

Event-Related Brain Dynamics I



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21st EEGLAB Workshop

Santa Margherita Ligure, Italy

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Human Functional Brain Imaging

Some human brain imaging milestones

1926 ~1st human EEG recordings

EEG era

1938 1st EEG spectral analysis

1962 ~1st computer ERP averaging (CAT)

ERP era

1979 1st event-related desynchronization

1993 1st fMRI BOLD recordings

fMRI era

1993 1st broadband ERSP

1995 1st multisource EEG filtering by ICA

2009 ~1st commercial dry electrode EEG toys

fEEG / BMI / MoBI era ...

FIGURE 1-2.—Sample of the first EEG tracing taken at the Bradley Hospital, E. Providence, Rhode Island, by H. Jasper and L. Carmichael. Subject: Carl Pfaffmann. Date: July 9, 1934. Record, which shows prominent alpha rhythm of about 11.5 per second, was made with a Westinghouse, galvanometer-type mirror oscillograph. Time line above: 25 Hz.

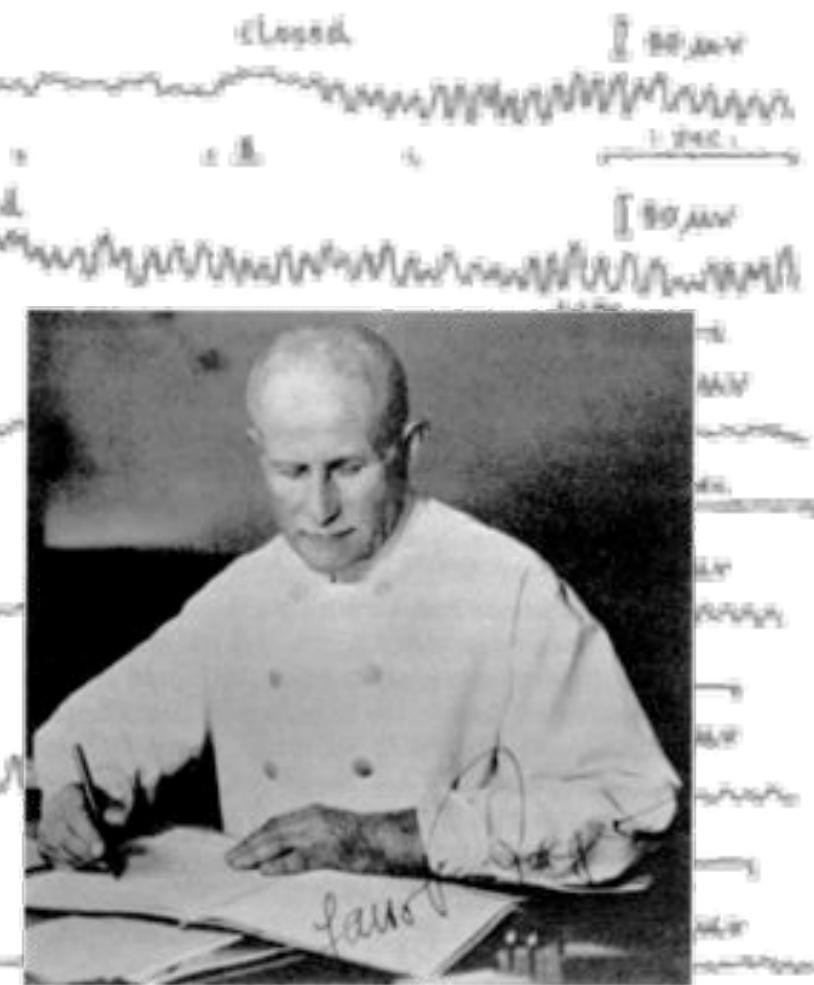
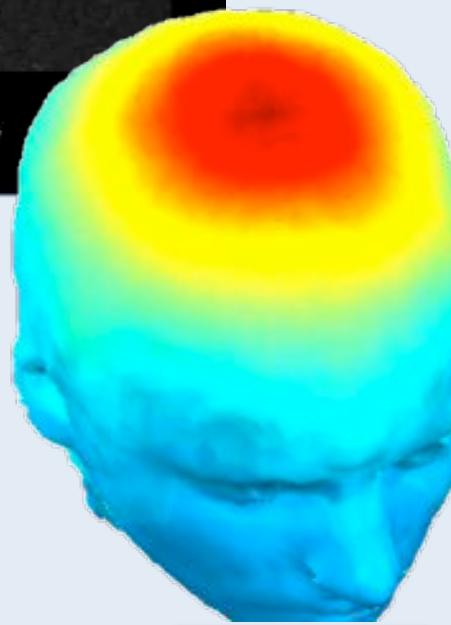


FIGURE 1-1.—Professor Hans Berger (1873–1941), neuro-psychiatrist, University of Jena, Jena, Germany, first to discover and describe in 1929 a unique kind of electrical activity recorded from the brain of man, which he named the electroencephalogram (Elektrenkephalogramm).

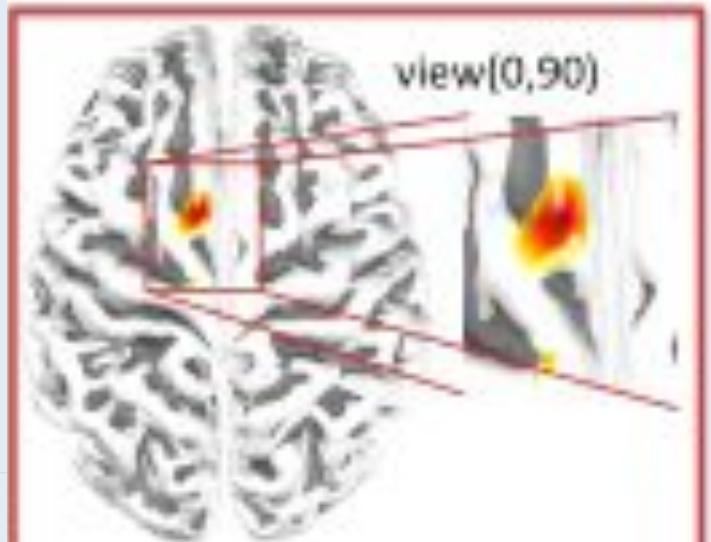


Whole Scalp 1957

1957



1997



view(-90,60)



2016

Functional Brain Imaging

Hemodynamic imaging

= imaging local brain

Energy

Direct 3-D inverse model,
but quite slow & indirect
as well as expensive
and heavy (non-portable)

Electromagnetic imaging

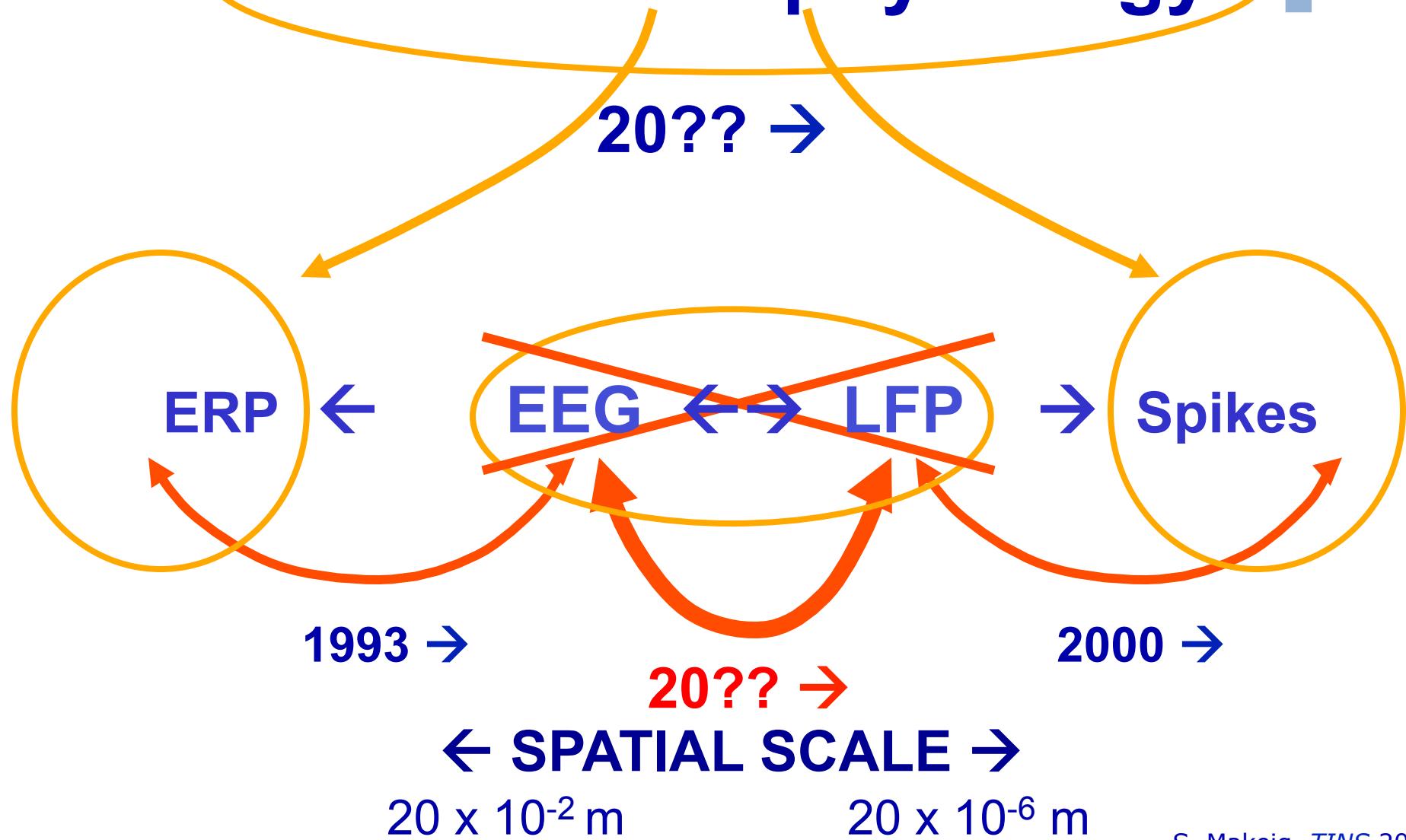
= imaging local cortical

Synchrony

3-D imaging requires model,
but quite fast & direct measure
of one aspect of cortical activity –
local spatial field coherence..

- **EEG is inexpensive**
- **EEG is well tolerated**
- **EEG is lightweight / wearable / mobile**

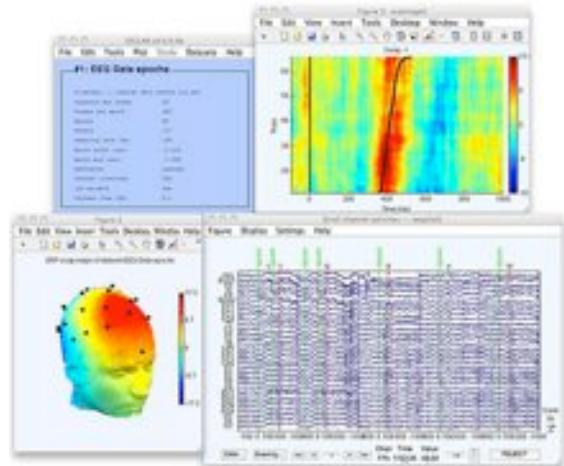
Brain Electrophysiology?

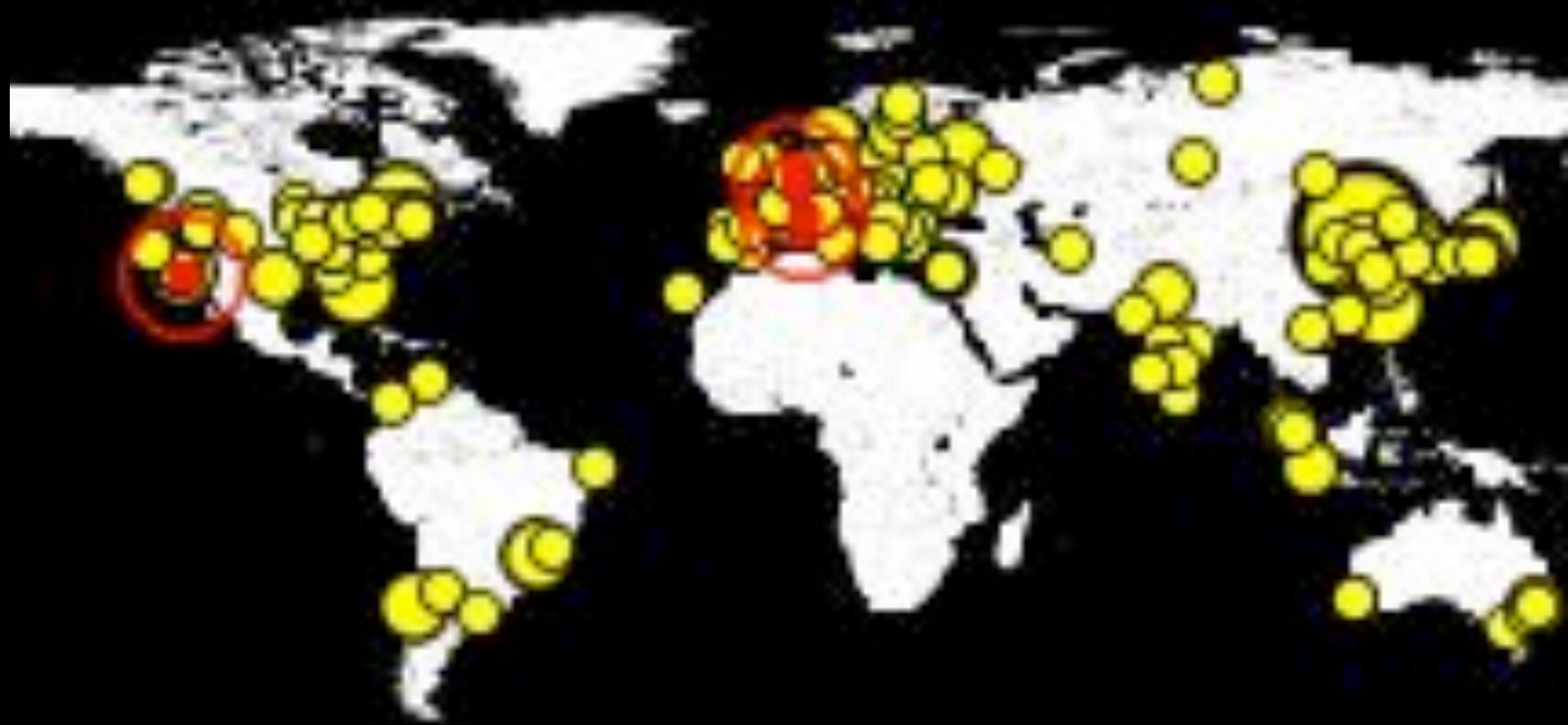




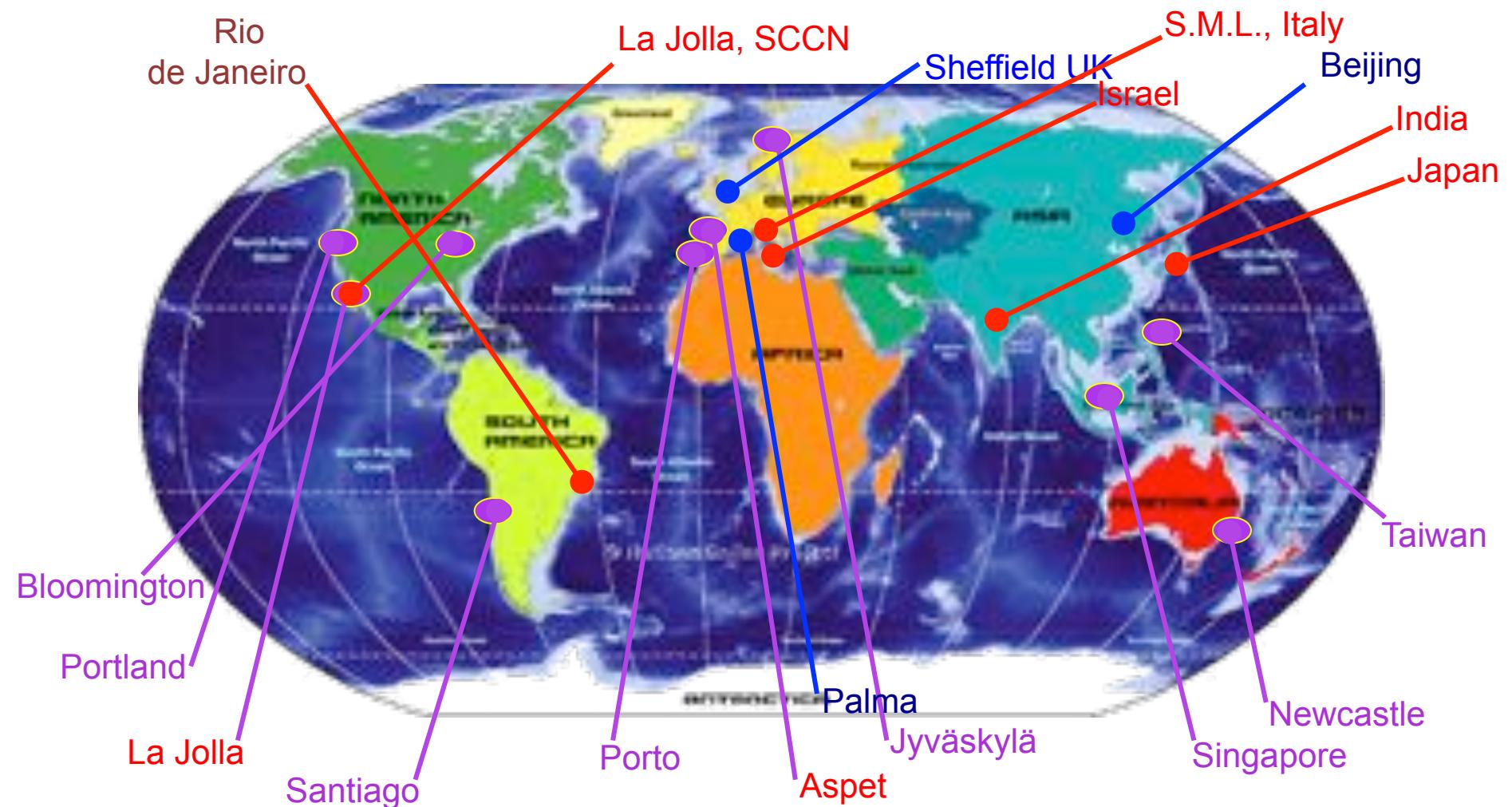
EEGLAB History

- 1993 – ERSP (Makeig)
- 1995 – Infomax ICA for EEG (Makeig, Bell, Jung, Sejnowski)
- 1997 - **EEG/ICA Toolbox (cnl.salk.edu), ITC & ERC**
- 1999 - ERP-image plots (Jung & Makeig)
- 2000 – EEGLAB GUI design (Delorme)
- 2002 – **1st EEGLAB (sccn.ucsd.edu)**
- 2004 - **1st EEGLAB support from U.S. NIH and reference paper**
- 2006 - **1st EEGLAB plug-ins, STUDY structure, and component clustering tools**
- 2009+ – **New toolboxes: NFT, SIFT, BCILAB, MPT, ...**
- 2011 – **EEGLAB, the most widely used EEG research environment !**
- 2012 - **ERICA (Experimental Real-time Interactive Control & Analysis) framework for Mobile Brain/Body Imaging (MoBI) (LSL, SNAP, XDF, MoBILAB)**
- 2013 – HeadIT.org online
- 2016 – LIMO / GLM integrated





