Elaborazione di Immagini - Laurea in Bioinformatica Prof. G. Menegaz

EEG SIGNAL PROCESSING

3. Electrical Source Imaging

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Measuring Neural Activity



Non-invasive imaging techniques

Measuring hemodynamic activity









PET (positron emission tomography)

SPECT (single photon emission computed tomography) fMRI (functional magnetic resonance imaging) ASL (arterial spin labellng)

Non-invasive imaging techniques





Imaging techniques





Temporal and Spatial resolution



Electrical Source Imaging

Electrophysiological imaging of brain activity



EEG/MEG signals are mainly generated from synchronized activation of cortical pyramidal neurons located within the cortical gray matter. When pyramidal neurons are excited, the synaptic currents flowing across the cell membranes induce local excitatory postsynaptic potentials as well as magnetic fluxes, which collectively form the sources for EEG and MEG, respectively. When cortical neurons in columnar vicinity are in synchronized activation, the synaptic current flow, at a macroscopic level, approximates a **current dipole** located on cortical surface and oriented perpendicular to the local cortical surface. The configuration (e.g. **location, magnitude, and orientation**) of such current dipole can be related with EEG or MEG signals through the modeling of head volume conduction.

He et al., IEEE Transaction on Biomedical Engineering, 2011

Electrical Source Imaging (ESI) SOURCE MODEL **HEAD MODEL Spherical** Dipole Scalp electric field Source ECD assumes the underlying neuronal sources to be focal Realistic Distributed The 3D grid of solution points is considered as a possible location of a brain activity source Pascual-Marqui et al., 1994 Michel et al., 2004

Electrical Source Imaging (ESI)

Electrophysiological source imaging (ESI) is a model-based approach for imaging electrical sources associated with brain activation from noninvasive EEG or MEG measurements.

ESI entails:

- 1) forward modeling of brain sources and head volume conduction to establish a linear source-to-measurement relationship.
- 2) inverse imaging of brain electrical sources from measured EEG, via various strategies, most commonly dipole localization and distributed source imaging.

Head models

1. Sphere-shaped head models

(uniform conductivity) \rightarrow computationally efficient



2. Realistic Head Model: these numerical models allow incorporating the realistic geometry of the head and brain after reconstruction of the anatomical structure from individual or standardized MRI data sets.

Boundary Element Method (BEM):

gathers a more realistic shape of brain compartments of isotropic and homogeneous conductivities by using closed triangle meshes



Meijs et al., 1987

Finite-Element Method (FEM): allow better accuracy than the BEM because they allow a better representation of the cortical structures





Bertrand et al., 1991, Awada et al., 1997



Forward Problem

EEG forward problem describes the distribution of electric potentials for given source locations, orientations, and signals. The relationship between EEG signals (Φ) and cortical current source dipoles (J) can be represented by a linear system:



Inverse Problem

The inverse problem is used to convert measured electric potentials (EEG) into current densities of the sources.

The inverse problem is **ill-posed** because an infinity of different source configurations can produce the same EEG scalp distribution Nurez and Srinivasan, 2006



Scherg and Von Cramon, 1985, Liu et al., 1998, Babiloni et al., 2003, Michel et al., 2004

Inverse solutions

1. DISCRETE

Equivalent current dipole (ECD) approach where the signals are assumed to be generated by few focal sources.



2. DISTRIBUTED

Linear distributed (LD) approaches

which consider that the dipoles are regularly distributed in cerebral volume according to a 3D grid and where all possible source locations are considered simultaneously.



Inverse Problem \rightarrow Inverse solutions

Equivalent current dipole approach

ECD assumes the underlying neuronal sources to be focal

Number of sources < number of sensors OVERDETERMINED PROBLEM

The lead field matrix has more rows (number of sensors) than columns (number of sources)

Source model and source waveforms



Linear distributed

The 3D grid of solution points is considered as a possible location of a brain activity source

Number of sources >> number of sensors UNDERDETERMINED PROBLEM

The lead field matrix has more columns than rows

3D volume image for each time point



Equivalent current dipole

Parameter Estimates

To determine the best location of the sources, the squared error between the surface electric potential map generated by dipoles using a certain forward model and the actual measured potential map is calculated.

