

# Signal Processing for Diagnosis and Treatment of Brain Disorders

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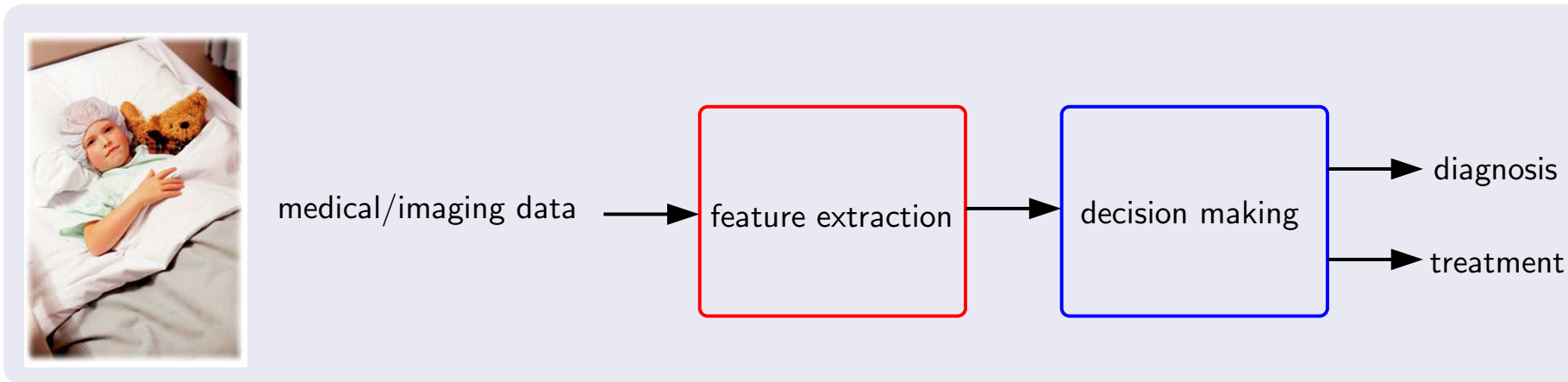
Laboratory for Information and Decision Systems (LIDS)  
Massachusetts Institute of Technology (MIT)

September 23, 2009

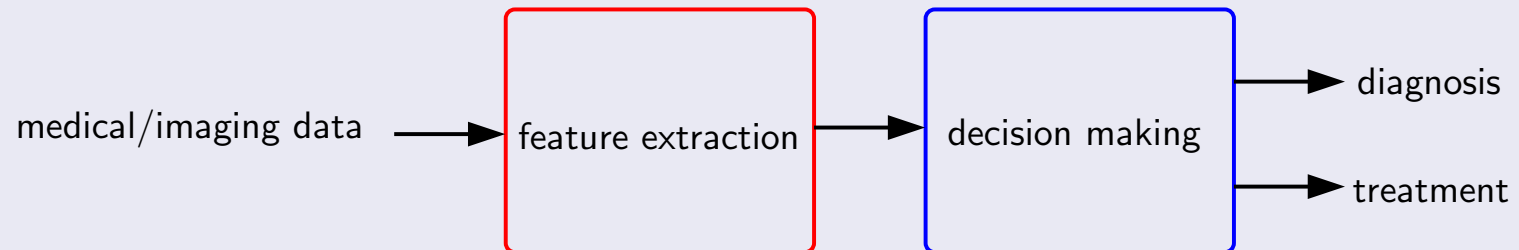
“With all the tools available to modern medicine—the blood tests, M.R.I.’s and endoscopes—you might think that **misdiagnosis** has become a rare thing. But you would be wrong.”

*Why doctors so often get it wrong*, David Leonhardt, New York Times, Feb. 22, 2006

# Proposed: Signal processing aided decision making



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- **Signal processing “microscope”**  
discovery of **features/patterns** in data associated with certain **disorders**
- **Algorithmic decision making**  
combine **evidence** “optimally” → **diagnosis** and **treatment**

- More **reliable** and **earlier** diagnosis
- More **effective** treatment (clinical outcome, time and cost)

# Synchrony of brain signals is an important feature

- Loss of synchrony of brain signals → brain disorder
- Normal synchrony of brain signals
- Increased synchrony of brain signals → brain disorder

Often subtle effects, requires signal processing

# Preview: Topics

## Signal Processing Aided Diagnosis

Loss in EEG synchrony → diagnosis of **early-Alzheimer's**

## Signal Processing Aided Treatment

Increased EEG synchrony → localization of **epileptic** brain tissue

## Future Work

Two **future** research projects

## Electroencephalogram (EEG)

= electrical activity along the **scalp** or **brain**

# Alzheimer's disease is complex, devastating, and common

- **Symptoms:** memory loss, language breakdown, loss of motor control, apathy, . . .
- **Causes:** not well understood, complex molecular mechanisms
- **2-5%** of people over **65** years old

# Our goal: early diagnosis of AD using EEG

## Early diagnosis of AD is important

- **medication** and other **therapies** most useful when used **early**

## More precisely: diagnosis of MCI and mild AD

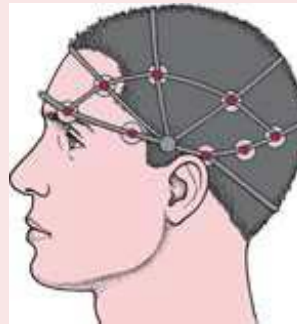
- **predementia** (a.k.a. **mild cognitive impairment**, MCI)
- **mild**–**moderate**–**severe** AD

# Our goal: early diagnosis of AD using EEG

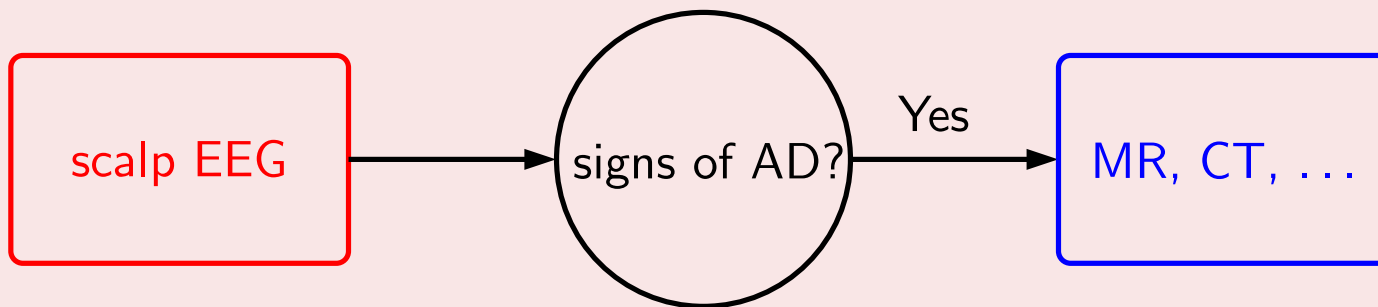
## Standard approach

mental testing, MR, CT, SPECT, PET, corticospinal fluid

## Scalp EEG is attractive complementary technology



- simple, affordable, mobile
- useful for screening



# EEG signals of AD patients are less coherent

- Loss of neurons → weaker brain connectivity

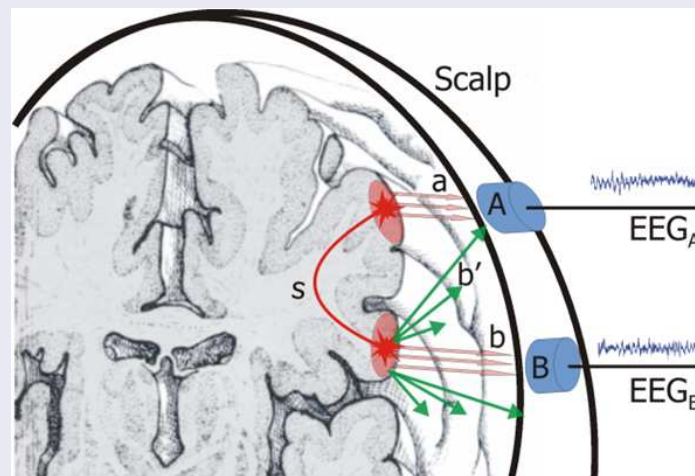
Healthy subject



AD patient



- EEG from different brain areas is less correlated in AD patients



- Loss in EEG synchrony → loss in brain connectivity → AD
- How to **detect** loss in EEG synchrony?

# How to measure EEG synchrony?

## Classical techniques

- correlation coefficient, coherence
- phase synchrony
- Granger causality
- information-theoretic measures
- ...

**Few** of them applied on **single** data sets of early-AD patients

# How to measure EEG synchrony?

## Systematic study

**30+** measures on **two** data sets (and more to follow)

# How to measure EEG synchrony?

## Systematic study

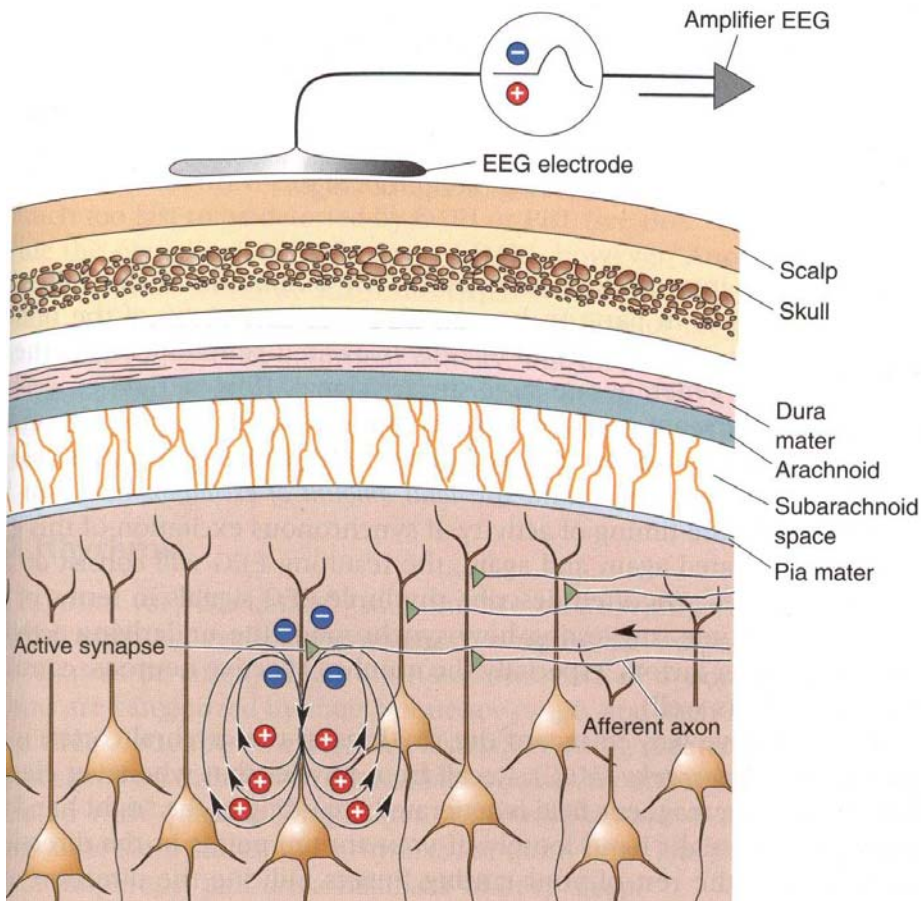
**30+** measures on **two** data sets (and more to follow)