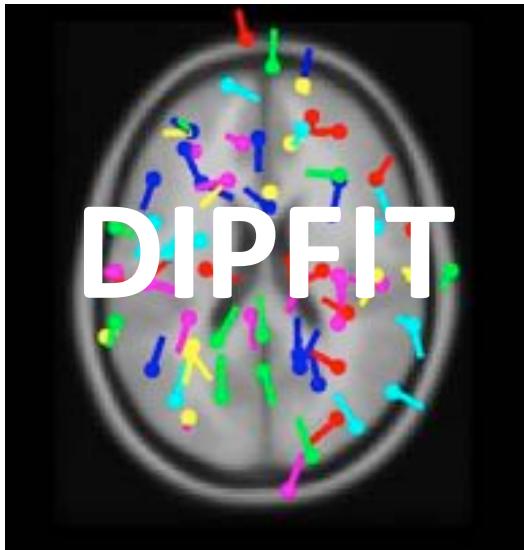
EEGLAB Workshops



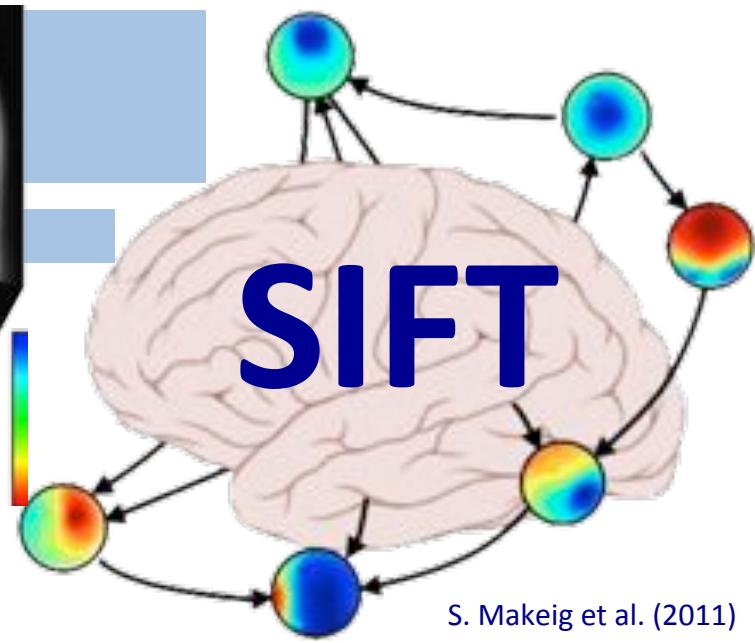
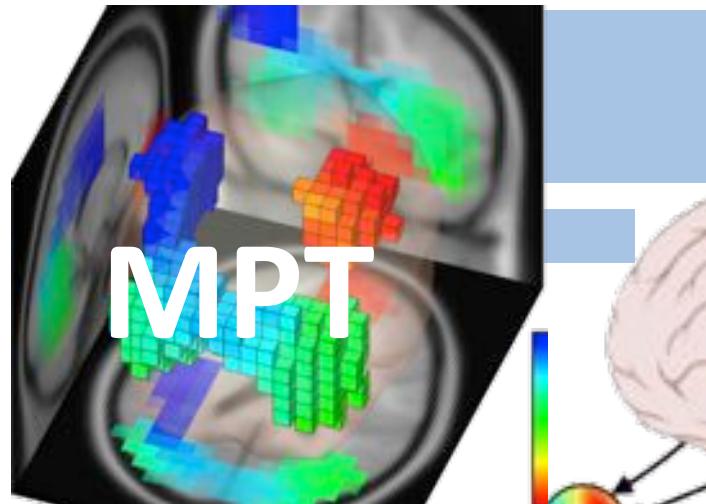
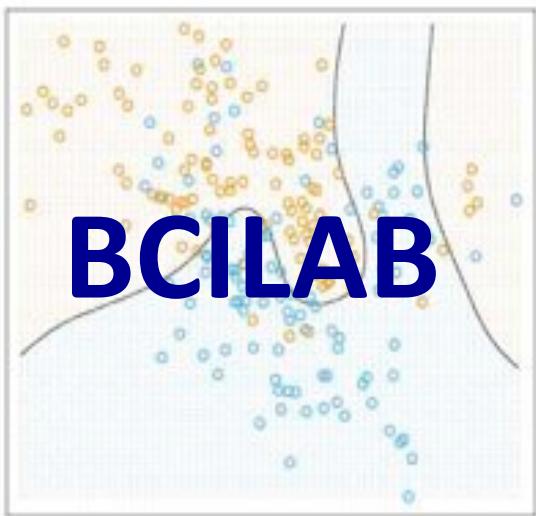
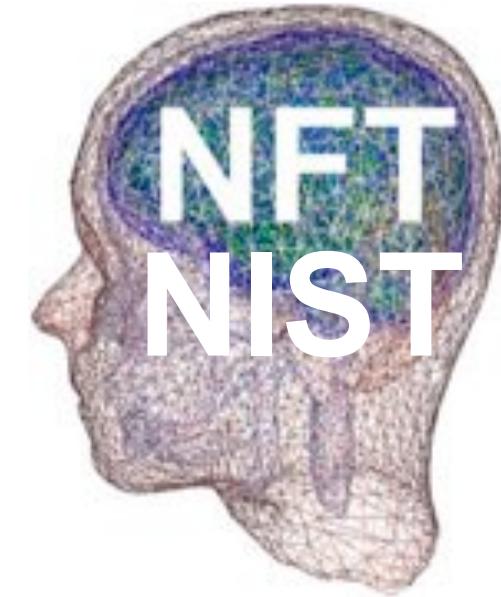
Swartz Center for Computational Neuroscience, UCSD



10th Anniversary SCCN Impromptu celebration 1/2/12



EEGLAB Plug-In Extensions



S. Makeig et al. (2011)

List of data import extensions

| Plug-in name | Version | Short plug-in description | Link | Contact | Comments |
|-----------------|---------|--|--------------------------|-----------------|---------------|
| MFFimport | 1.00 | Import MFF files from the EGI company | Download | S. Chennu | User comments |
| ANTeepimport | 1.10 | Import ANT .cnt data and trigger files | Download | M. van de Velde | User comments |
| BCI2000import | 0.36 | Import BCI2000 data files | Download | C. Boulay | User comments |
| BDFimport | 1.10 | Import BDF data files | Download | A. Delorme | User comments |
| biopac | 1.00 | Import BIOPAC data files | Download | A. Delorme | User comments |
| ctfimport | 1.04 | Import CTF (MEG) data files | Download | D. Weber | User comments |
| erpssimport | 1.01 | Import ERPSS data files | Download | A. Delorme | User comments |
| INSTEPascimport | 1.00 | Import INSTEP ASCII data files | Download | A. Delorme | User comments |
| neuroimaging4d | 1.00 | Import Neuroimaging4d data files | Download | C. Wienbruch | User comments |
| ProcomInfinity | 1.00 | Import Procom Infinity data files | Download | A. Delorme | User comments |
| WearableSensing | 1.09 | Import Wearable Sensing files | Download | S. Pillen | User comments |
| NihonKoden | 0.10 | Import Nihon Koden M00 files (beta) | Download | M. Miyakoshi | User comments |
| xdfimport | 1.12 | Import files in XDF format | Download | C. Kothe | User comments |
| bva-io | 1.5.12 | Import Brain Vision Analyser data files | Download | A. Widmann | User comments |
| Fileio | Daily | Import multiple data files formats | Download | R. Oostenveld | User comments |
| Biosig | 2.88 | Import multiple data files formats | Download | A. Schloegl | User comments |
| Cogniscan | 1.1 | Import Cogniscan data files | Download | P. Sajda | User comments |
| NeurOne | 1.0.3.2 | Import NeurOne data files | Download | Support | User comments |
| loadhdf5 | 1.0 | Load hdf5 files recorded with g.recorder | Download | Simon L. Kappel | User comments |

List of data processing extensions

| Plug-in name | Version | Short plug-in description | Link | Contact | Comments |
|------------------------|---------|---|--------------------------|----------------------|---------------|
| iERP | 0.4 | Estimate overlapping ERPs using multiple regression | Download | M. Burns | User comments |
| LIMO | 1.5 | Linear MOdeling of EEG data | Download | C. Pernet | User comments |
| commap | 2.02 | Cluster ICA components using correlation of scalp maps | Download | S. Dobszay | User comments |
| bioelectromag | 1.01 | Uses Bioelectromagnetism toolbox for ERP peak detection | Download | D. Weber | User comments |
| Valid | 1.05 | Add/Edit dataset events | Download | J. Desjardins | User comments |
| loreta | 1.10 | Export and import data to and from LORETA software | Download | A. Delorme | User comments |
| irfit | 1.02 | Non linear filtering using IIR filter | Download | M. Pazo | User comments |
| std_envtopo | 2.39 | Plot STUDY ICA cluster contribution to ERP | Download | M. Miyakoshi | User comments |
| std_selectICsByCluster | 0.10 | Forward-project clustered ICs to channels (beta) | Download | M. Miyakoshi | User comments |
| std_dipoleDensity | 0.23 | Plot STUDY ICA cluster dipole density (beta) | Download | M. Miyakoshi | User comments |
| std_ErpCalc | 0.11 | Test and visualize simple effects on ERP (beta) | Download | M. Miyakoshi | User comments |
| pvaltopo | 0.10 | Plot topography of percent variance accounted for (beta) | Download | M. Miyakoshi | User comments |
| trimOutlier | 0.16 | Trim outlier channels and datapoints interactively (beta) | Download | M. Miyakoshi | User comments |
| clean_rawdata | 0.31 | Cleans continuous data using Artifact Subspace Reconstruction | Download | Miyakoshi and Kotche | User comments |
| AFITStudio | 0.10 | Cleans spiky artifacts using AFIT (beta) | Download | Miyakoshi and Mullen | User comments |
| Mutual_Info_Clustering | 1.00 | Group single dataset ICA components by Mutual Information | Download | N. Bigdely | User comments |
| mass_univ | 130502 | Mass Univariate ERP Toolbox | Download | D. Groppe | User comments |
| REGICA | 1.00 | ICA regression based EEG removal | Download | M. Klaas | User comments |
| MARA | 1.1 | Multiple Artifact Rejection Algorithm | Download | I. Winkler | User comments |
| Mrfit | 1.6.1 | Routines for designing linear filters | Download | A. Widmann | User comments |
| PACT | 0.17 | Computes phase-amplitude coupling for continuous data | Download | M. Miyakoshi | User comments |
| IMRi | 2.00 | Remove fMRI artifacts from EEG | Download | J. Dien & R. Niazy | User comments |
| SIFT | 1.33 | Analysis and visualization of multivariate connectivity | Download | T. Mullen | User comments |
| AAR | 131130 | ICA-based Automatic Artifact Removal | Download | G. Gomez-Hernero | User comments |
| Adjust | 1.1 | Automatic Detector - Joint Use of Spatial and Temporal features | Download | Adjust Support | User comments |
| Cleanline | 1.02 | Removes sinusoidal artifacts (line noise) | Download | T. Mullen | User comments |
| Fieldtrip-like | Daily | Adds source localization and statistics tools to EEGLAB | Download | R. Oostenveld | User comments |
| EYE-EEG | 0.41 | Open source MATLAB tool for simultaneous eye tracking & EEG | Download | O. Dimigen | User comments |
| Interpolator | 151127 | Performs 2D/3D linear interpolation | Download | M. Miyakoshi | User comments |

Who
am I?

Embodied Agency

Brain processes have evolved and function *to optimize the outcome of our behavior –* which the brain organizes in response to *perceived challenges and opportunities.*

Brains meet the challenge of the moment!



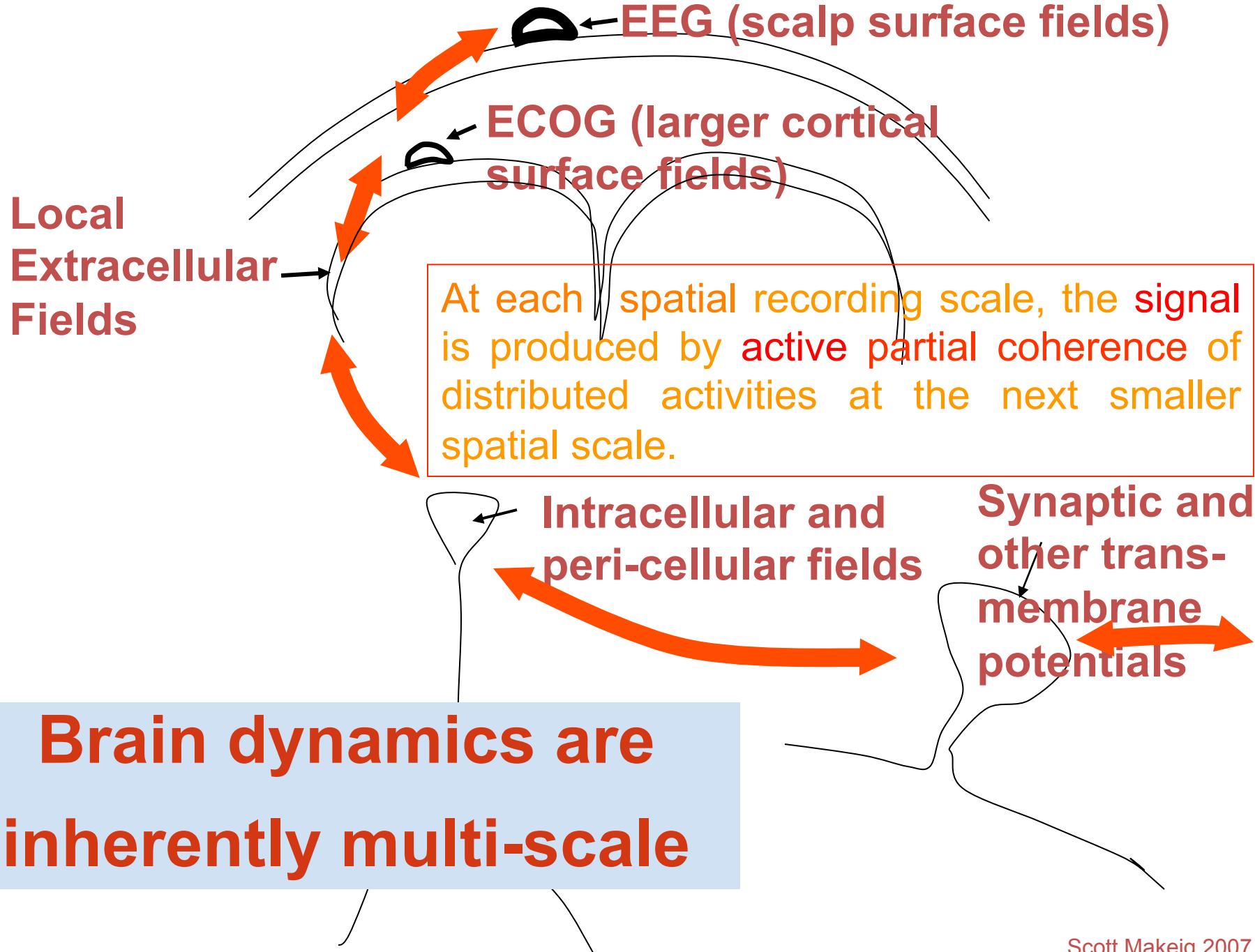
Brain imaging natural cognition -- actions & interactions

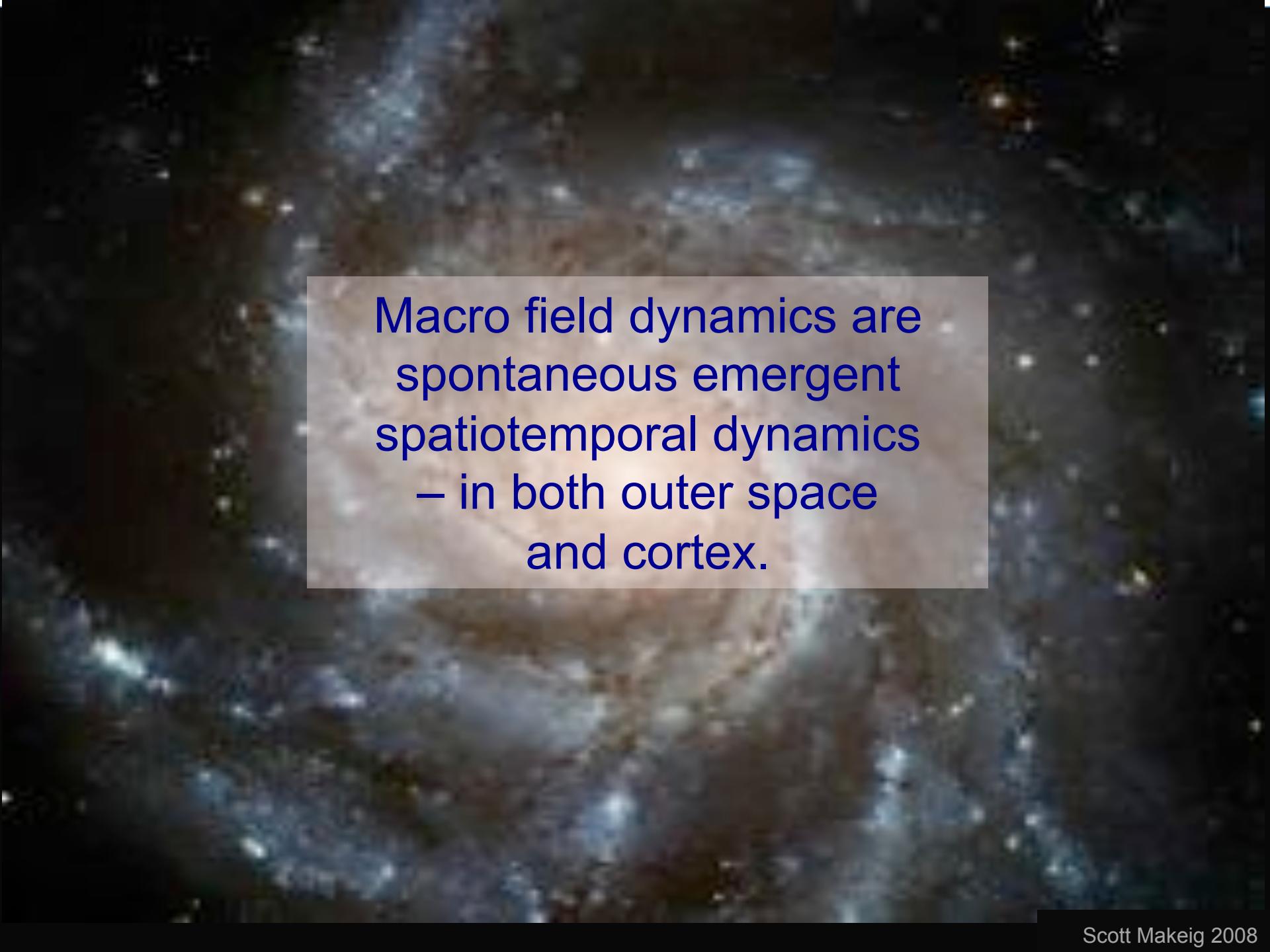




What is EEG?

