

Bioelectromagnetism in Neuroscience (Gehirnstromkrams und so)

Skiseminar 2012



Outline

Introduction

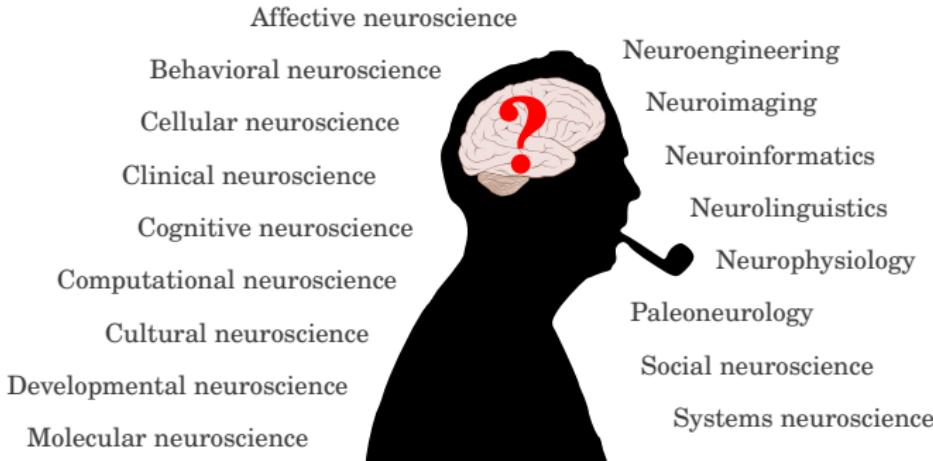
Forward Computation

Head Model Generation

Data Analysis / Inverse Problem

“The human brain undoubtedly constitutes the most complex system in the known universe” (Wolf Singer, Director of the MPI for Brain Research)

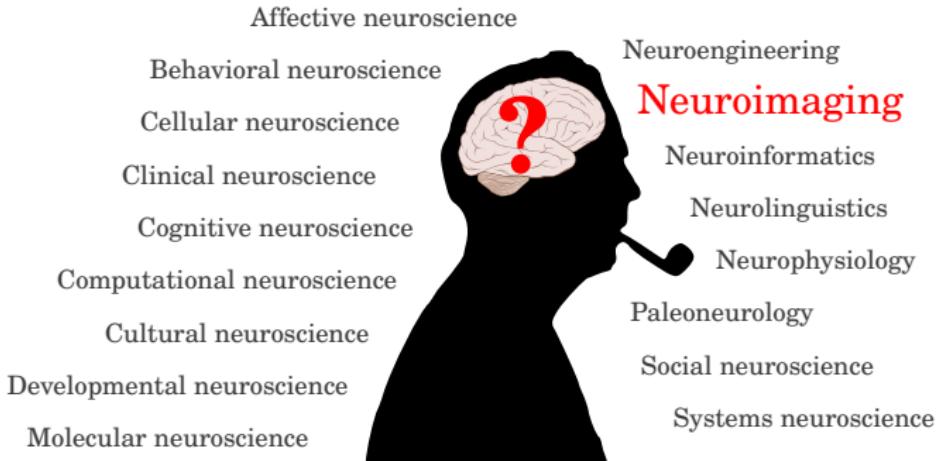
Major branches of neuroscience (by Wikipedia):



Needs people from: Biology, chemistry, computer science, engineering, linguistics, mathematics, medicine, philosophy, physics and psychology.

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Major Modalities for Neuroimaging

X-ray imaging

- ▶ Projectional Radiography
- ▶ Computed Tomography (CT)

Nuclear imaging

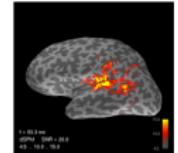
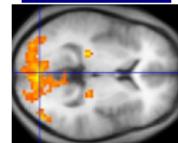
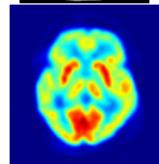
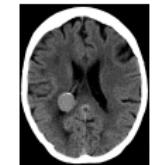
- ▶ Planar Scintigraphy
- ▶ Positron emission tomography (PET)
- ▶ Single photon emission computed tomography (SPECT)

Magnetic resonance imaging (MRI)

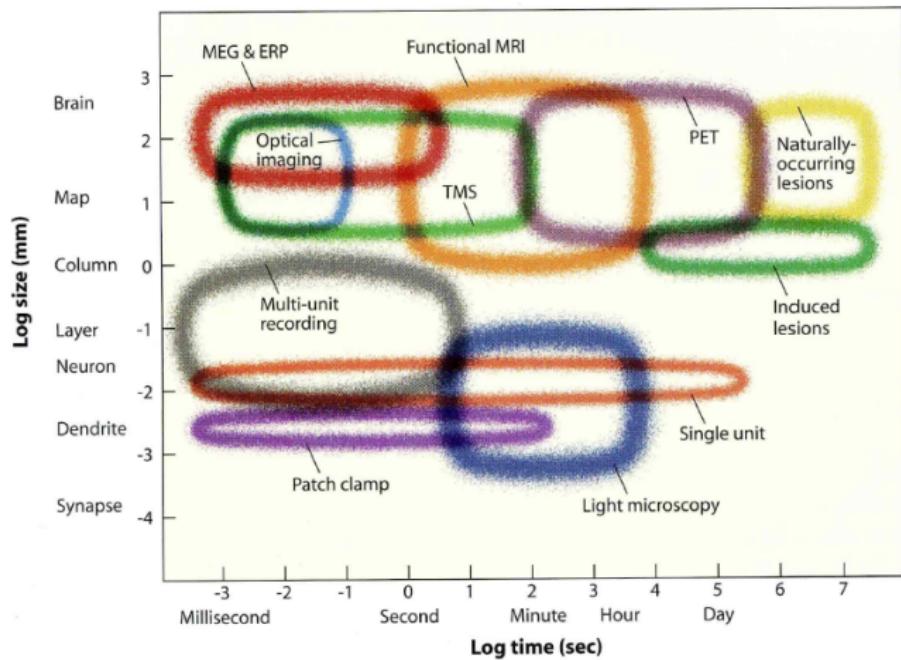
- ▶ Basic structural scans
- ▶ Functional (fMRI)
- ▶ Diffusion weighted (DW-MRI)

Bioelectromagnetic imaging:

- ▶ Electroencephalography (EEG)
- ▶ Magnetoencephalography (MEG)



Spatio-Temporal Resolution in Neuroimaging



source: Gazzaniga, Ivry & Mangun, Cognitive Neuroscience, 2nd ed., W.W.Norton & Company, 2002

Electroencephalography (EEG) and Magnetoencephalography (MEG)

Aim: Reconstruction of brain activity by **non-invasive** measurement of induced electromagnetic fields (**bioelectromagnetism**) outside of the skull.



source: Wikimedia Commons



source: Wikimedia Commons

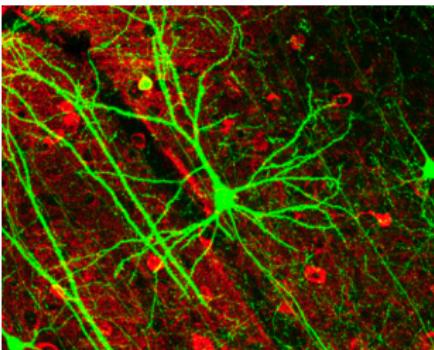


Neural Generators of EEG/MEG Signals

Signals derive from the net effect of ionic currents flowing in the dendrites of neurons during correlated synaptic transmission.

EEG: **Extracellular volume currents** produced by postsynaptic potentials.
→ strongly dependent on tissue's conductivity.

MEG: **Intracellular currents** associated with these postsynaptic potentials.
→ less dependent on tissue's conductivity.



source: Wikimedia Commons

Basics of Mathematical Modeling I

Maxwell's equations (differential, microscopic form):

$$\operatorname{div}(\mathbf{E}) = \rho/\sigma \quad \operatorname{rot}(\mathbf{E}) = -\partial_t \mathbf{B}$$

$$\operatorname{div}(\mathbf{B}) = 0 \quad \operatorname{rot}(\mathbf{B}) = \mu_0 \cdot (\vec{j} - \sigma \partial_t \mathbf{E})$$

...4 coupled, (non-linear) time-dependent PDEs!



source: Wikimedia Commons

Simplifying assumptions:

- ▶ *Linearity*: Body \approx passive conductor
- ▶ *Superposition*: Elementary sources don't interact.
- ▶ *Quasistatic approximation*: Temporal changes \ll spatial propagation velocity; Tissue is time-independent and has no inductance.
- ▶ *Charge-free*: No macroscopic charge aggregation.
- ▶ *Primary- and volume currents*: Separate current into a primary and resulting volume current.

Basics of Mathematical Modeling II

Forward/Direct Problem of EEG/MEG

Let $\sigma(\vec{r})$ be the **conductivity** and $\vec{j}^{pri}(\vec{r})$ a **primary current density** in $\Omega \subset \mathbb{R}^3$. The **electric potential** u on $\partial\Omega$ is given by::

$$\nabla \cdot (\sigma \nabla u) = \nabla \cdot \vec{j}^{pri} \quad \text{in } \Omega$$

$$n \cdot (\sigma \nabla u) = 0 \quad \text{on } \partial\Omega \text{ (no-penetration condition)}$$

$$\int_{\partial\Omega} u \cdot dS = 0 \quad \text{(fix ground potential)}$$

The **magnetic field** \mathbf{B} can be conducted by (Biot-Savart):

$$\mathbf{B}(\vec{r}) = \frac{\mu_0}{4\pi} \int_{\Omega} \left\{ \vec{j}^{pri}(\vec{r}') - \sigma(\vec{r}') \cdot \nabla u(\vec{r}') \right\} \times \frac{\vec{r} - \vec{r}'}{\|\vec{r} - \vec{r}'\|^3} d\vec{r}' \quad \text{for } \vec{r} \in \mathbb{R}^3 \setminus \bar{\Omega}$$

Solving the forward problem necessitates concerning 3 things:

- ▶ A **source-model** for \vec{j}^{pri} : How can we model the macroscopic current-flows?
- ▶ A **volume-conductor-model** of $\sigma(\vec{r})$: How can we model the dielectric properties of the different tissues?
- ▶ A **numerical method** for solving the PDE w.r.t. to source and volume conductor model; mostly FEM or BEM approaches.

Basics of Mathematical Modeling III

Inverse Problem of EEG/MEG Source Reconstruction

Given

- ▶ **measurements b** of the electric potential u and/or of the normal-component of the magnetic field $\langle n, \mathbf{B} \rangle$ on the surface $\partial\Omega$;
- ▶ a **volume-conductor-model of $\sigma(\vec{r})$** ;
- ▶ a **source model $\mathcal{J} \subset \mathcal{D}'(\Omega, \mathbb{R}^3)$** ;

estimate the **primary current $\vec{j}^{pri} \in \mathcal{J}$** (source) that is consistent with b and the neurophysiological constraints of brain activity.

Solving the inverse problem (source reconstruction) necessitates concerning 3 things:

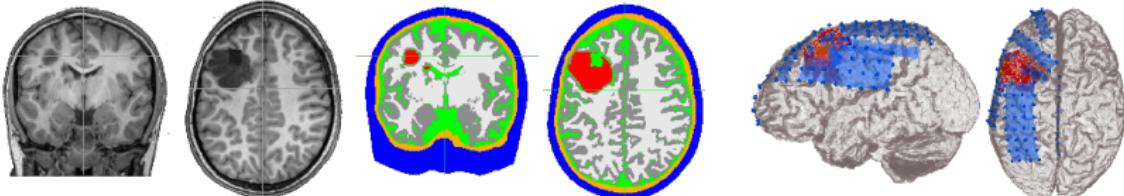
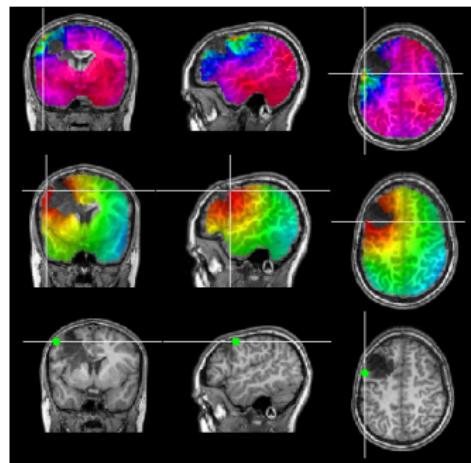
- ▶ **Data preprocessing:** How can we clean/filter the data from external sources (artifacts), noise, unwanted brain activity components?
- ▶ **A-prior modeling:** How much and which assumptions on brain activity do we need to incorporate and how do we model them to stabilize the inverse problem?
- ▶ **Implementation:** How do we solve the inverse problem practically?

Work-flow in EEG/MEG Source Reconstruction

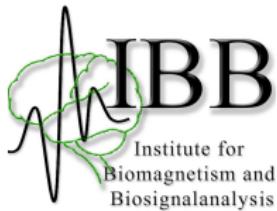
1. Head modeling;
2. Forward computation based on Head modeling;
3. Data preprocessing;
4. Source reconstruction based on forward computation and data preprocessing;

Applications of EEG/MEG

- ▶ Diagnostic tool in neurology, e.g., Epilepsy.
- ▶ Scientific applications:
 - ▶ Examination tool in several fields neuroscience.
 - ▶ Validation of therapeutic approaches in clinical neuroscience.
 - ▶ Examination tool for neurophysiology.



