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# Possible Projects

## Barnabás Póczos

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UofA, Edmonton

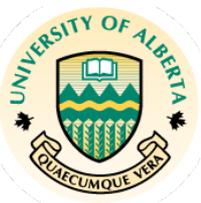
Sept 21, 2009



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# Blind Subspace Deconvolution

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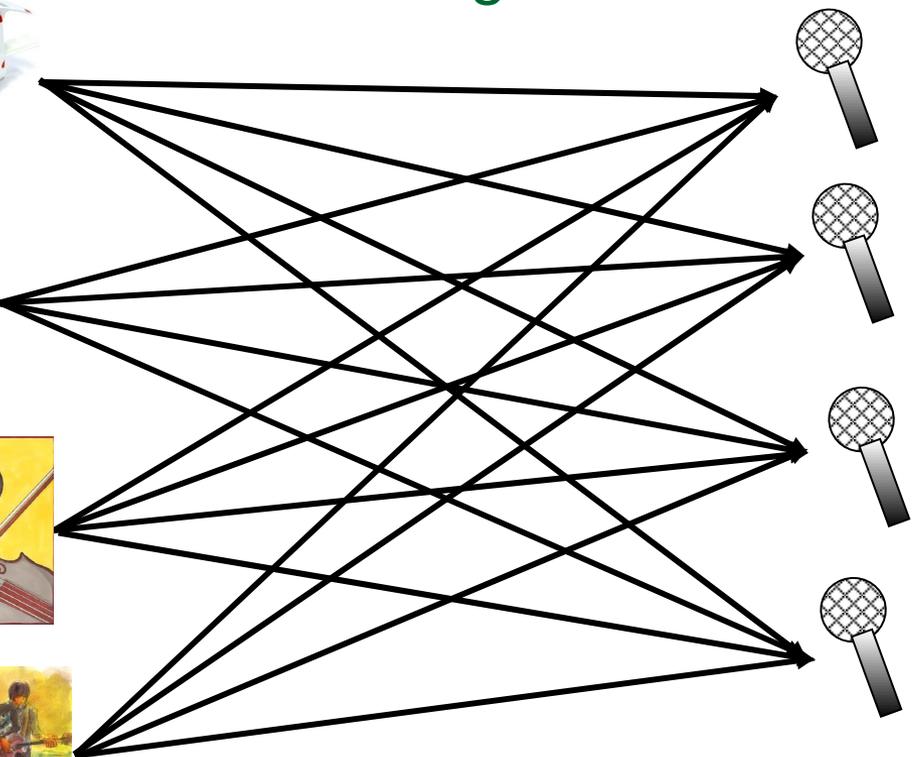
# Independent Component Analysis

Sources

Mixing

Observation

Estimation



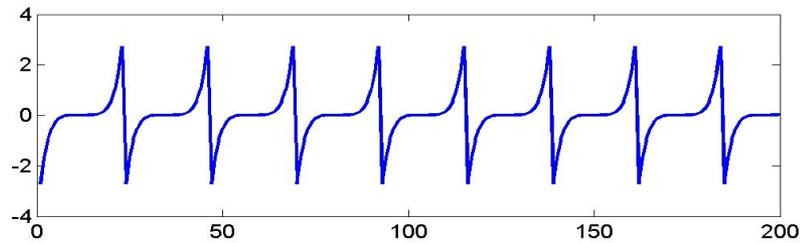
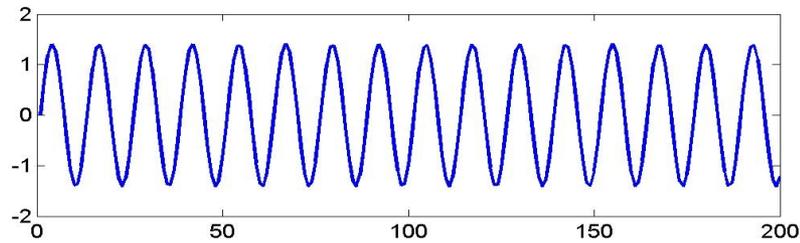
$$y(t) = Wx(t)$$