- A small portion of *cortical* brain electrical activity
- An even smaller portion of *total* brain electrical activity
- **But *which* portion?**
- **Triggered and modulated *how*?**
- **With *what* functional significance?**





Macro field dynamics are
spontaneous emergent
spatiotemporal dynamics
– in both outer space
and cortex.

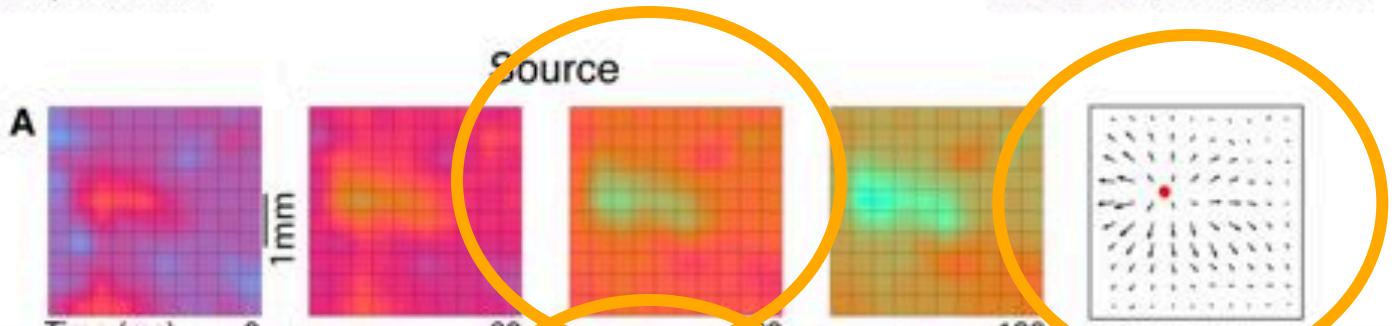
Phase cones (Freeman)

Avalanches (Plenz)

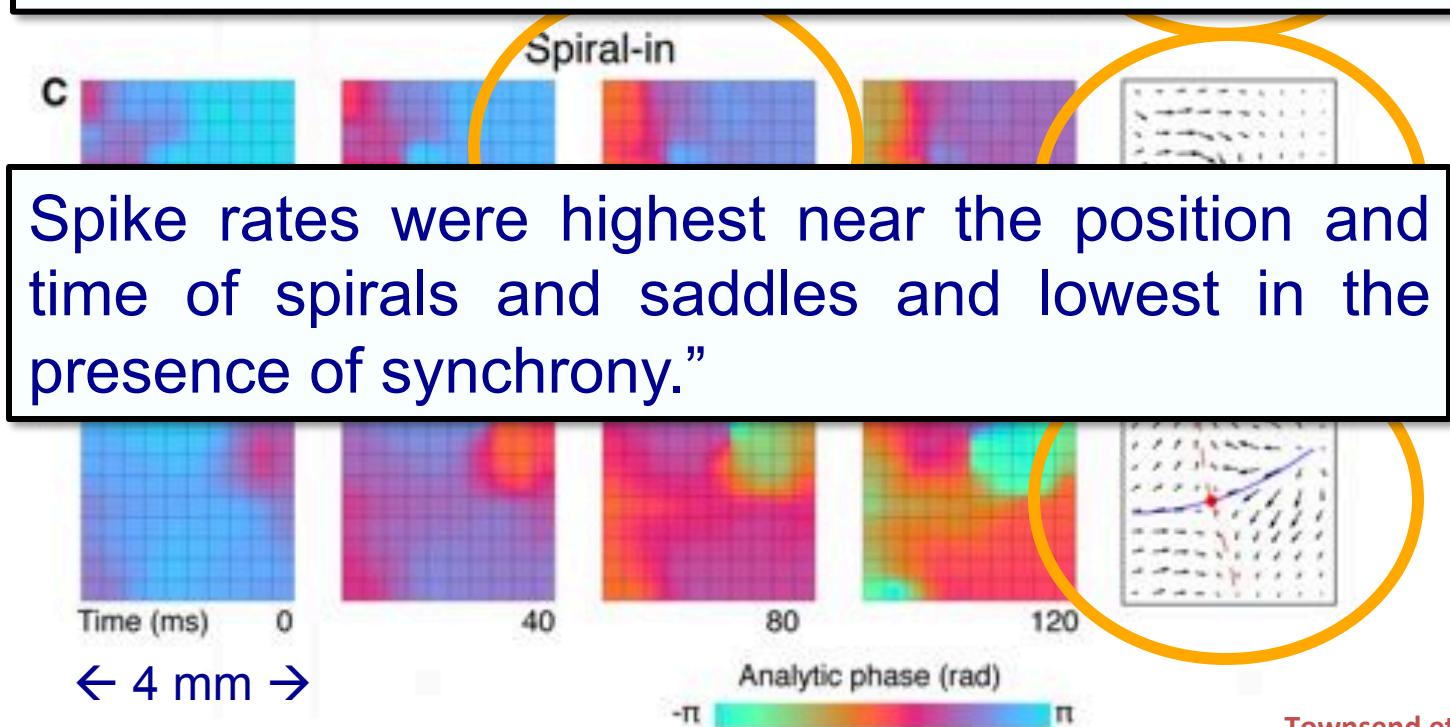


**Local Cortical Synchronies
are
Effective EEG Sources**

Delta band
(1-4 Hz)
in
anesth.
animals

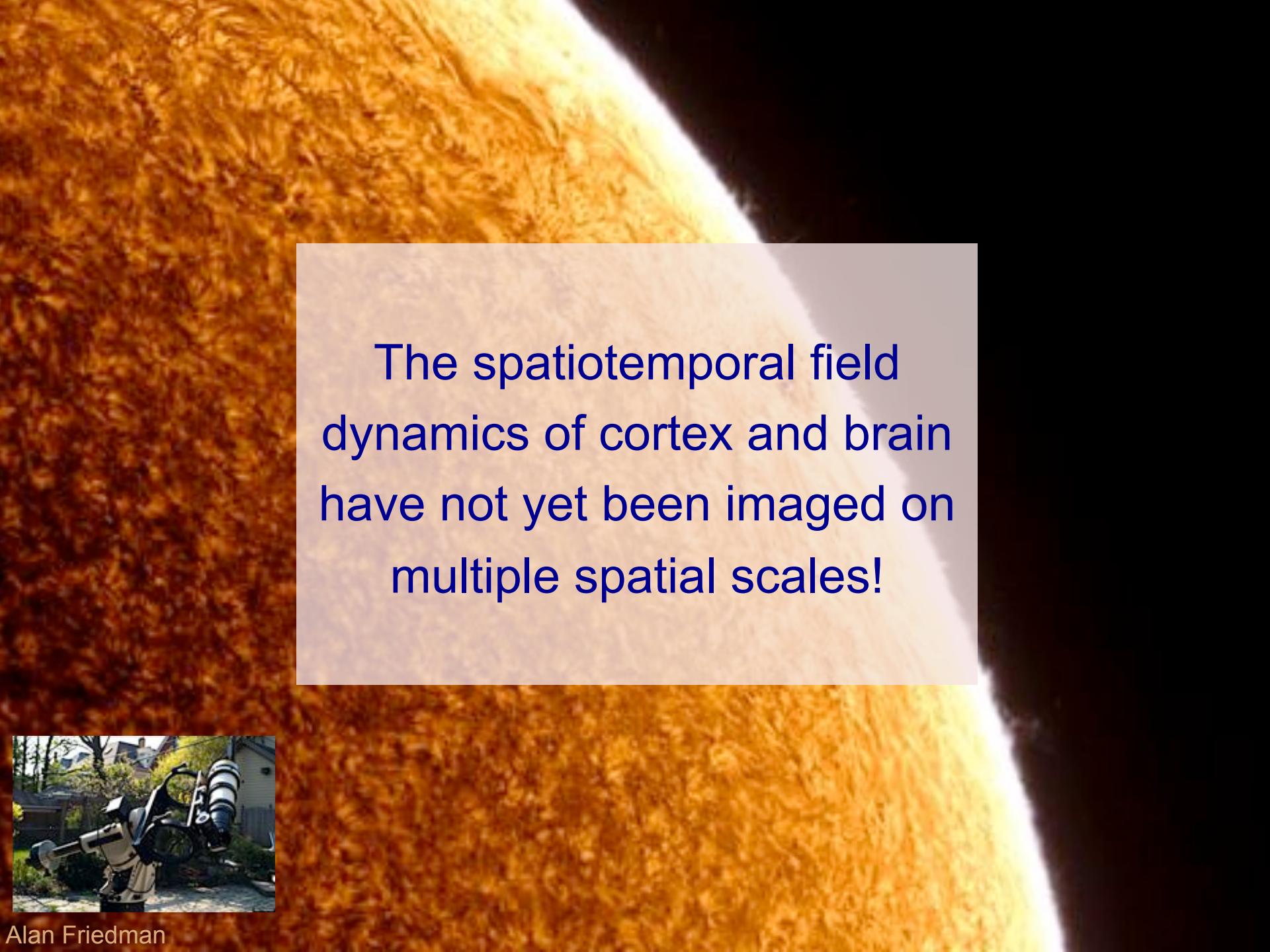


“Synchrony was associated with high delta-band amplitude (averaged across the recording array), whereas complex waves were associated with low average delta-band amplitude. ...”



Simple patterns

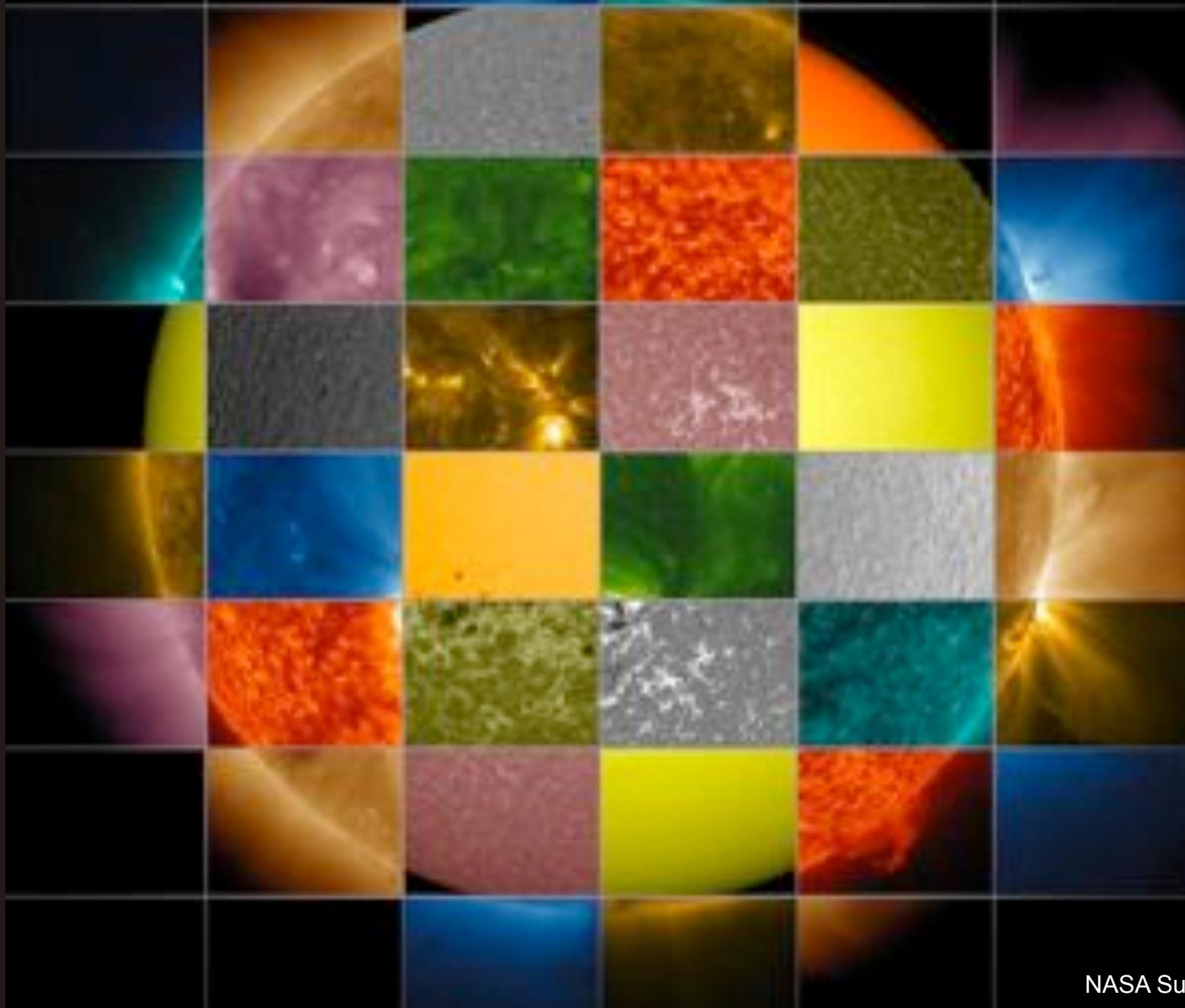
Complex patterns



The spatiotemporal field
dynamics of cortex and brain
have not yet been imaged on
multiple spatial scales!

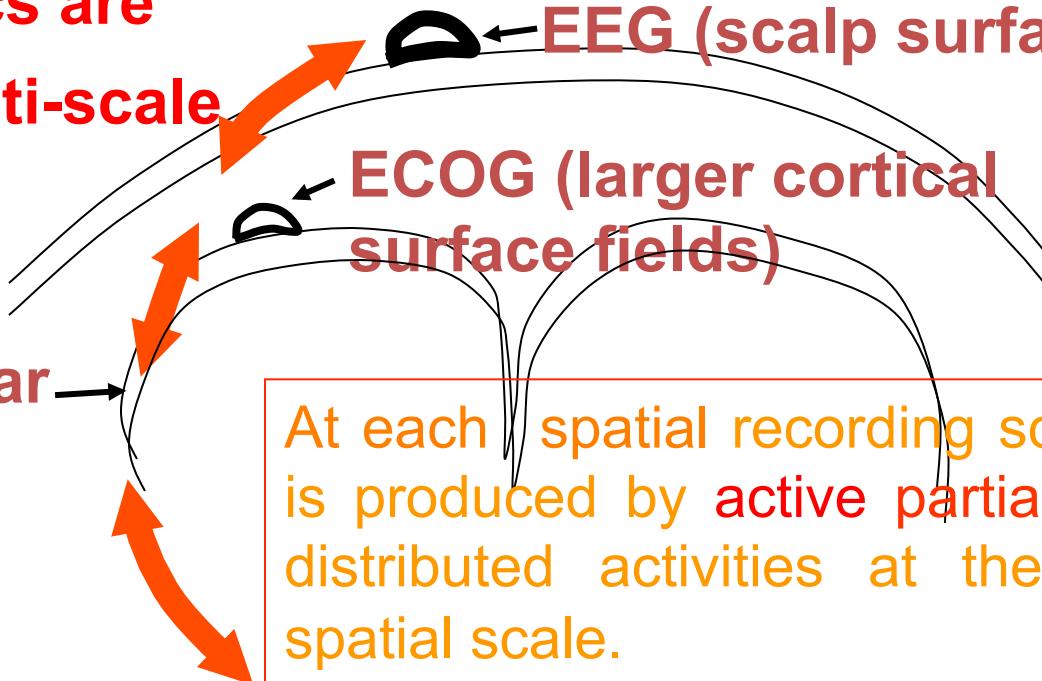


Spatial complexity depends on frequency



Brain dynamics are inherently multi-scale

Local
Extracellular
Fields



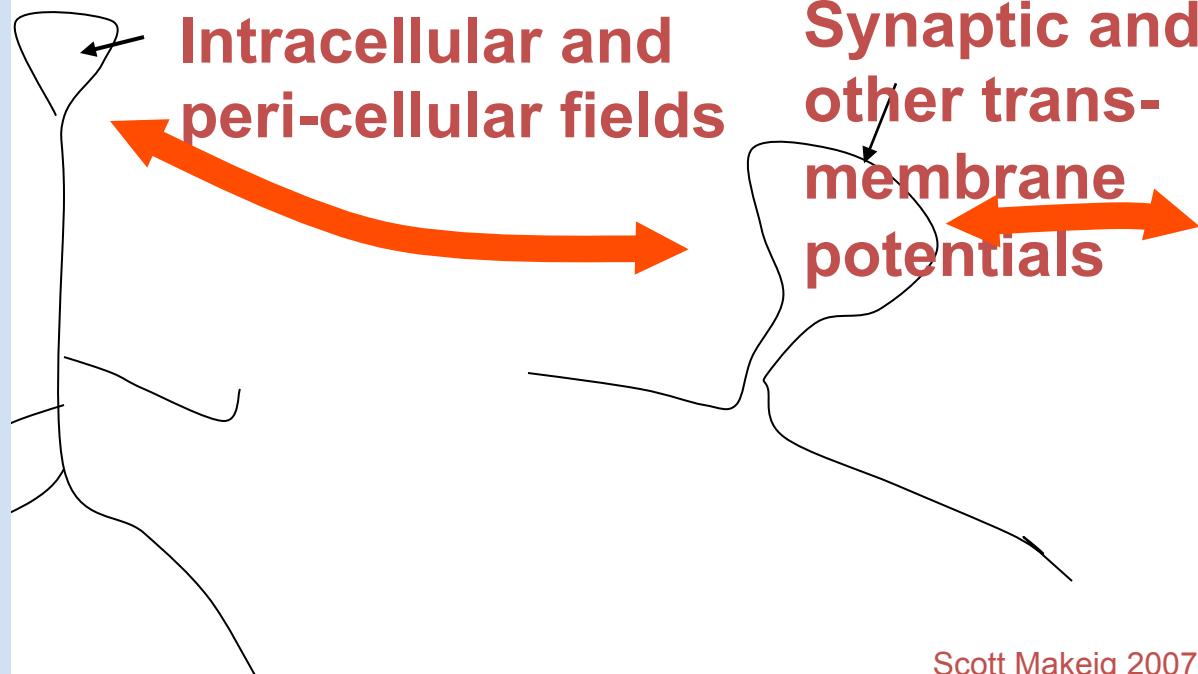
At each spatial recording scale, the signal is produced by active partial coherence of distributed activities at the next smaller spatial scale.

