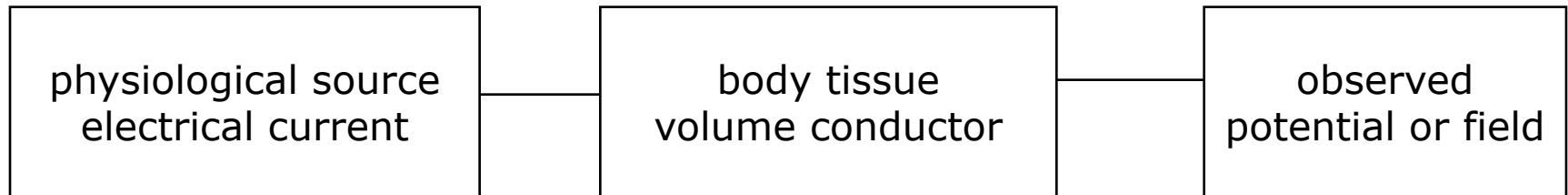
- Beamforming methods

Inverse modeling - independent components

Summary

Biophysical source modelling: overview

forward model



inverse model



Inverse methods

Single and multiple dipole models

Minimize error between model and measured potential/field

Distributed source models

Perfect fit of model to the measured potential/field

Additional constraint on source smoothness, power or amplitude

Spatial filtering

Scan the whole brain and compute filter output at every location

Beamforming (e.g. LCMV, SAM, DICS)

Multiple Signal Classification (MUSIC)

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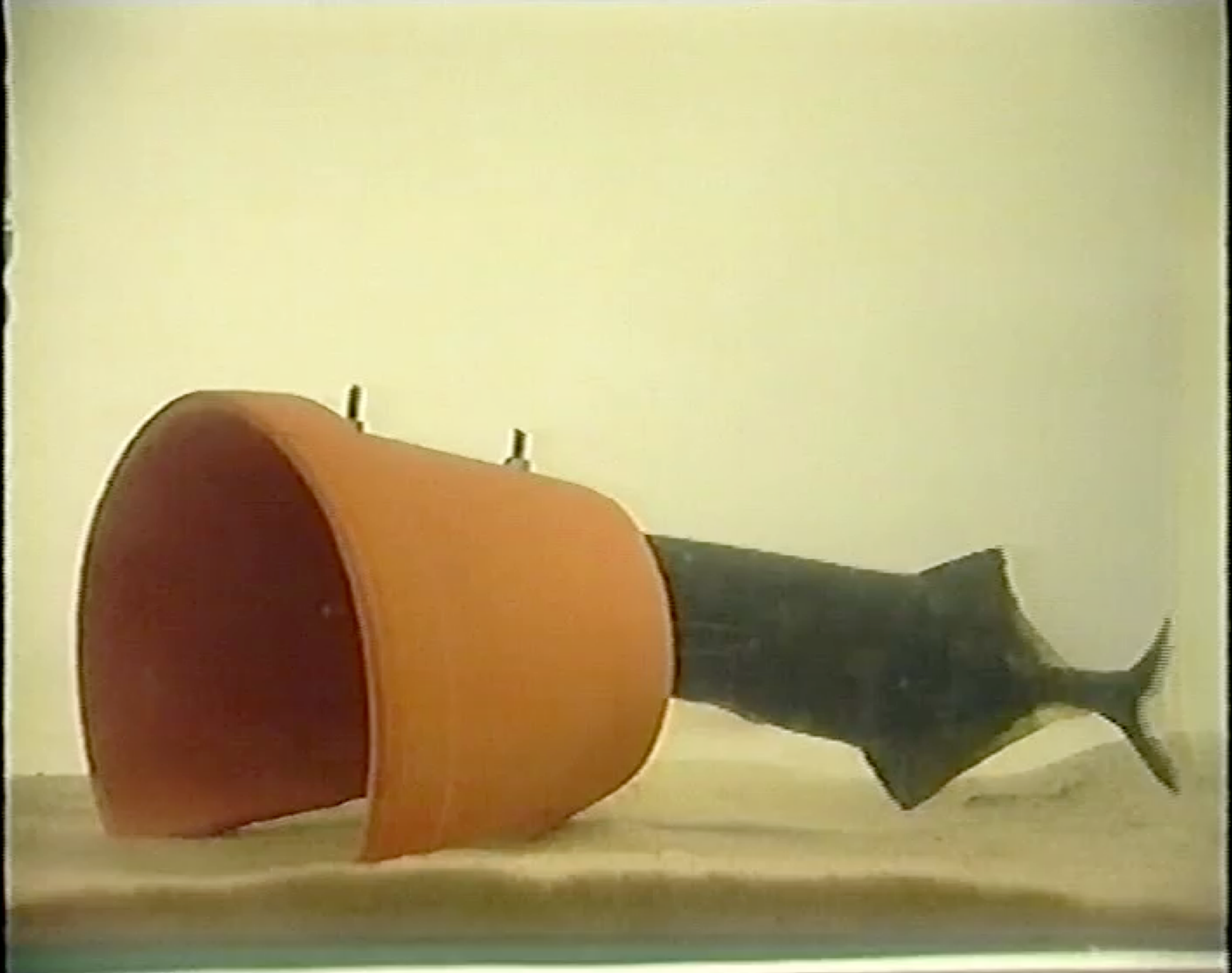
- Distributed source models