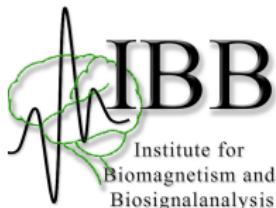
Institute for Biomagnetism and Biosignalanalysis



Focus on:

- ▶ Affective nsc
- ▶ Behavioral nsc
- ▶ Cognitive nsc
- ▶ Neuroimaging
- ▶ Clinical nsc
- ▶ Developmental nsc
- ▶ Neurolinguistics

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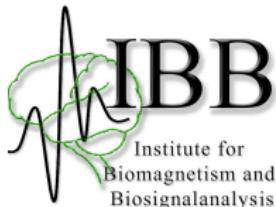
Experimental devices used:

- ▶ MEG & EEG
- ▶ Behavioral laboratory
- ▶ MRI (Basic, fMRI, DW-MRI)
- ▶ tDCS & TMS

Current fields of research:

- ▶ Auditory system: Tinnitus, neuroplasticity;
- ▶ Emotion, attention and affection;
- ▶ Language & speech: Plasticity, cochlea implantation, aphasia;
- ▶ Visual system: Conscious vision;
- ▶ Neuromuscular disorders in stroke patients.
- ▶ Methodical development

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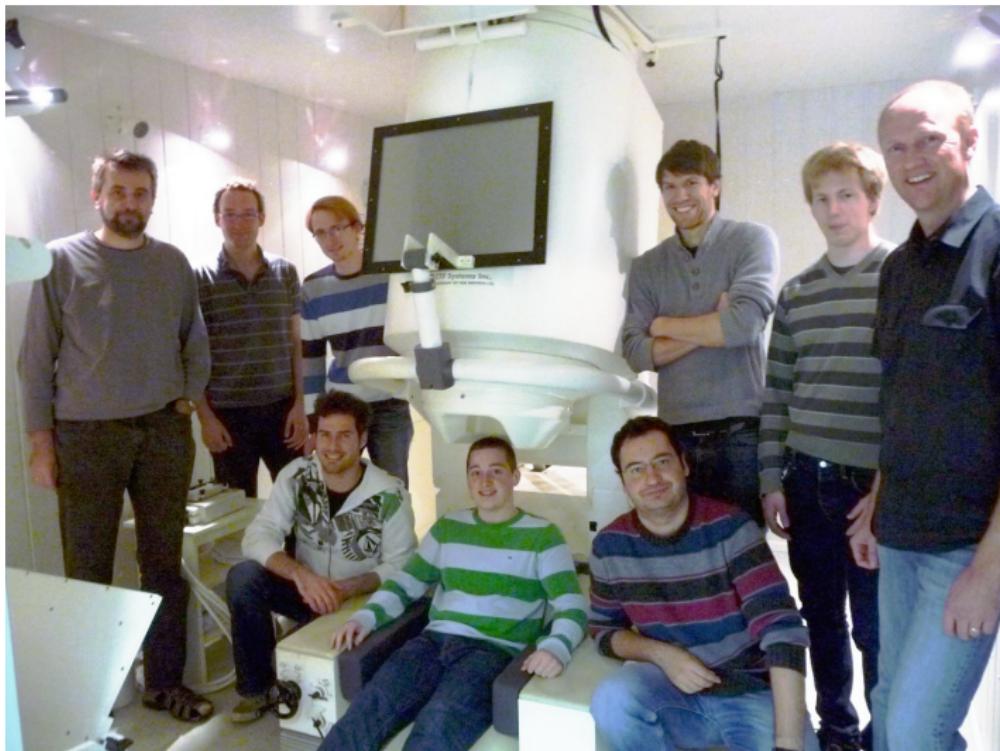
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- ▶ **Methodical development**

Workgroup "Methods in Bioelectromagnetism"



Aim: Improve quality, applicability and reliability of EEG/MEG based source reconstruction in the presurgical diagnosis of epilepsy patients.

Current Major Projects

- ▶ Cooperation with epilepsy centers in Erlangen, Bochum, Kiel.
- ▶ COMESAPED: Reconstruction of epilepsy-characteristic sources by means of a simultaneous evaluation of EEG- and MEG- data using calibrated realistic head models.
- ▶ KONNEKFEM: Development, validation and application of methods for the determination of connectivity between brain structures.
- ▶ Cooperation with BESA: Integration of realistic head modeling and FEM computation into "end user" software.
- ▶ Hierarchical Bayesian modeling for EEG/MEG source reconstruction of brain networks involving deep-lying sources using multimodal integration.



Current Cooperations with Mathematical/Physical Departments of the WWU



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theoretische physik

- ▶ Co-supervision of Diploma/Master/PhD thesis by Martin Burger.
- ▶ Lectures/seminars in applied mathematics.
- ▶ Dynamical causal modeling with Christian Himpe / Mario Ohlberger.
- ▶ Modeling of brain tumor cell dynamics with Markus Knappitsch / Christina Surulescu
- ▶ FEM mesh adaptation techniques (in DUNE?) with Mario Ohlberger.
- ▶ Nonlinear dynamics of epileptic activity with Rudolf Friedrich.

Epilepsy

- ▶ Epileptic seizures: Transient symptoms of “abnormal excessive or synchronous neuronal activity in the brain” (prevalence: 4%).
- ▶ Epilepsy: Long term risk of recurrent seizures (prevalence: 0.5-1%).
- ▶ No “standard” type of seizure, no “standard” respond to treatment.
- ▶ 30% of patients are resistant to medication
 ⇒ Surgery may be considered for *focal* epilepsies.



source: Wikimedia Commons

Presurgical Epilepsy Diagnosis

- ▶ Aim: Localize epileptic zones. Focal *irritative/seizure-onset zone*?
- ▶ Basic diagnostics: EEG, (MEG), MRI, PET, SPECT, neuropsychology.
- ▶ Phase 1, non-invasive: Video EEG-monitoring, ictal/interictal EEG/MEG, ictal SPECT.
- ▶ Phase 2, invasive: Subdural depth-recording (ECoG), WADA test, angiography

⇒ Interdisciplinary case conference ⇒ Surgery decision.

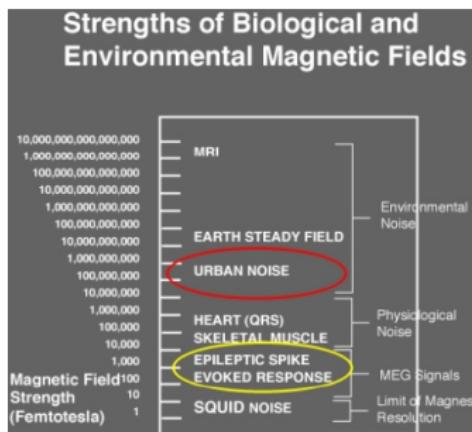


source: Wikimedia Commons

EEG & MEG in Presurgical Epilepsy Diagnosis

EEG and MEG are **non-invasive but direct** measurements of epileptic activity.

- ▶ EEG is standard for various diagnosis tasks.
- ▶ MEG is expensive.
- ▶ MEG data processing is as complicated as for EEG.
- ▶ The health insurance in Germany does not cover MEG (contrary to USA).



EEG & MEG in Presurgical Epilepsy Diagnosis

So why should MEG be used in presurgical epilepsy diagnosis?

- ▶ EEG and MEG are complementary to each other.
- ▶ Different sensitivity profiles.
- ▶ Recent studies: Stefan et al. 2003, Iwasaki et al. 2005, Knake et al. 2006).

modality	tissue geometry & conductivity	sensitivity to	
		superficial sources	deep sources
EEG	high	tang: moderate radial: high	moderate
MEG	low	tang: very high radial: very low	low

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Even better than retrospective comparison (*converging evidence*):

Symmetric data fusion of simultaneous measurements.

Challenges of Combined EEG/MEG Source Reconstruction

- ▶ Practical and technical challenges.
- ▶ Different noise characteristics.
- ▶ Different sensitivity to tissue conductivity/forward modeling.

⇒ We need:

- ▶ Realistic and individual modeling of the electrophysiological properties of the human head.
- ▶ Fast and accurate numerical forward computation methods.
- ▶ Advanced data preprocessing techniques and inverse methods for symmetric data fusion.

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Forward Computation

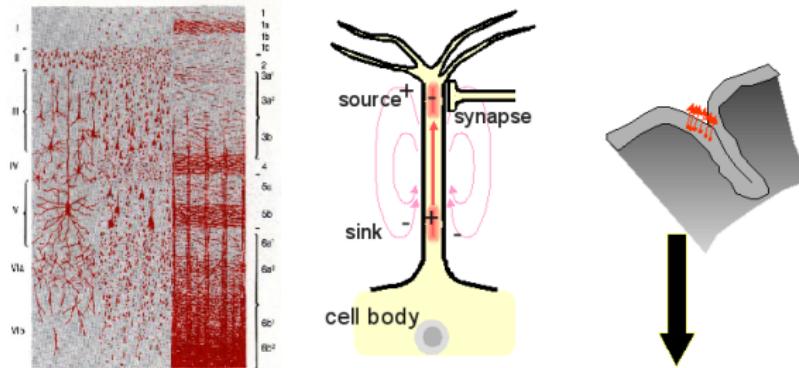
Head Model Generation

Data Analysis / Inverse Problem

Forward Computation I: Source Model

Reminder: We need 3 things: A source-model for \vec{j}^{pri} , a volume-conductor-model of $\sigma(\vec{r})$ and numerical method for solving the PDE w.r.t. to source and volume conductor model.

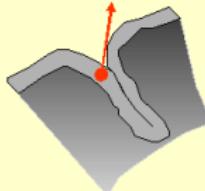
Common source model: **Equivalent current dipoles**, $\vec{j}^{pri}(x) = \sum_i M_i \delta(x - x_i)$



Equivalent Current Dipole (Primary current) (~ 50 nAm)

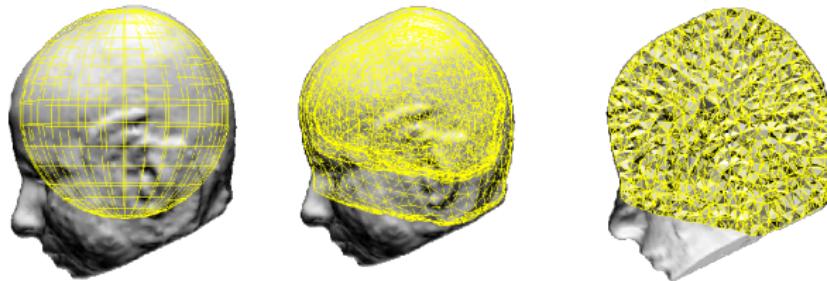
parameters:

position : x_0
moment : M



Size of Macroscopic Neural Activity
 $\sim 30 \text{ mm}^2 = 5.5 \times 5.5 \text{ mm}^2$

Forward Computation II: Numerical Methods



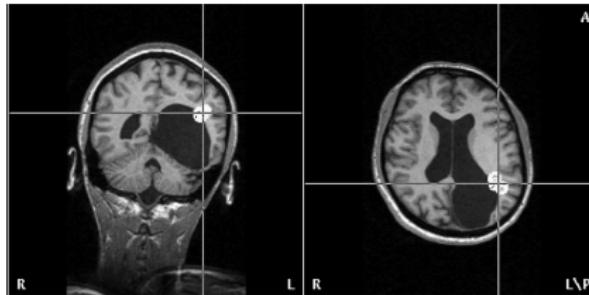
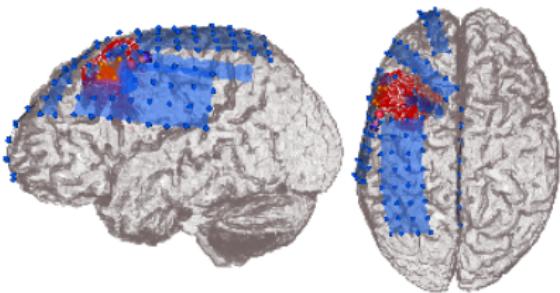
Sphere: Under the assumption of modeling the head by a multi-layer sphere model, a (quasi-)analytic solution exists.

BEM: Assuming a nested shell topography **boundary element methods** can be used, demanding the discretization of the compartment boundaries.

FEM: **Finite element methods** are based upon a discretization of the whole volume conductor.

Why FEM?

- ▶ Possibility to incorporate nearly arbitrary complex geometries and arbitrary number of compartments:
 - ▶ CSF, gray/white matter, cerebellum, brain steam, muscles, dura mater, blood vessels;
 - ▶ Realistic skull modeling: Skull holes, three-layeredness.
 - ▶ Anatomical anomalies from surgeries / brain damages.
- ▶ Modeling of invasive recording devices (ECoG, depth-electrodes)
- ▶ Inclusion of anisotropic conductivities, e.g., white matter anisotropy



FE Forward Approaches

Problem: Singular source model; $D^\alpha \delta(r) \in H^{-3/2 - |\alpha| - \varepsilon}(\Omega) \forall \varepsilon > 0$

FEM approaches to deal with this:

- ▶ Subtraction approach
- ▶ Venant approach
- ▶ Partial Integration approach

Alternatives: Source model $j^{imp}(r) \in H(\text{div}, \Omega, ; \mathbb{R}^3)$ and use **Whitney elements** for FEM.

Subtraction Approach

Split the potential and the conductivity into two parts:

$$u = u^\infty + u^{\text{cor}}, \quad (1)$$

$$\sigma = \sigma^\infty + \sigma^{\text{cor}}. \quad (2)$$

Now calculate the analytic solution u^∞ for a source in a medium with constant homogeneous conductivity σ^∞ . Afterwards, only a correction potential has to be calculated numerically:

$$-\nabla \cdot (\sigma \nabla u^{\text{cor}}) = f \quad \text{in } \Omega, \quad f := \nabla \cdot (\sigma^{\text{cor}} \nabla u^\infty), \quad (3)$$

$$\sigma \partial_n u^{\text{cor}} = g \quad \text{on } \Gamma, \quad g := -\sigma \partial_n u^\infty. \quad (4)$$

This equation can now be discretized and solved without having to deal with the singularity numerically.

Venant Approach

Approximation of the dipolar source by a distribution of electrical monopoles (i.e., current sinks and sources/electrical charges), placed on the FE-nodes in the vicinity of the source position.

Then, try to match the moments of the distribution to those of a dipolar source:

$$({}^l T)_j = \sum_{i=1}^k (\Delta x_i)_j^l q_i, \quad j = 1, 2, 3. \quad (5)$$

For $l = 0, 1, 2$ we have

$${}^l T = \frac{1 - (-1)^l}{2^l} \cdot p. \quad (6)$$

This leads to a sparse RHS-vector (≈ 30 non-zero entries).

Partial Integration Approach

Take the weak formulation of the PDE and apply integration by parts to both sides:

$$\int_{\Omega} \nabla(\sigma \nabla u) \cdot \phi_i dx = \int_{\Omega} f \cdot \phi_i dx = \int_{\Omega} \nabla j^{imp} \cdot \phi_i dx. \quad (7)$$

$$-\int_{\Omega} (\sigma \nabla u) \cdot \nabla \phi_i dx + \int_{\partial\Omega} \sigma \partial_n u \cdot \phi_i d\gamma(x) = -\int_{\Omega} j^{imp} \cdot \nabla \phi_i dx + \int_{\partial\Omega} \partial_n j^{imp} \cdot \phi_i d\gamma(x). \quad (8)$$

We use the Neumann boundary condition and the fact that the current density vanishes at the head surface and obtain

$$\int_{\Omega} (\sigma \nabla u) \cdot \nabla \phi_i dx = \int_{\Omega} j^{imp} \cdot \nabla \phi_i dx = p \nabla \phi_i(x_0), \quad (9)$$

which yields a sparse RHS again (4 non-zero entries for tetrahedral, 8 non-zero entries for hexahedral meshes).