## New approach: “Stochastic Event Synchrony (SES)”

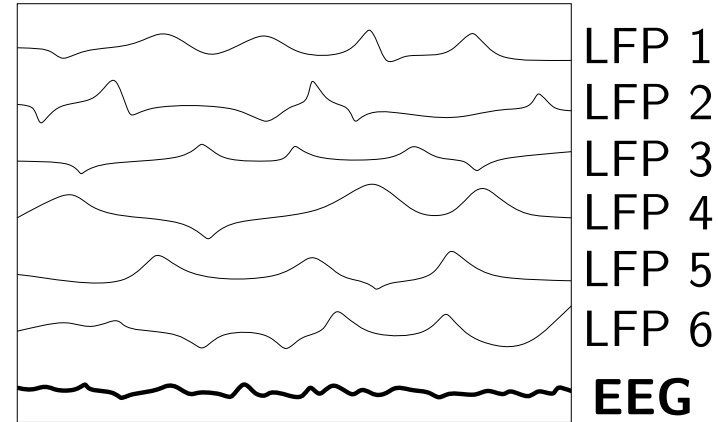
- Extracts “events” from EEG signals and **aligns** those events
- Provides **complementary** information about synchrony
- **Promising** results for diagnosing AD
- **General** formalism for **similarity** of **point processes**

In this talk: **pairs** of **two**-dimensional point processes (EEG)

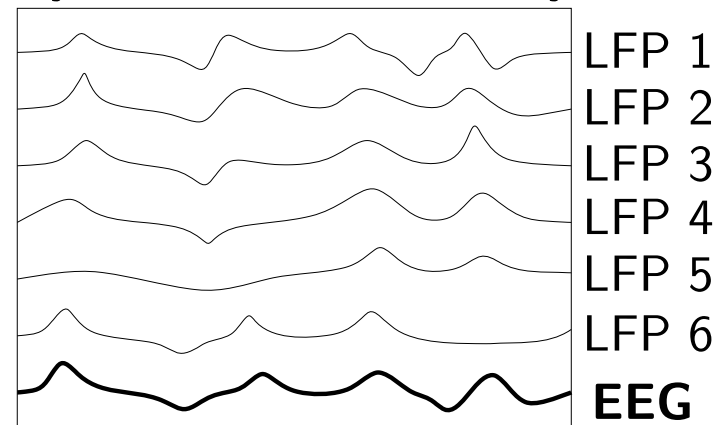
# Local synchronous activity leads to "events" in EEG