- Beamforming methods

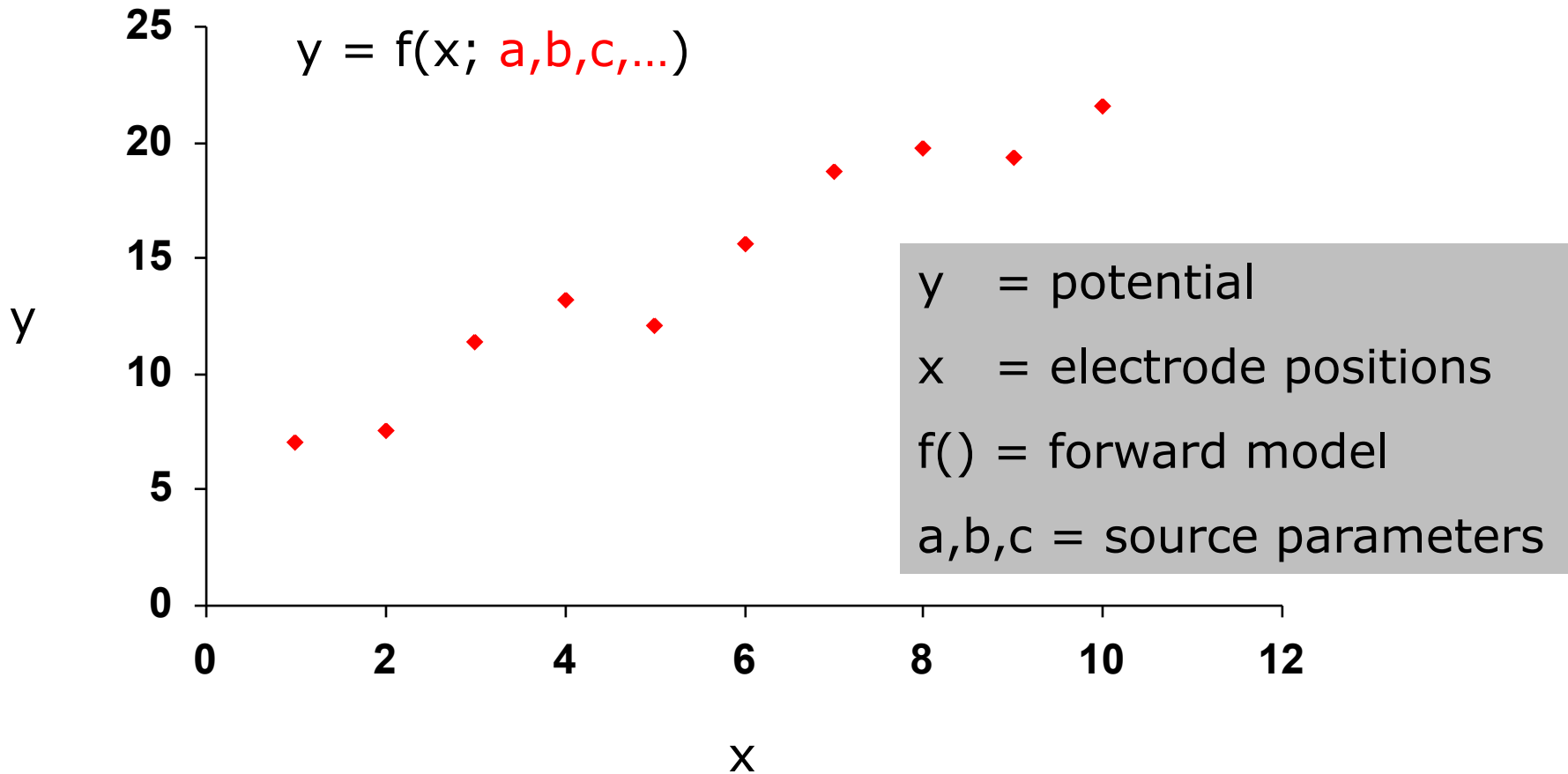
Inverse modeling - independent components

Summary

Inverse localization: demo



Single or multiple dipole models - Parameter estimation



Parameter estimation: dipole parameters

source model with
few parameters

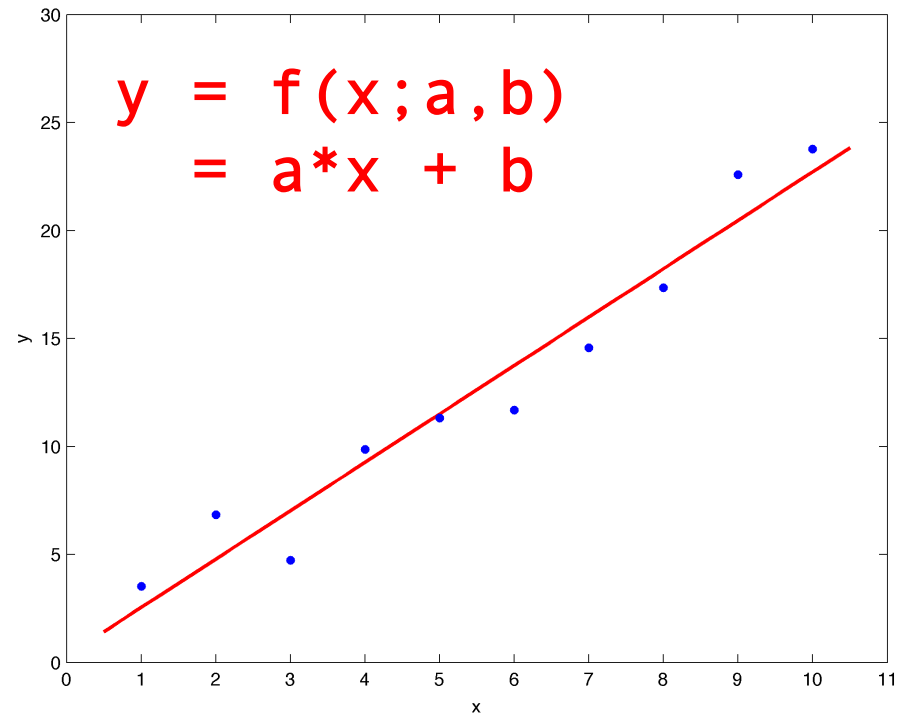
position

orientation

strength

compute the model
data

minimize difference
between actual and
model data



Non-linear parameters: grid search

One dimension, e.g. location along medial-lateral

100 possible locations

Two dimensions, e.g. med-lat + inf-sup

$100 \times 100 = 10.000$

Three dimensions

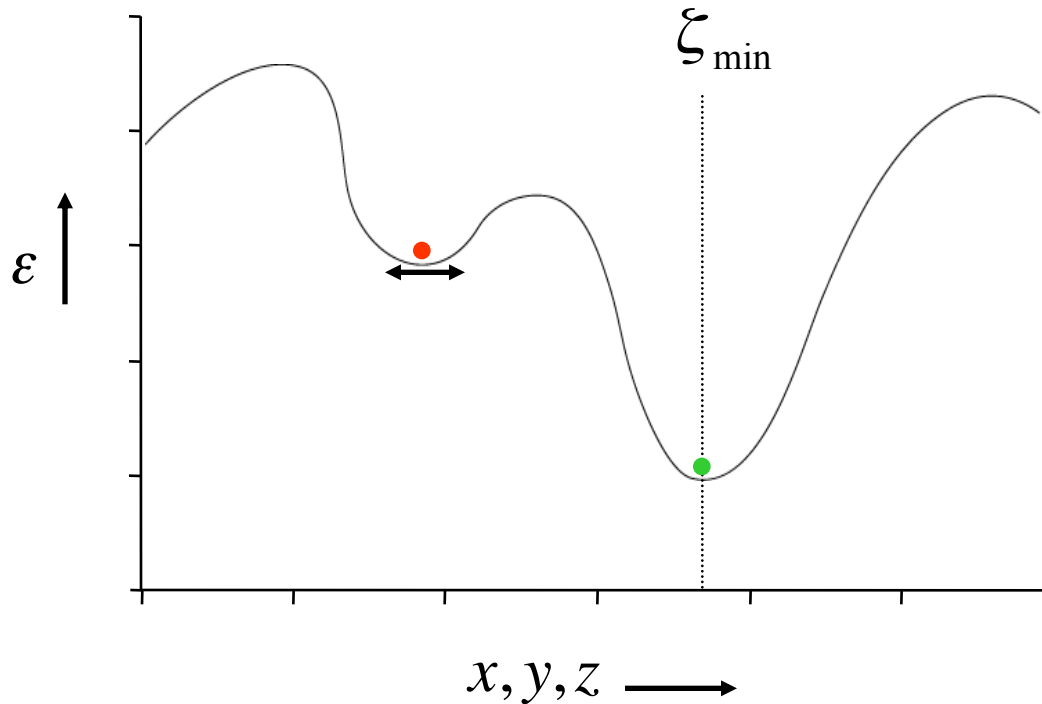
$100 \times 100 \times 100 = 1.000.000 = 10^6$