Cross-scale coupling is bi-directional!

Larger



Smaller



Brain dynamics are inherently multi-scale

Local
Extracellular
Fields

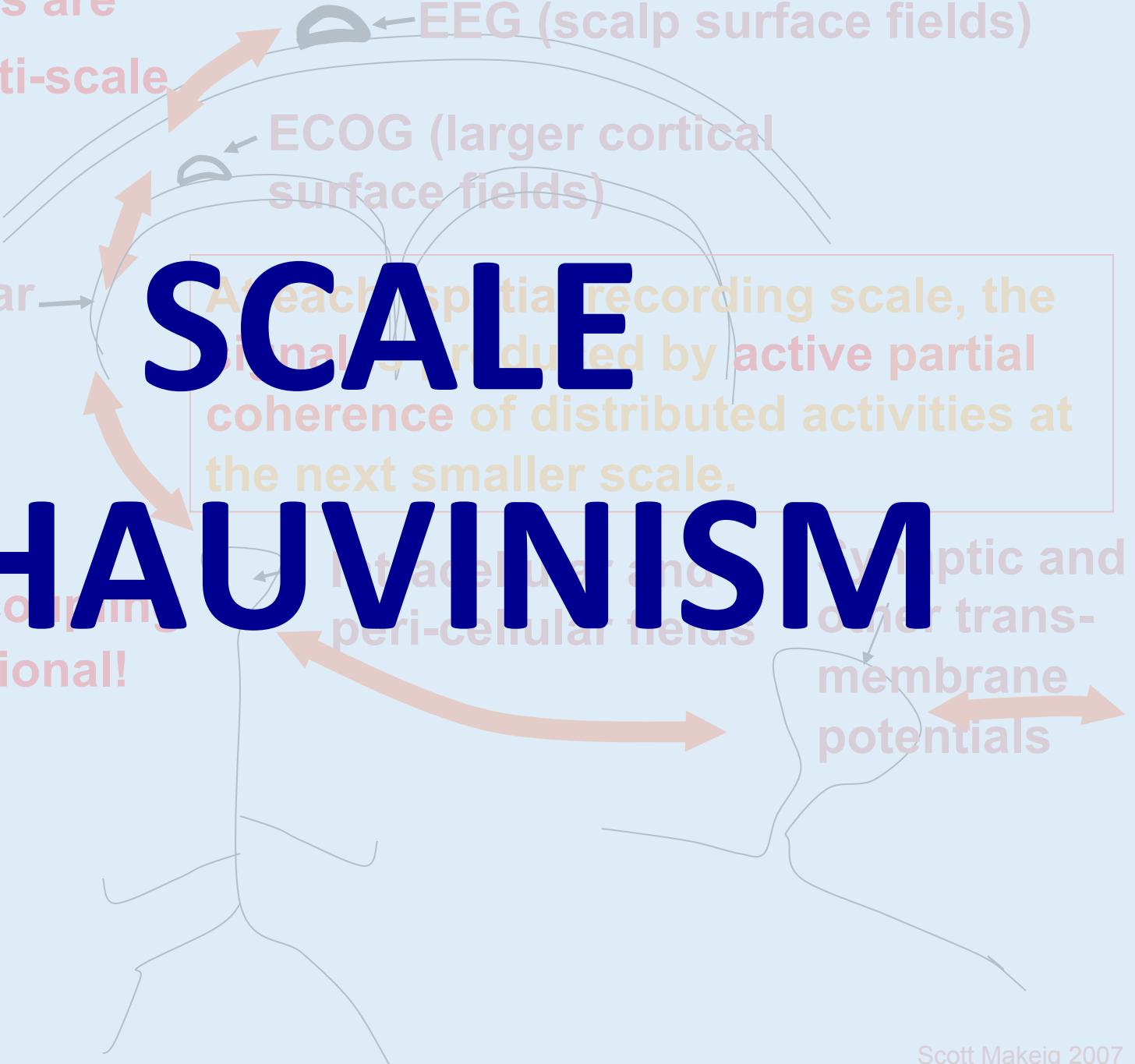
Cross-scale coupling
is bi-directional!

Larger

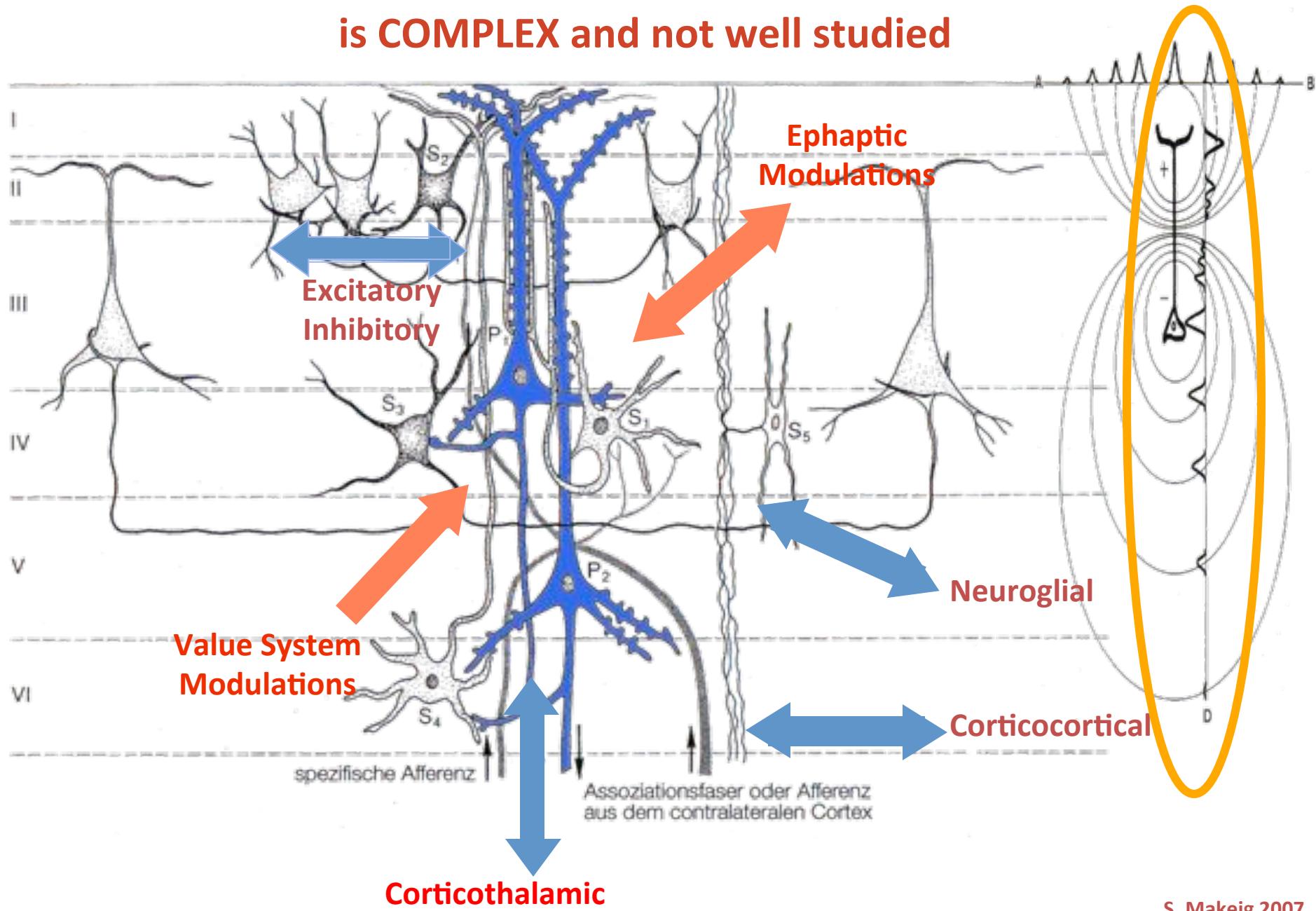
Smaller

SCALE CHAUVINISM

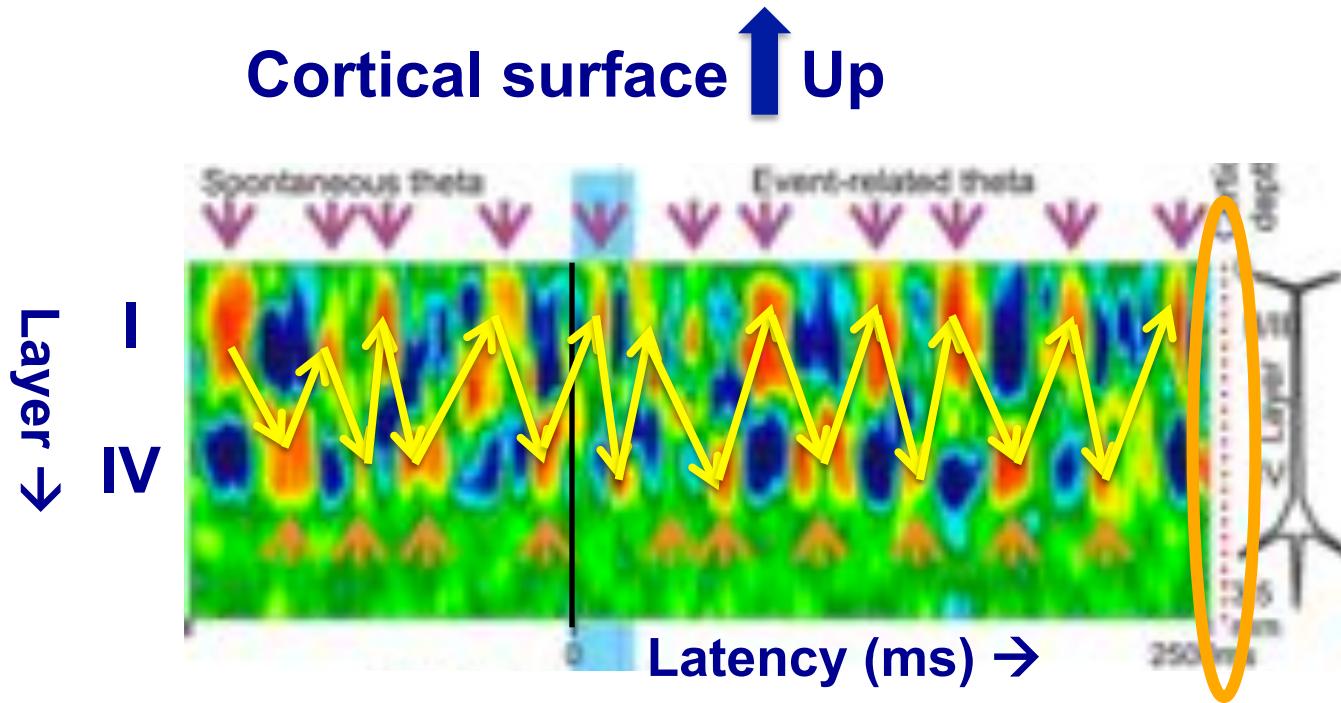
At each spatial recording scale, the spatial resolution is reduced by active partial coherence of distributed activities at the next smaller scale.



The generation and modulation of EEG is COMPLEX and not well studied



In Cortex: Up ≠ Down and $+μV \neq -μV$



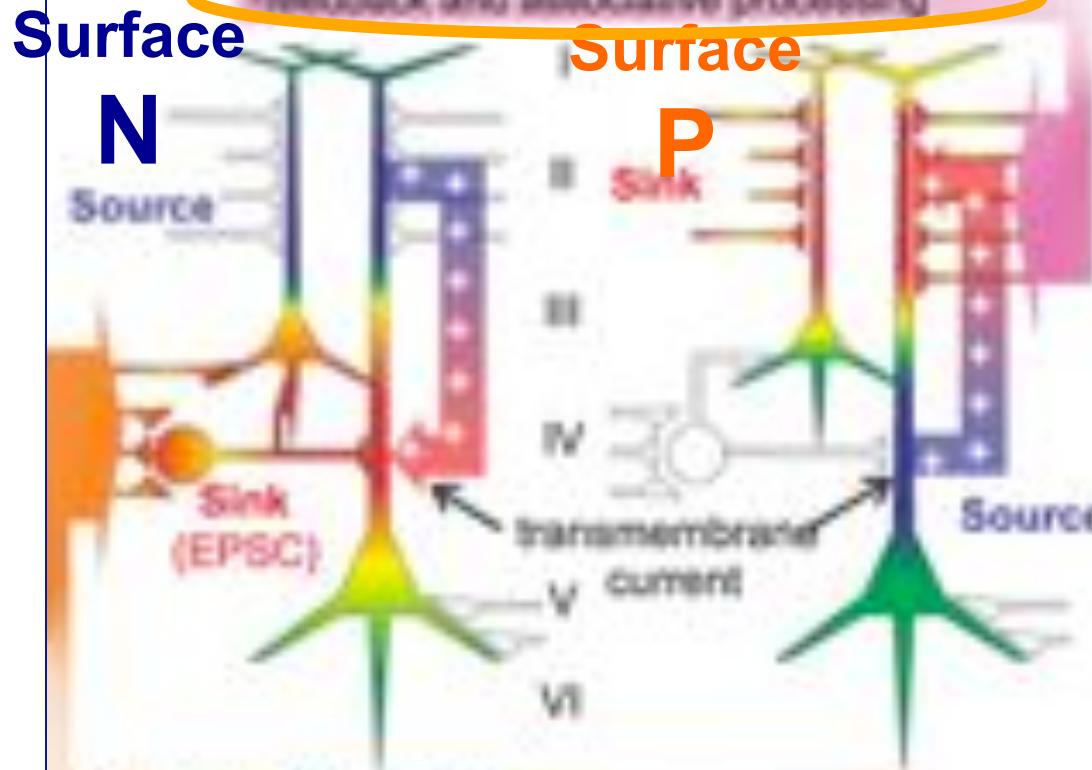
Thalamus
Down



C Neuronal information-processing stages

Interpreting information:

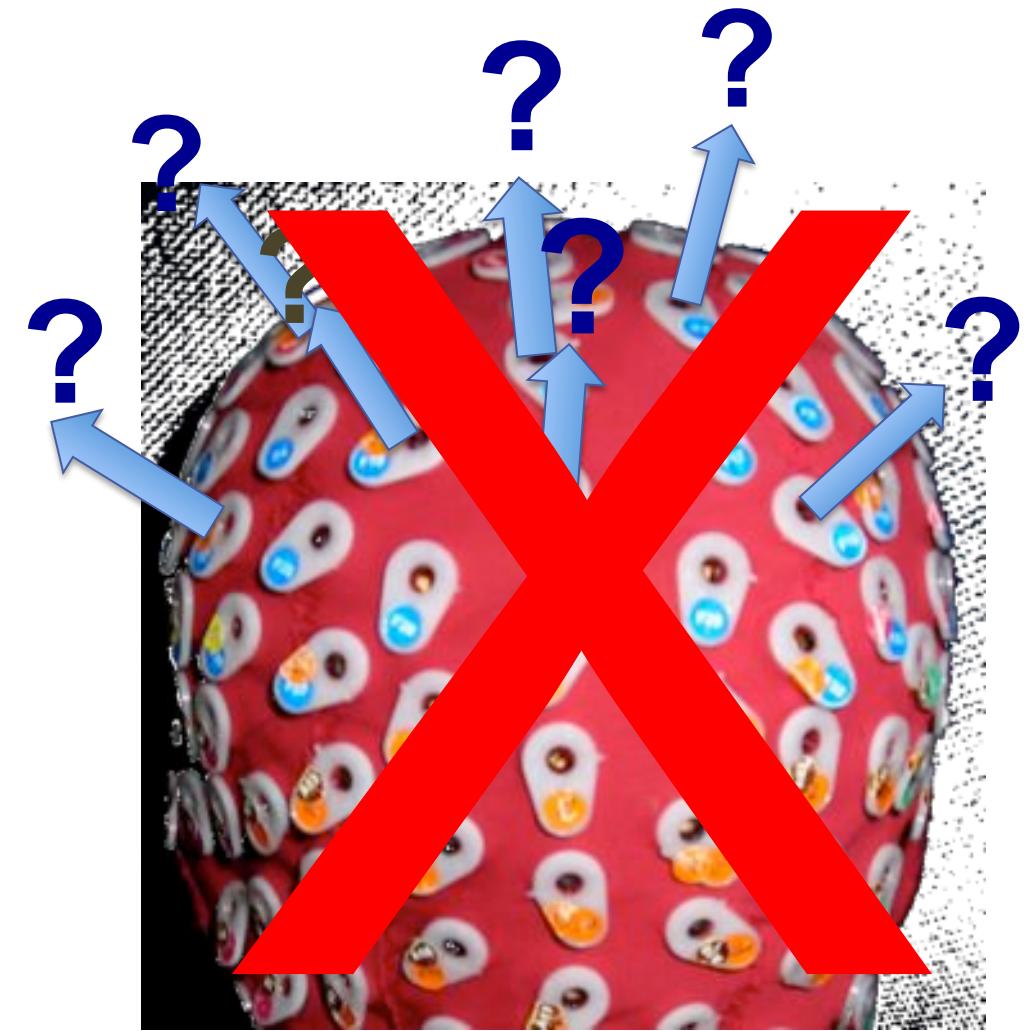
- Upper layer sink
- Input from lower layers & higher cortical areas
- Feedback and associative processing



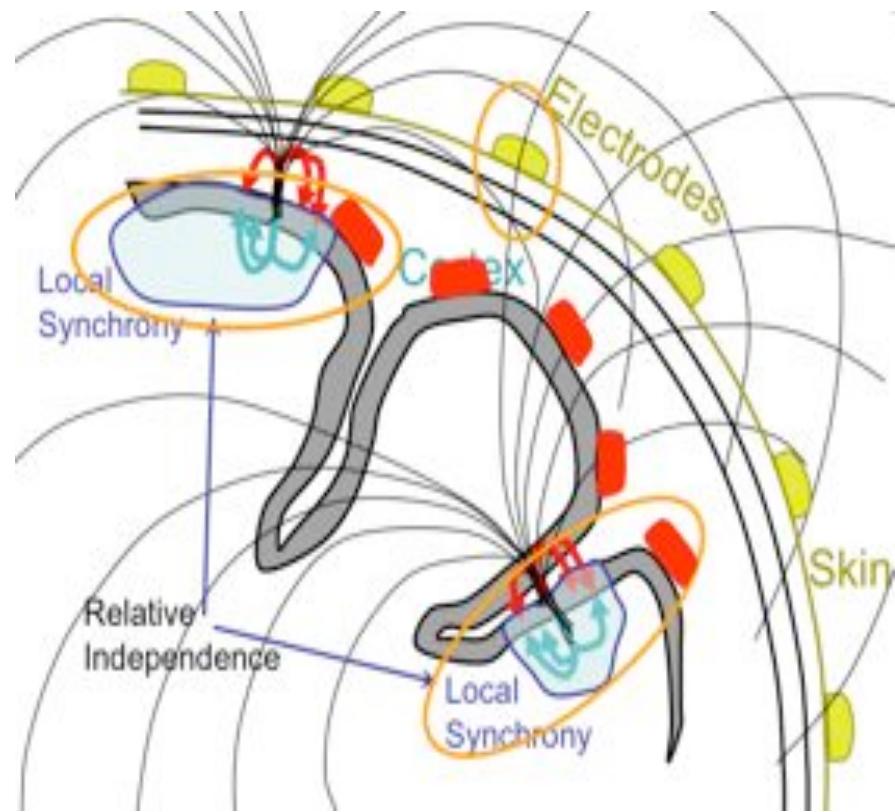
Acquiring information:

- Middle layer sink
- Input from previous and lower cortical areas
- Feedforward processing

Naïve 2-D interpretation of EEG signals?