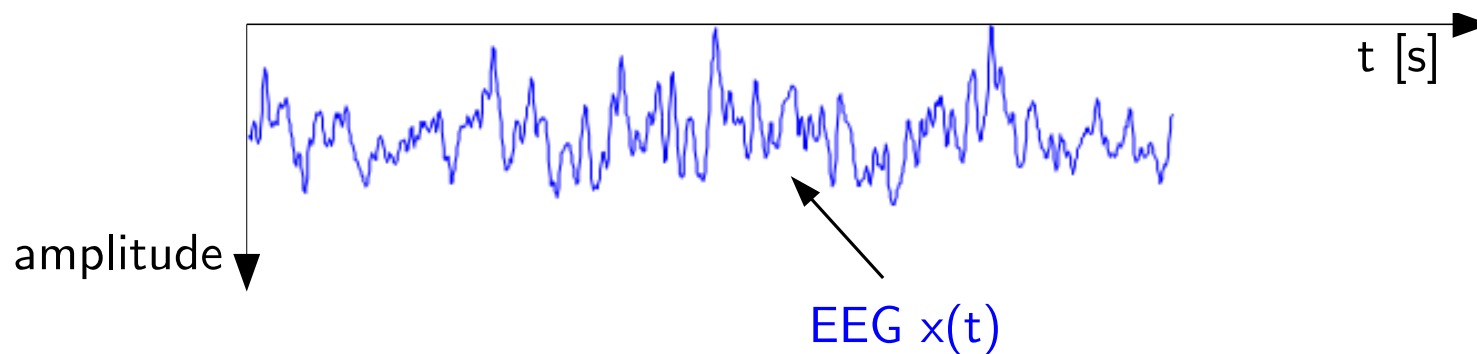
## Non-synchronized activity



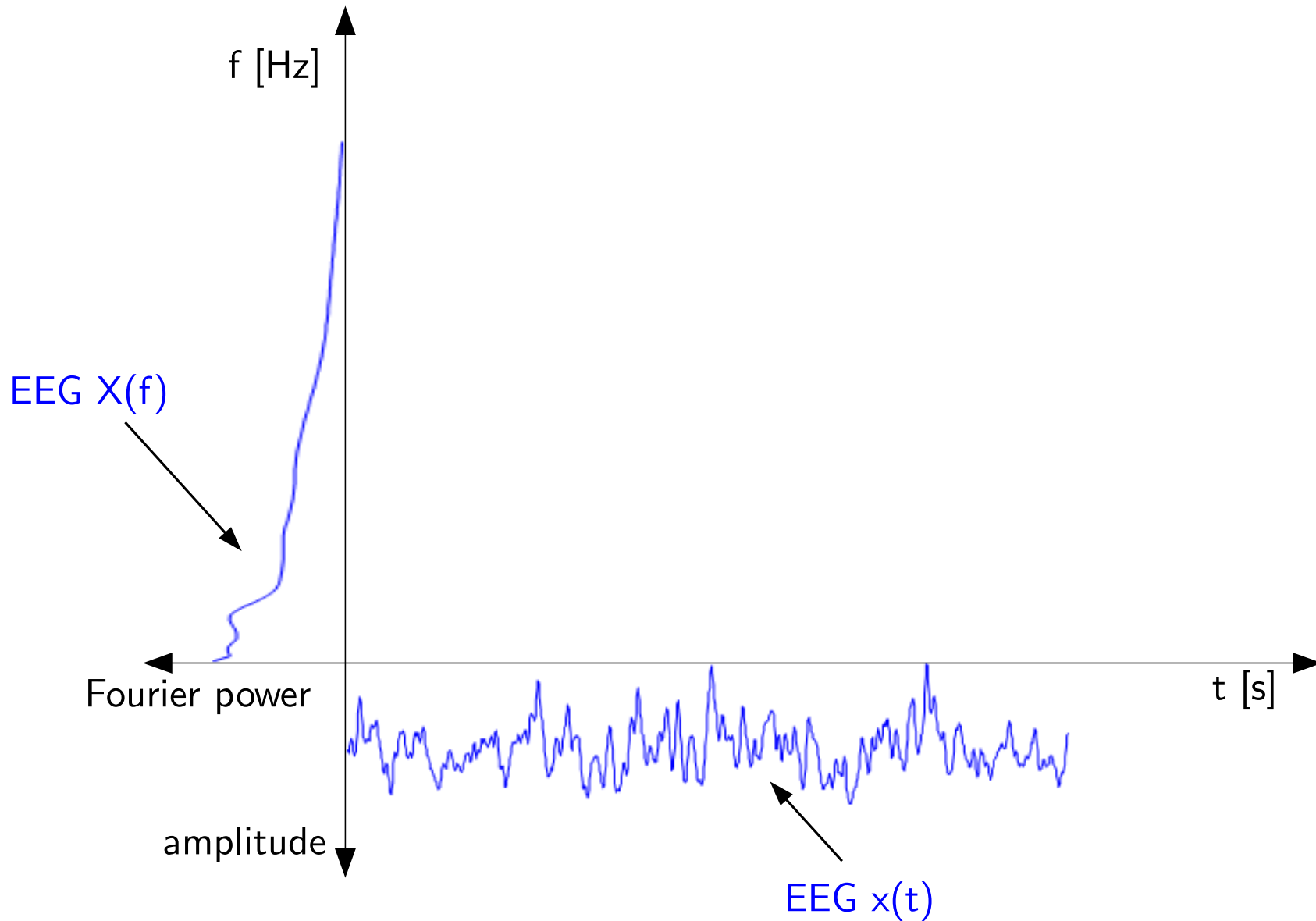
## Synchronized activity



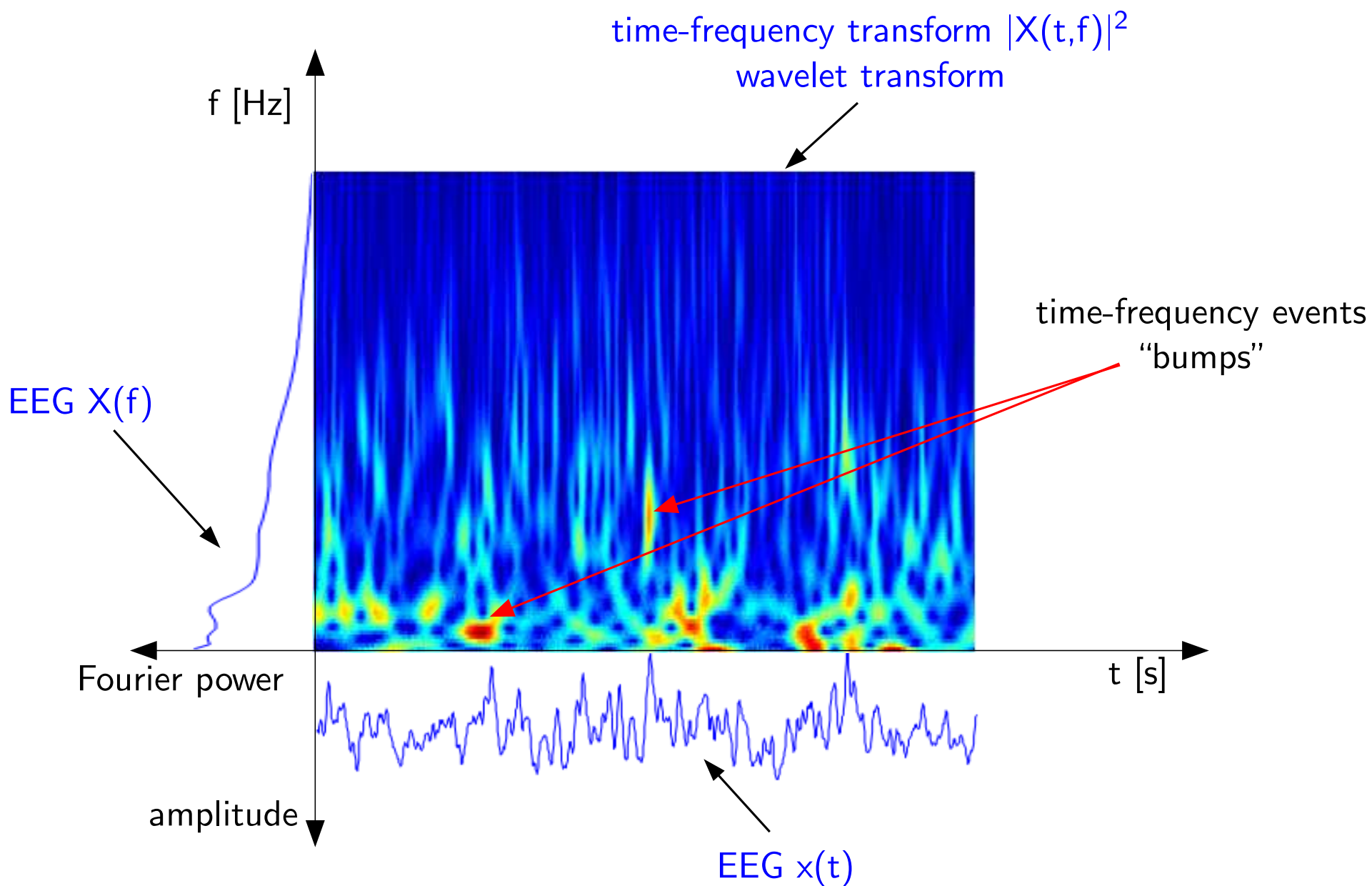
# Typical EEG signal in time-domain



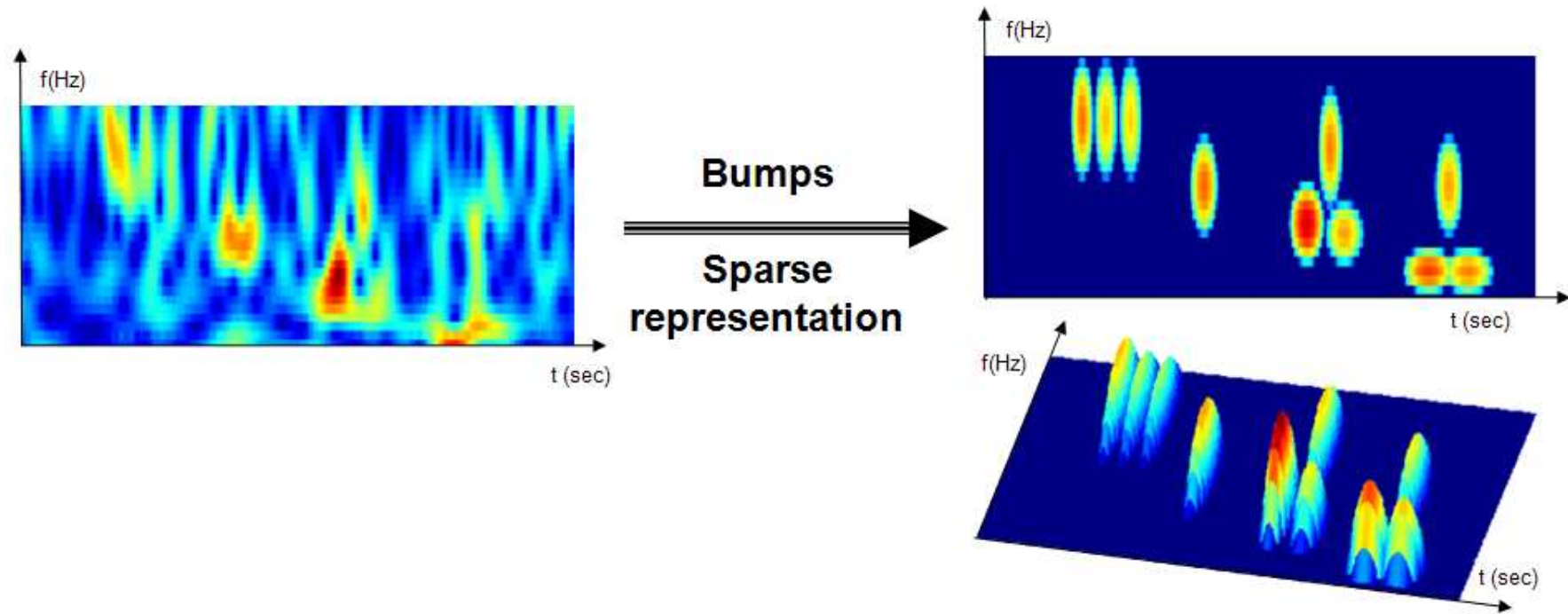
# Frequency-domain representation of the same signal



# Events can clearly be observed in time-frequency domain



# Oscillatory events extracted from time-frequency map



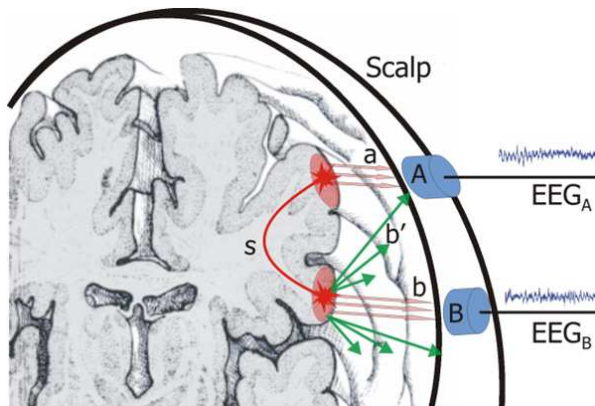
10.000 coefficients

→

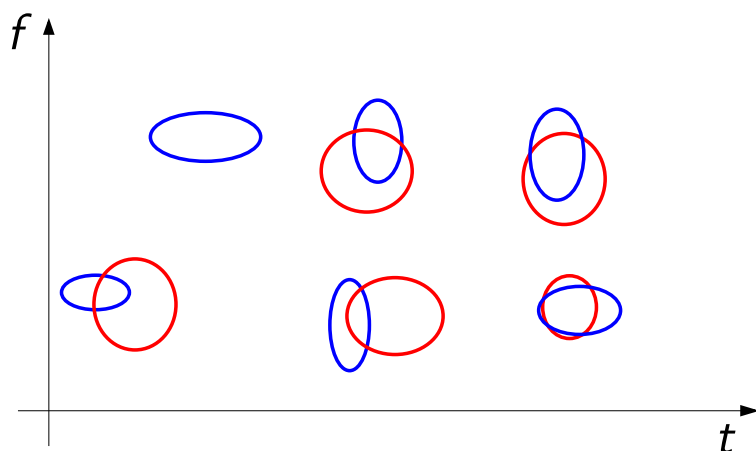
50 parameters

- Bump models are **point processes** on time-frequency plane
- **5 bump parameters**:  
timing, frequency, width in time and frequency, amplitude

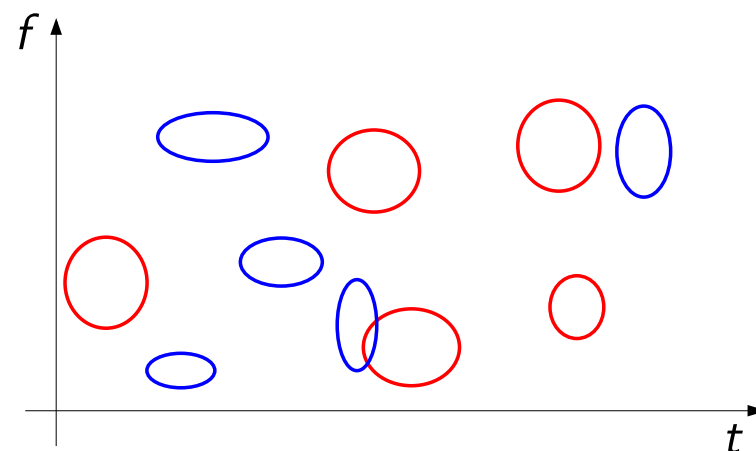
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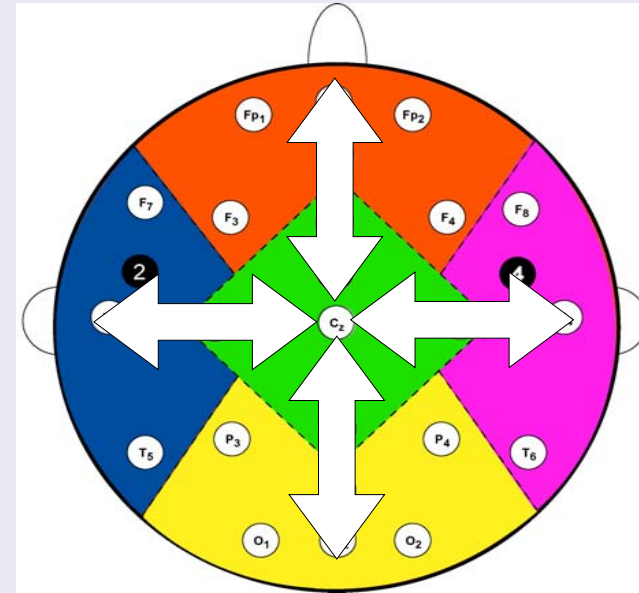
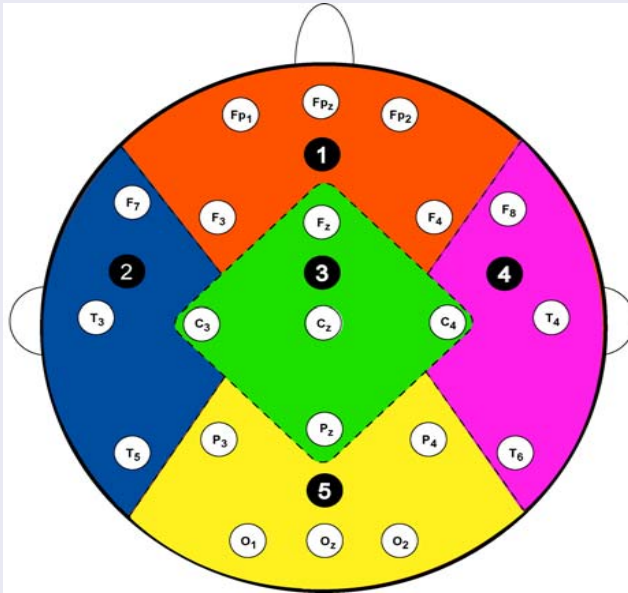
Healthy subject



AD patient



# How do we compute synchrony of events?

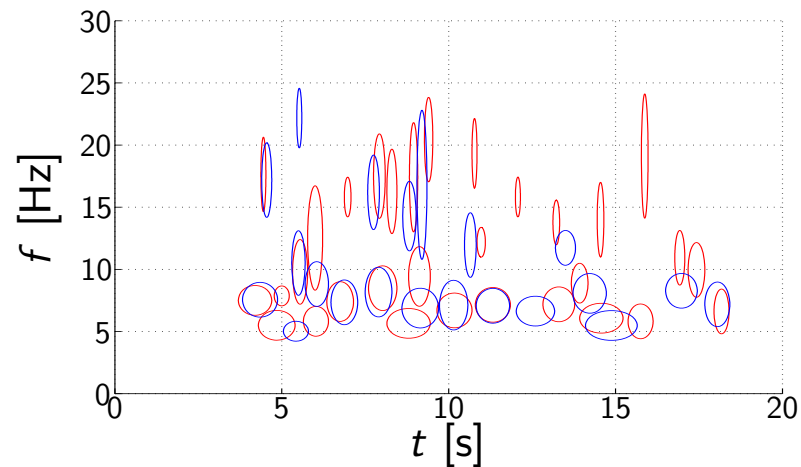


- 1 Group electrodes in regions
- 2 Extract bump model for each region
- 3 Determine similarity of each pair of bump models
- 4 Average similarity over all pairs

Misalignment of bump models → loss of brain connectivity → AD

# Stochastic Event Synchrony

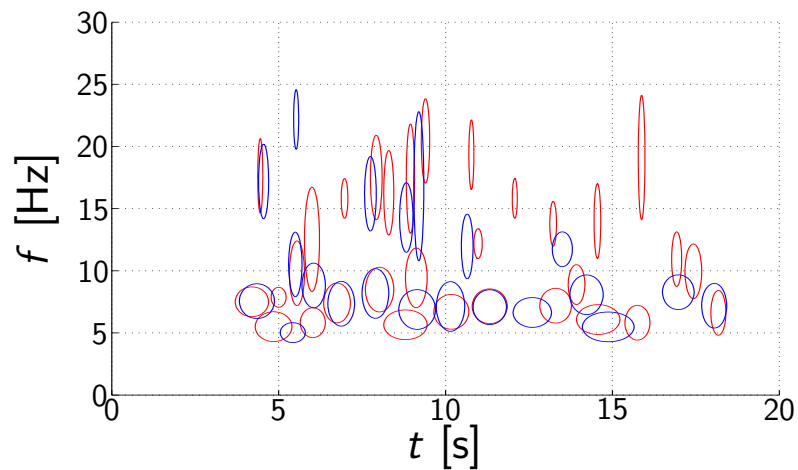
## Bump models of two EEG signals



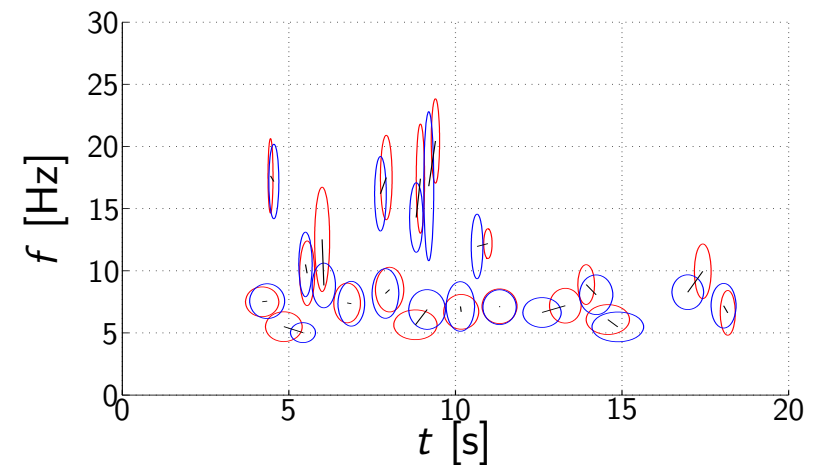
① How many bumps coincide?