Two dipoles, each with three dimensions

$100 \times 100 \times 100 \times 100 \times 100 \times 100 = 10^{12}$

Optimization of non-linear parameters

$$\text{error}(x, y, z) = \sum_{i=1}^N (Y_i(x, y, z) - V_i)^2 \Rightarrow \min_{x, y, z} (\text{error}(x, y, z))$$



Single or multiple dipole models - Strategies

Single dipole:

scan the whole brain, followed by iterative optimization

Two dipoles:

scan with symmetric pair, use that as starting point for iterative optimization

More dipoles:

sequential dipole fitting

Sequential dipole fitting for ERPs

Assume that activity starts “small”

explain earliest ERP component with single equivalent current dipole

Assume later activity to be more widespread

add ECDs to explain later ERP components

estimate position of new dipoles

re-estimate the activity of all dipoles

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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

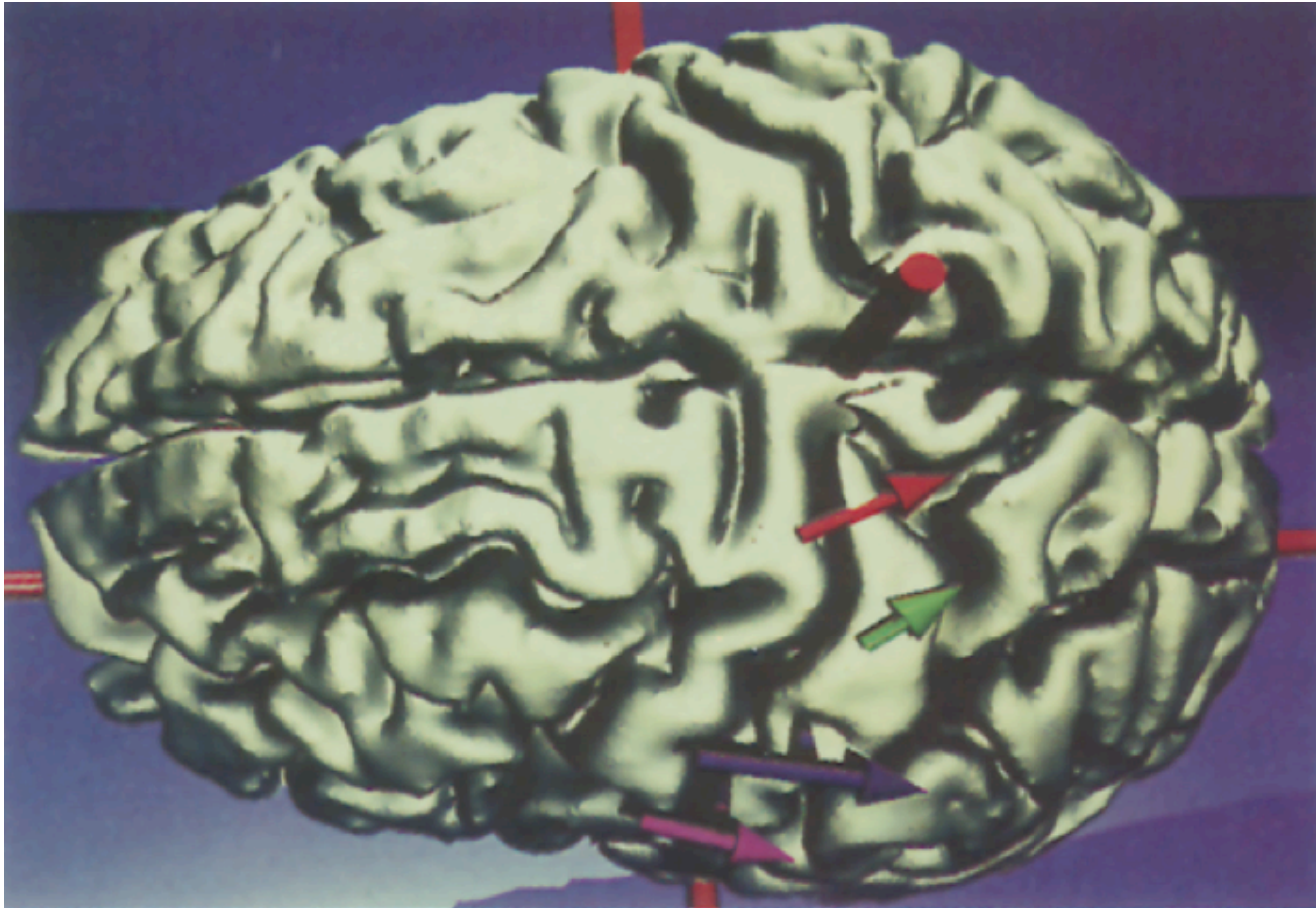
In principle easy to solve, however...

More “unknowns” (parameters) than “knowns”
(measurements)