$A$

$$x(t) = As(t)$$

$s(t)$

# Independent Component Analysis

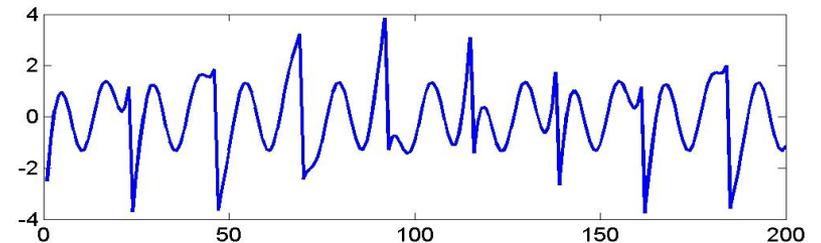
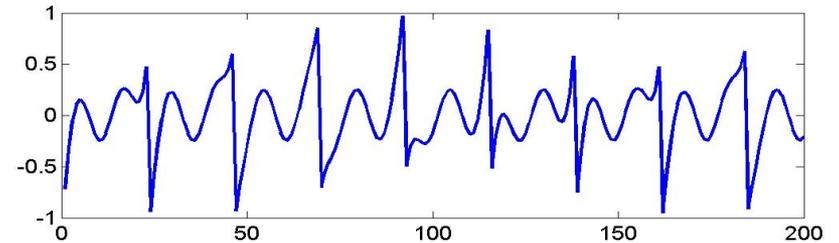


Independent signals

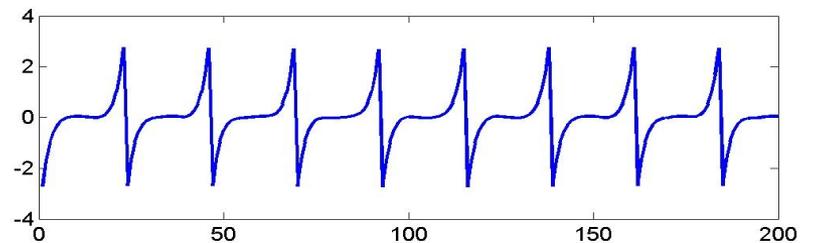
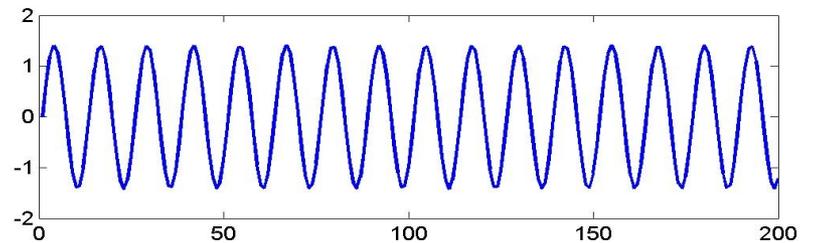
$$x_1(t) = a_{11}s_1(t) + a_{12}s_2(t)$$

$$x_2(t) = a_{21}s_1(t) + a_{22}s_2(t)$$

$$WA = \begin{bmatrix} \text{white} & \text{black} \\ \text{black} & \text{white} \end{bmatrix}$$



Mixtures



ICA estimation

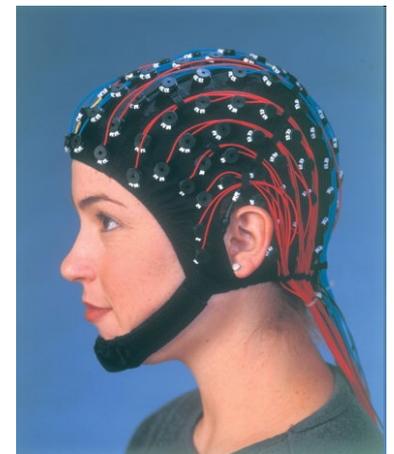
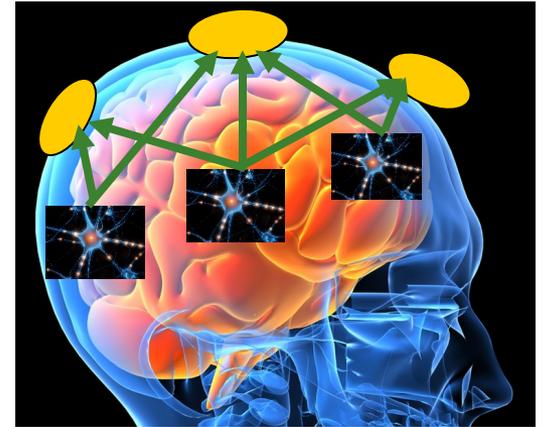
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# Some ICA Applications