# Stochastic Event Synchrony

Bump models of two EEG signals



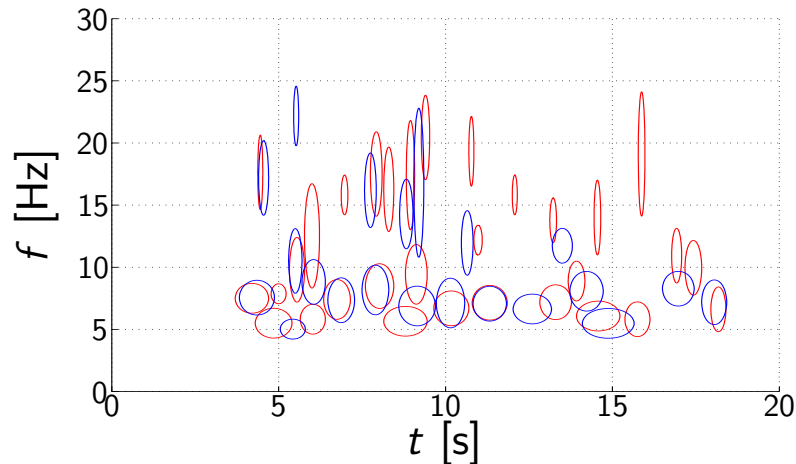
Coincident bumps



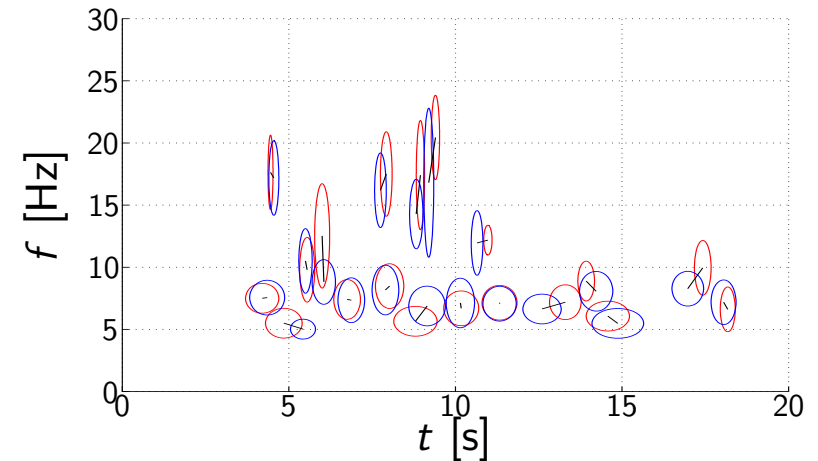
- 1 How **many** bumps **coincide**?
- 2 How **well** do the coincident bumps **align**?

# Stochastic Event Synchrony ( $\rho, \delta_t, \delta_f, s_t, s_f$ )

## Bump models of two EEG signals



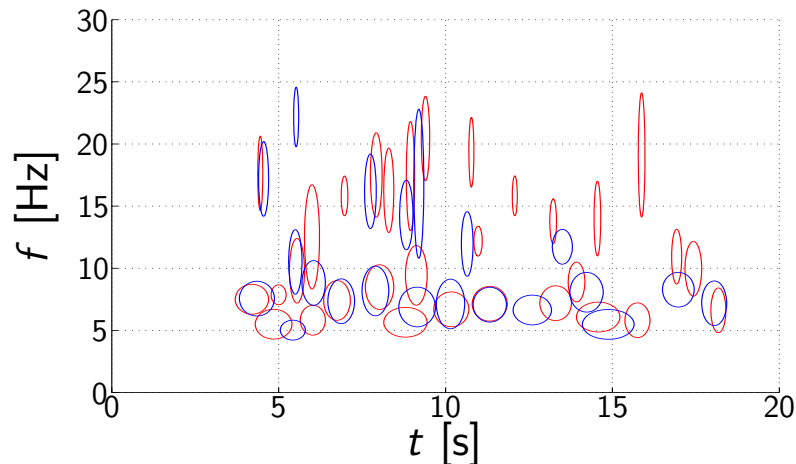
## Coincident bumps



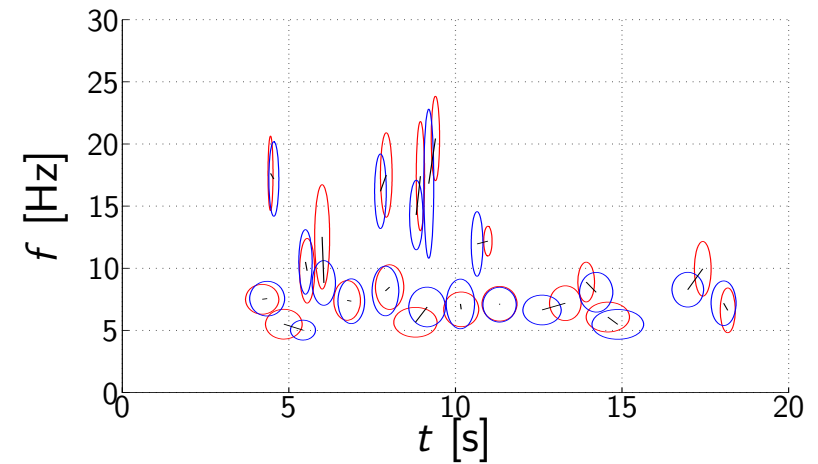
- 1 How **many** bumps **coincide**?
  - fraction of **non-coincident** bumps  $\rho$
- 2 How **well** do the coincident bumps **align**?
  - average  $t$  and  $f$  offset of **coincident bumps** ( $\delta_t$  and  $\delta_f$ )
  - $t$  and  $f$  jitter of **coincident bumps** ( $s_t$  and  $s_f$ )

# Stochastic Event Synchrony ( $\rho, \delta_t, \delta_f, s_t, s_f$ )

Bump models of two EEG signals



Coincident bumps



① How **many** bumps **coincide**?

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② How **well** do the coincident bumps **align**?

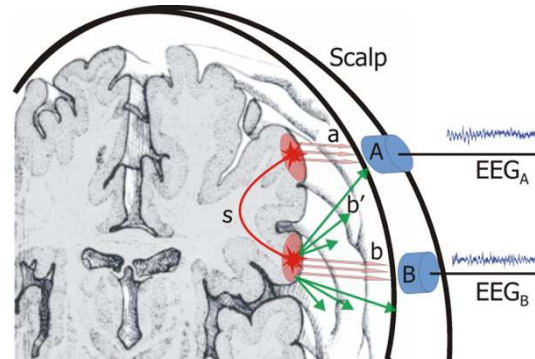
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It is natural to **iterate** Step 1 and 2

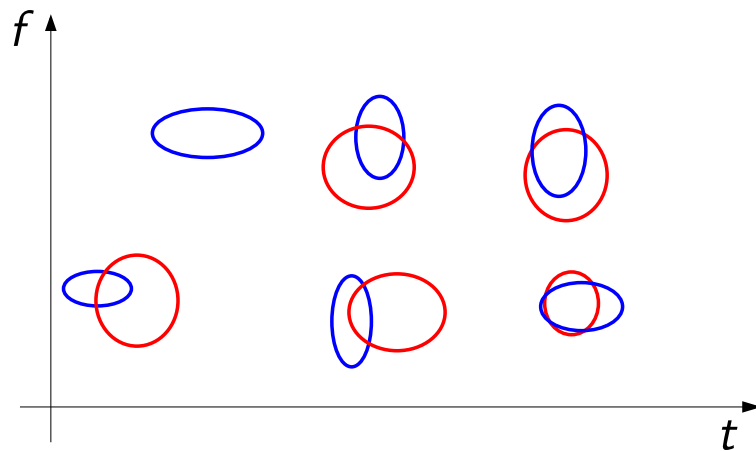
Want to know more? Want to try it out?

<http://www.dauwels.com/SESToolbox/SES.html>

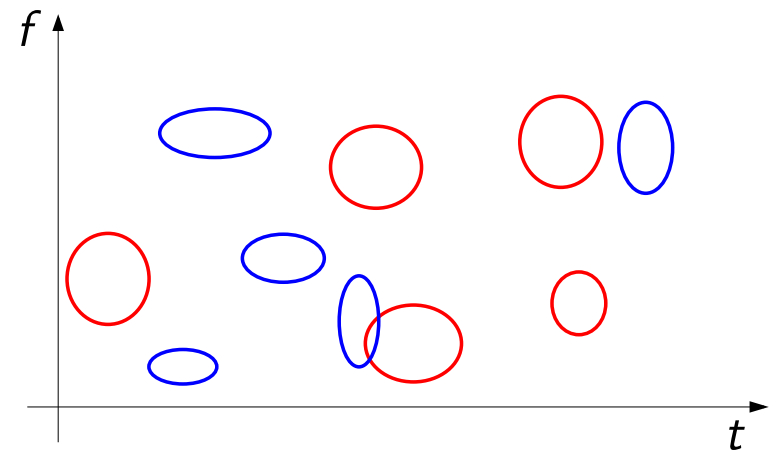
# Hypothesis: events are less correlated in AD patients



Healthy subject



AD patient



- larger jitter of coincident bumps ( $s_t$ )
- larger fraction of non-coincident bumps ( $\rho$ )

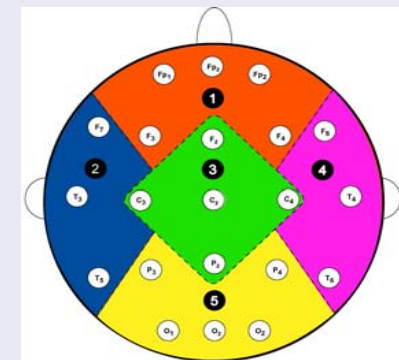
# We applied SES to EEG of healthy and early-AD patients

## 2 data sets, **different** patients/hospitals

- 22 MCI and 38 Control subjects
- 17 Mild AD and 24 Control subjects

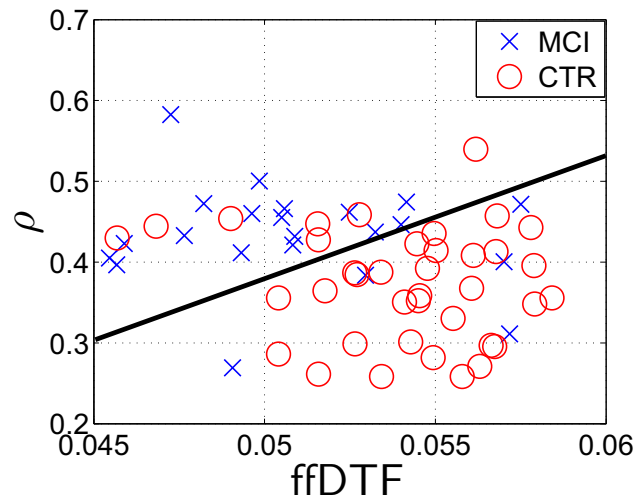
## Setup

- spontaneous EEG, in rest with eyes closed
- 21 electrodes (10-20 international system)
- electrodes grouped in 5 zones
- bandpass filtered between 4 and 30Hz
- 20s of artifact-free EEG
- 30+ synchrony measures, including SES