Pros and Cons of Approaches

Direct Approaches (Venant, Partial Integration, Whitney)

- ✓ Extremely fast computation after unique setup of a lead-field basis (some ms per RHS)
- ! Achieved accuracy depends on local mesh structure
- ! Only heuristic derivation (esp. Partial Integration)

Subtraction Approach

- ✓ Mathematically well-understood; existence, uniqueness and convergence of a solution can be proven
- ! High computational effort for the setup of a single RHS (some s)

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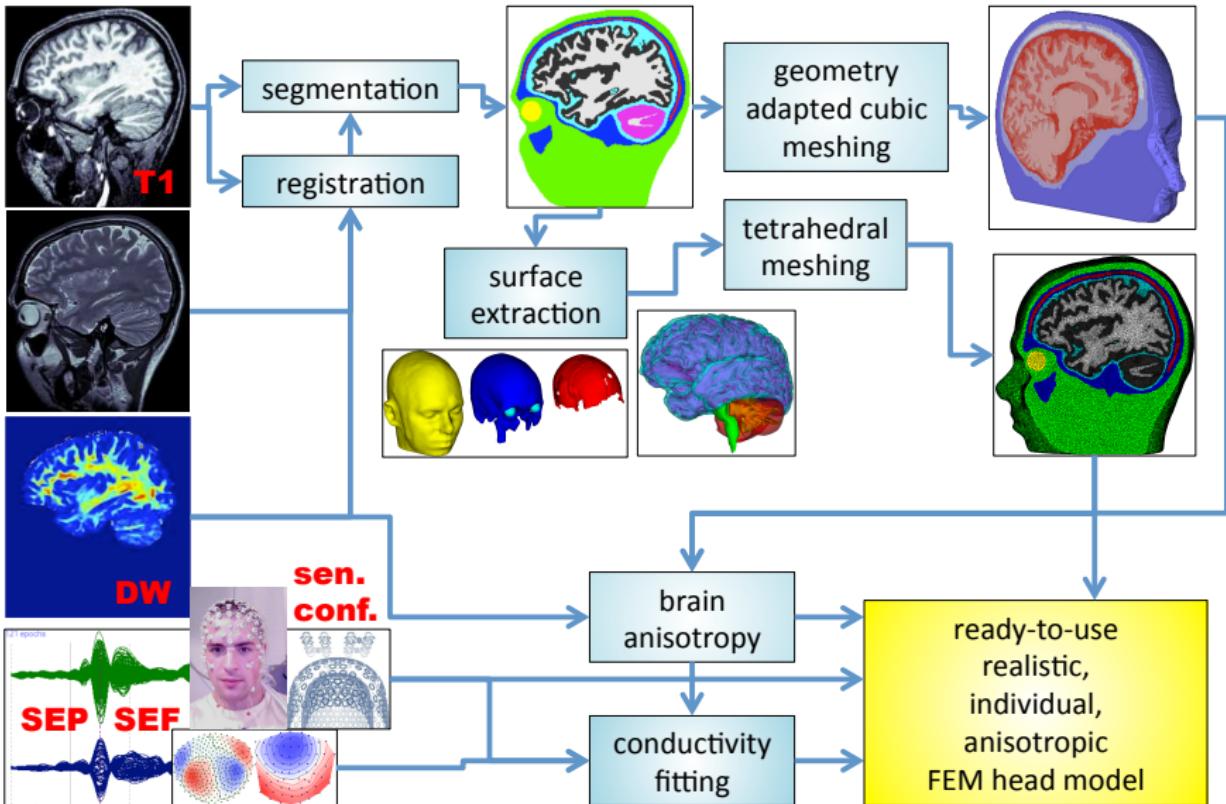
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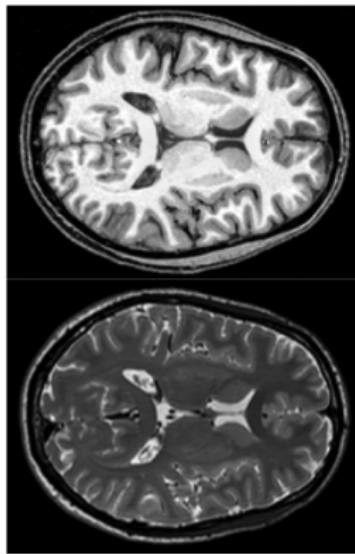
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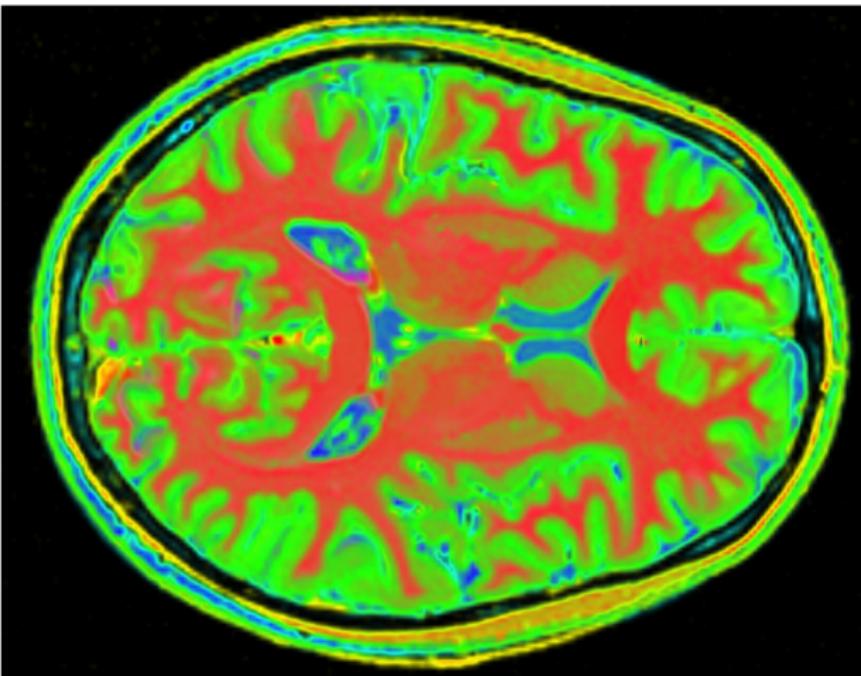
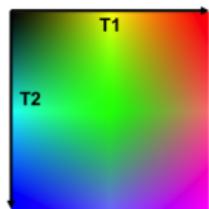
Realistic, individual head modeling for bioelectromagnetic applications



Part 1: MRI Processing, Structural Scans

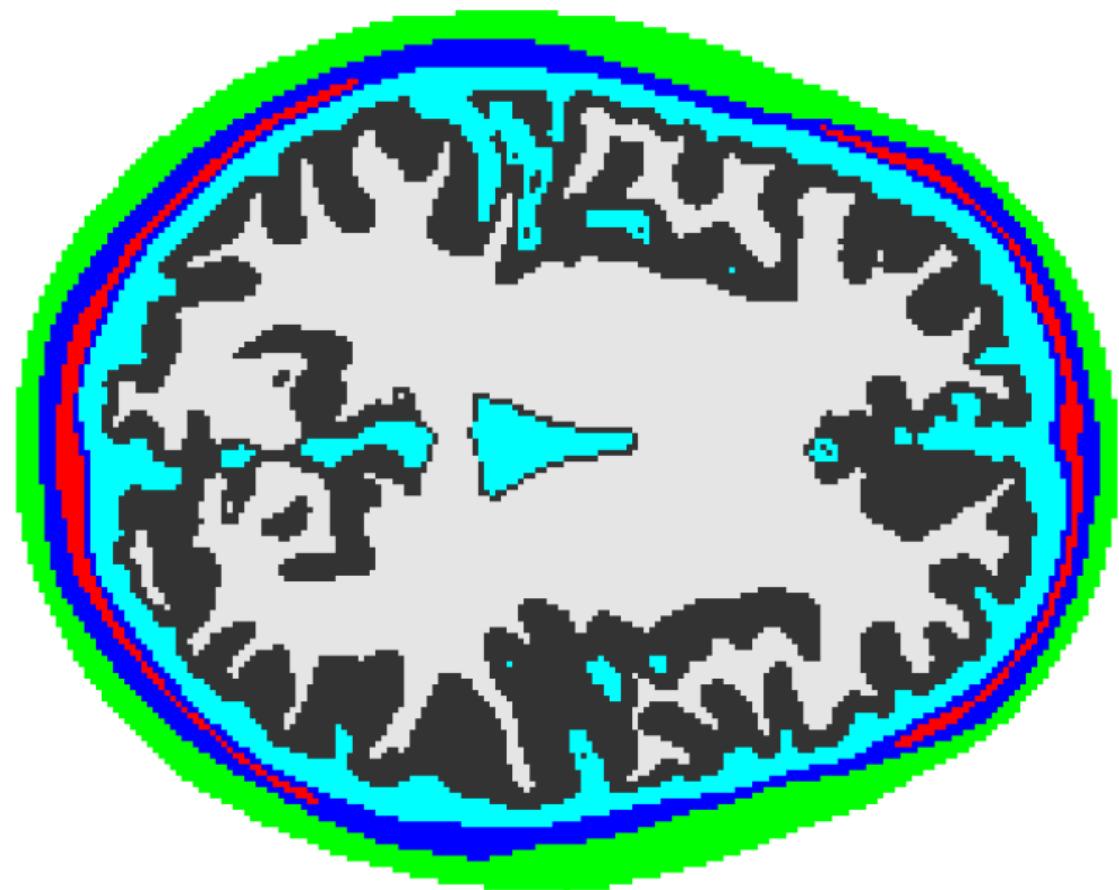


RGB
map



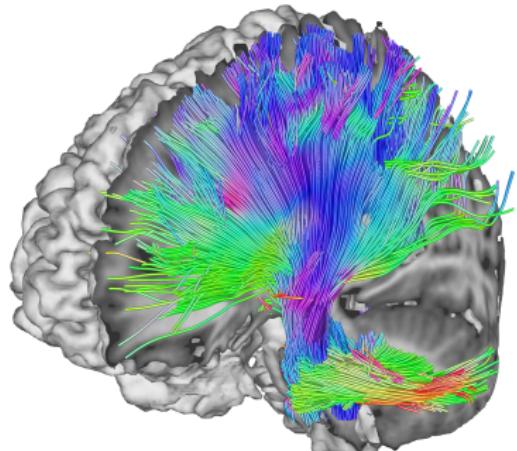
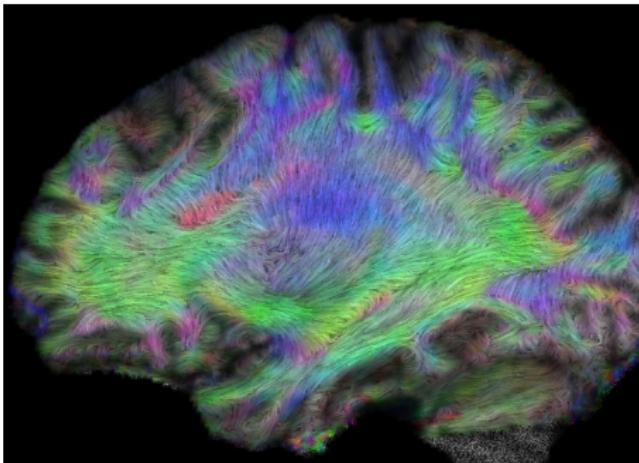
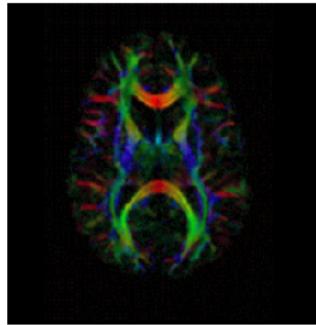
RGB composite of T1 and T2 MRI scan

Part 1: MRI Processing, Segmentation



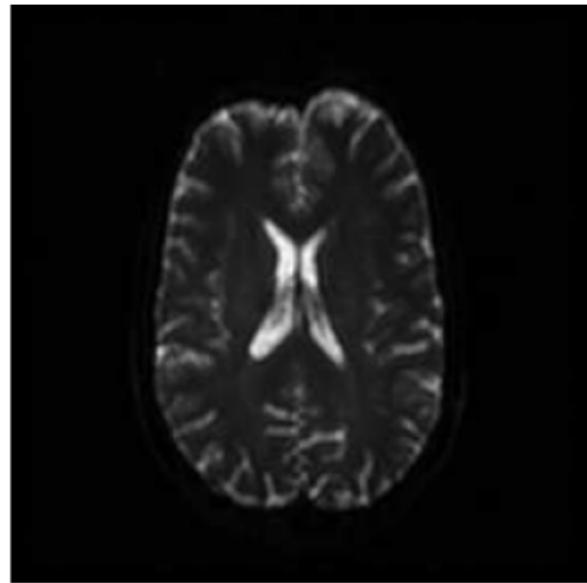
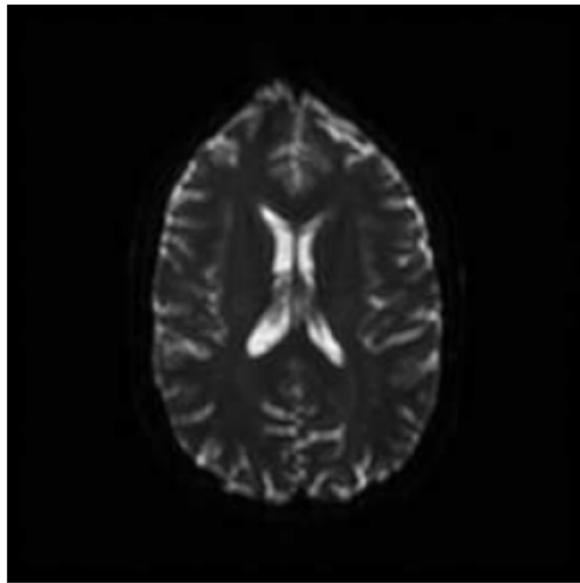
Part 1: MRI Processing, Diffusion Weighted MRI

- ▶ DW-MRI allows the mapping of diffusion processes of molecules in biological tissues, *in vivo*.
- ▶ Clinical application: Localization of white matter lesions in stroke patients, surgical planning.
- ▶ Key imaging modality to assess **connectivity** via tractography .
- ▶ We use it to compute **conductivity tensors**.



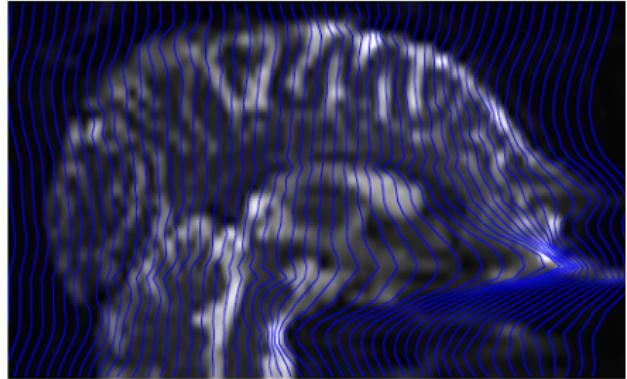
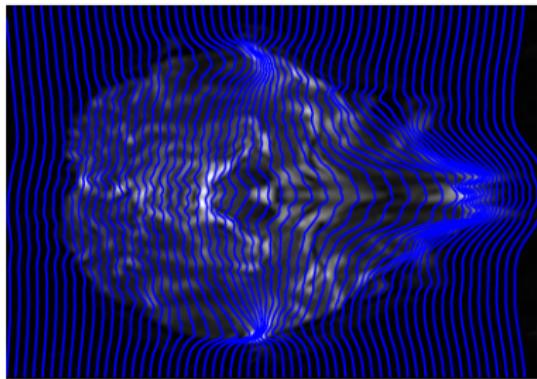
Part 1: MRI Processing, Diffusion Weighted MRI

Fast Echo-Planar Imaging (EPI)