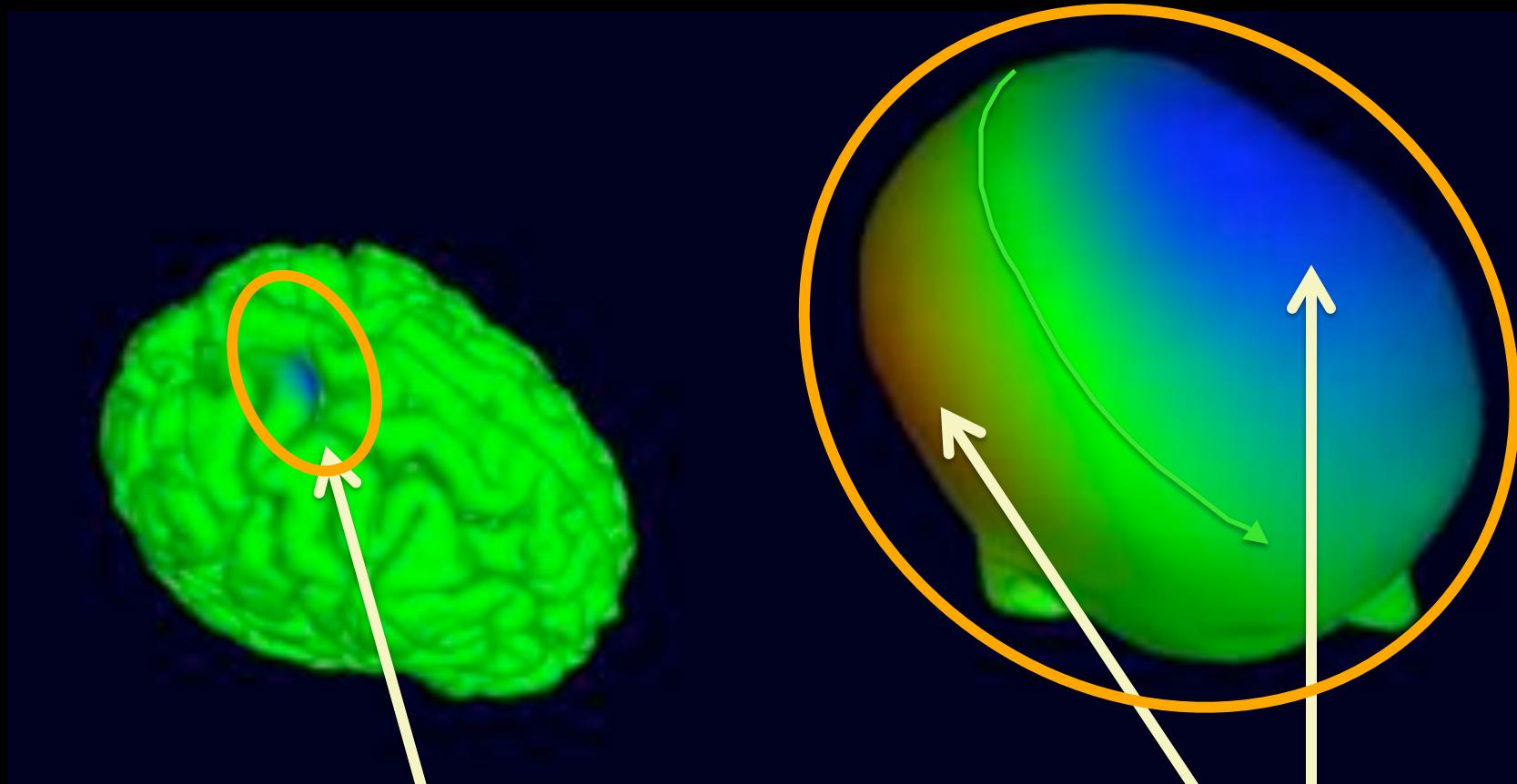


Cortical EEG signal projection
patterns as point processes



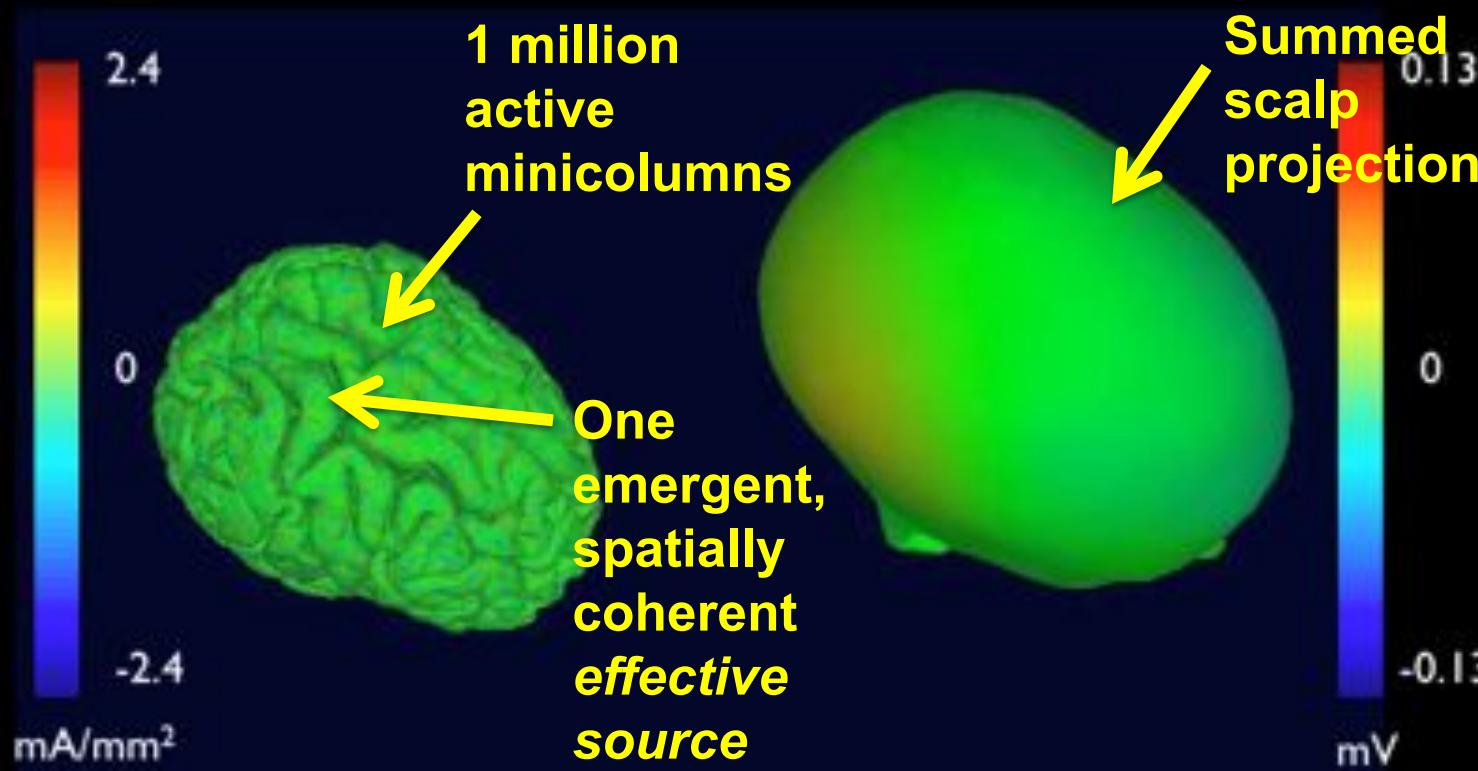
Cortical source current volume
conduction patterns

The very broad EEG point-spread function

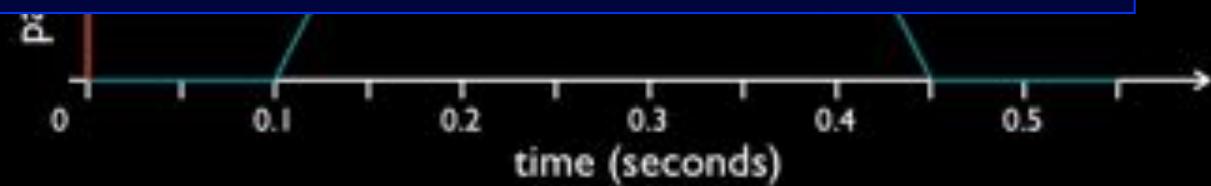


Simulated parietal effective source

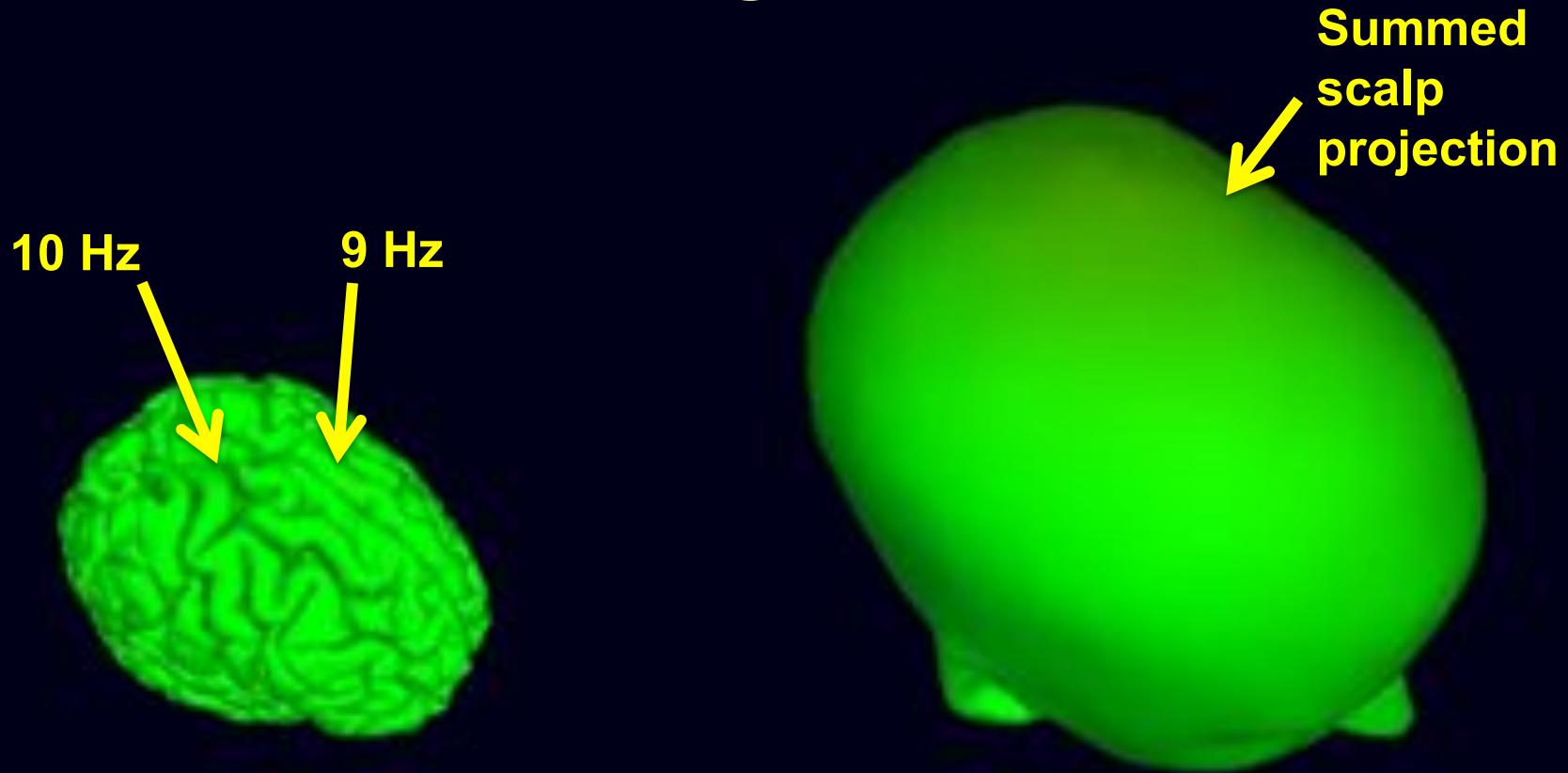
Very broad projected scalp potentials



The **effective sources** of the scalp EEG & MEG are emergent islands of local synchrony / near-synchrony.



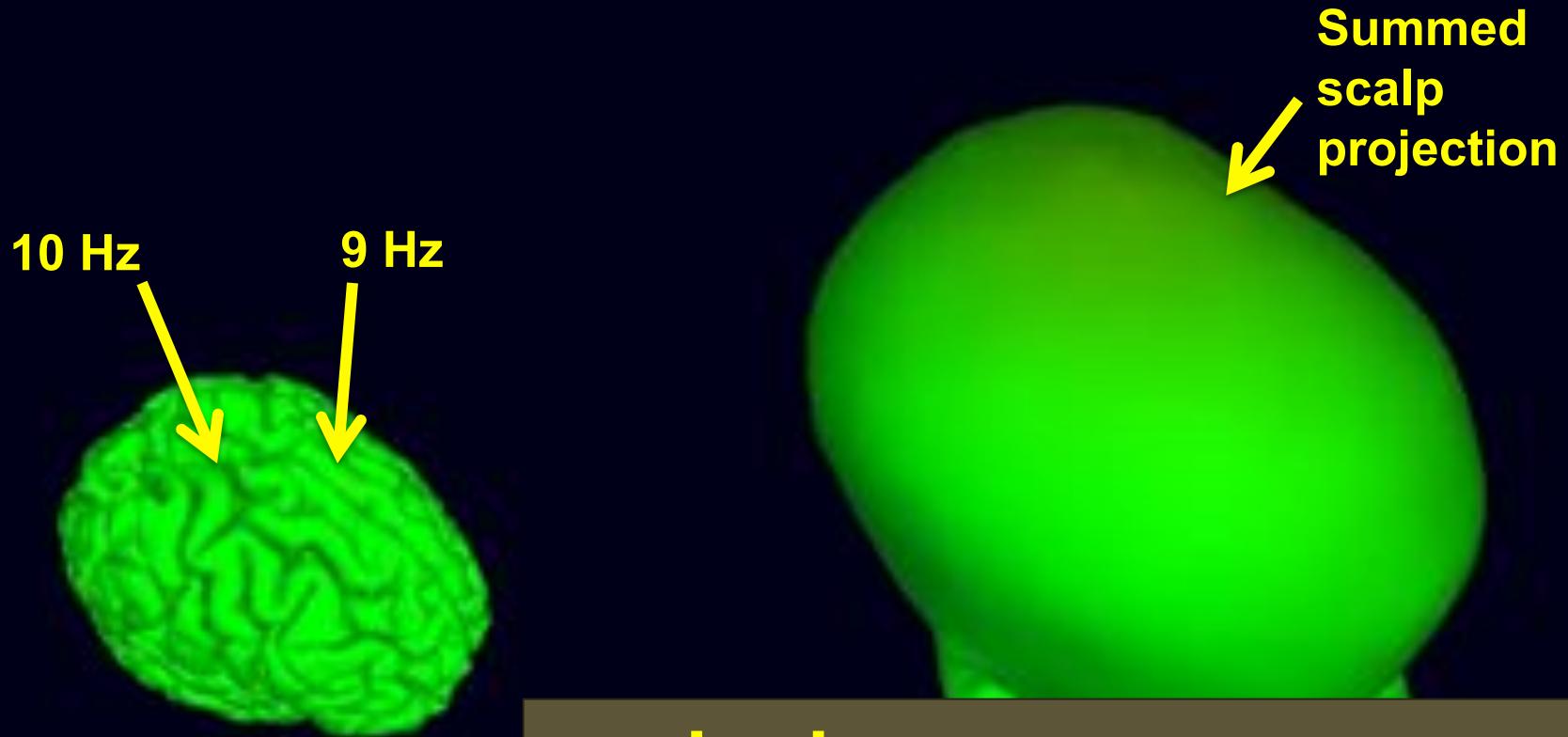
EEG Scalp Signal Illusion



Two Effective Sources

Their summed
scalp projection

Scalp EEG Illusion

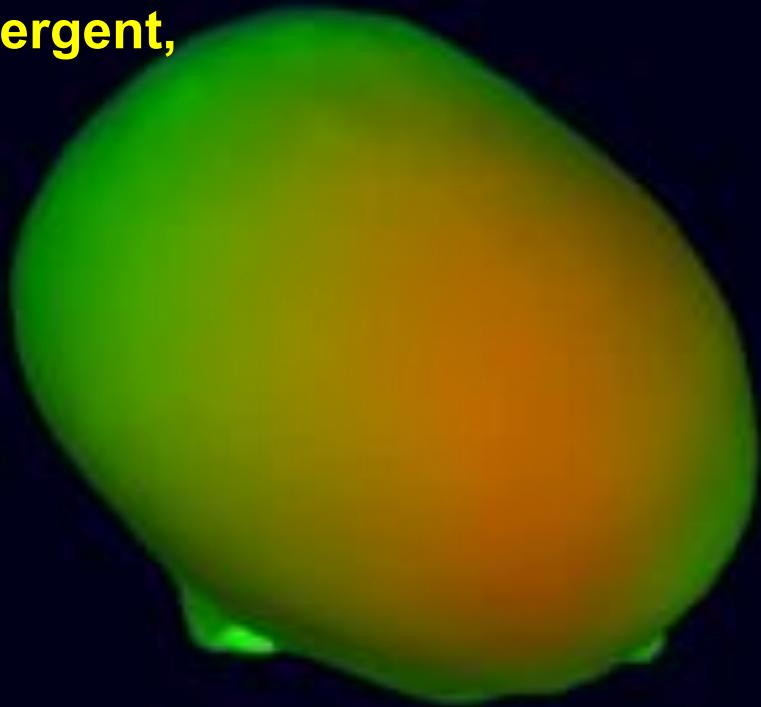


ep·i·phe·nom·e·non: a secondary effect or byproduct that arises from but does not causally influence a process.