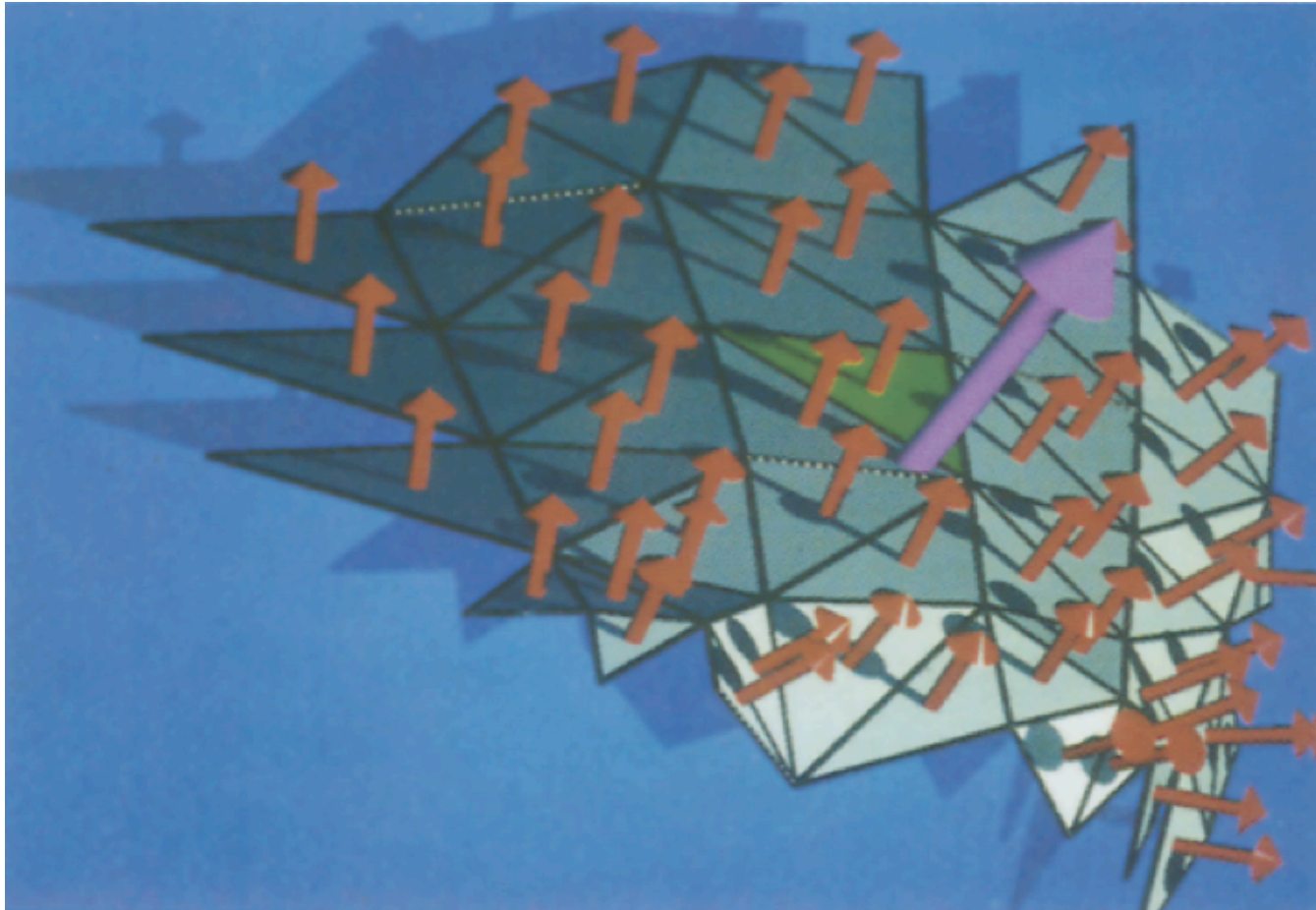
Infinite number of solutions can explain the data
perfectly

Additional constraints required

Distributed source model



Distributed source model

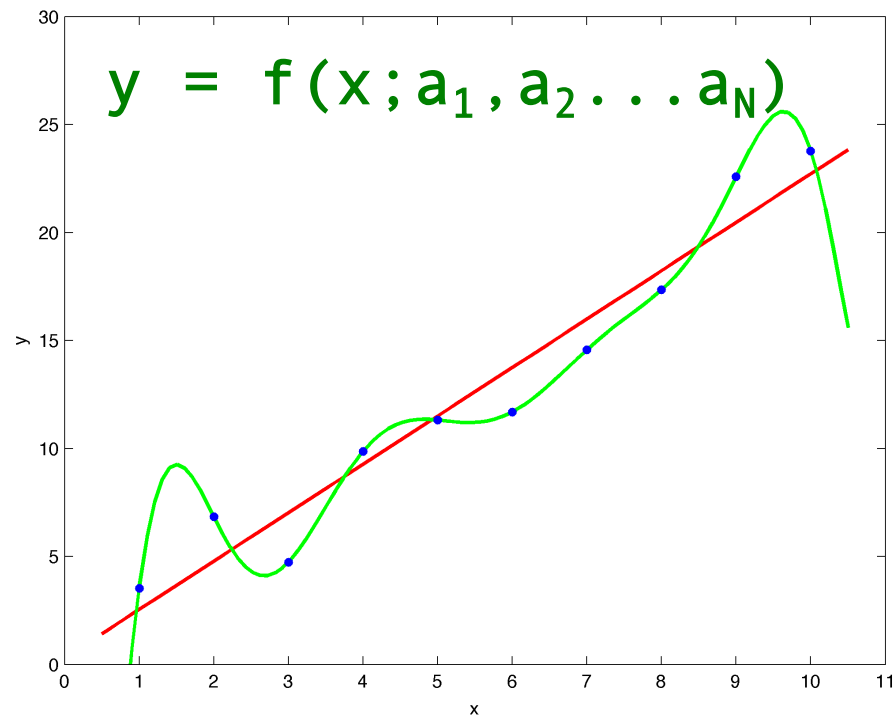


Distributed source model: linear estimation

distributed source model
with **many dipoles**
throughout the whole
brain

estimate the strength of
all dipoles

data and noise can be
perfectly explained



Distributed source model: regularization


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

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

Regularized linear estimation:

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

 mismatch with data

 mismatch with prior assumptions

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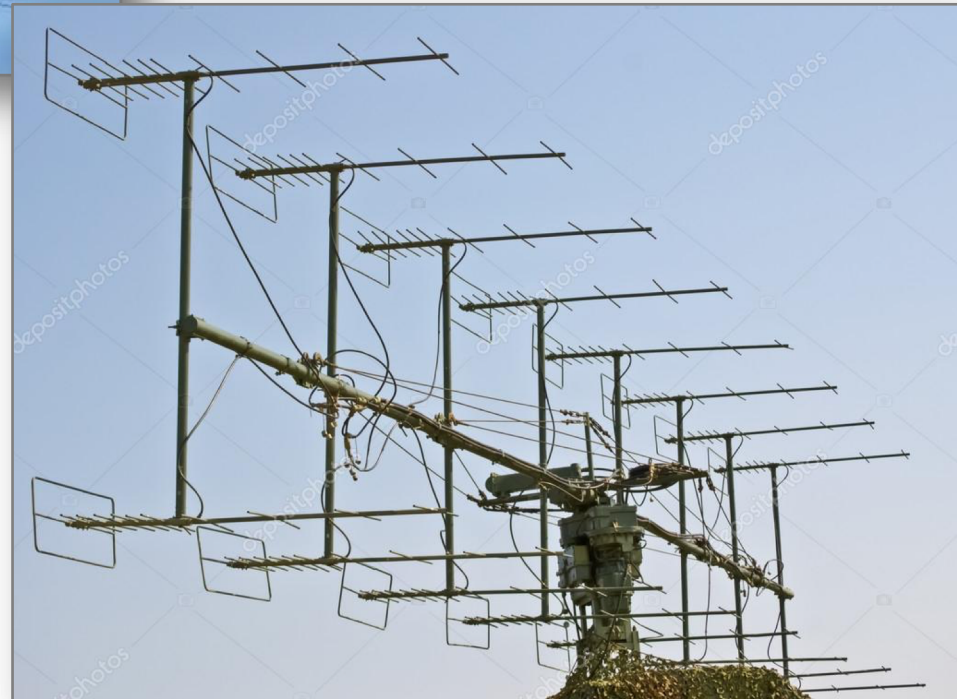
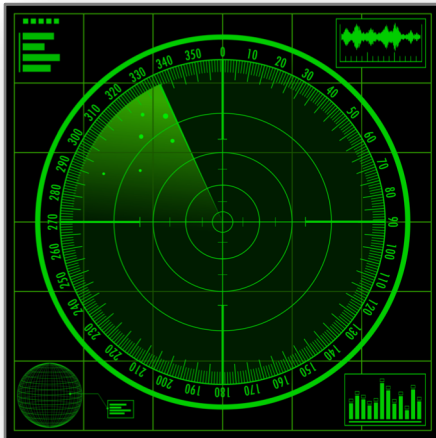
- Distributed source models