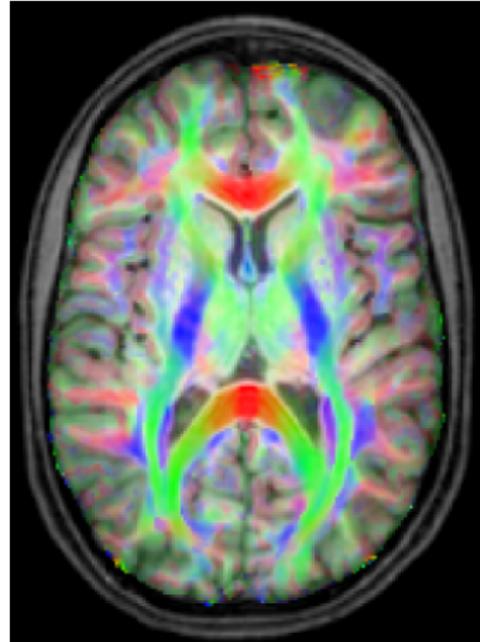
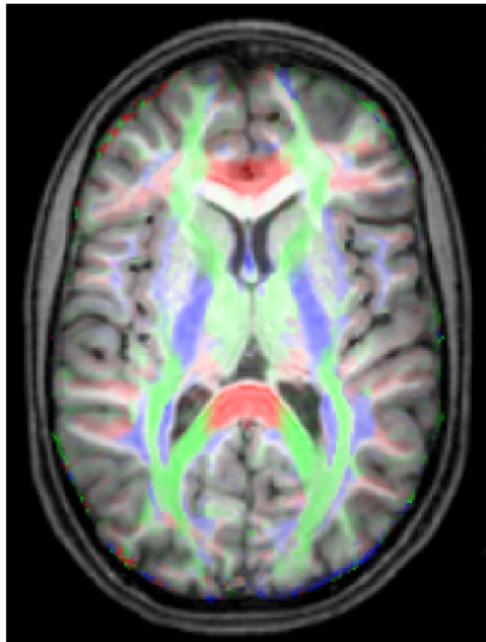


Part 1: MRI Processing, Diffusion Weighted MRI

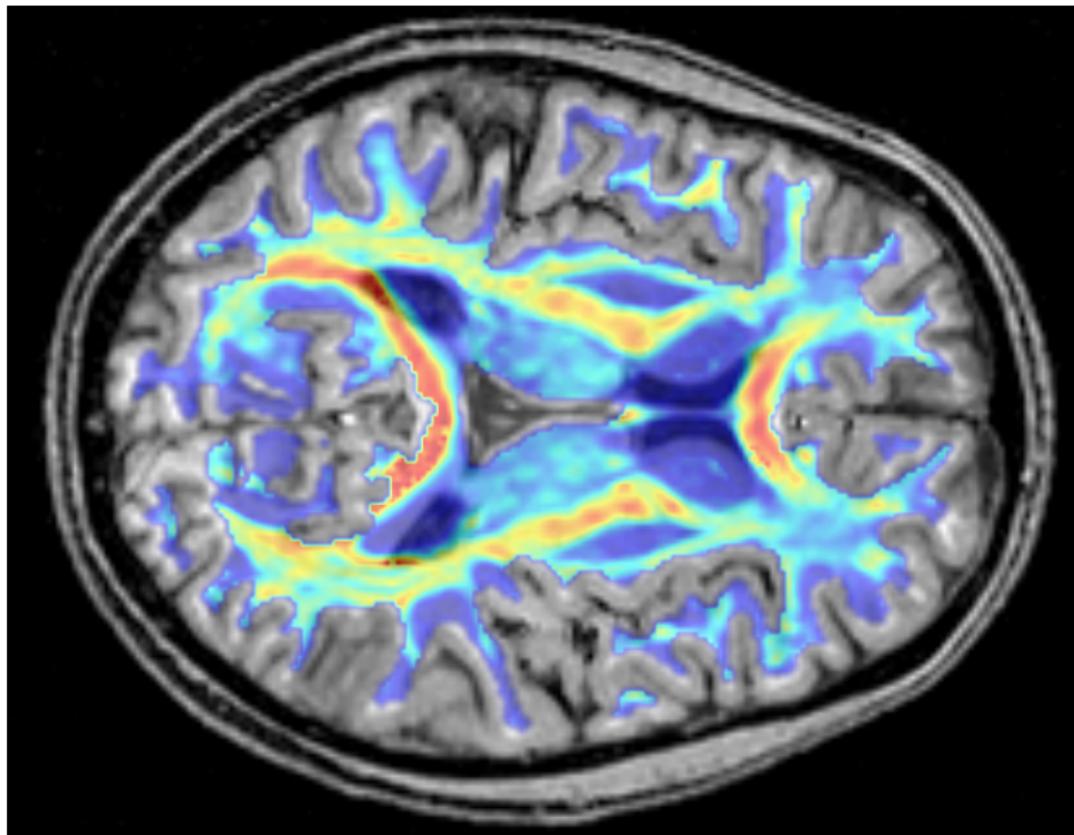
New, non-linear, variational registration approach (DA and PhD by Lars Ruthotto):



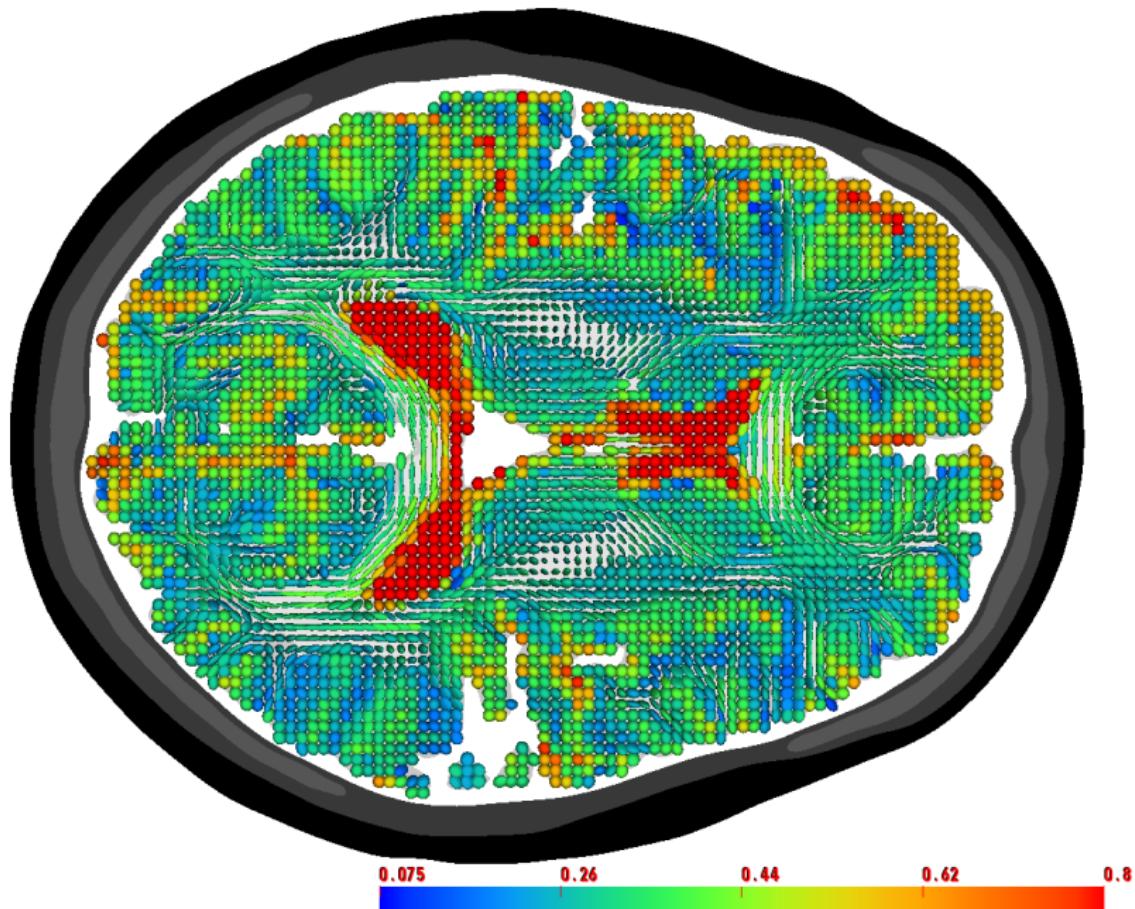
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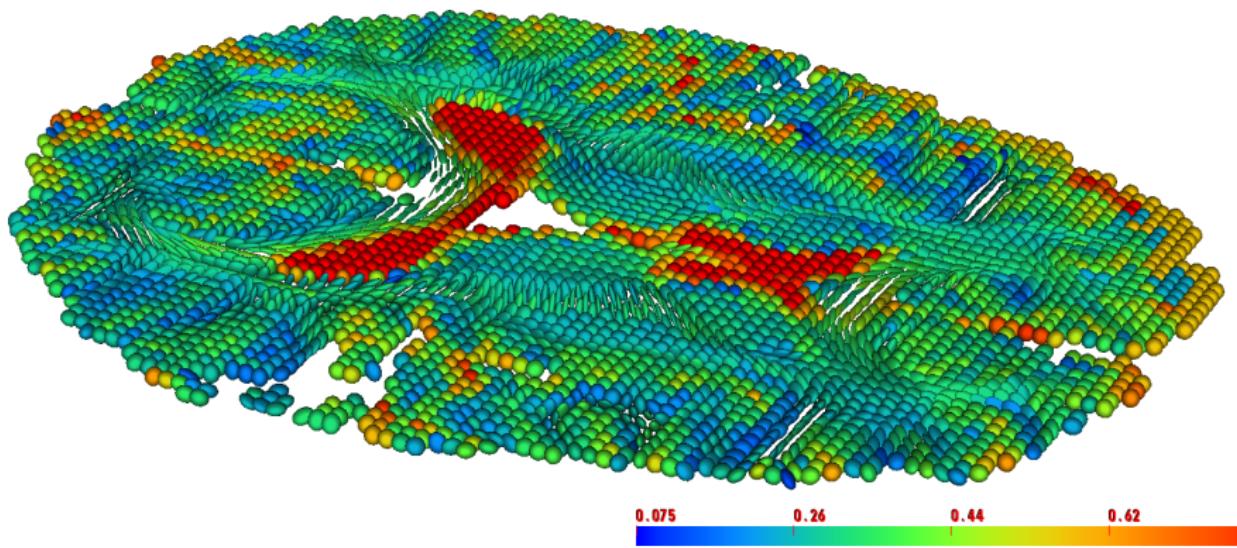
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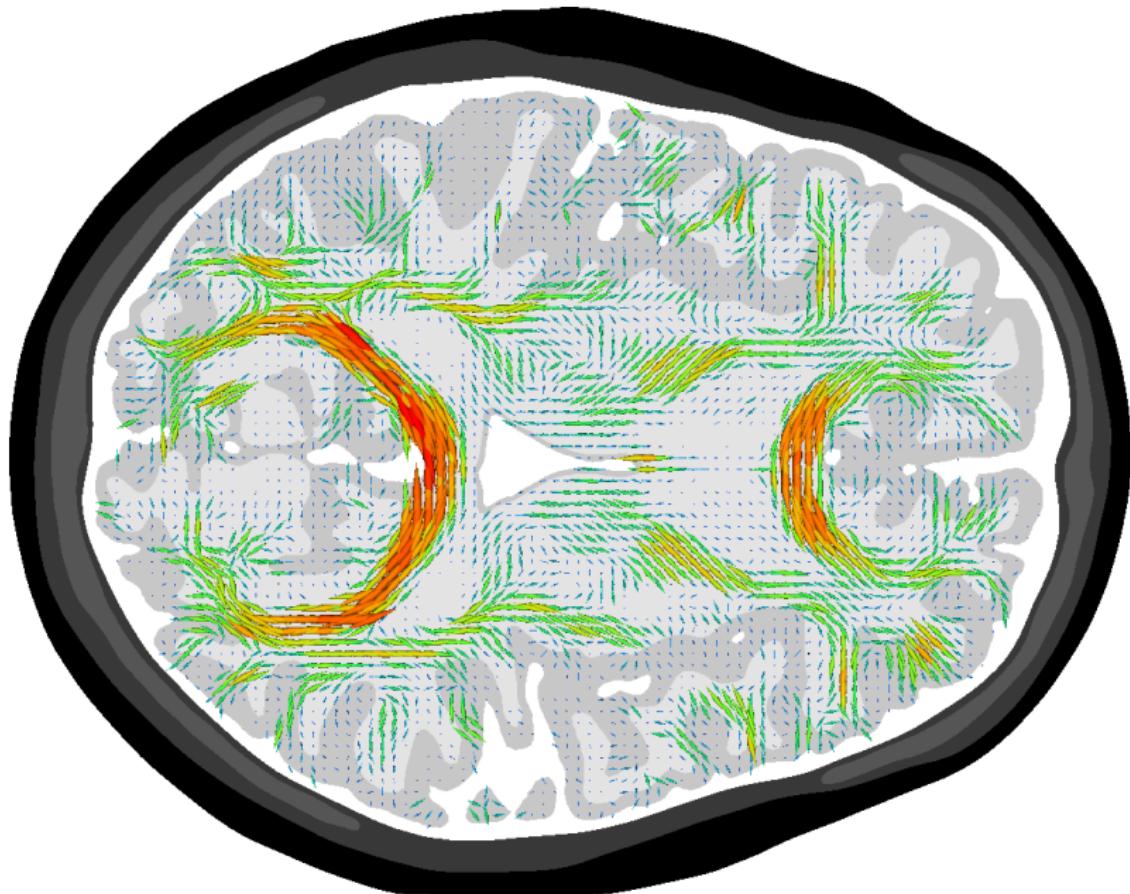
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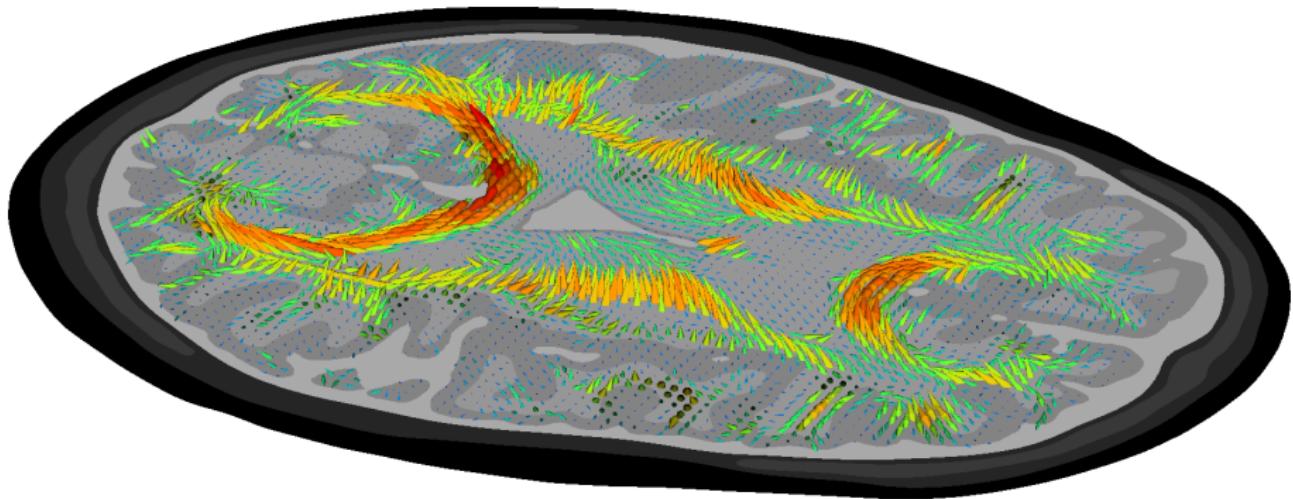
Part 1: MRI Processing, Diffusion Weighted MRI