The very broad EEG point-spread function

30

**spontaneously emergent,
spatially coherent
*effective sources***



Simulated EEG summing 30 cortical effective sources (animation at 1/5th real time)

Non-brain ‘artifact’ contributions to scalp EEG



Brain sources only

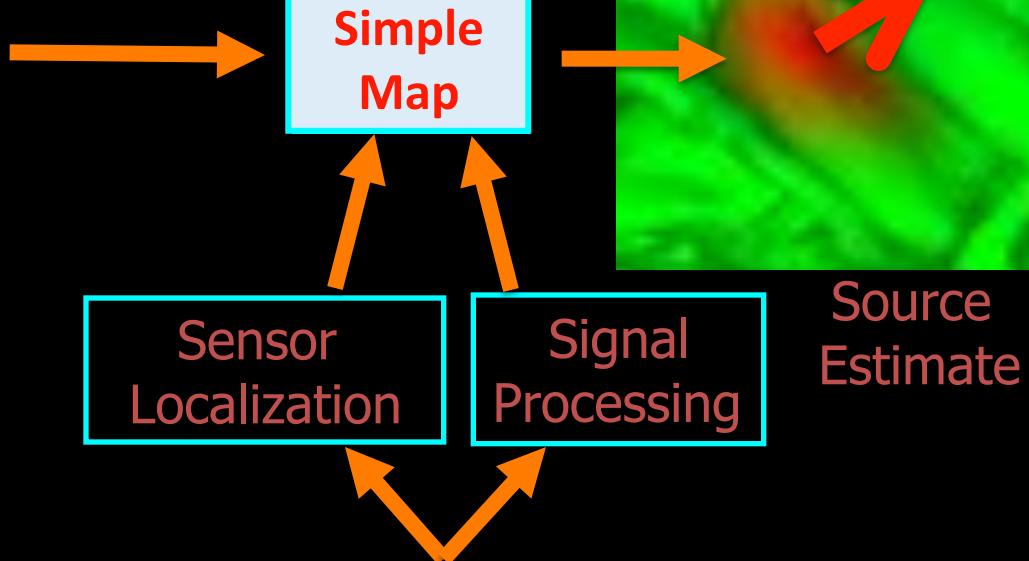
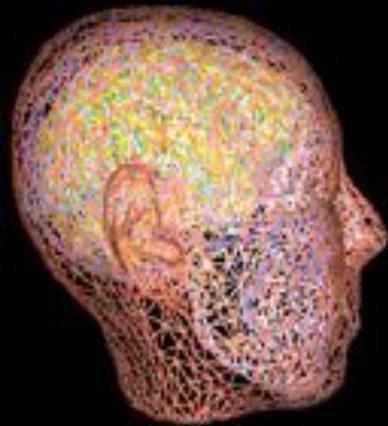


PLUS non-brain sources

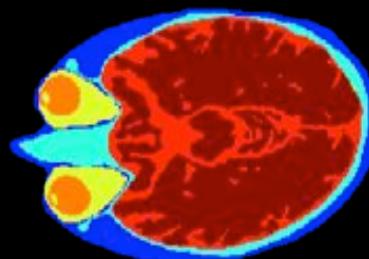
Electromagnetic source localization using realistic head models



Solve the forward problem using realistic head models (BEM)



MRI



Segmentation



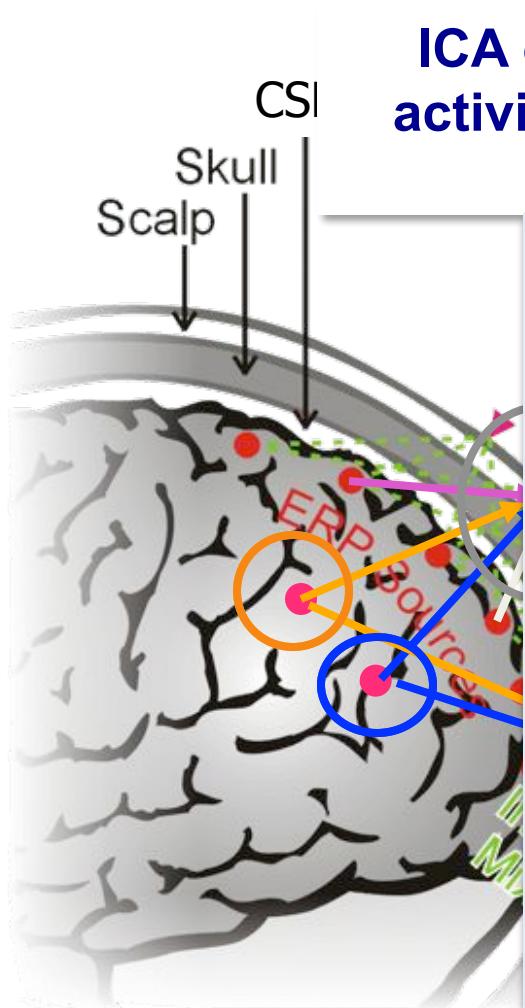
EEG/MEG

But how to find
EEG simple maps
?

Blind EEG Source Separation by Independent Component Analysis



Tony Bell,
developer of
Infomax ICA



ICA can find distinct EEG source activities -- and their 'simple' scalp maps!

Independent Component Analysis of Electroencephalographic Data

Steve Makeig
Kern National Research Center
P.O. Box 80220
San Diego, CA 92108-0220
smakeig@ucsd.edu, www.sccn.ucsd.edu

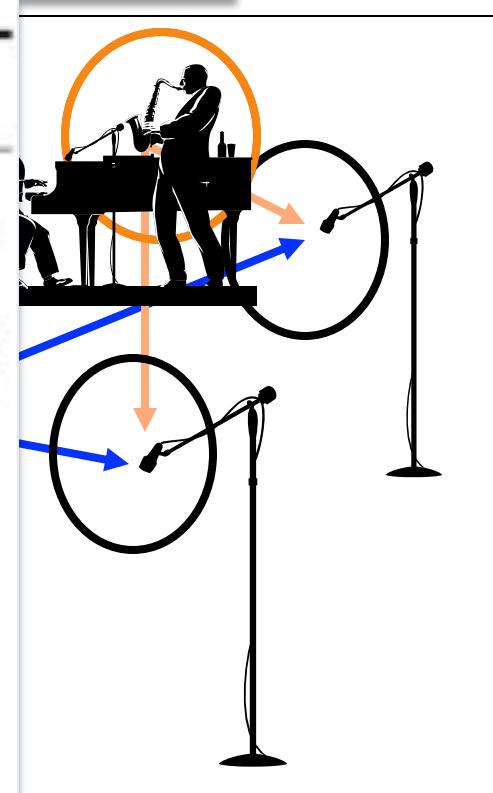
Anthony J. Bell
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The Salk Institute, P.O. Box 8400
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Willy-Ping Jung
Kern National Research Center and
Computational Neurophysiology Lab
The Salk Institute, P.O. Box 8400
San Diego, CA 92108-0400
jung@kern.edu

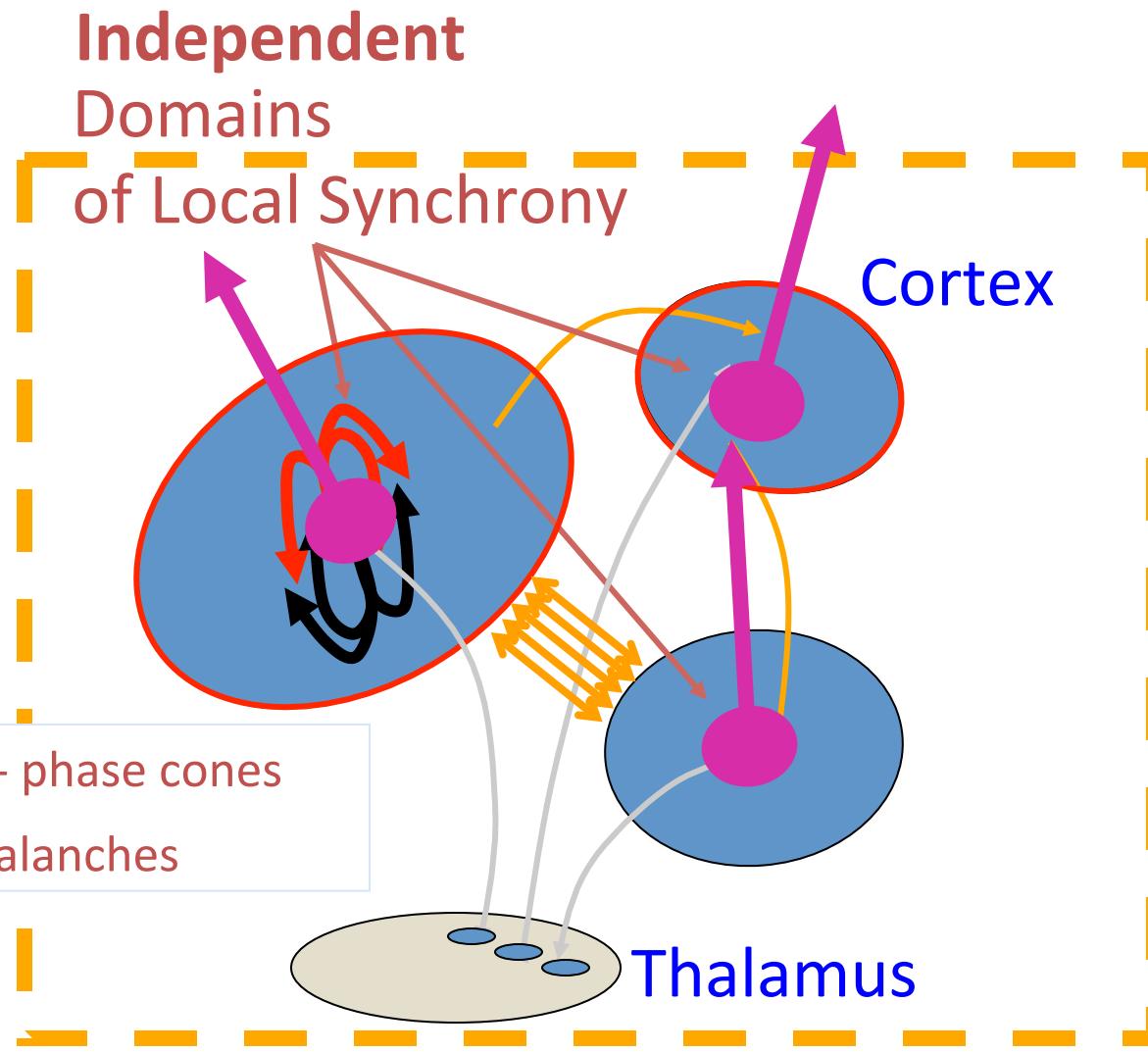
Brian D. Sejnowski
Kern National Research Center and
Computational Neurophysiology Lab
The Salk Institute, P.O. Box 8400
San Diego, CA 92108-0400
www.salk.edu/~sejnowski

Abstract:

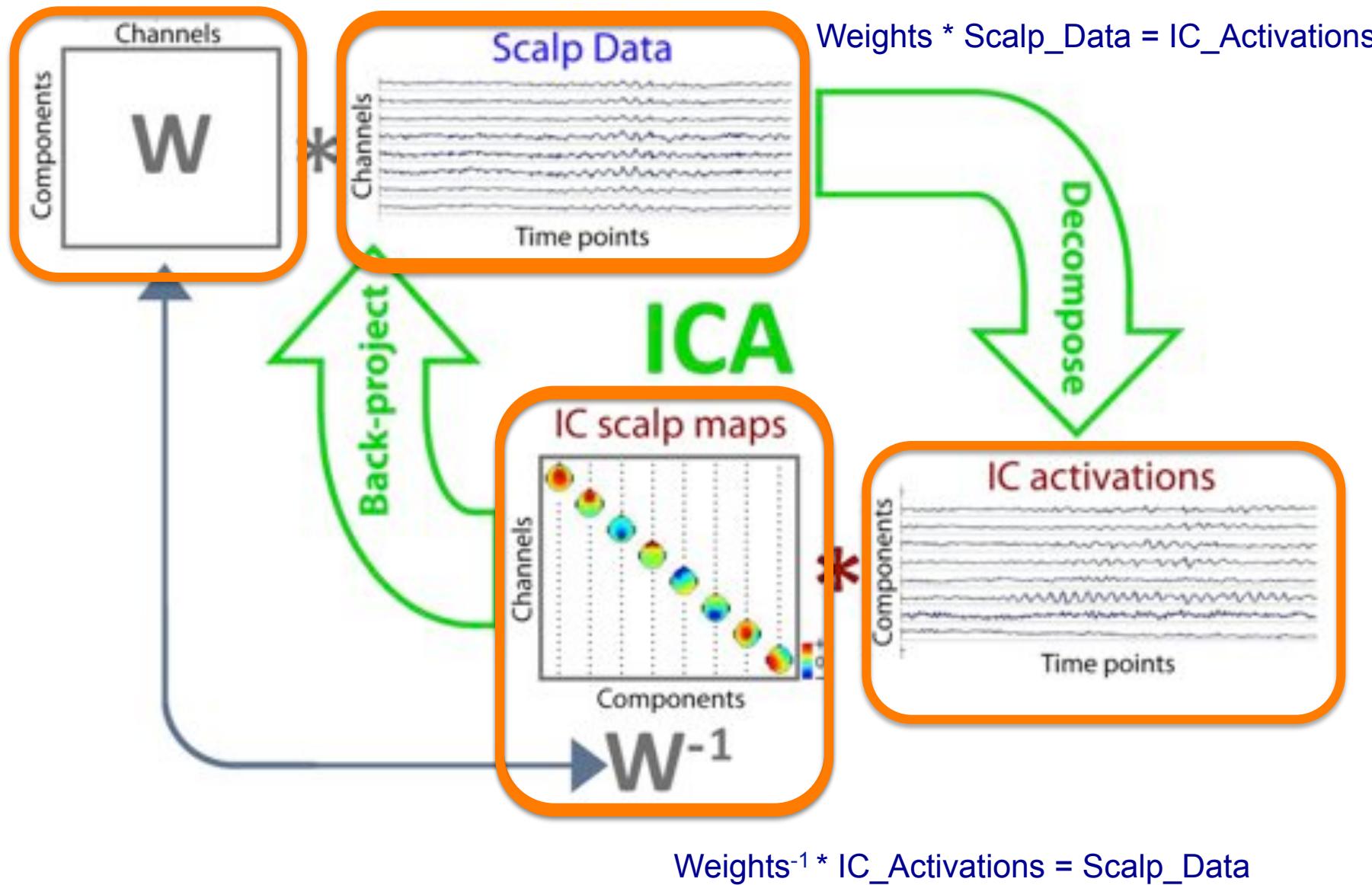
Because of the distance between the brain and the scalp, different activities, electroencephalography (EEG) data collected from any point on the human scalp include activity generated within a large brain area. This spatial smearing of EEG data by volume conduction does not induce significant time delays, however, suggesting that the Independent Component Analysis (ICA) algorithm of Bell and Sejnowski¹ is well-suited for performing blind source separation on EEG data. The ICA algorithm separates the problem of source identification from that of source localization. That makes it easy to apply the ICA algorithm to EEG and evoked potential (EP) data collected during a mixed auditory stimulus task. We show: (1) ICA working is insensitive to different volume conduction; (2) ICA may be used to segregate various auditory EP components (like and mismatch tones) from other sources; (3) ICA is capable of isolating overlapping EP phenomena, including alpha and theta tones and spike-frequency EP components, to separate ICA channels; (4) Mismatch tones in EPs and behavioral data can be tracked using ICA via change in the amount of statistical correlation between ICA-derived component channels.