Methods

- Dipole fitting methods [Scherg, 1990]
- linear constrained minimum variance (LCMV) beamformers [Van Veen et al., 1997]
- the multiple signal classification (MUSIC) [Mosher and Leahy, 1998]

- ...

Limitations

ECD models have some limits in estimating in advance the number of dipoles and localizing extended sources.

The center of mass of the cortical activity is localized, but the distribution and the extension of the activity remain to be determined [He et al., 2011].

Linear distributed approches

The estimation of the dipole source configuration \mathbf{J} is provided by the solution of the linear system:

 $\Phi = \mathbf{K} \mathbf{J} + c\mathbf{1}$

$$\Phi \in \mathbb{R}^{N_E \times 1}$$
 with $\Phi = (\Phi_1, \Phi_2, ..., \Phi_{N_E})^T$: is a NE × 1 known matrix of measurements of scalp electric potential differences

NE: number of electrodes

 $1 \in \mathbb{R}^{N_E \times 1}$: is a vector of ones

 $\mathbf{J} \in \mathbb{R}^{(3N_V \times 1)}$: matrix of current densities at N_V points within the brain volume

c : accounts for the physical nature of electric potential

 $K \in R^{N_E \times (3N_V)}$: transfer matrix or lead field matrix

$$\mathbf{K} = \begin{pmatrix} \mathbf{k}_{1,1}^{T} & \mathbf{k}_{1,2}^{T} & \cdots & \mathbf{k}_{1,N_{V}}^{T} \\ \mathbf{k}_{2,1}^{T} & \mathbf{k}_{2,2}^{T} & \cdots & \mathbf{k}_{2,N_{V}}^{T} \\ \cdots & & & \\ \mathbf{k}_{N_{E},1}^{T} & \mathbf{k}_{N_{E},2}^{T} & \cdots & \mathbf{k}_{N_{E},N_{V}}^{T} \end{pmatrix} \qquad \mathbf{k}_{e,v} \in \mathbb{R}^{3\times 1} : \text{ are determined by all proprieties of the head, i.e. geometry and conductivity profile.}$$

Pascual-Marqui, Seikihara, Brandeis and Michel, Electrical Neuroimaging

SOLUTION SPACE

Particular inverse solutions

Minimum norm least square (MN) solution

$$\begin{split} \min_{\mathbf{J},c} \Psi & \text{with } \Psi = \left\| \Phi - \mathbf{K}\mathbf{J} - c\mathbf{1} \right\|^2 + \lambda \mathbf{J}^T \mathbf{J} & \text{[Hamalainen and Ilmoniemi, 1984]} \\ \text{Solution: } \hat{\mathbf{J}} = \mathbf{T}\Phi & \text{with } \mathbf{T} = \mathbf{K}^T \mathbf{H} (\mathbf{H}\mathbf{K}\mathbf{K}^T\mathbf{H} + \lambda\mathbf{H})^+ \\ \mathbf{H} = \mathbf{I} - \frac{1}{N_E} \mathbf{1} \mathbf{1}^T & \text{: denotes the } N_E \mathbf{x} N_E \text{ average reference operator} & \lambda & \text{: Tikhonov regularization parameter} \\ & \mathbf{I} & \mathbf{I} N_E \mathbf{x} N_E \text{ identity matrix} \\ & \mathbf{I} & \mathbf{I} N \mathbf{x} \mathbf{1} \text{ matrix comprised of ones} \\ & N_E & \text{: number of electrodes} \end{split}$$