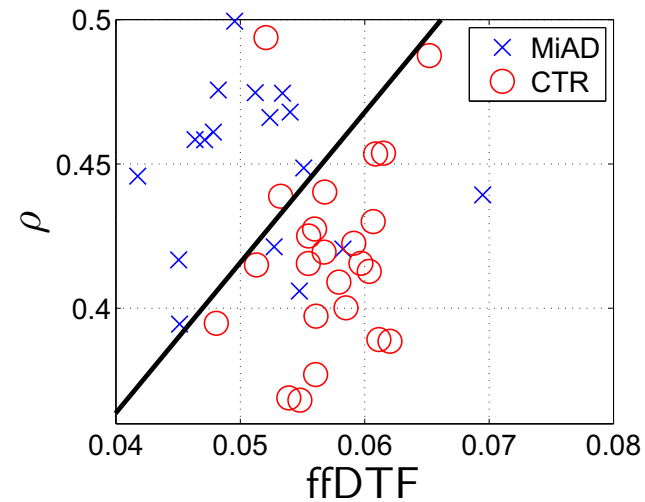


# Promising and consistent numerical results

## MCI vs. Control



## Mild AD vs. Control



- Only **full-frequency direct transfer function** (ffDTF) and  $\rho$  strongly significant ( $p < 0.001$ ), for **BOTH** data sets
- MCI vs. Control **harder** than Mild AD vs. Control

# Promising and consistent numerical results

## Discriminant analysis with leave-one-out crossvalidation

MCI vs. CTR	Linear DA	Quadratic DA
ffDTF	70.0%	70.0%
$\rho$	68.3%	75%
ffDTF and $\rho$	<b>83.3%</b>	<b>83.3%</b>
MiAD vs. CTR	Linear DA	Quadratic DA
ffDTF	82.9%	75.6%
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- MCI vs. Control **harder** than Mild AD vs. Control

- **EEG synchrony loss** seems to be strong indicator of early **AD**
- **SES improves** diagnosis of AD (for those two data sets)

# Preview: Topics

## Signal Processing Aided Diagnosis

Loss in EEG synchrony → diagnosis of early-Alzheimer's

We needed to develop **novel** measures of synchrony

## Signal Processing Aided Treatment

Increased EEG synchrony → localization of epileptic brain tissue

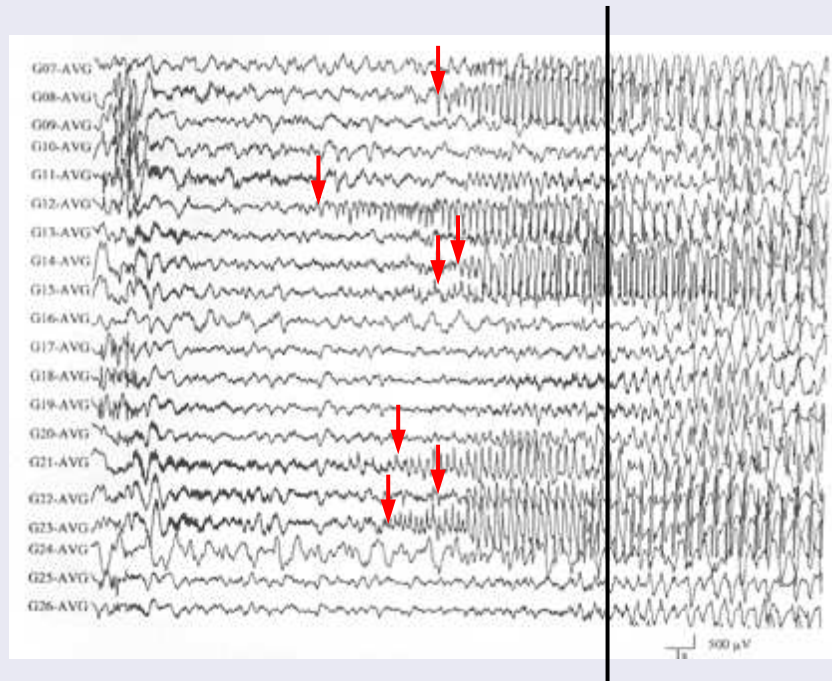
**Existing** synchrony measures proved to be effective

## Future Work

Two future research projects

# Epilepsy is common, complex, and devastating

- 2.5 million patients in U.S., 300'000 in U.K., 50 million world
- **Heterogeneous group** of central nervous system disorders characterized by **recurrent unprovoked seizures**
- **Symptoms**: disturbance of consciousness or awareness, alterations of bodily movement, sensation or posture, . . .
- **Causes**: genetic/metabolic abnormalities, tumors, . . .



# Medication is usual treatment but not always effective

- **Medication** is most common treatment
- However, medication alone not effective for **30%** of patients

# If medication alone fails, surgery may be solution

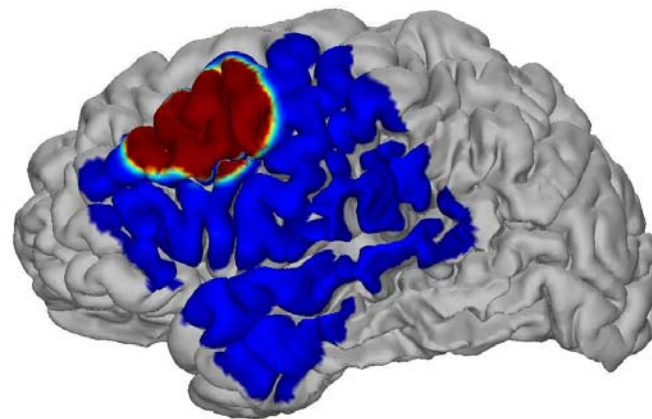
## Resection if medication alone fails

- **Resection** may be alternative for **location-related** epilepsy
- **Objective**: remove **seizure onset zone**
- **How to determine the seizure onset zone?**
  - from **clinical behavior** (e.g., left arm shaking) or **scalp EEG**
  - from **intracranial EEG**, if everything else fails  
risky, costly, uncomfortable

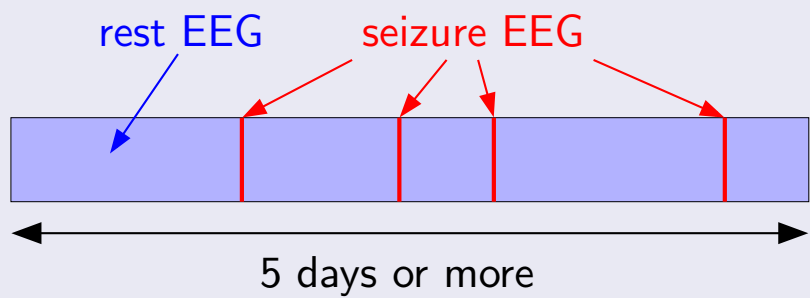
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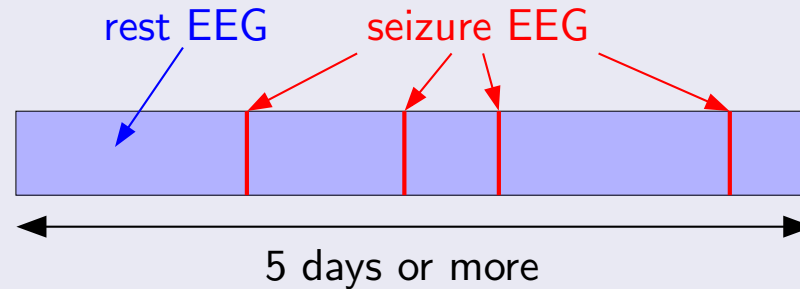
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# Standard procedure relies mostly on “seizure” EEG

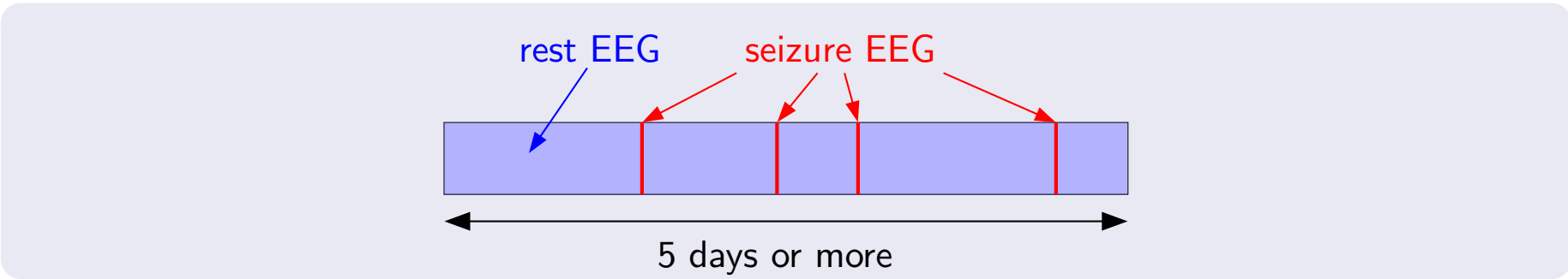


# Our approach: exploit rest EEG to shorten hospitalization

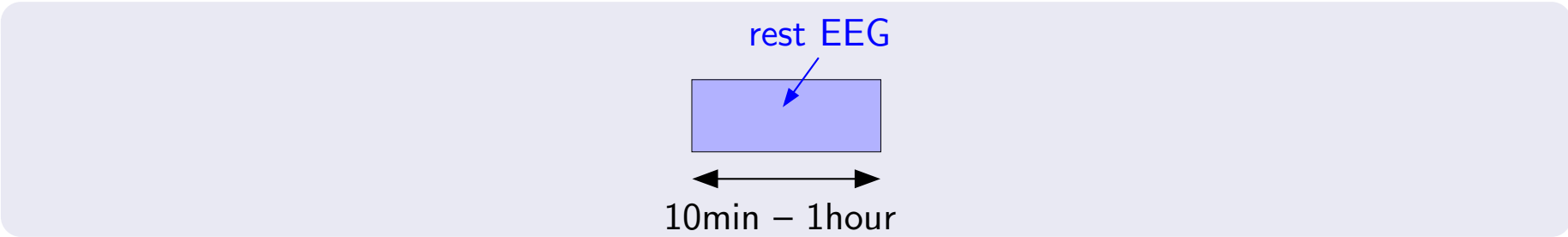


We wish to make intracranial recordings as **short** as possible

# Our approach: exploit rest EEG to shorten hospitalization



We wish to make intracranial recordings as **short** as possible



- Objective: Seizure Focus Localization**
- Use **rest** EEG only → **shorter** hospitalization
  - **Automated** procedure → **faster** and potentially more **reliable**

# Differences with diagnosis of AD project

- Intracranial EEG vs. scalp EEG
- Signal processing to guide surgery vs. diagnosis
- Increased EEG synchrony vs. EEG synchrony loss

# Interictal characteristics of seizure onset zone

- Locally **enhanced** synchrony (“hypersynchrony”)
- **Slowing** is signature of damaged cortex
- Interictal **spikes**
- Interictal **high-frequency bursts**

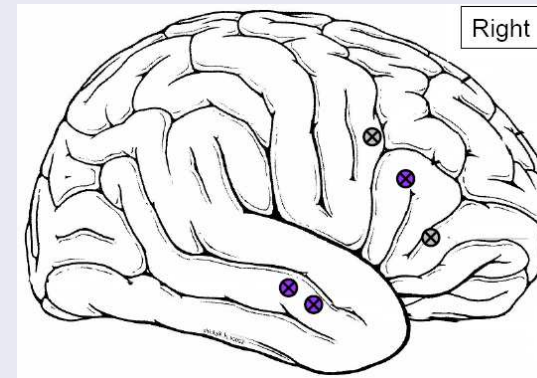
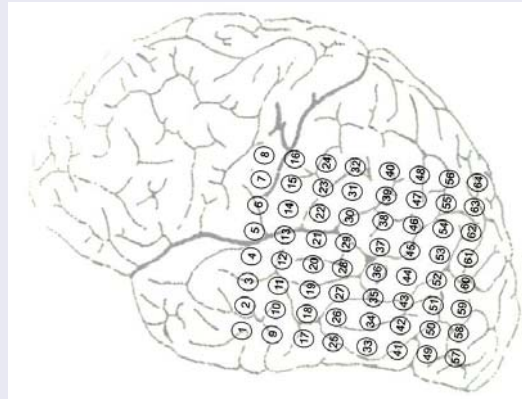
# Three main questions

## Hypothesis

**Seizure onset zone** is characterized by **hypersynchronous** activity even while at **rest**

- 1 Does **hypersynchronous** activity occur in rest EEG?
- 2 Do hypersynchronous areas correlate with **seizure onset zone**?
- 3 How to determine **seizure onset zone** using hypersynchrony?