- Medical signal processing – fMRI, ECG, EEG
  - Brain Computer Interfaces
  - Image denoising
  - Modeling of the hippocampus, place cells
  - Modeling of the visual cortex
  - Microarray data processing
  - Decomposing the spectra of galaxies
  - Blind deconvolution
  - Feature extraction
  - Face recognition
  - Time series analysis
  - Financial applications
  - Clustering
  - Classification
-

# ICA Applications, Removing Artifacts from EEG

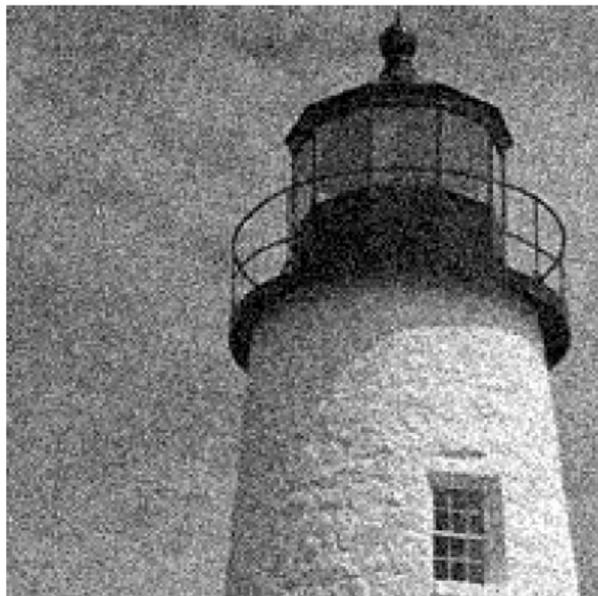
- EEG ~ *Neural cocktail party*
- Severe **contamination** of EEG activity by
  - eye movements,
  - blinks,
  - muscle,
  - heart, ECG artifact
  - vessel pulse
  - electrode noise
  - line noise, alternating current (60 Hz)
- ICA can effectively **detect, separate and remove** activity in EEG records from a wide variety of artifactual sources.  
(Jung, Makeig, Bell, and Sejnowski)



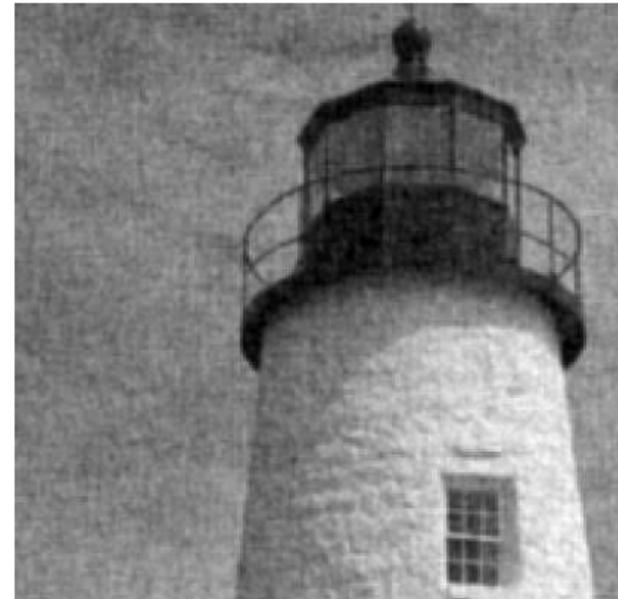
# ICA for Image Denoising (Hoyer, Hyvarinen)