Minimum norm solutions favors superficial sources and misplaces deep sources

Weighted minimum-norm least squares (WMN) solution

$$\min_{\mathbf{J},c} \Psi_D \quad \text{with} \quad \Psi_D = \left\| \Phi - \mathbf{K}\mathbf{J} - c\mathbf{1} \right\|^2 + \lambda \mathbf{J}^T \mathbf{D}\mathbf{J} \qquad \text{[Pascual-Marqui et al., 1994,} \\ \text{Gorodnitsky et al., 1995,} \\ \text{Grave de Peralta and Gonzalez,} \\ \text{1998]} \\ \text{Solution:} \quad \hat{\mathbf{J}}_D = \mathbf{T}_D \Phi \qquad \text{with} \quad \mathbf{T}_D = \mathbf{D}^{-1} \mathbf{K}^T \mathbf{H} (\mathbf{H}\mathbf{K}\mathbf{D}^{-1}\mathbf{K}^T \mathbf{H} + \lambda \mathbf{H})^+$$

D is used to "re-weight" the solution, i.e. to incorporate some prior knowledge about the spatial distribution of the source activity

Particular inverse solutions

Low-resolution electromagnetic tomography algorithm (LORETA)

 $\min_{\mathbf{J},c} \Psi_{W} \text{ with } \Psi_{W} = \| \Phi - \mathbf{K}\mathbf{J} - c\mathbf{1} \|^{2} + \lambda \mathbf{J}^{T}W\mathbf{J} \qquad \text{[Pascual-Marqui et al., 1994, 1999]}$ Solution: $\hat{\mathbf{J}}_{W} = \mathbf{T}_{W}\Phi \text{ with } \mathbf{T}_{W} = \mathbf{W}^{-1}\mathbf{K}^{T}\mathbf{H}(\mathbf{H}\mathbf{K}\mathbf{W}^{-1}\mathbf{K}^{T}\mathbf{H} + \lambda\mathbf{H})^{+}$ $\mathbf{J}^{T}\mathbf{W}\mathbf{J} = \sum_{v} \| \mathbf{j}_{v} - AveNeighb(\mathbf{j}_{v}) \|^{2} \qquad AveNeighb \text{ : average of current densities in the immediate neighborhood of point v, excluding point v}}$

LORETA minimizes the squared norm of the Laplacian of the weighted 3D current-density vector field. It incorporates the "smoothness assumption" selecting the inverse solution of the measured data with the smoothest distribution in space.

Local Autoregressive Average (LAURA)

$$\mathbf{J}^{T}\mathbf{W}_{Laura}\mathbf{J} = \sum_{v} \left\| \mathbf{j}_{v} - WeightedAveNeighb(\mathbf{j}_{v}) \right\|^{2}$$

[Grave de Peralta and Gonzalez, 2002]

The estimated activity at one point depends on the activity at neighboring points according to electromagnetic laws (i.e. the strength of the source declines with the inverse of the squared distance of the potential field).

Effect of the electrode distribution on the estimation of the source



 \rightarrow Occipital activation

Michel et al., Clinical Neurophysiol 2004

Effect of the number of electrodes on source localization



Michel et al., Clinical Neurophysiol 2004

Identification of spontaneous EEG activity:

Interictal activity of epileptic patients
Brain rhythm in resting state (e.g. alpha rhythm)
Sleep waves (spindle)

And evoked:

4. Evoked potential (EP)

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Drug resistant focal epilepsy

The feature of partial seizures is the presence of abnormal electrical activity that originates from an epileptic foci.

Seizures prevent healthy development and may cause brain damage.

Treatment

35% of focal epilepsy patients do not respond to medication, and must undergo surgical resection of the epileptic focal points.

Surgery requires accurate localization of the foci.