# Our approach



- 6 patients with **grid electrodes**, 5 with **depth electrodes**
- **1 hour** segment of intracranial **interictal EEG**
- **48 hours** between segment and seizures
- **bandpass filter** between 1–200Hz, **notch filter** at 60Hz
- **no** further preprocessing
- **several** synchrony measures  
correlation coefficient, phase synchrony, Granger causality, ...
- **actual** seizure focus determined by clinicians from **seizure EEG**

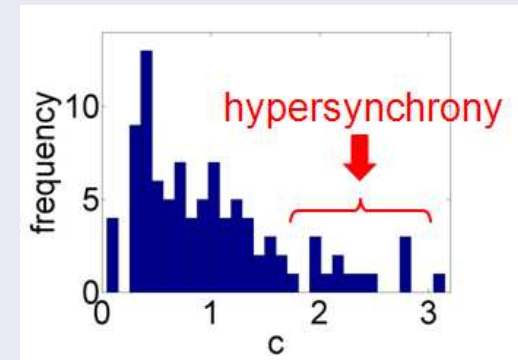
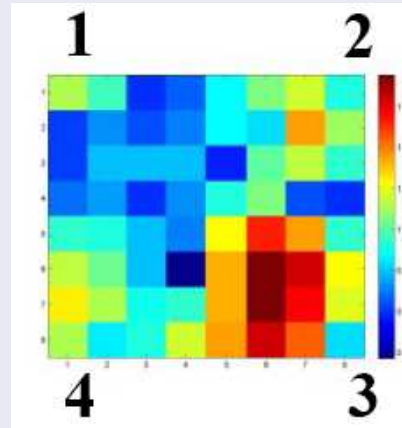
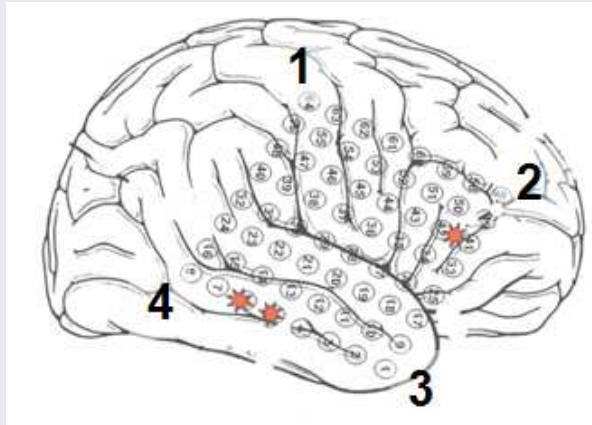
# Some brain areas are consistently hypersynchronous

## Strong local correlation in anterior temporal lobe

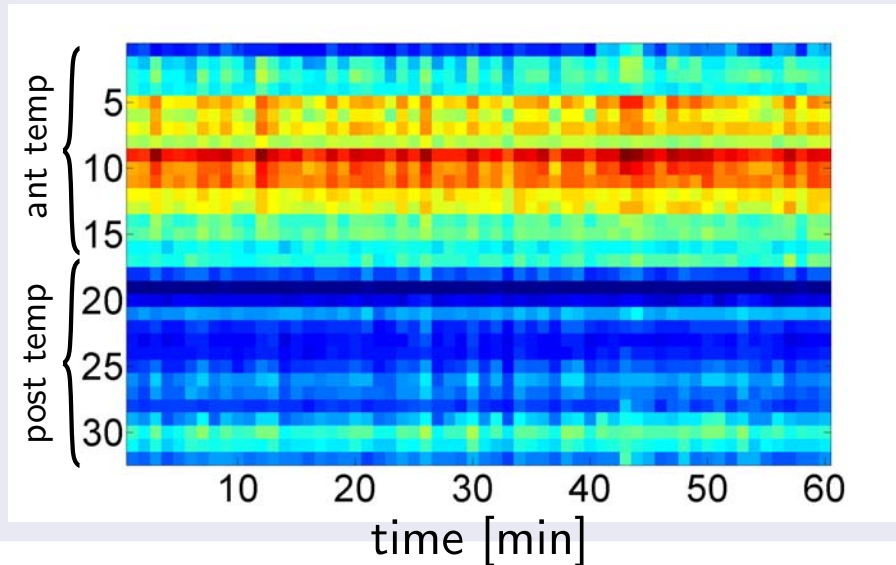
Patient 1

correlation coeff  $c$

histogram



## Variability over time is small (less than 15%)



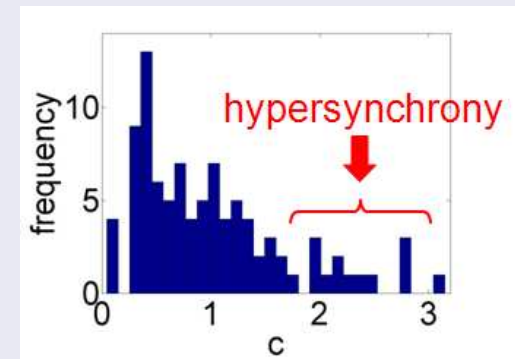
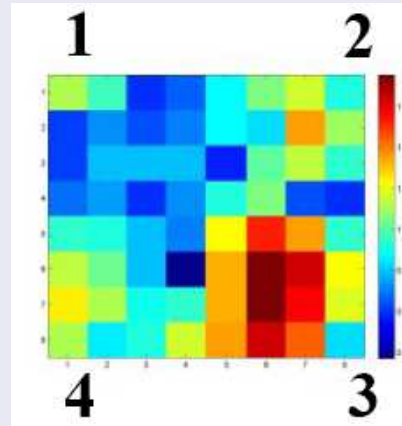
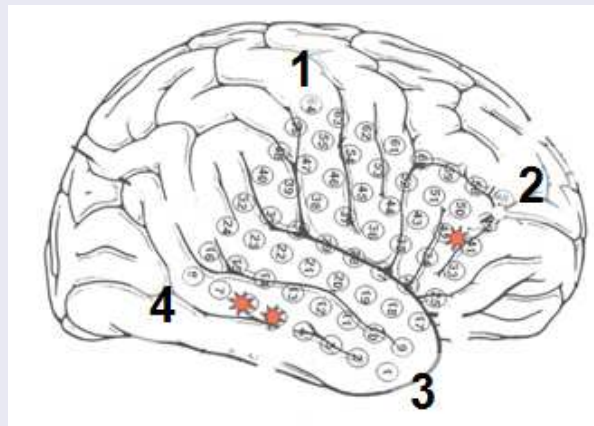
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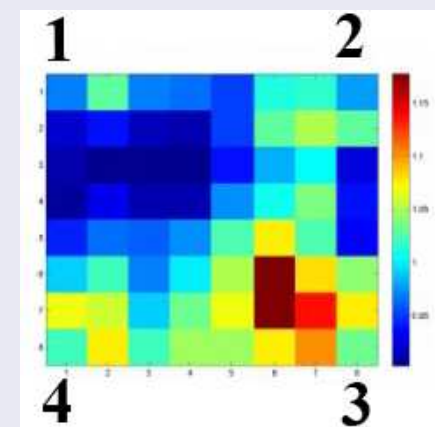
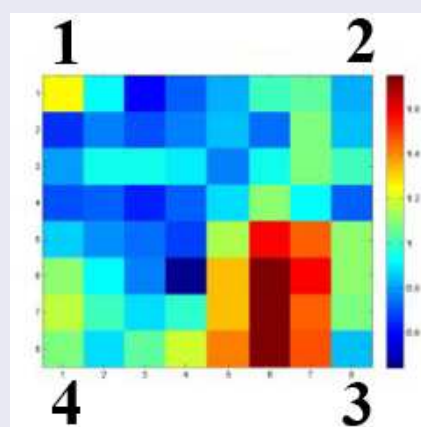
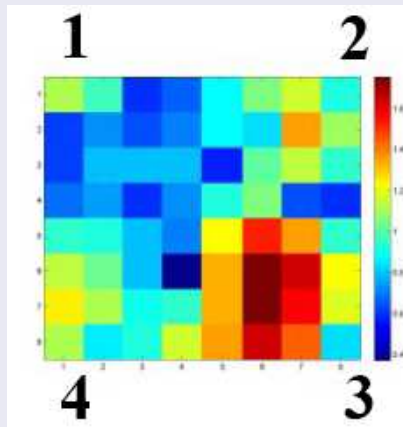


## Very similar results for other synchrony measures

Phase Synchrony

Coherence

Granger causality (DTF)



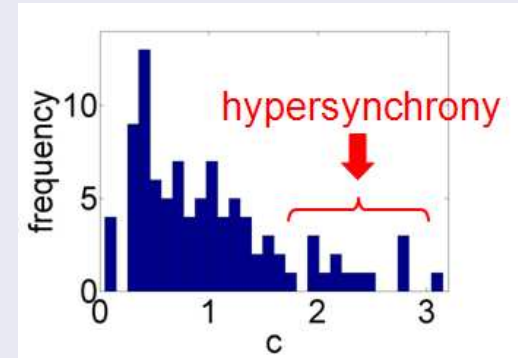
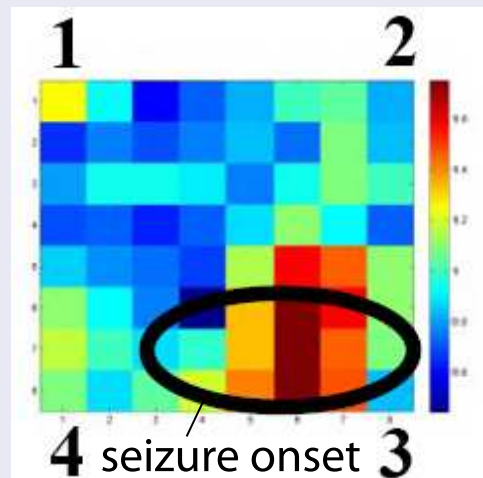
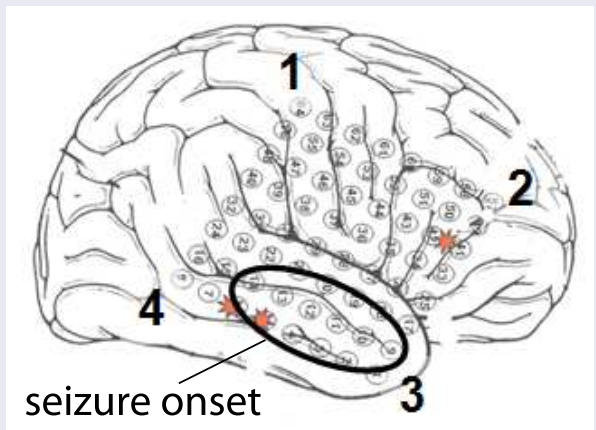
# Hypersynchrony correlates strongly with seizure onset zone