- Beamforming methods**

Inverse modeling - independent components

Summary

Scanning with a beamformer filter



Spatial filtering with beamforming

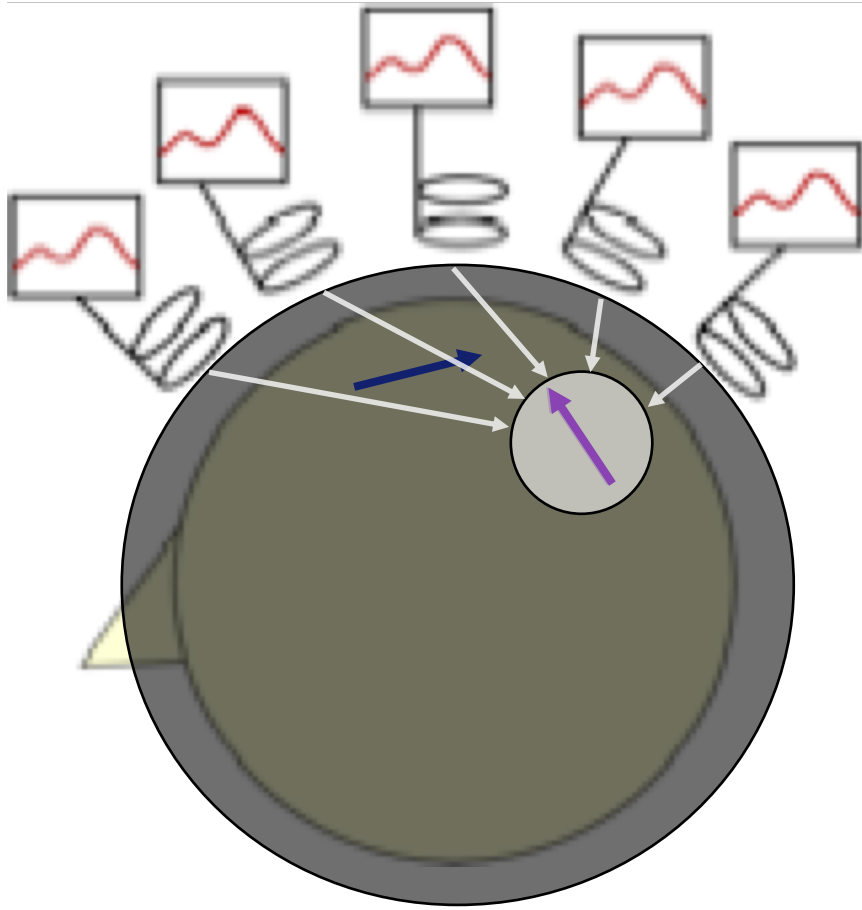
Position of the source is **not estimated** as such

Manipulate filter properties, not source properties

No explicit assumptions about source constraints
(implicit: single dipole)

Assumption that sources that contribute to the data
should be uncorrelated

Spatial filtering with beamforming



$$\mathbf{w}^T * \mathbf{G} = 1$$

$$\mathbf{w}^T * \mathbf{G} = 1$$

$$\mathbf{w}^T * \mathbf{G} = 0$$

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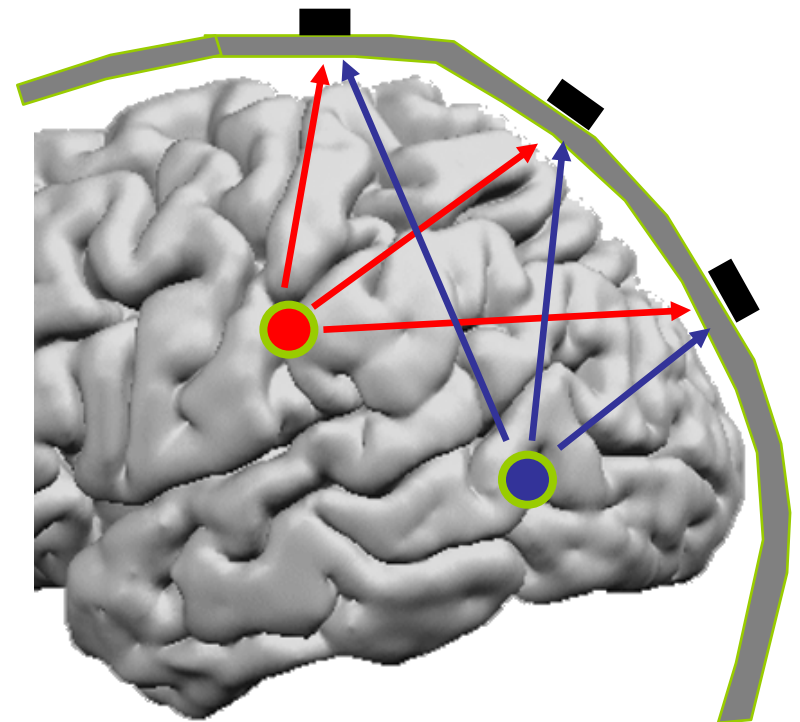
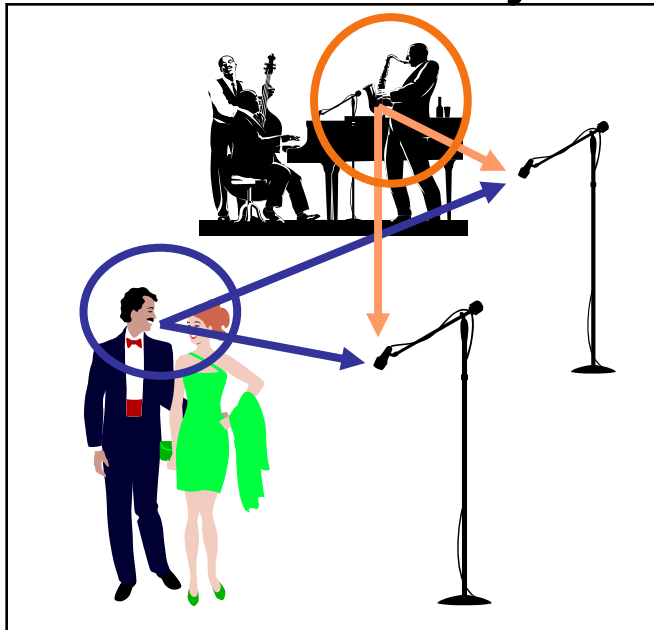
Inverse modeling - independent components