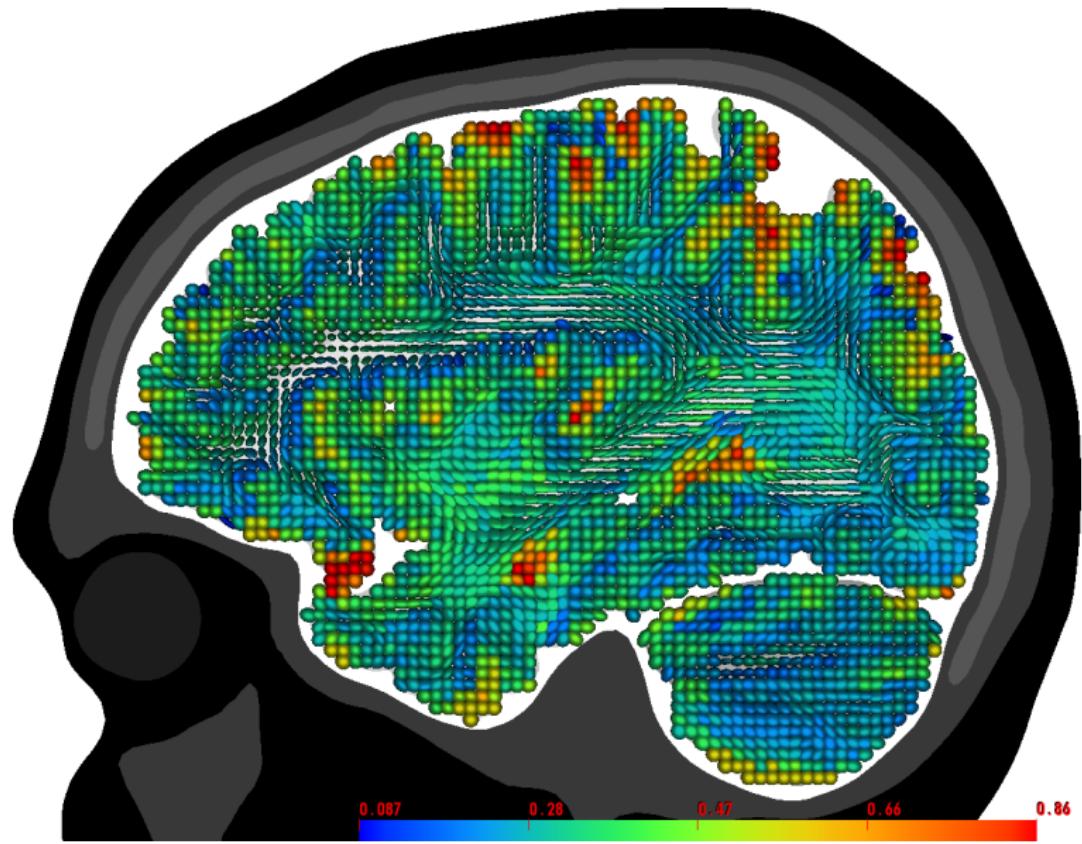
Part 1: MRI Processing, Diffusion Weighted MRI



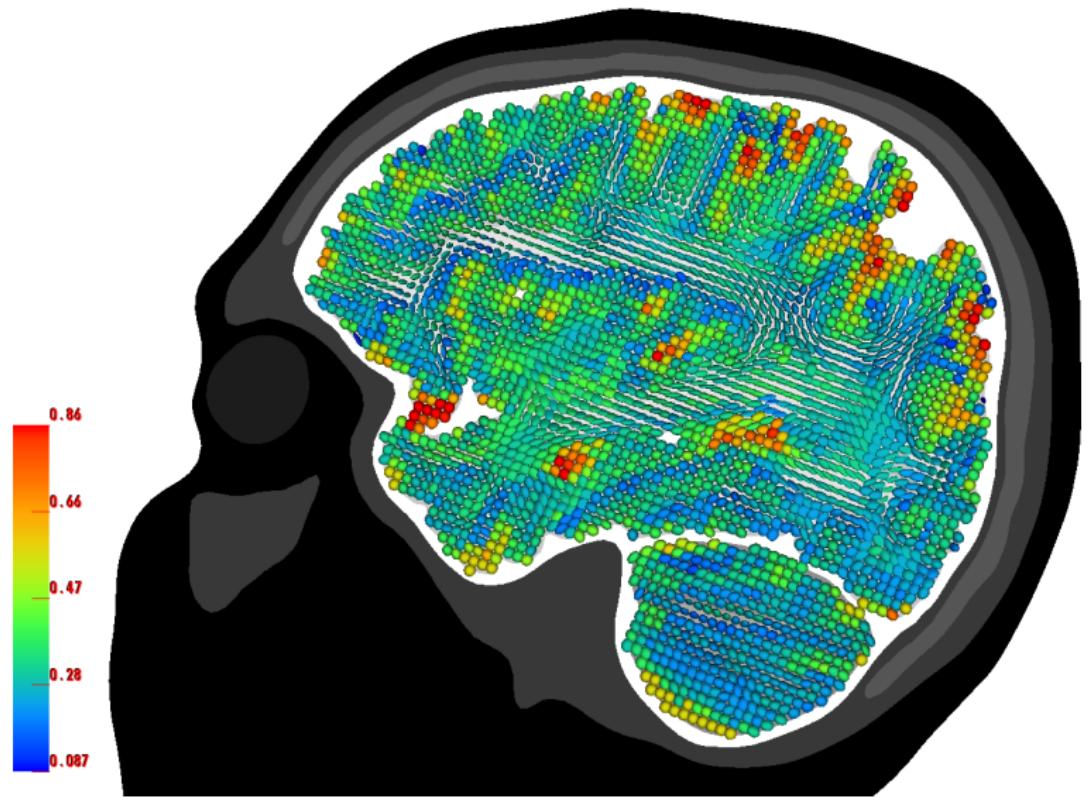
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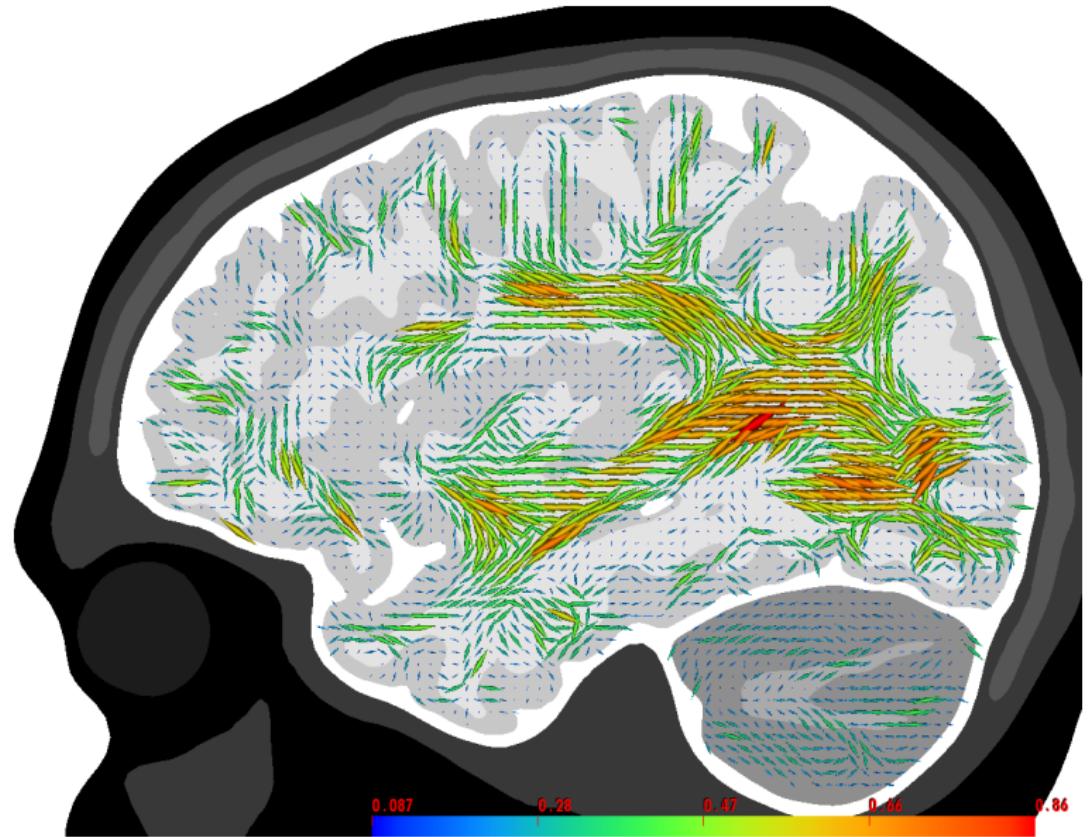
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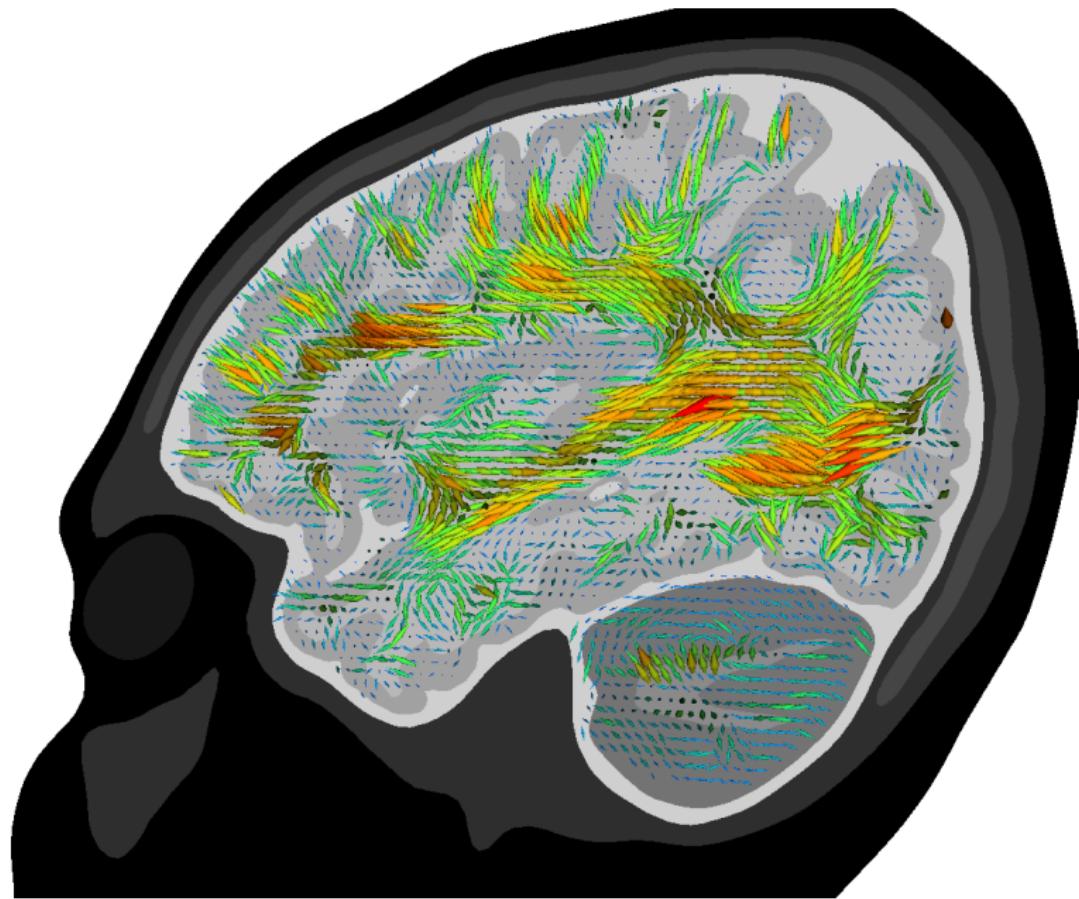
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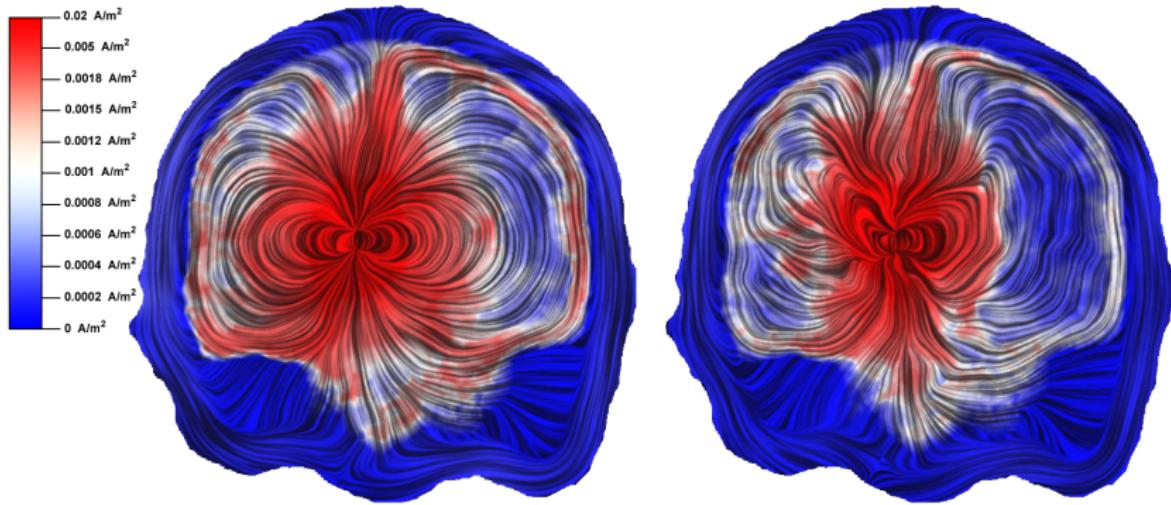


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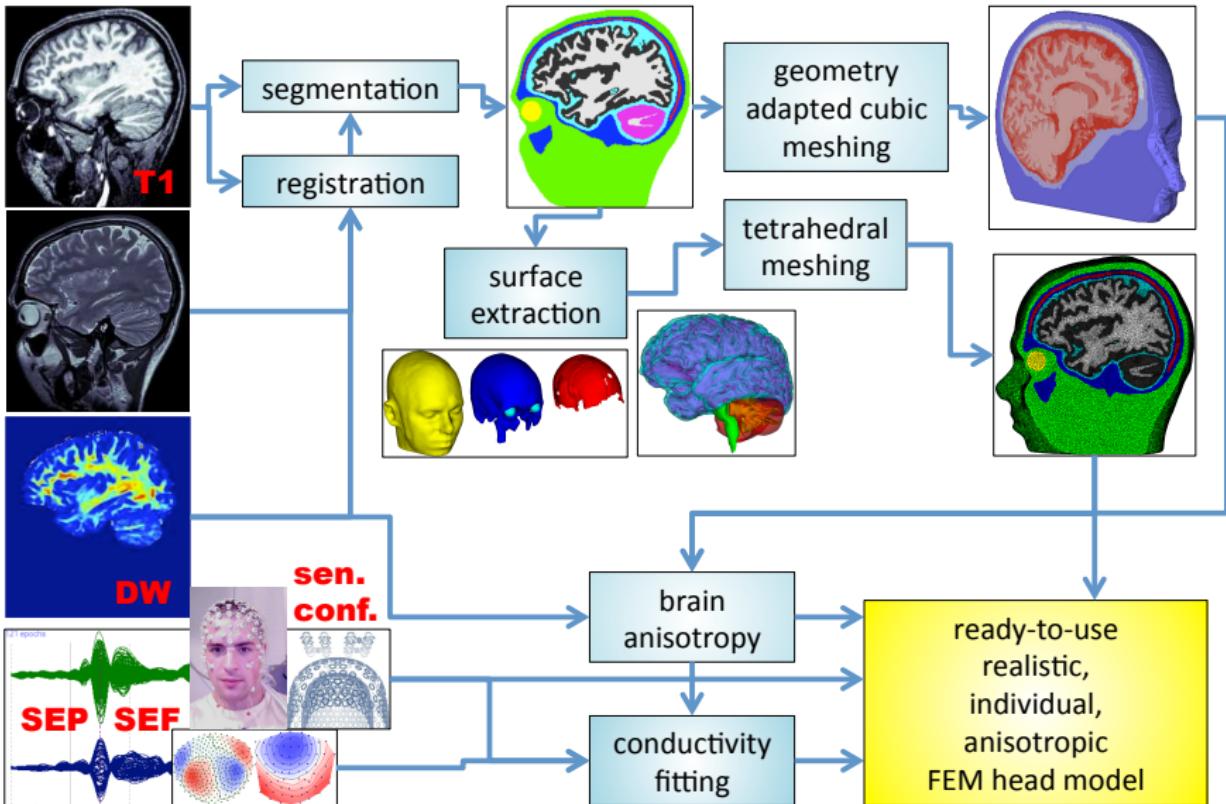