## Strong local correlation in anterior temporal lobe

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correlation coeff  $c$

histogram

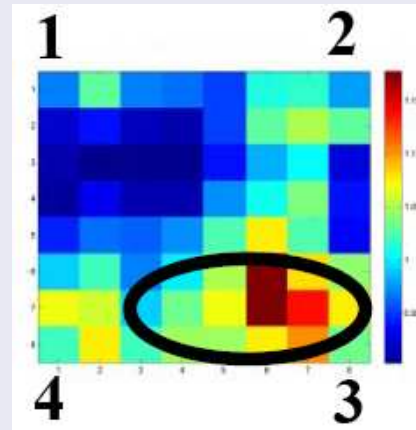
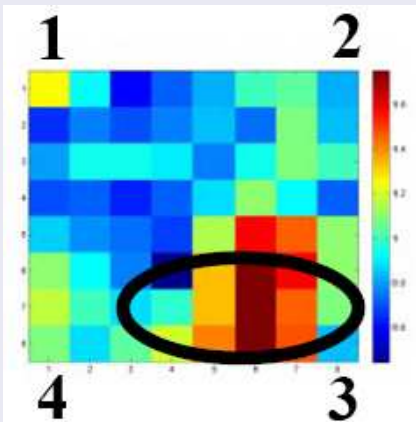
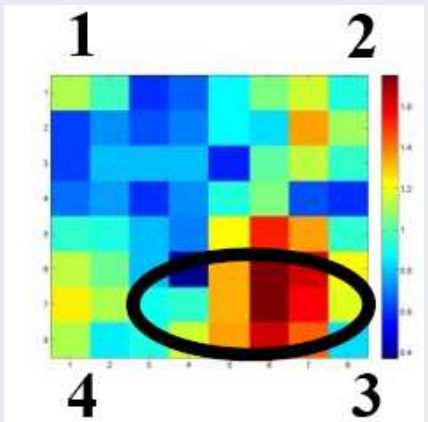


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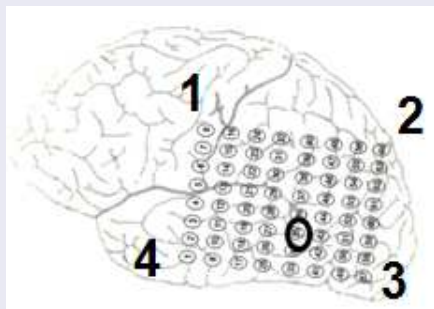
Coherence

Granger causality (DTF)

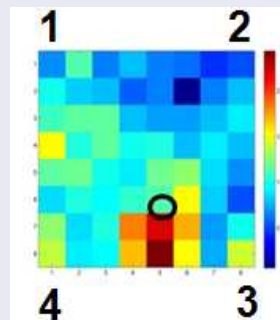


# Same phenomenon also occurs in most other patients

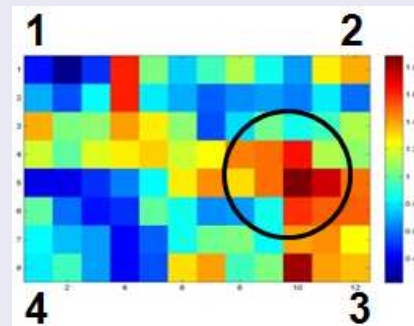
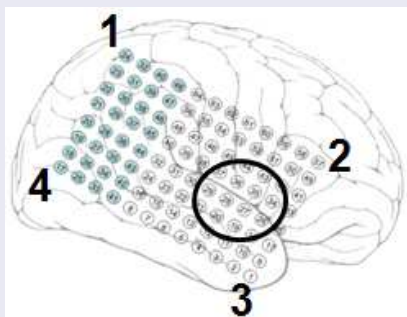
### Patient 2 (left hemisphere)



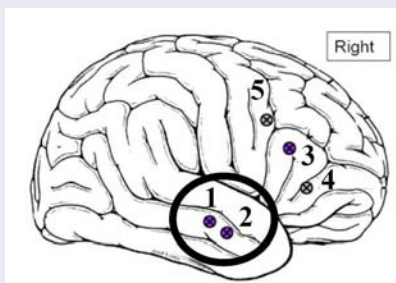
### correlation coefficient



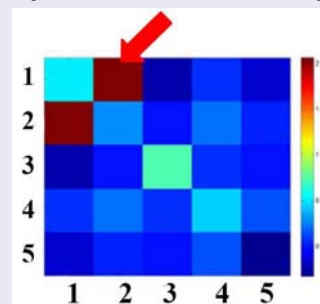
### Patient 3 (right hemisphere)



### Patient 4 (right hemisphere)



### (right - left)



# Three main questions

① Does **hypersynchronous** activity occur in rest EEG?

**Yes!** (in **9** of the 11 patients analyzed so far)

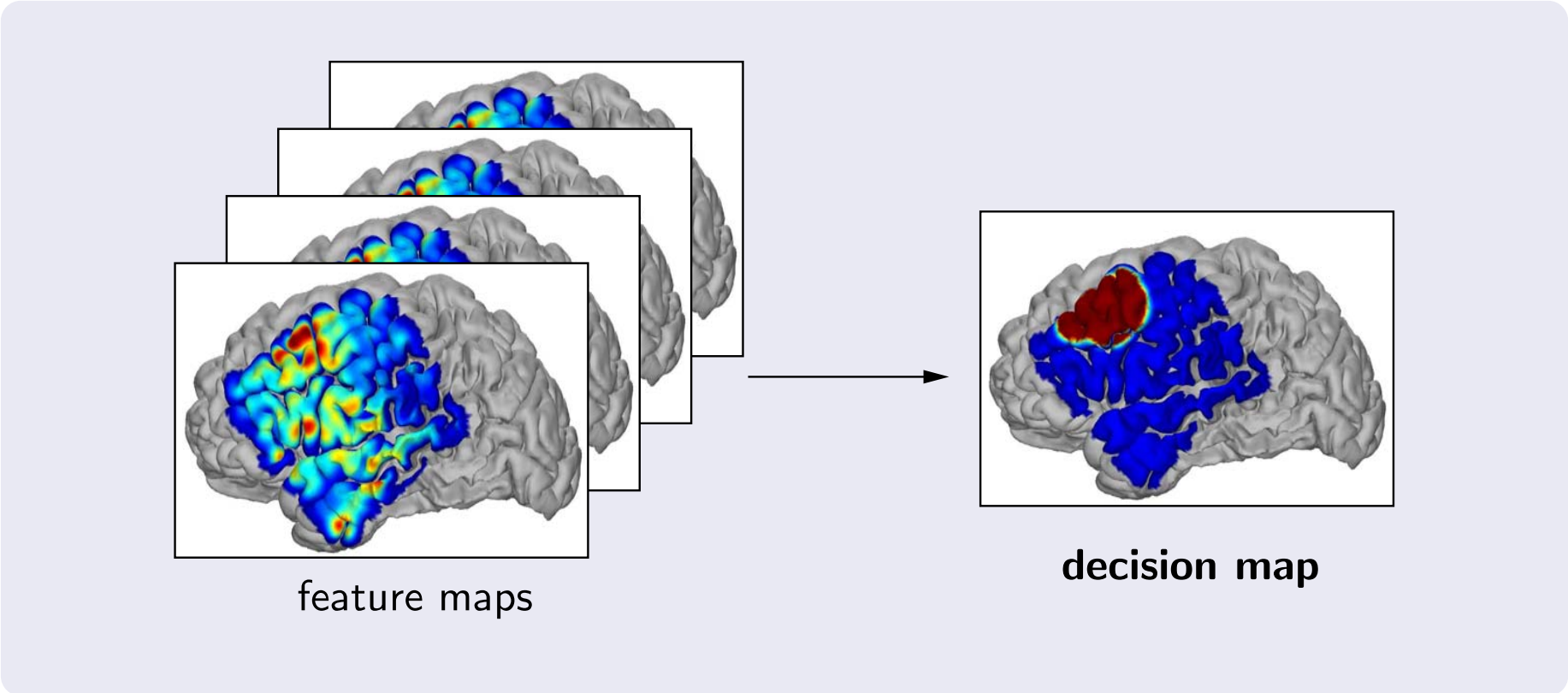
② Do hypersynchronous areas correlate with **seizure onset zone**?

**Yes!** (in **8** of the 11 patients analyzed so far)

③ How to determine **seizure onset zone** using hypersynchrony?

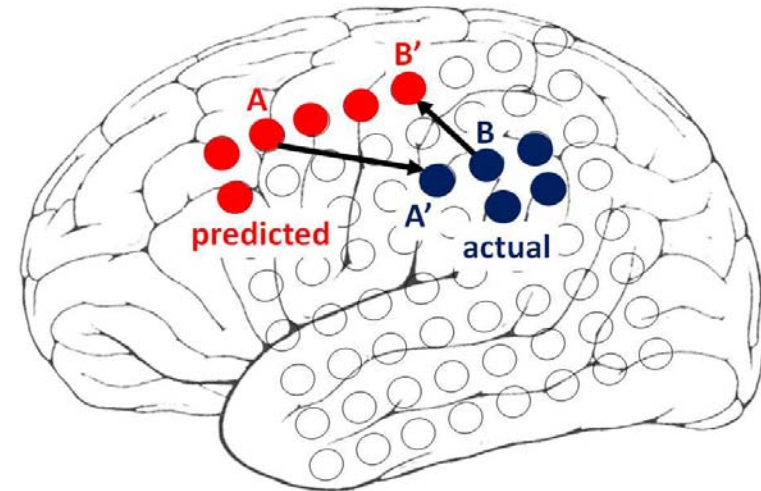
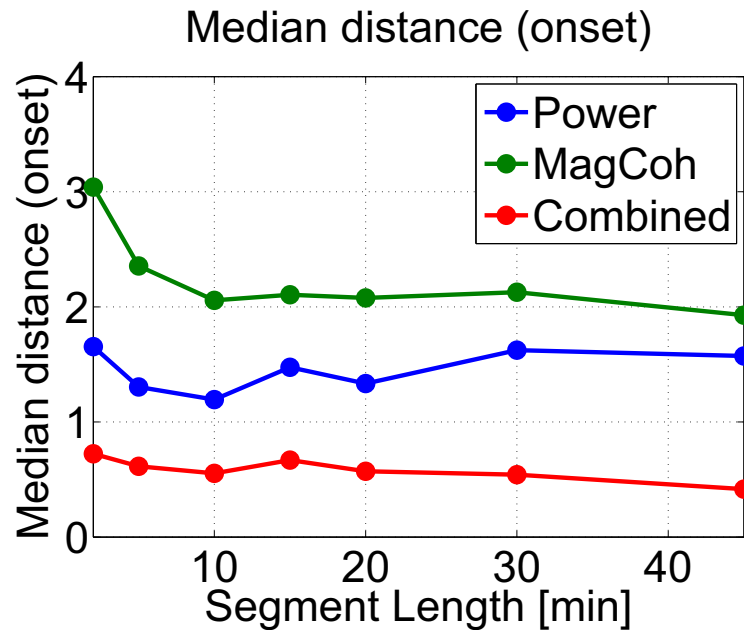
**Statistical decision making, using a graphical model**

# Exploit hypersynchrony to determine seizure onset zone



- Many feature maps  
different frequency bands, different measures
- Noisy!