Summary

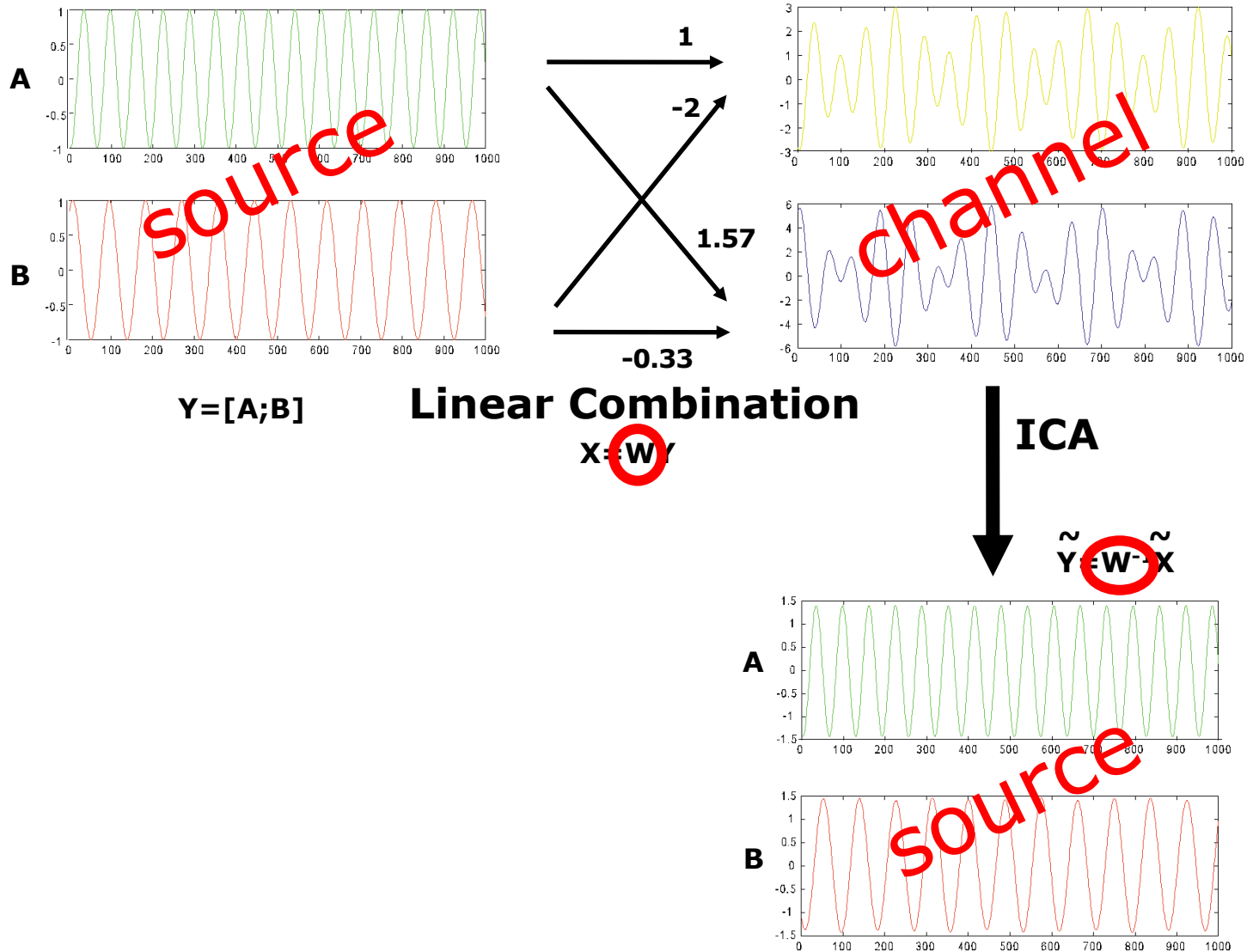
Independent component analysis

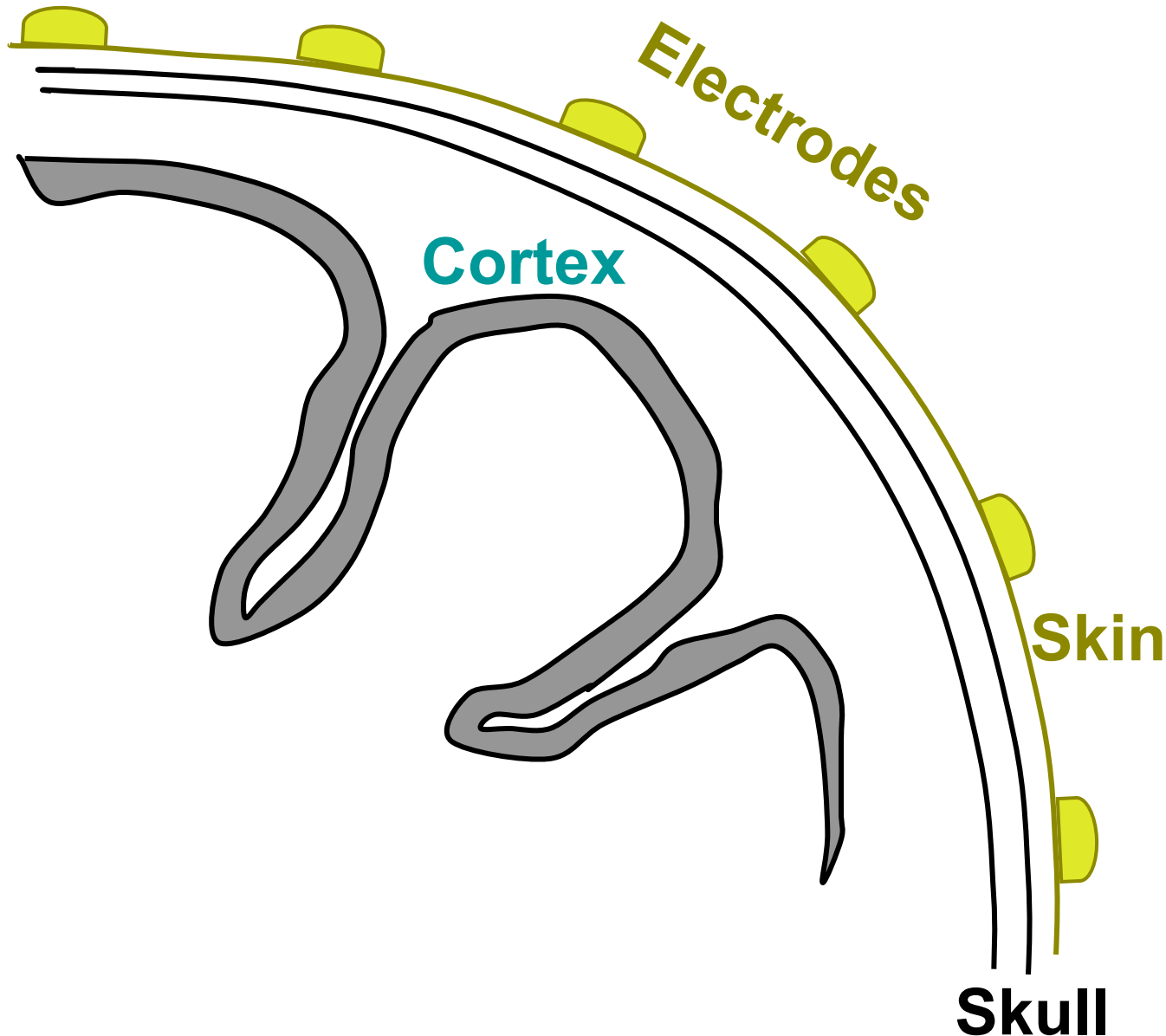
Mixture of Brain source activity

Cocktail Party



Independent component analysis



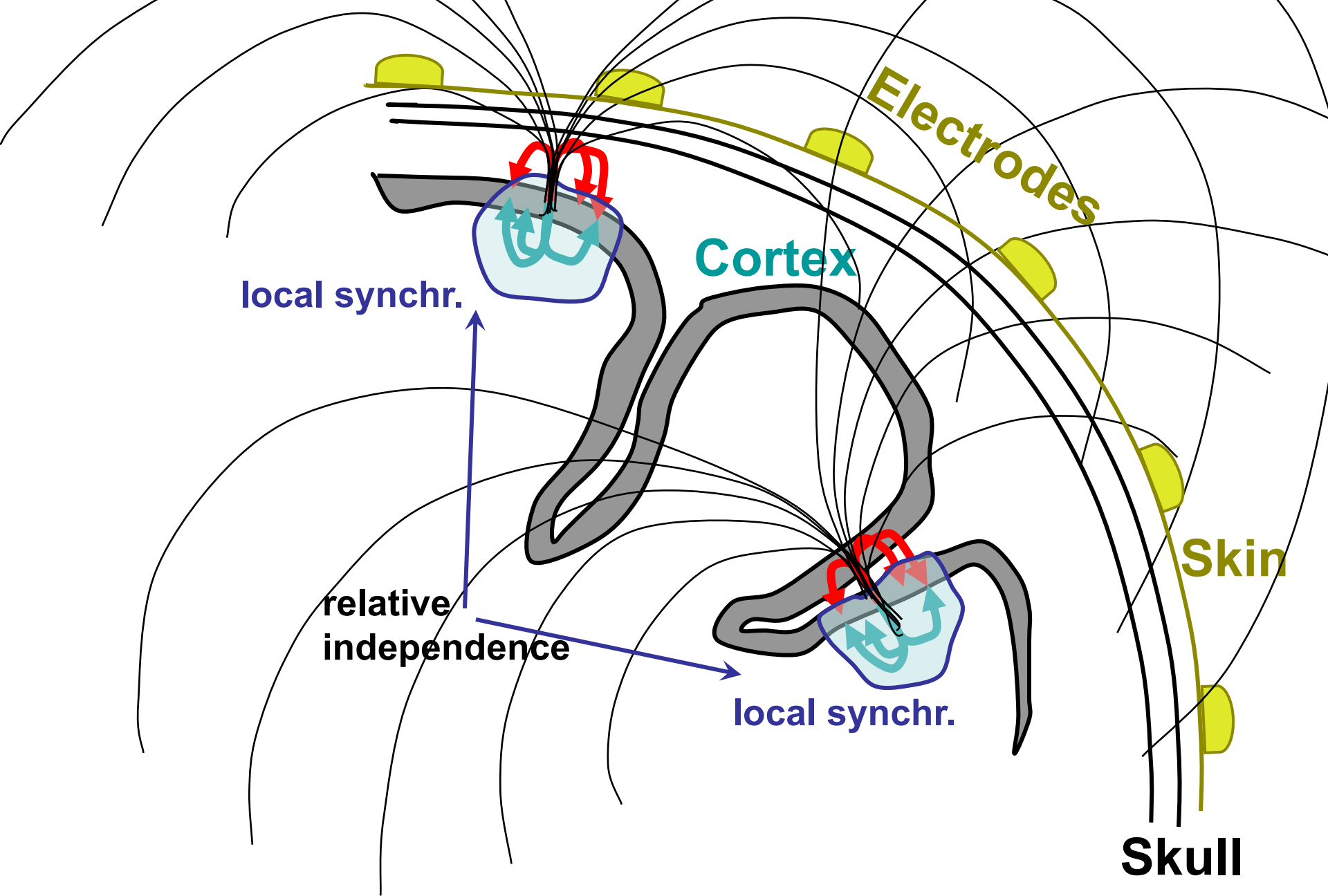


Electrodes

Cortex

Skin

Skull



Electrodes

Cortex

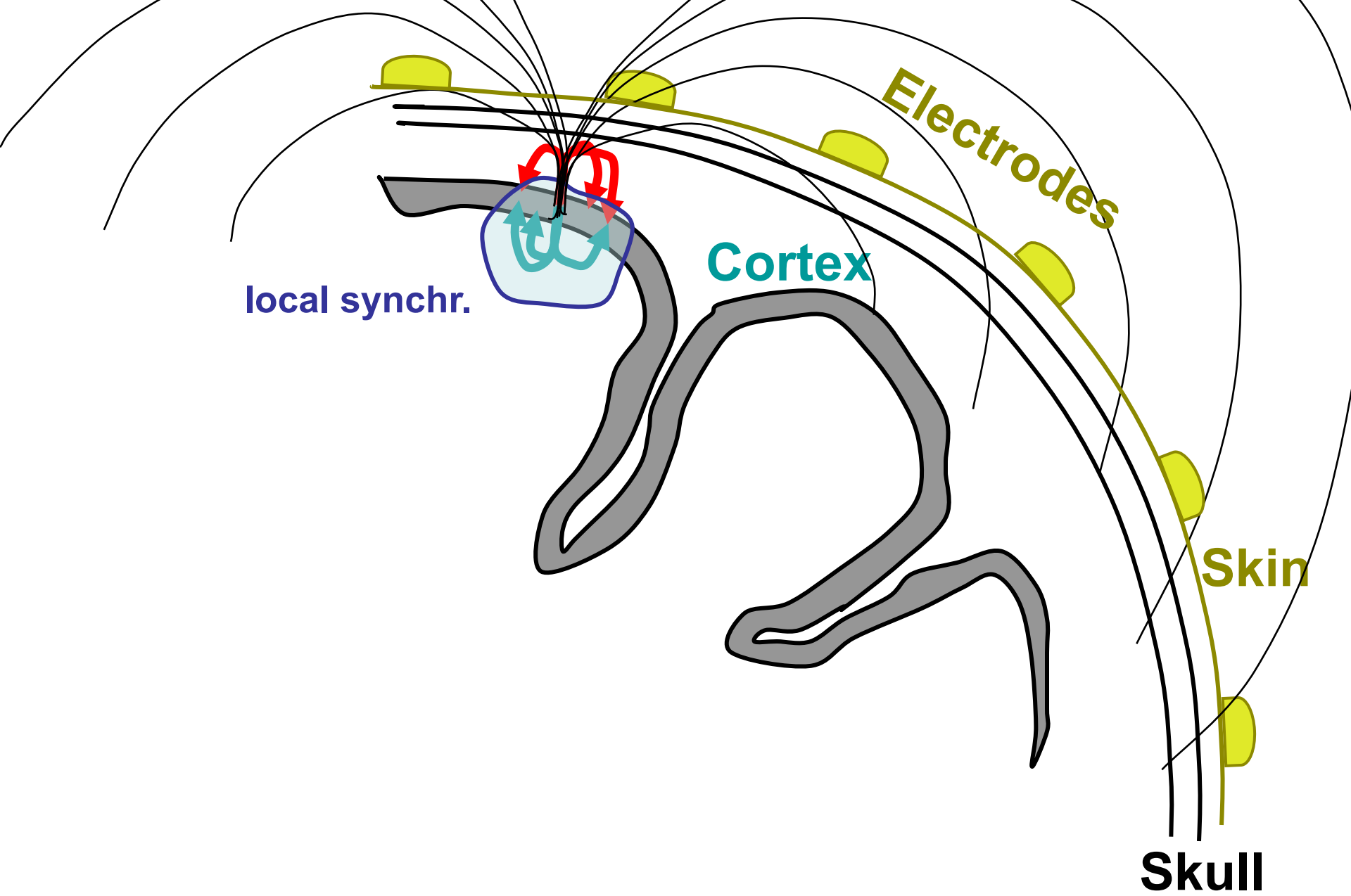
local synchr.

Skin

relative independence

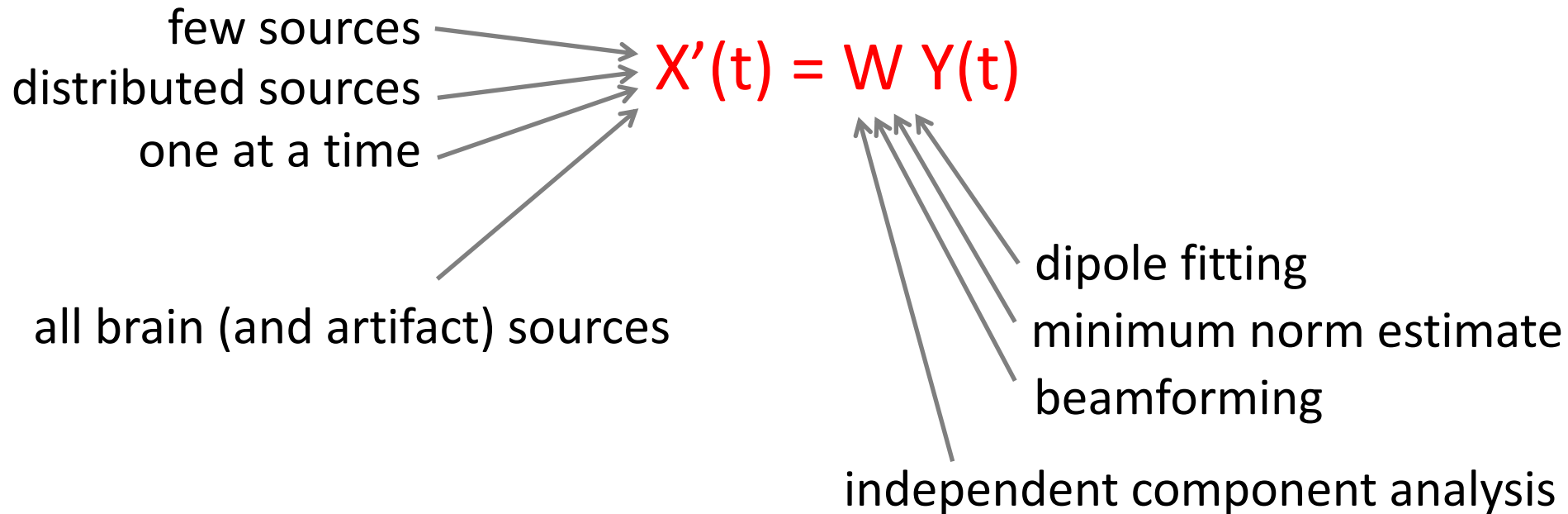
local synchr.

Skull



Estimating source timecourse activity

$$Y = G_1 X_1 + G_2 X_2 + \dots + G_n X_n + \text{noise}$$



Source modelling of independent components

ICA takes care of unmixing of timeseries

Source analysis to take care of the location

Assumption: components correspond to compact spatial patches (or bilateral patches)

Use simple dipole models to model the spatial component topographies

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- Spatial filtering

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Forward modelling

Required for the interpretation of scalp topographies

Different methods with varying accuracy

Inverse modelling

Estimate 1) location and 2) timecourse

Assumptions on source locations

Single or multiple point-like source

Distributed source

Assumptions on source timecourse

Uncorrelated (and dipolar)

Independent

Summary 2

Independent component analysis

separates topography and timecourse

Inverse methods to interpret topography

Single or multiple point-like source

Distributed source

Temporally independent component topographies
are often dipolar

Summary 3

EEGLAB dipfit plugin

head model

grid search

non-linear optimization

Results in equivalent current dipole location for each component topography