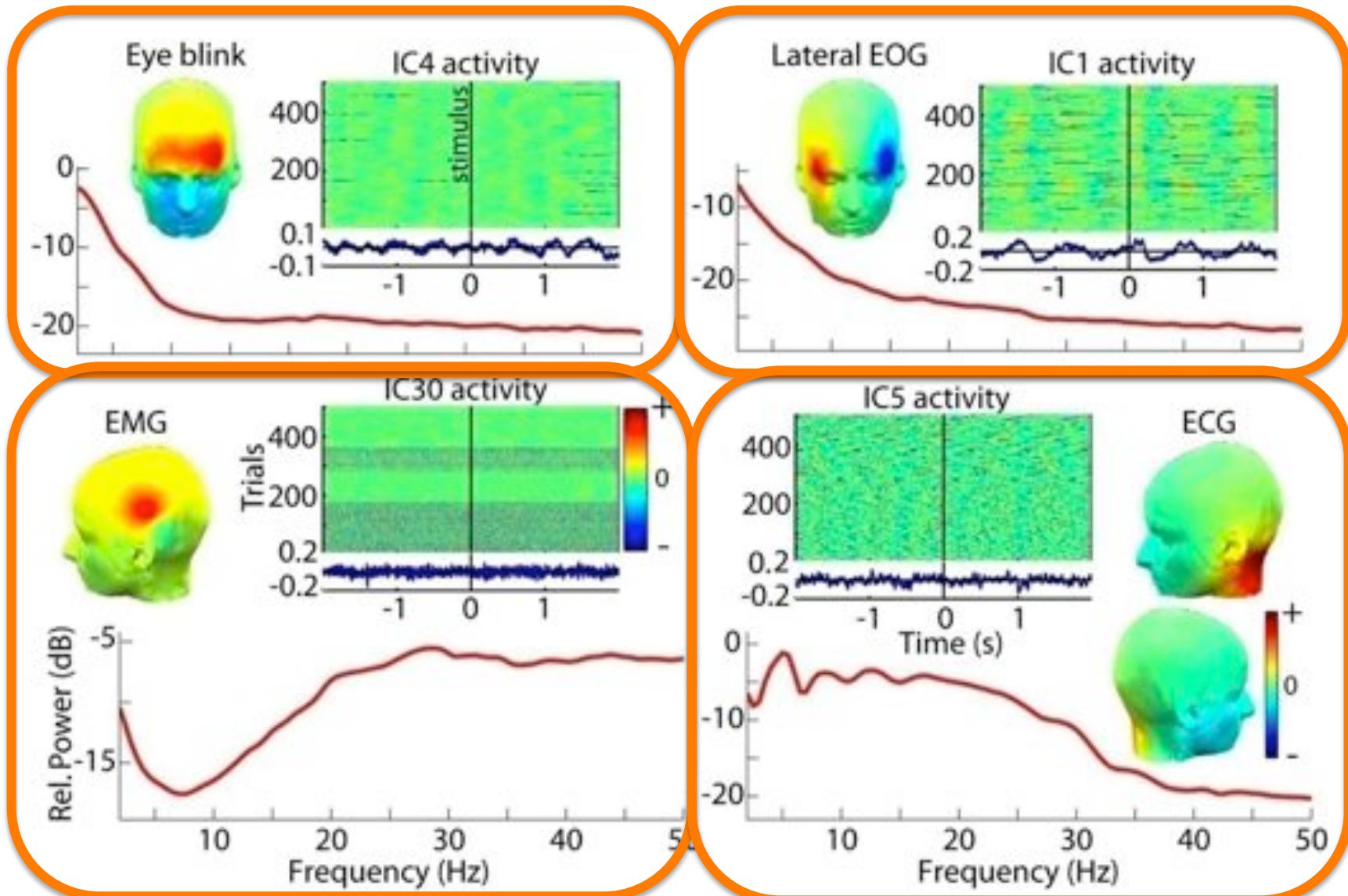
Are EEG source signals ~independent?



ICA is a linear data decomposition method

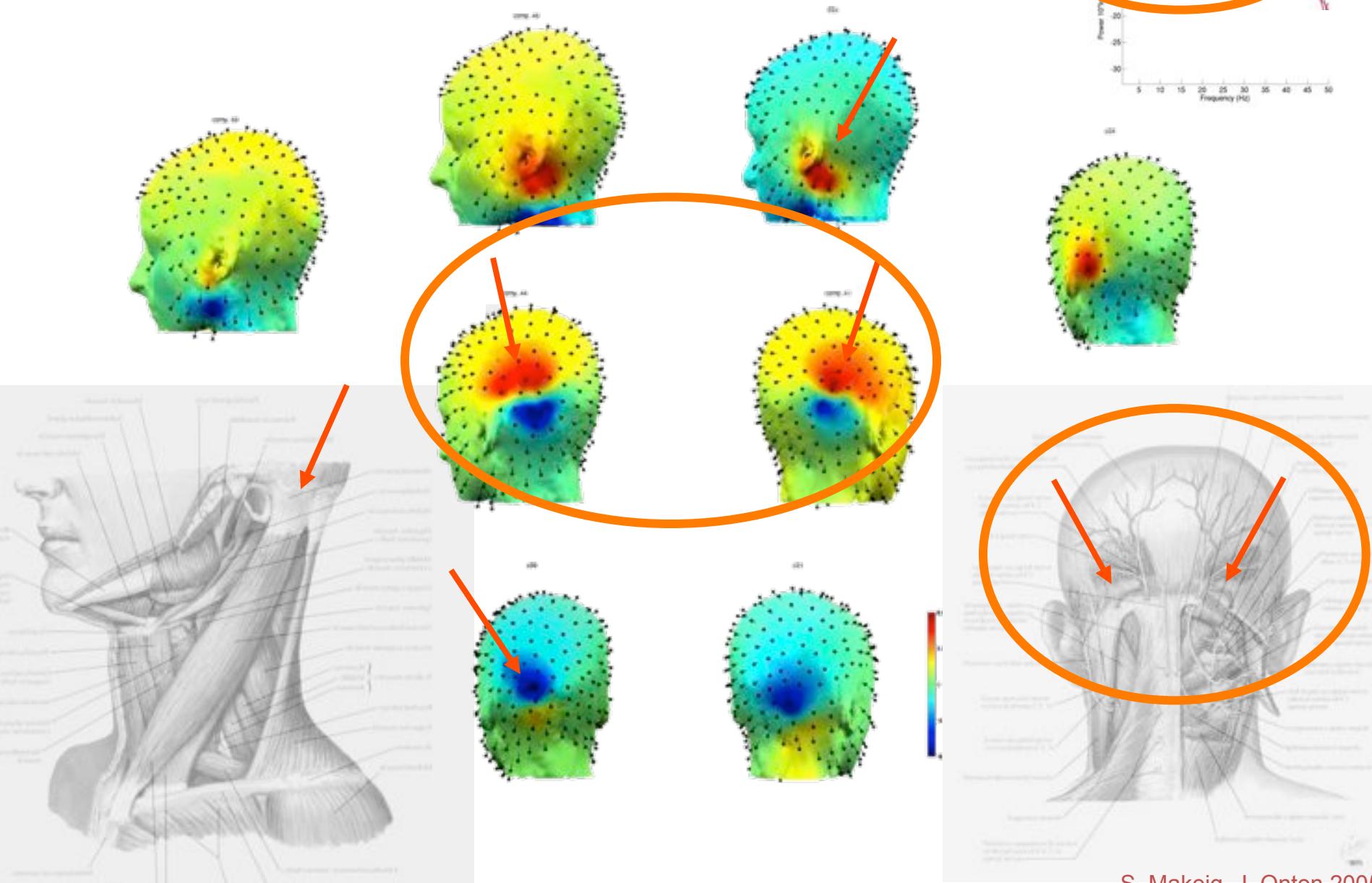


ICA finds Non-Brain Independent Component (IC) Processes ...

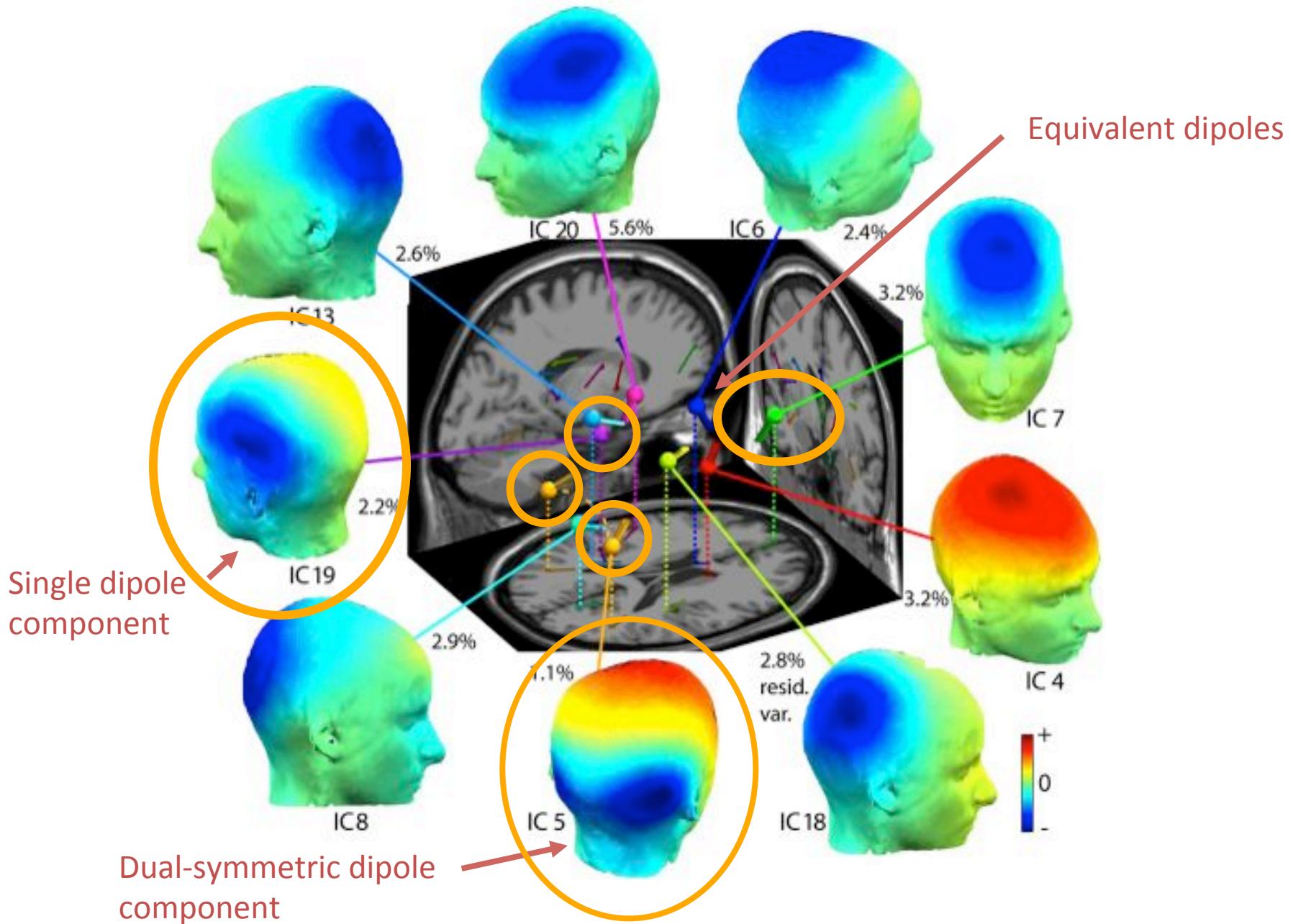


... separates them from the remainder of the data ...

Independent muscle signals



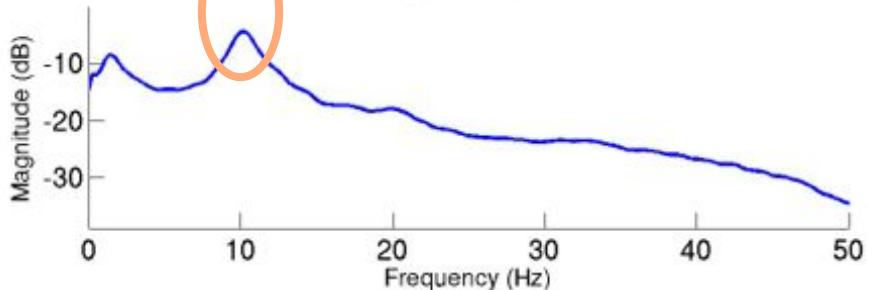
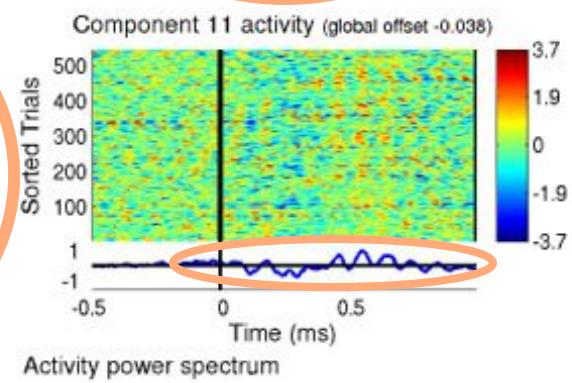
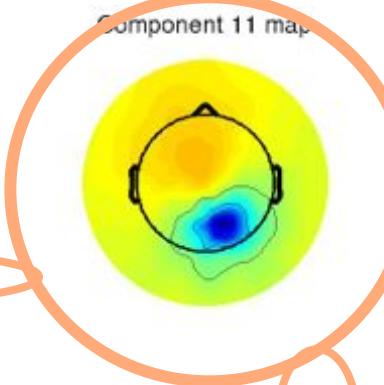
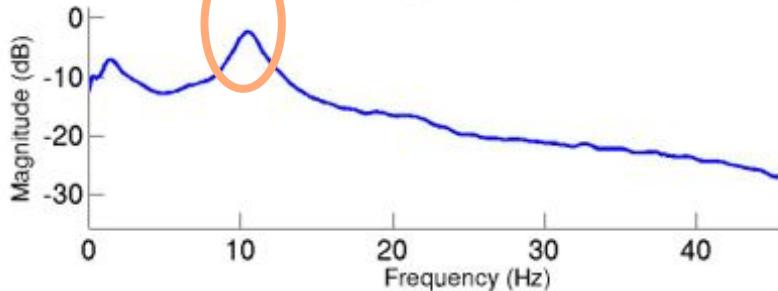
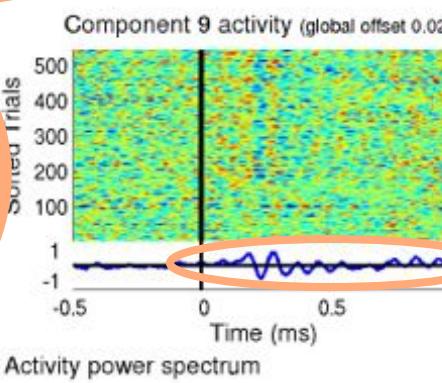
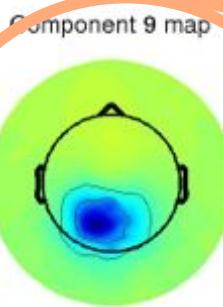
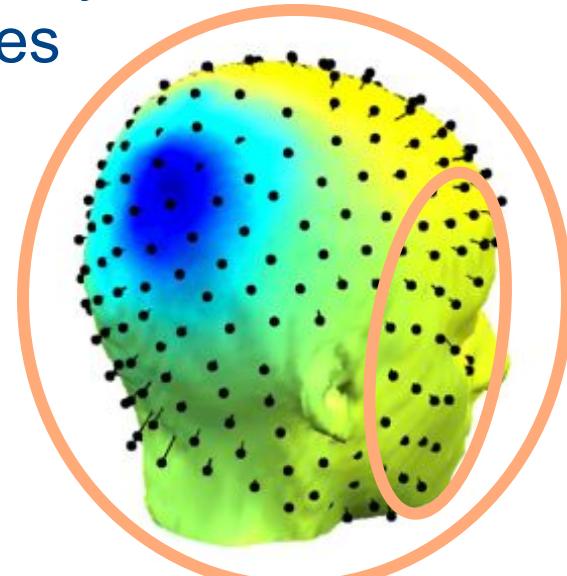
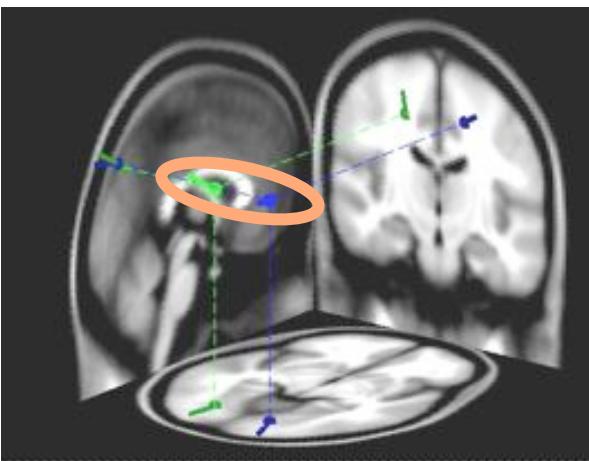
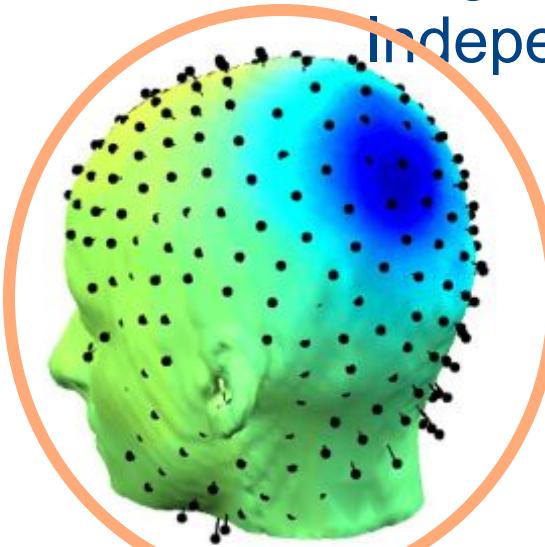
... and also separates cortical brain IC processes



IC9

IC11

Single Session - Two Maximally Independent Alpha Processes



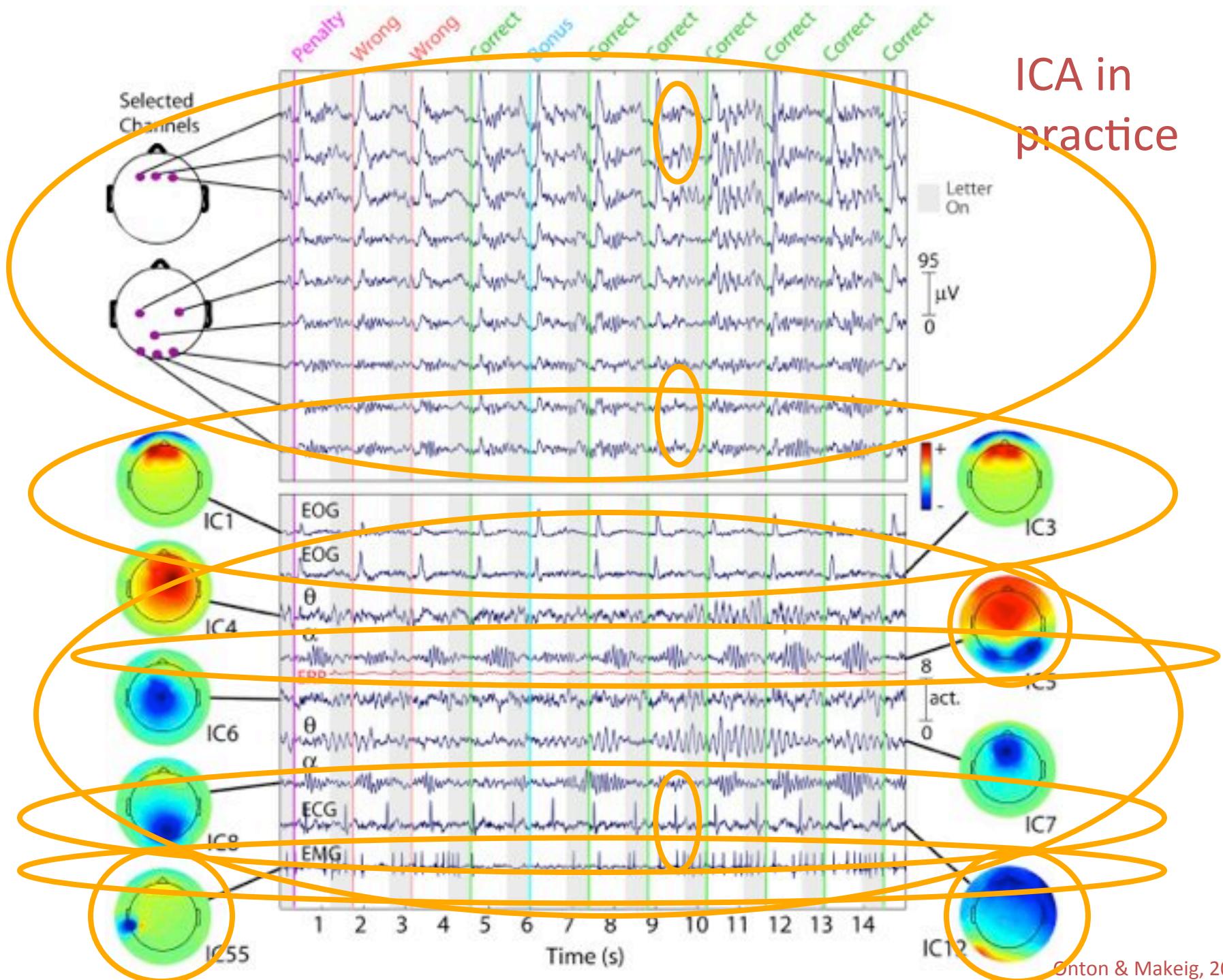
No more than
~30% of any scalp
channel variance
is produced by any
one brain source!