# Reliable localization of seizure focus



- **Magnitude coherence** in several frequency bands (1–5Hz, 5–20Hz, 20–35Hz, . . . , 80–95Hz)
- **Relative power** within 1–5Hz band
- **Average** over segments of length 2min, 5min, . . . , 45min
- **Short** segment seems to suffice (20 min)

# Summary

- **First-ever** automated procedure to detect seizure focus from interictal recordings
- Exploits hypersynchrony and slowing effect
- Straightforward to include additional features
- Reliable estimates of seizure focus
- **Shorter** hospitalization  
5 days to several weeks → 1 day or less.

# Topics

## Signal Processing Aided Diagnosis

Loss in EEG synchrony → diagnosis of early-Alzheimer's

We needed to develop **novel** measures of synchrony

## Signal Processing Aided Treatment

Increased EEG synchrony → localization of epileptic brain tissue

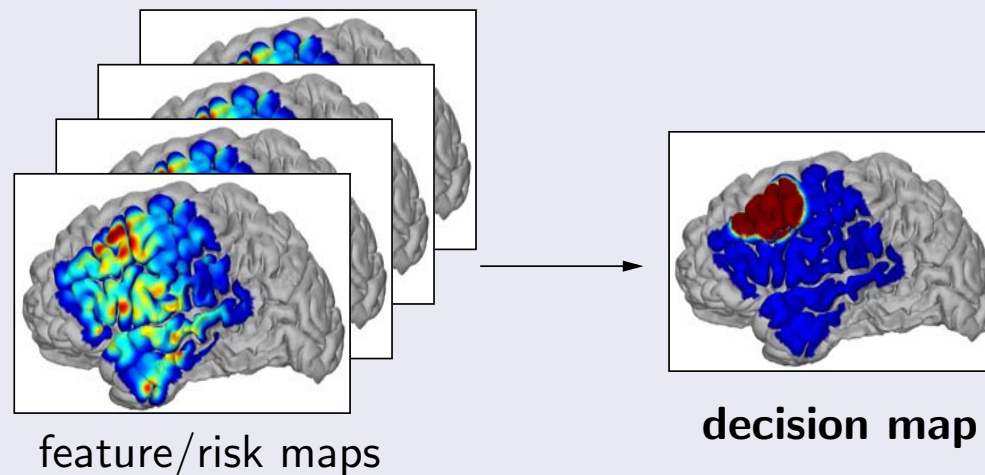
**Existing** synchrony measures proved to be effective

## Future Work

Two future research projects

# Localization of Seizure Focus w/o Relying on Seizures

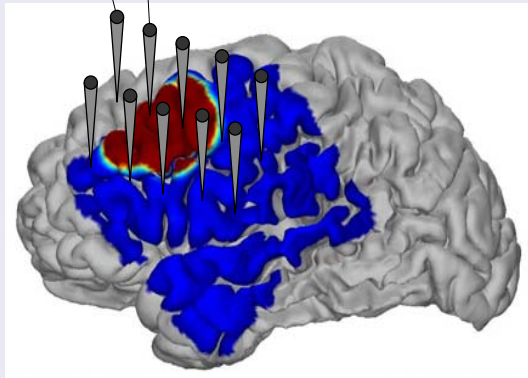
- **Multiple information sources**
  - Brain imaging (EEG, MEG, SPECT, PET, CT, MR, DTI)
  - Clinical behavior
  - Biochemical markers (e.g., neurotransmitters)
  - Genetic data
  - Biophysical models
- **Risk factors**



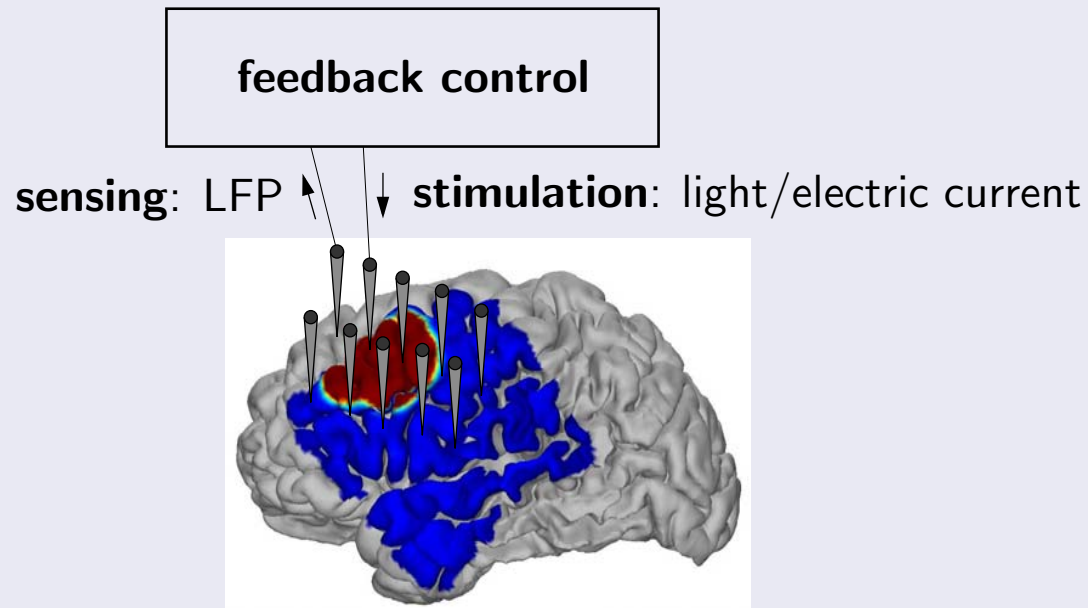
# Brain Implants for Seizure Suppression

feedback control

sensing: LFP ↑      ↓ stimulation: light/electric current

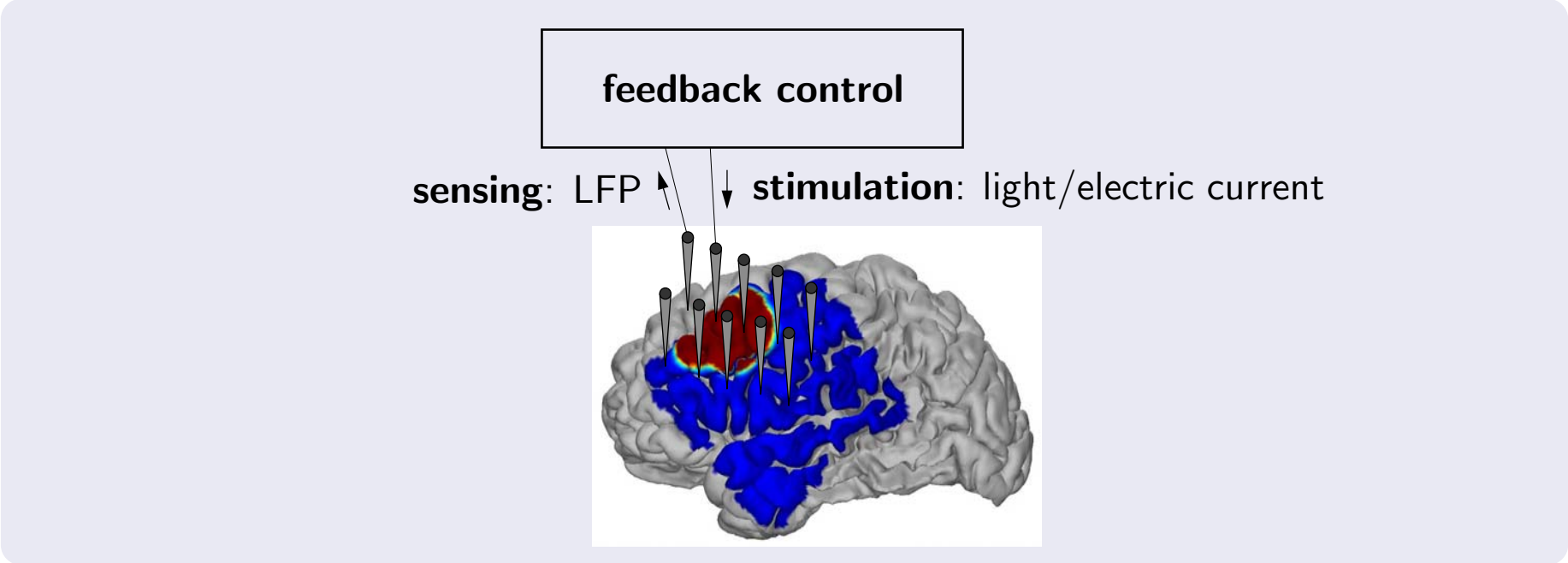


# Brain Implants for Seizure Suppression



- **Electrical** stimulation: treatment of PD, recently also epilepsy
  - **Photonic** stimulation: optogenetics
- Advantages:** better resolution, more specific, less side effects

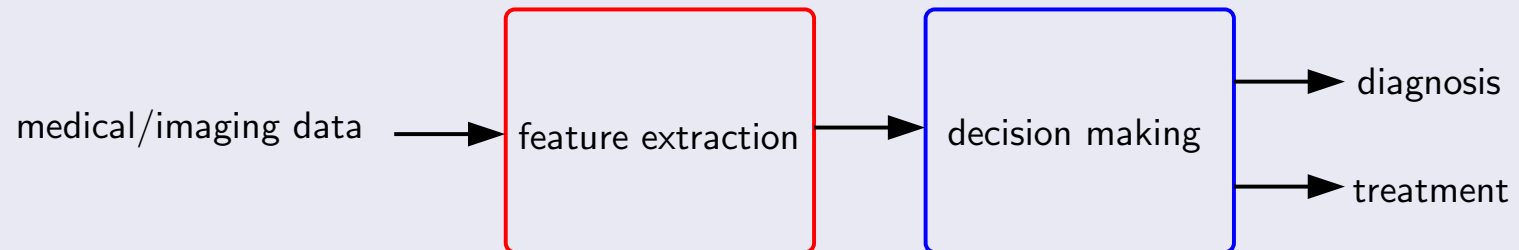
# Brain Implants for Seizure Suppression



- **Electrical** stimulation: treatment of PD, recently also epilepsy
  - **Photonic** stimulation: optogenetics
- Advantages:** better resolution, more specific, less side effects

**How to design feedback control?**

# Back to the big picture . . .



- **Signal processing “microscope”**  
discovery of **features/patterns** in data associated with certain **disorders**
- **Algorithmic decision making**  
combine **evidence** “optimally” → **diagnosis and treatment**

- More **reliable** and **earlier** diagnosis
- More **effective** treatment (clinical outcome, time and cost)

# Acknowledgements

## The SES Team

- François Vialatte (RIKEN)
- Theo Weber (MIT)
- Jordi Solé-Casals (U Vic)
- Andrzej Cichocki (RIKEN)

## Diagnosis of AD

- Toshimitsu Musha (Tokyo Tech)
- Jaeseung Jeong (KAIST)
- Corinna Hänschel  
(MPI&U Frankfurt/U Bangor)
- Johannes Pantel  
(MPI&U Frankfurt)

## SES: Applications beyond AD

- Monique Maurice (RIKEN)
- Tomek Rutkowski (RIKEN)
- Danilo Mandic (Imperial)
- Shin Ishii (Kyoto University)
- Yuichi Sakumura (NAIST)

## Epileptic seizures

- Sydney Cash (MGH/Harvard)
- Emad Eskandar (MGH/Harvard)
- Ed Boyden, J.J. Slotine (MIT)
- AI Willsky, Emery Brown (MIT)
- ...

# Analogy: Waiting for a train



- Train may **not** arrive (e.g., mechanical problem)  
= **event reliability**  $\rho$
- Train may or may not be on **time**  
= **timing precision**  $s_t$