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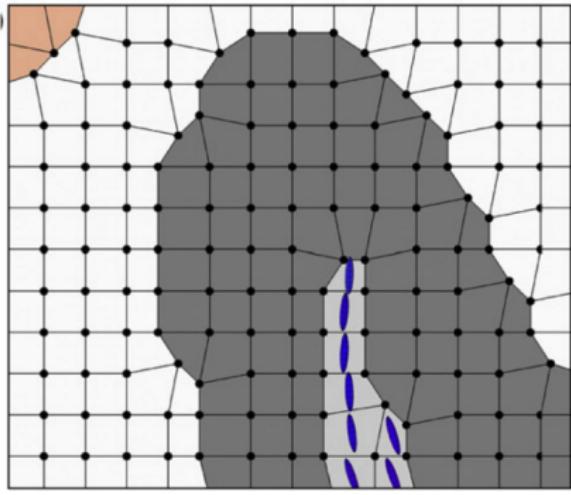
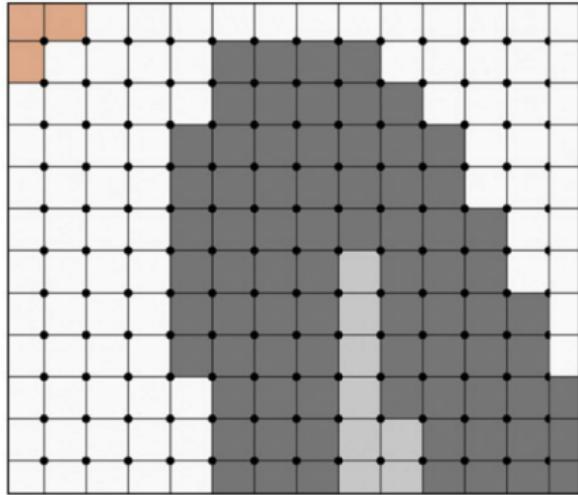
Effects of white matter anisotropy on thalamic source:



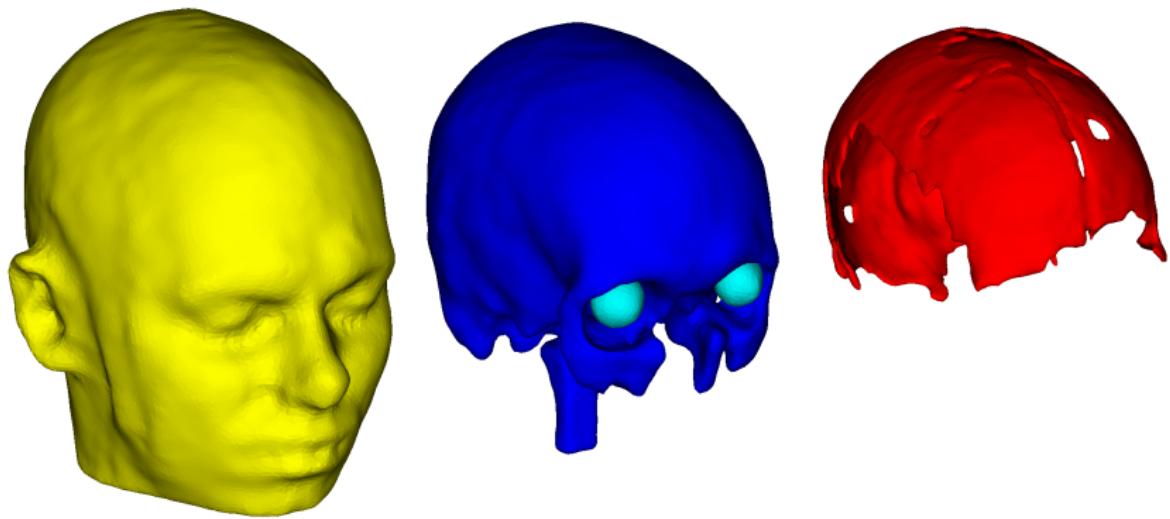
Realistic, individual head modeling for bioelectromagnetic applications



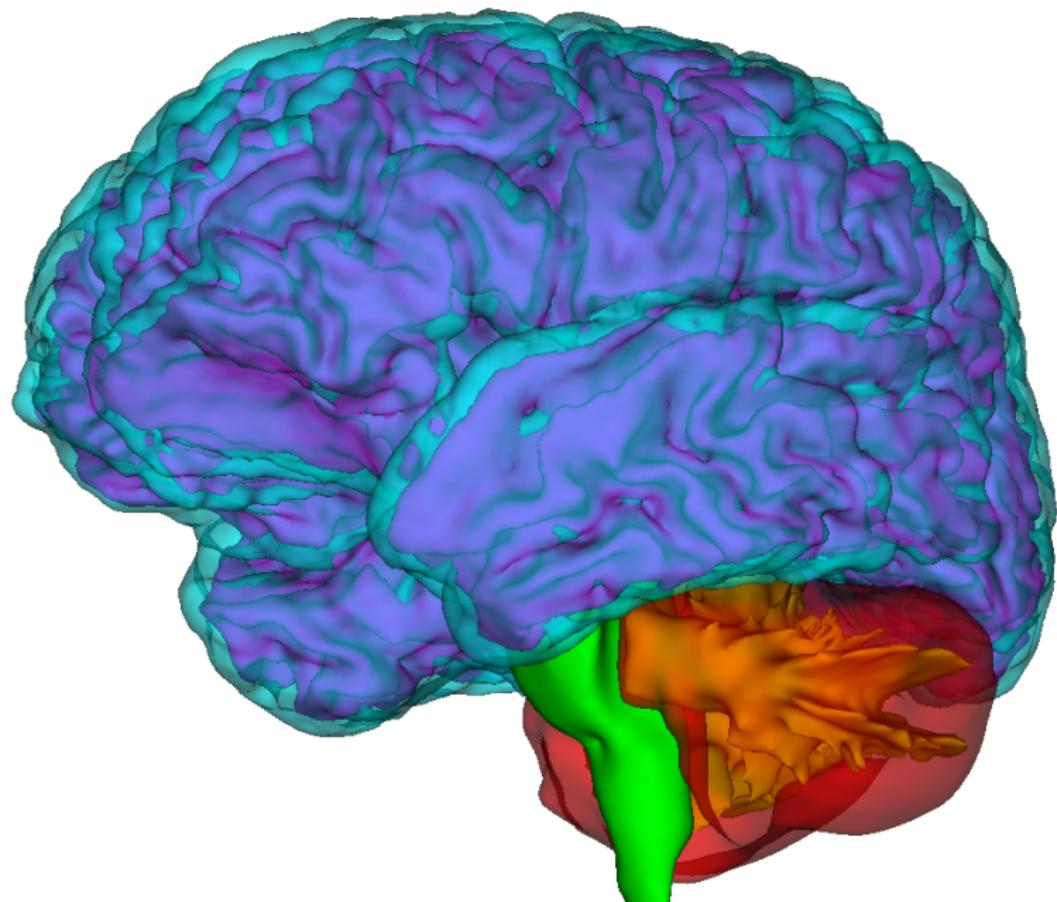
Part 2: FEM Meshing, the Cubic Way...



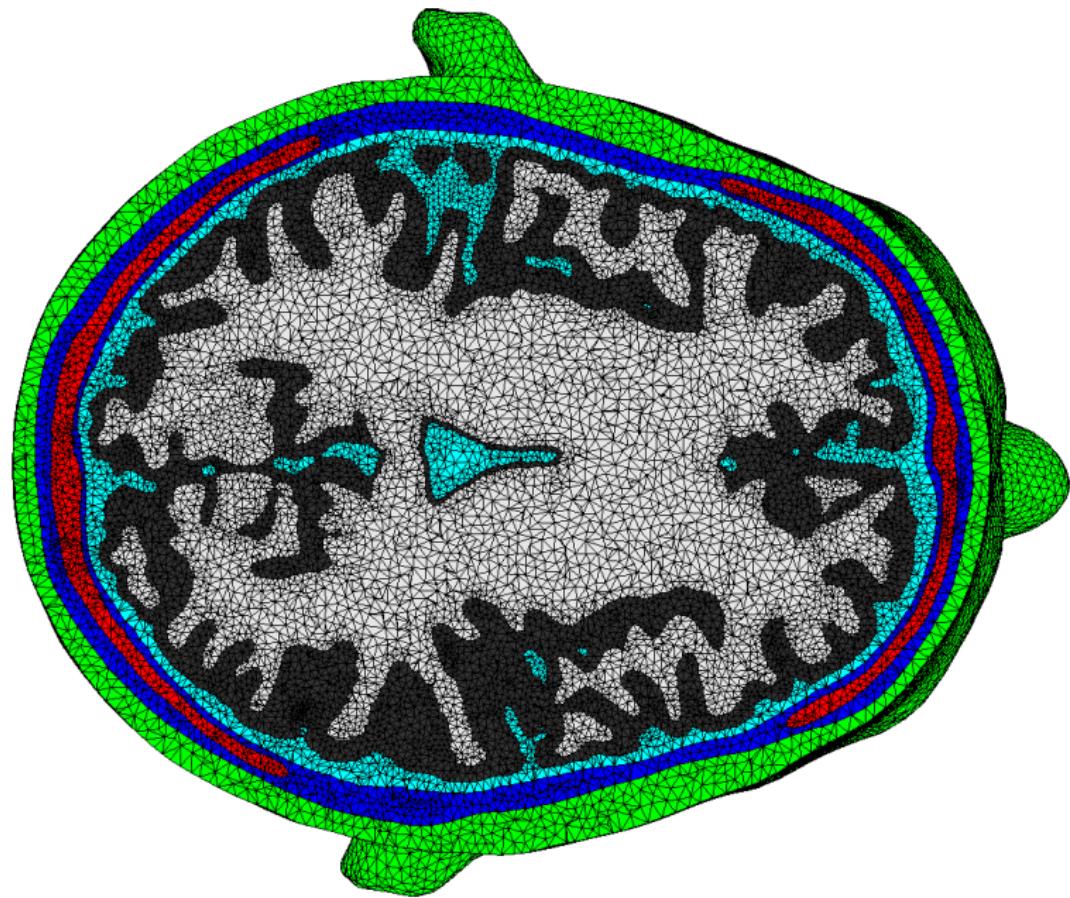
Part 2: FEM Meshing, the Tetrahedral Way...



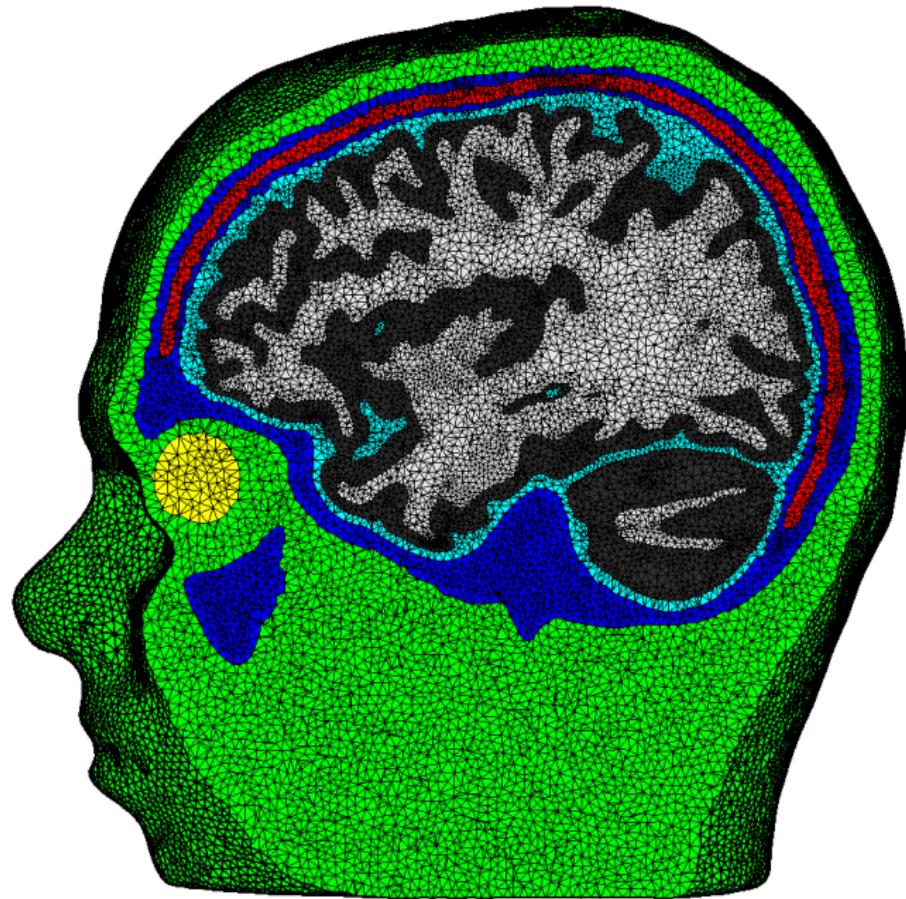
Part 2: FEM Meshing, the Tetrahedral Way...



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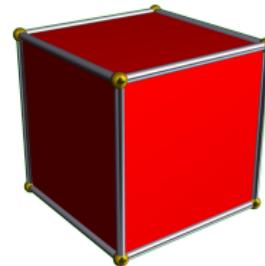
Part 2: FEM Meshing, the Tetrahedral Way...



Part 2: FEM Meshing, Pros and Cons

Cubic meshing:

- ✓ Mesh generation is simple and fast.
- ! Surfaces are still “blocky”/“staircase like”.
- ! Mesh refinement is complicated; involves *hanging-nodes*.