Candidate for epilepsy surgery

Persistent seizures despite appropriate pharmacological treatment
Impairment of quality of life due to ongoing seizures



They do not always provide the localizing accuracy required for surgical planning

Candidate for epilepsy surgery

- Standard presurgical workup does not always provide the localizing accuracy and precision required for surgical planning
- Non-invasive imaging methods are useful to correctly identify the activity before the surgery treatment



NON-INVASIVE IMAGING TECHNIQUES IN EPILEPSY M

MULTIMODAL APPROACH

Each bears limitations that can be partly overcome by combining their results!



Bagshaw et al., 2006; Brodbeck et al., 2010; Groening et al., 2009; Storti et al., 2012; Vulliemoz et al., 2010ab

Invasive examinations

Stereo EEG

Intracerebral electrodes are implanted into the selected brain area to record the electrical activity during epileptic seizures, thus contributing to define with accuracy the boundaries of the epileptogenic zone, i.e. the area of brain generating the seizures which should be eventually surgically resected to achieve freedom from epileptic attacks.

- The implantation of intracerebral electrodes is carried out on the basis of non-invasive examinations.

- It is used in patients with epilepsy not responding to drug treatment, and who are potential candidates to receive brain surgery in order to control seizures.



High-density EEG and ESI



Electrodes registration





Electrodes registration : practical issues

1. Electrodes floating in the air below the MRI cut





2. Registering electrodes to a corrupted MRI (irregular surface) \rightarrow some electrodes could land inside the head





MRI (brain and grey matter extraction): issues

3. Skull-stripped brain should contain grey and white matters and void of cerebellum



MRI (brain and grey matter extraction): issues

3. Skull-stripped brain should contain grey and white matters and void of cerebellum







Effect of the number of electrodes on the estimation of the source



The source localization can be biased by a low number of electrodes [Michel et al., 2004] \rightarrow increasing the number of electrodes the localization can be improved.

Electrical source imaging – rising phase





Electrical source imaging – peak



ESI and Clinical Validation

Differences between inverse solution algorithms and the post-surgical MRI



Michel et al., Electrical Neuroimaging, 2009

Identification of spontaneous EEG activity:

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Electrical Source Imaging of Alpha Rhythm

Scalp EEG activity shows oscillations at a variety of frequencies. Several of these oscillations have characteristic frequency ranges, spatial distributions and are associated with different states of brain functioning (e.g. waking and the various sleep stages). These oscillations represent synchronized activity over a network of neurons.

The localization of EEG rhythms in normal subjects, without any paradigm (attention, visual and auditory stimuli) are obtained from the EEG signal filtered for specific frequency bands.

Electrical Source Imaging of Brain Rhythms





Identification of spontaneous EEG activity:

Interictal activity of epileptic patients
Brain rhythm in resting state (e.g. alpha rhythm)
Sleep waves (spindle)

And evoked:

4. Evoked potential (EP)

Sleep

Stage	Frequency (Hz)	Amplitude (micro Volts)	Waveform type
awake	15-50	<50	
pre-sleep	8-12	50	alpha rhthym
1	4-8	50-100	theta
2	4-15	50-150	splindle waves
3	2-4	100-150	spindle waves and slow waves
4	0.5-2	100-200	slow waves and delta waves
REM	15-30	<50	

A **sleep spindle** is a burst of oscillatory brain activity visible on an EEG that occurs during stage 2 sleep. Sleep spindles are bursts of waxing and waning oscillations in the frequency of 10 to 15 Hz and last from 0.5 to 2 s.



Electrical Source Imaging of Brain Rhythm

Spindle



Alpha rhythm

Mu rhythm



Manshanden et al., Clin Neurophysiol 2002

Identification of spontaneous EEG activity:

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Electrical Source Imaging of EP

256-channel somatosensory evoked potential (SSEP) after right median nerve stimulation 256-channel visual evoked potential (VEP) after fullfield checkerboard reversal. 192-channel auditory evoked potential (AEP) after short tones. 64-channel olfactory evoked potential after unilateral nostril stimulation with hydrogen sulfide.









Lascano et al., J. Clin. Neurophysiol 2009