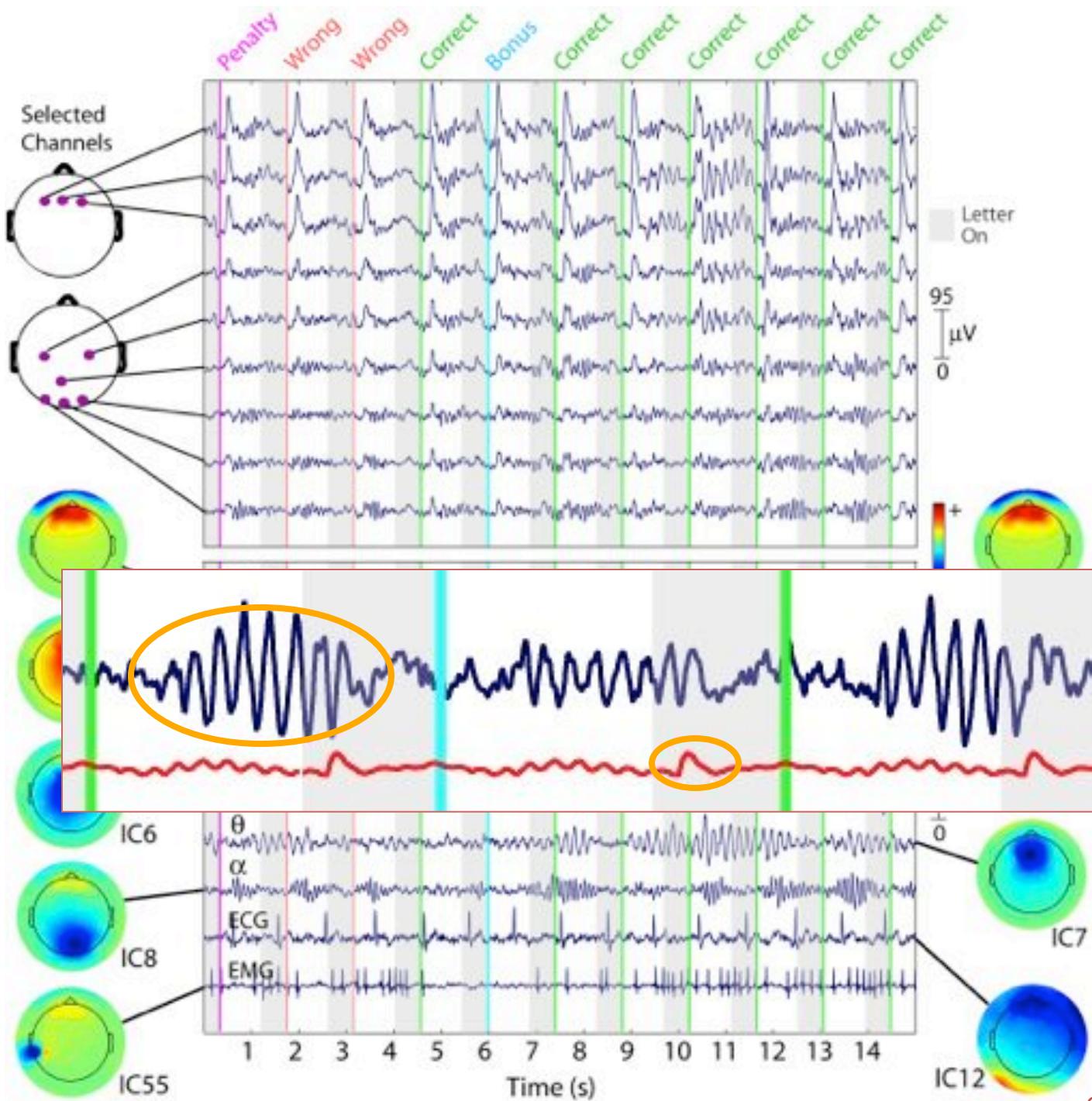
Scalp EEG signals are
strong mixtures
of brain sources.

In this sense scalp
channel signals are
epiphenomena.

Source signals are the EEG
phenomena of interest.

ICA in practice





Important Recent Result

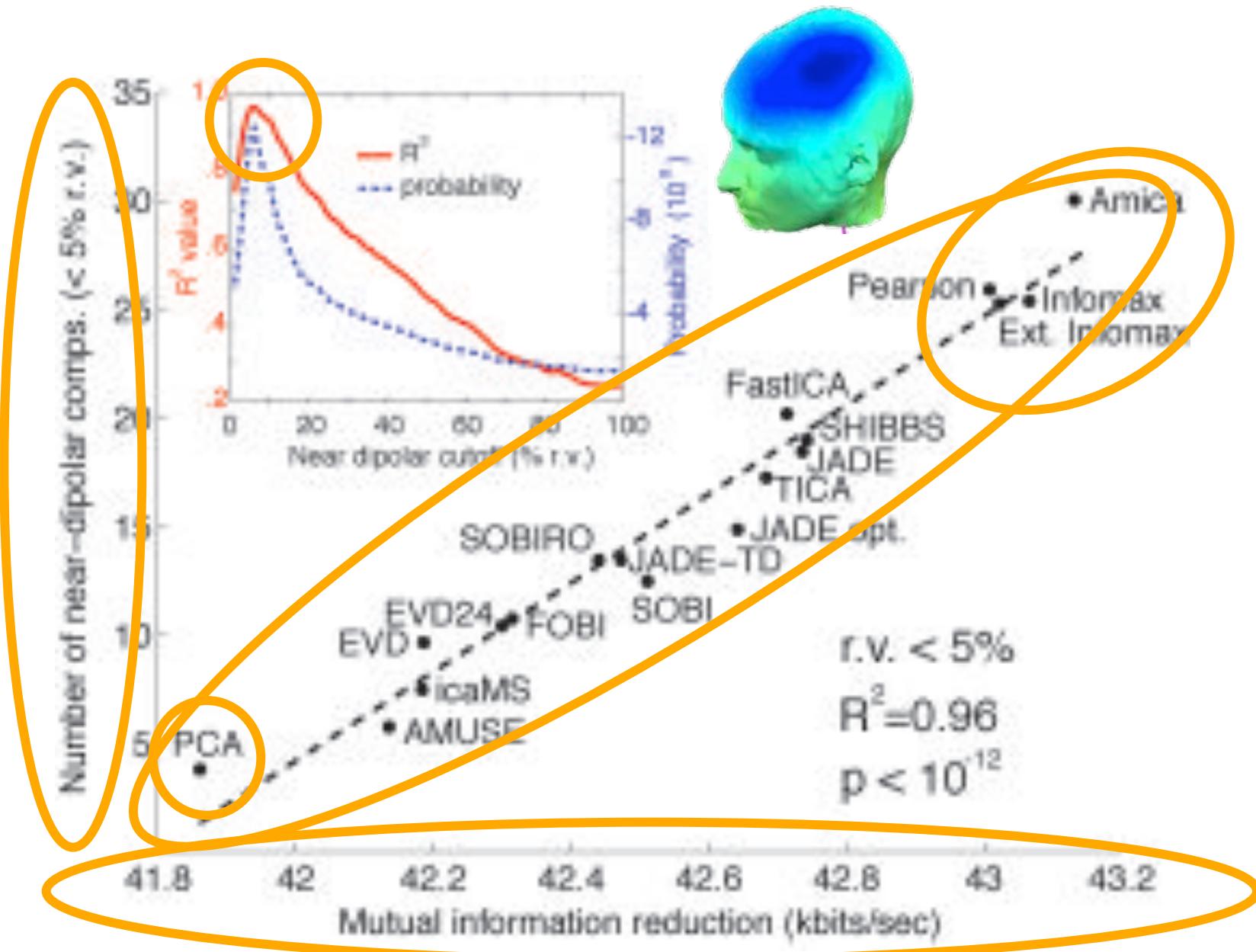
Those linear decompositions of multi-channel EEG data that find ICs whose time courses are more temporally independent

Also find more ICs whose scalp maps are highly ‘dipolar’ – i.e., ICs compatible with the spatial projection of a single local cortical (or non-brain, artifactual) source process – whose location can be identified.

More independent time courses \leftrightarrow Larger number of dipolar ICs

Dipolar ICs = Localized cortical source processes

Delorme et al., *PLOS One*, 2012

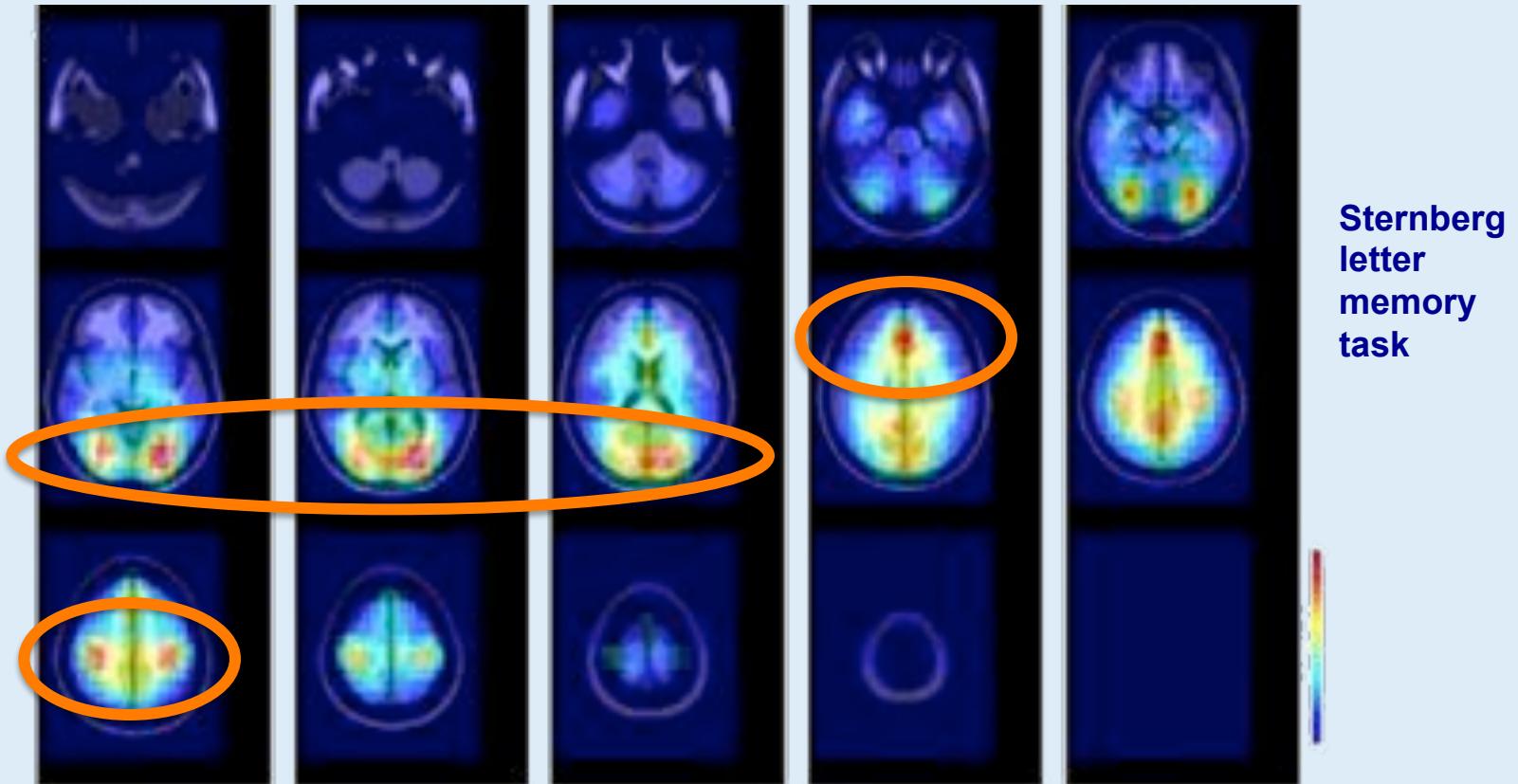


Delorme et al., *PLOS One*, 2012

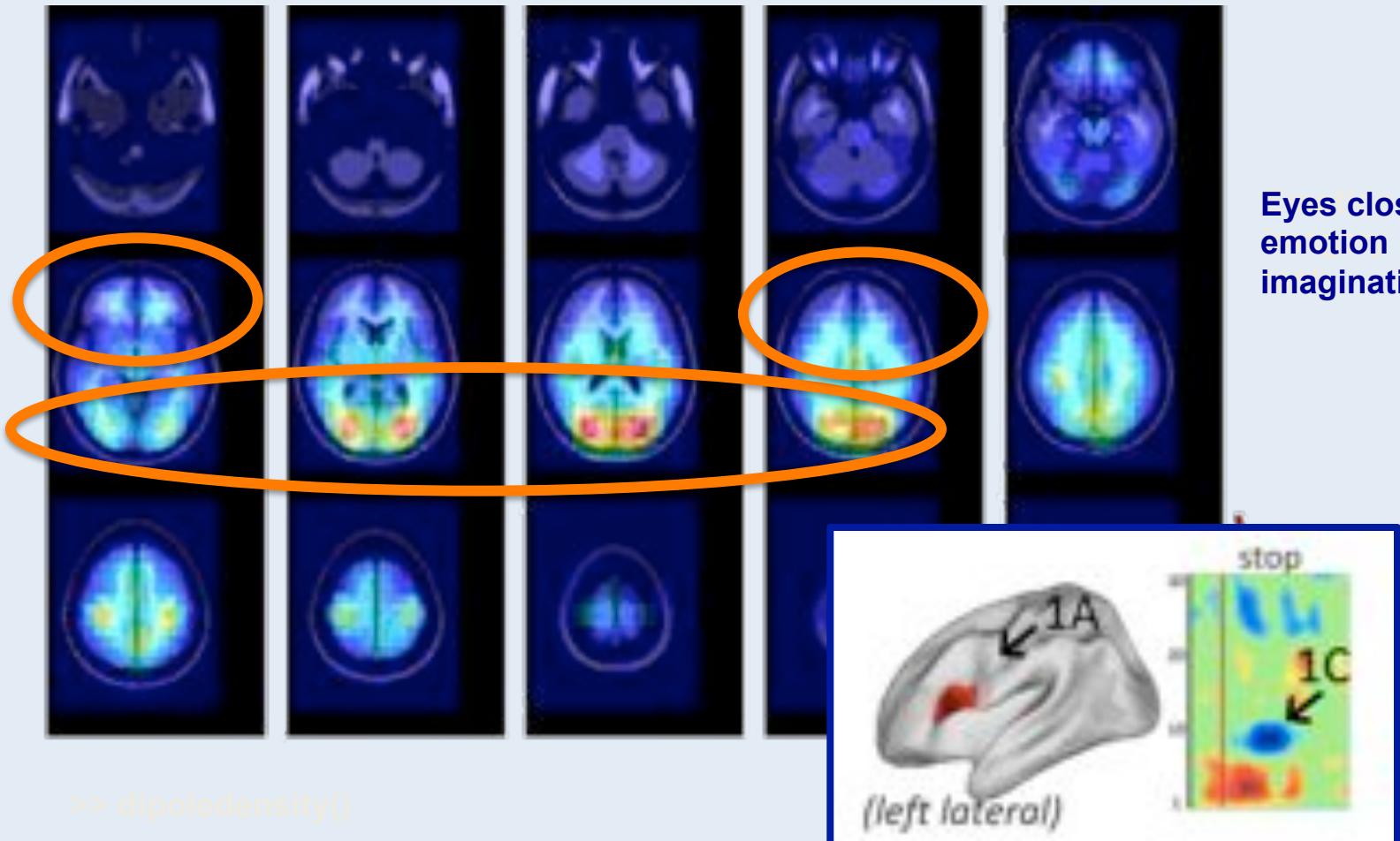
S. Makeig, 2011

Equivalent dipole density

Visual Working Memory



Equivalent dipole density





Questions ... ?