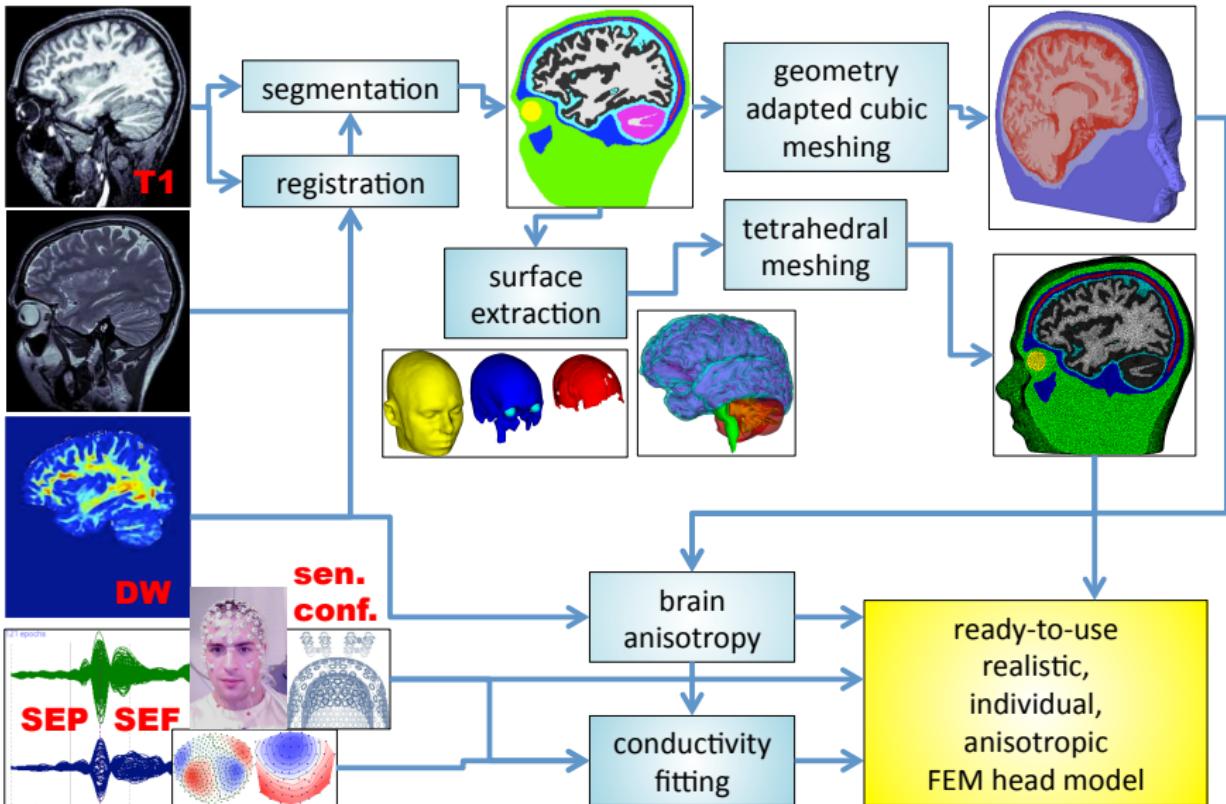


Tetrahedral meshing:

- ✓ Possibility to model thin compartments.
- ✓ Better for complex geometries.
- ✓ Mesh refinement is simple.
- ! Mesh generation is complicated and time consuming.



Realistic, individual head modeling for bioelectromagnetic applications



Part 3: Head Model Calibration

Problem: Tissue conductivity variations / uncertainty.

- ▶ **Approximation error modeling / marginalization:** Account for uncertainty explicitly (propagation of uncertainty, *Fehlerrechnung*).
- ▶ **Model calibration:** Reduce uncertainty by fitting model parameters to match known source and data.

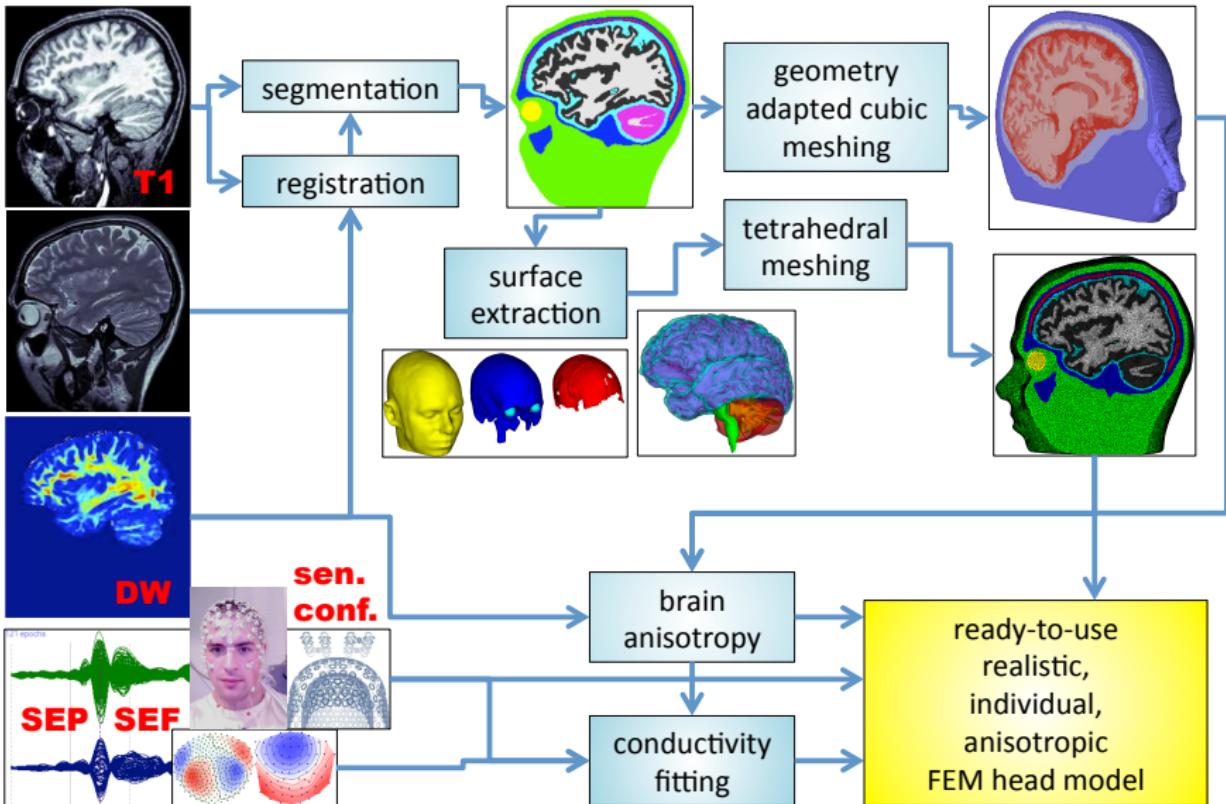
Problem for calibration: We don't really know the sources.

~~ Use complementary characteristics of EEG and MEG?

modality	sensitivity to		
	tissue geometry & conductivity	superficial sources	deep sources
EEG	high	tang: moderate radial: high	moderate
MEG	low	tang: very high radial: very low	low

- ▶ Low resolution conductivity estimation (LRCE) with SEP/SEF data.
- ▶ Work in progress: Bayesian approaches

Realistic, individual head modeling for bioelectromagnetic applications

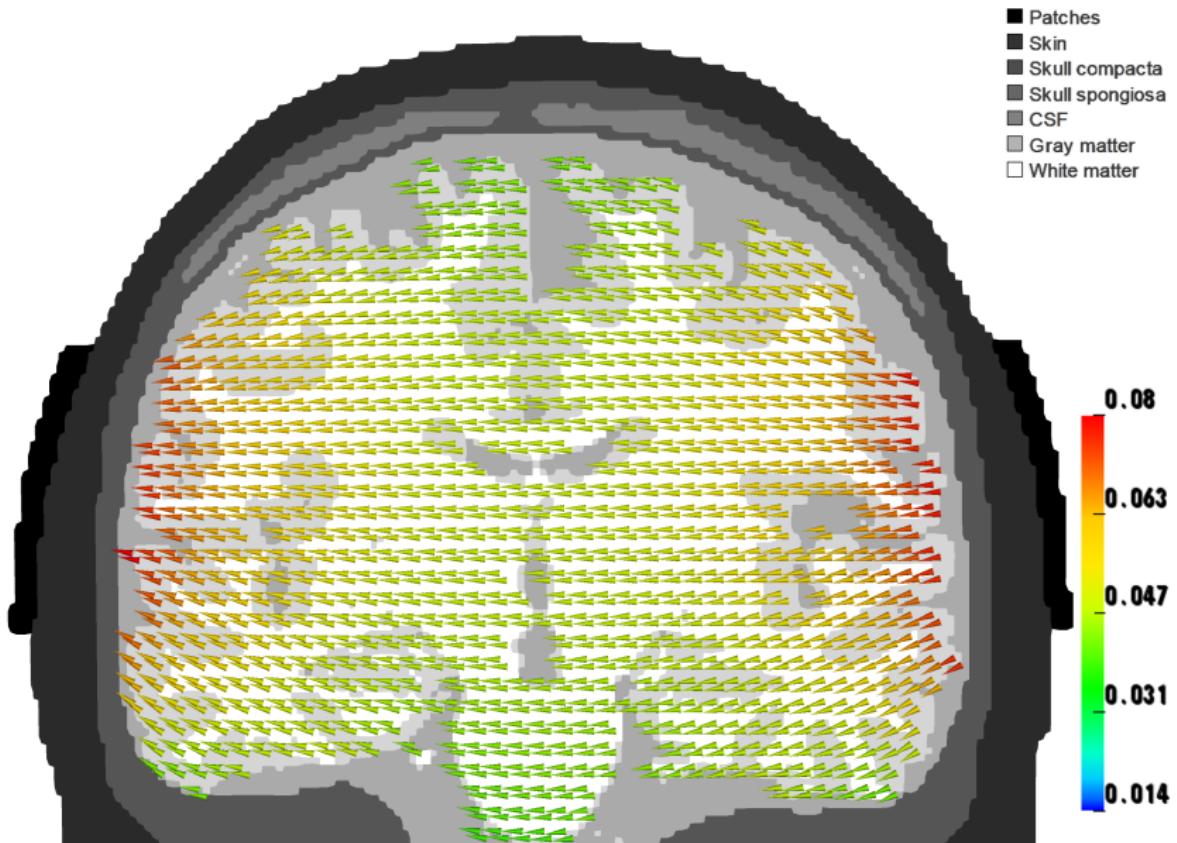


Modeling tDCS & TMS

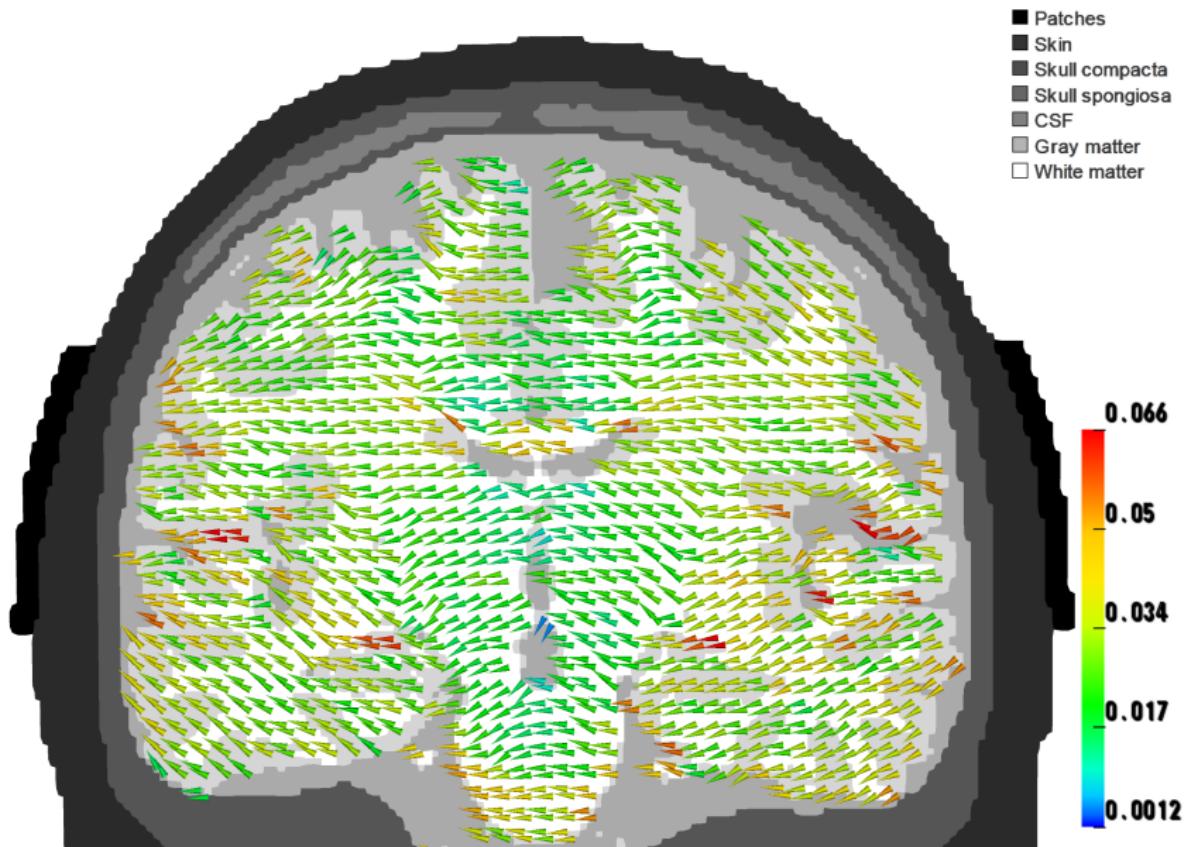
“Transcranial direct current stimulation” (tDCS) and “Transcranial Magnetic Stimulation” (TMS) \Rightarrow Means of **non-invasive** brain stimulation.

- ▶ Huge potential (alternative to invasive brain stimulation / hard medication) for therapeutic use in
 - ▶ Major depression
 - ▶ Chronic pain
 - ▶ Rehabilitation of aphasia and motor disability after stroke
 - ▶ Tinnitus
 - ▶ Parkinson's disease
 - ▶ Schizophrenia
- ▶ Neurophysiological effects are poorly understood.
- ▶ EEG/MEG source analysis and tDCS/TMS are linked via **Helmholtz reciprocity**.
 - \Rightarrow We can model it.
 - \Rightarrow It helps to understand and visualize the effects of volume conduction and volume conductor modeling.

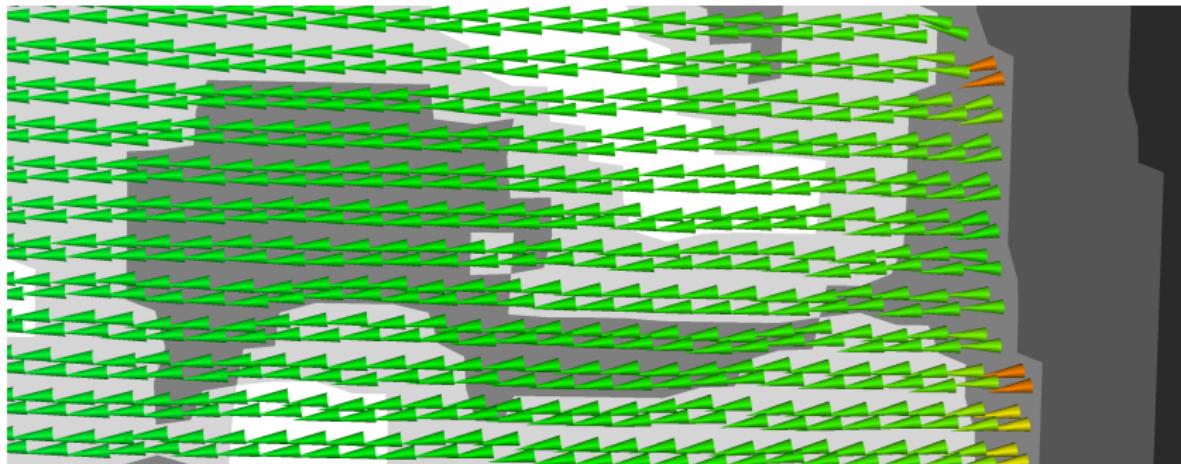
Modeling tDCS & TMS: Skin, skull, brain



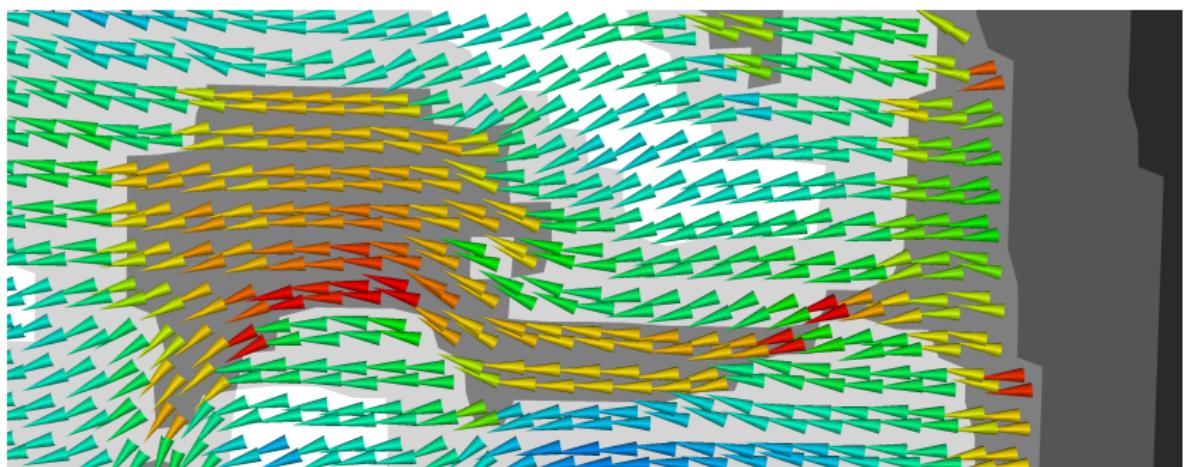
Modeling tDCS & TMS: Skin, skull comp., skull spon., CSF, GM, WM (aniso)



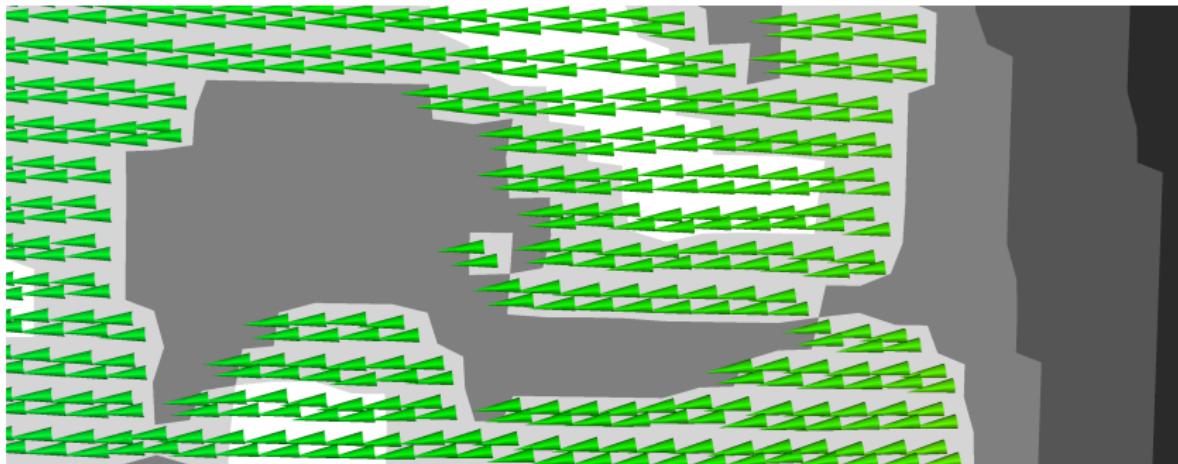
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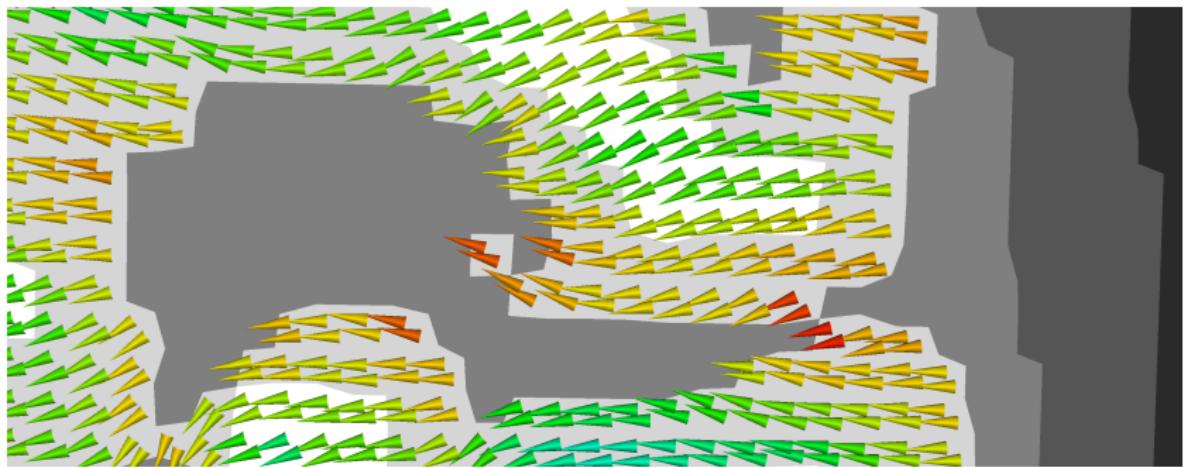
Modeling tDCS & TMS: Skin, skull comp., skull spon., CSF, GM, WM (aniso)



Modeling tDCS & TMS: Skin, skull, brain



Modeling tDCS & TMS: Skin, skull comp., skull spon., CSF, GM, WM (aniso)





Outline

Introduction

Forward Computation

Head Model Generation

Data Analysis / Inverse Problem

Ways to use the EEG/MEG Data...

1. Analysis/preprocessing in sensor space: Infer information from indirect measurements.
2. Source reconstruction: Infer information about basic underlying activity from preprocessed sensor data.
3. Higher order analysis: Infer information about processes behind the reconstructed source activity.

Data Preprocessing

EEG/MEG measurements are a **mixture** of different signals:

- ▶ Signals from different brain activities: Task/stimulus related and periodic “background” activity in different frequency bands.
- ▶ Signals from muscular activity (eye blink, heart, swallowing).
- ▶ Signals from implants, piercings, make-up (e.g., magnetic nail polish)...
- ▶ Signals from measurement/stimulus devices.
- ▶ External fields like 50Hz fields or a helicopter flying over the building.

⇒ **Unmix** by filtering, artifact removal, trial averaging, PCA, ICA, and activity specific detection algorithms (e.g., for automatic epileptic spike detection)

Then: Analysis in sensor space.

Examine event-related potentials/fields (ERP/ERF):

- ▶ Temporal latencies of components like N100, P200,
- ▶ Mismatch negativity (MMN)
- ▶ Group differences

But we want more!

Use the data to infer information about underlying brain activity in a more direct fashion.

Reminder:

Inverse Problem of EEG/MEG Source Reconstruction

Given

- ▶ **measurements b** of the electric potential u and/or of the normal-component of the magnetic field $\langle n, \mathbf{B} \rangle$ on the surface $\partial\Omega$;
- ▶ a **volume-conductor-model of $\sigma(\vec{r})$** ;
- ▶ a **source model $\mathcal{J} \subset \mathcal{D}'(\Omega, \mathbb{R}^3)$** ;

estimate the **primary current $\vec{j}^{pri} \in \mathcal{J}$** (source) that is consistent with b and the neurophysiological constraints of brain activity.

Characteristic Features of Inverse Problems

Hadamard's definition of *well-posed* problems:

1. A solution exists.
2. The solution is unique.
3. The solution depends continuously on the data.

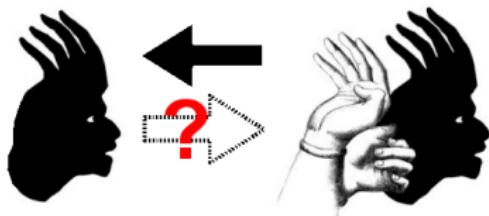
If one of the conditions does not hold, the problem is called **ill-posed**.

Inverse problems are typically ill-posed.

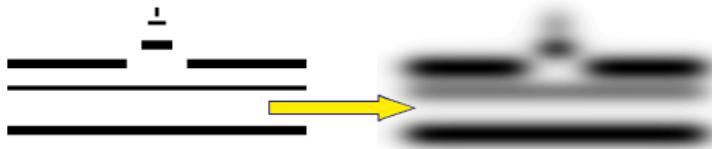


Jacques Salomon Hadamard
(1865-1963)

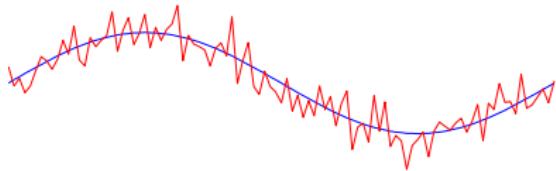
What About the Inverse Problem of EEG/MEG?



► (Presumably) under-determined



► Severely ill-conditioned



► Low SNRs

Summary: The problem is **severely ill-posed**.

Measurements **alone** are insufficient and unsuitable to determine solution.

⇒ Incorporation of **a-priori information** about the solution in an explicit or implicit way:

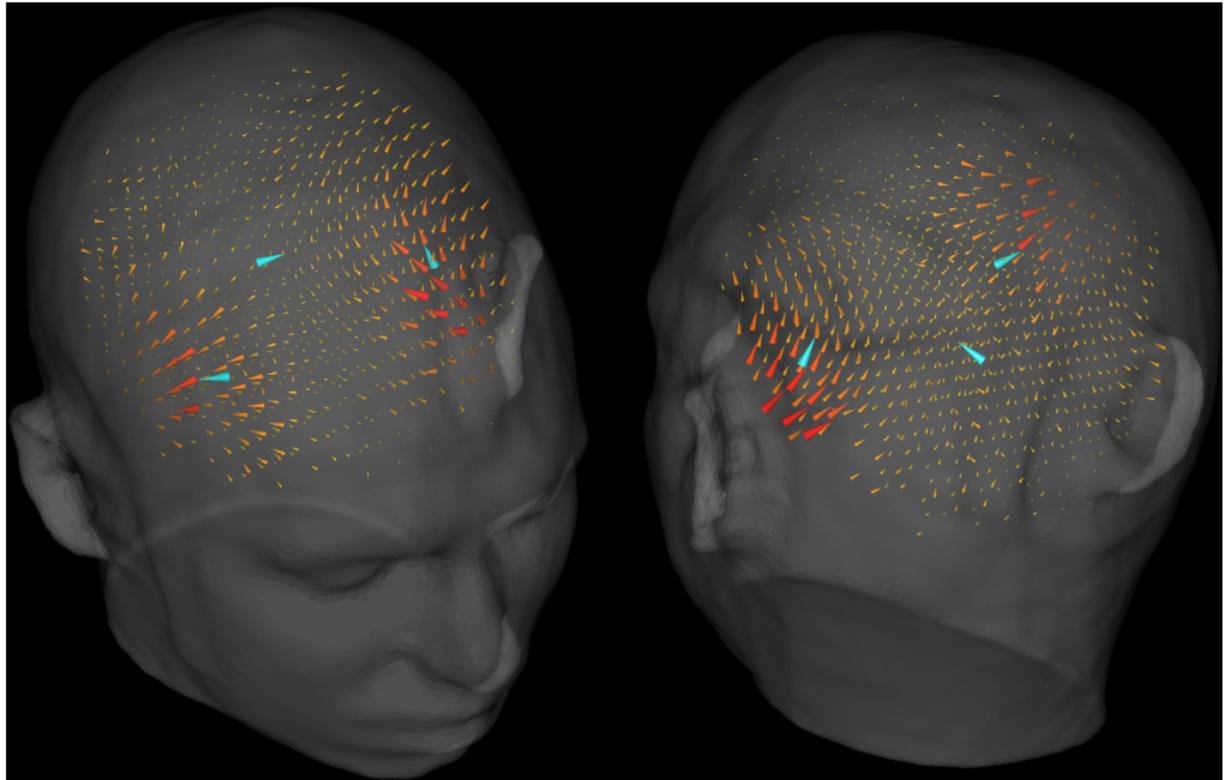
- ▶ Knowledge about general/specific brain activity?
- ▶ Integration of multimodal information (fMRI, DW-MRI, PET)?
- ▶ Mathematical formulation?
- ▶ Computational implementation?

⇒ Variety of inverse methods for EEG/MEG

My focus: Hierarchical Bayesian inference for current density reconstruction (CDR).

Current Density Reconstruction

Discretization of an underlying continuous current distribution by large number of **current dipoles** with fixed location and orientation.



Current Density Reconstruction

Lead-field matrix concept:

- ▶ $L \in \mathbb{R}^{m \times n}$; columns represent measurements at m sensors caused by the n single current dipoles.
- ▶ Linear combination of the dipoles is represented by **source vector** $s \in \mathbb{R}^n$.
- ▶ Measurements $b \in \mathbb{R}^m$ caused by s can then be calculated via:

$$b = L s$$

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Infer s from b ? Apparently ill-posed problem:

- ▶ $n \gg m \implies b = L s$ is under-determined.
- ▶ L inherits the bad condition of the continuous problem.
- ▶ Noise $\mathcal{E} \sim \mathcal{N}(0, \sigma^2 \text{Id})$ is added to signal.

Common approaches:

- ▶ Variational regularization
- ▶ (Hierarchical) Bayesian inference
- ▶ Spatial scanning methods/beamforming



We still want more! Dynamic causal modeling (DCM)

Christan Himpe oder Mario Ohlberger fragen.

Thank you
for
your attention!

Software used by our group:

- ▶ Registration: FSL, FAIR;
- ▶ Segmentation: FSL, CURRY;
- ▶ FEM Meshing: Tetgen, vgrid, iso2mesh;
- ▶ FEM Computation: **SimBio**;
- ▶ Data Preprocessing: CURRY, BESA;
- ▶ Inverse computation: Matlab;
- ▶ Volume Visualization: SCIRun;
- ▶ Everything else & software integration: Matlab;