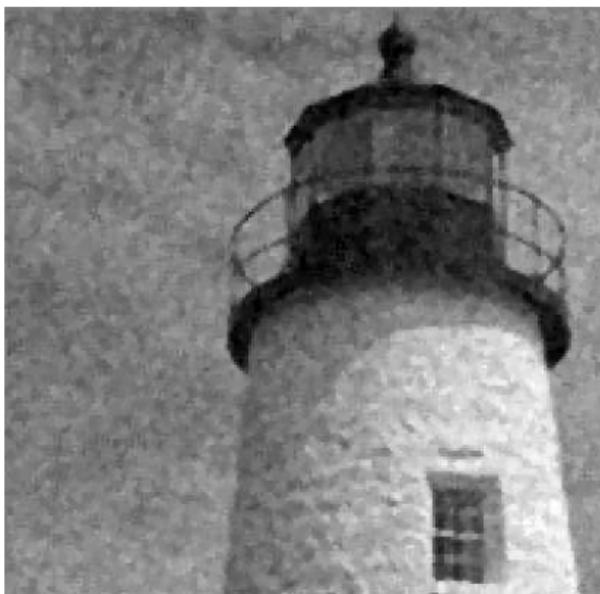
original



noisy



Wiener filtered

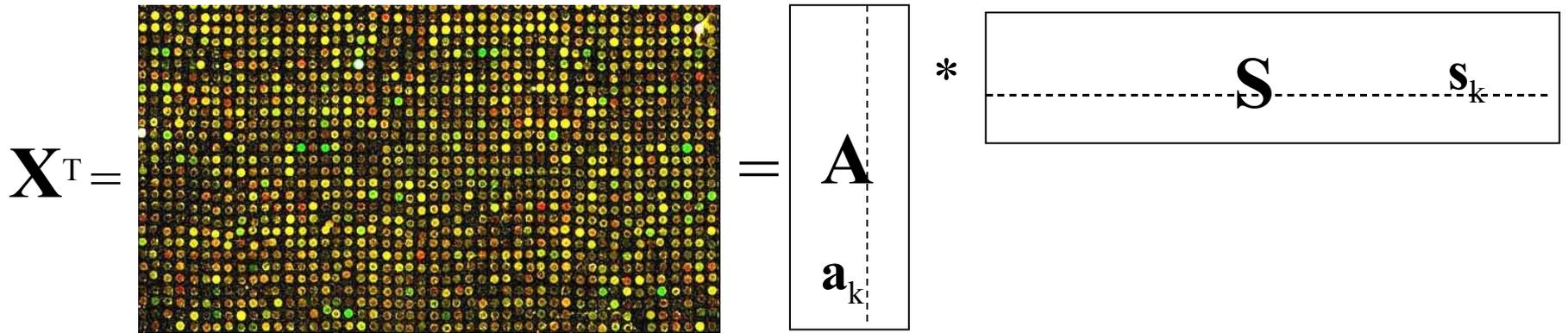


median filtered

ICA denoised



# ICA for Microarray data processing



$$\mathbf{X}^T \in \mathbb{R}^{M \times N}$$

$M$  = number of experiments

$N$  = number of genes

## Assumption:

- each experiment is a mixture of **independent expression modes** ( $\mathbf{s}_1, \dots, \mathbf{s}_K$ ).
- some of these modes (e.g.  $\mathbf{s}_k$ ) can be related to the difference between the classes.
- $\rightarrow \mathbf{a}_k$  correlates with the class labels

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# Generalizations

- **Blind Deconvolution**

- $x(t) = A_1 s(t-1) + A_2 s(t-2) + \dots + A_k s(t-k)$
- After FFT we have to do ICA in the frequency space

- **Independent Subspace Analysis**

- ***Blind Subspace Deconvolution using FFT???***

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# Manifold learning using time processes

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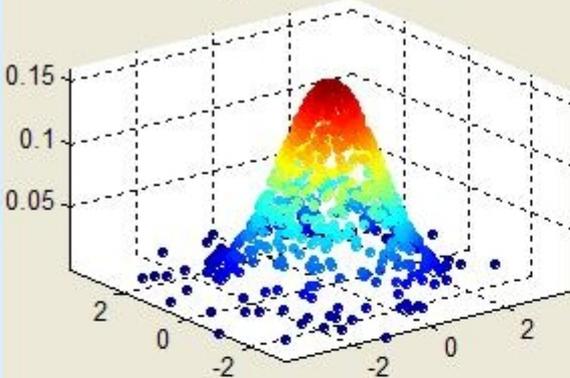


# Manifold learning

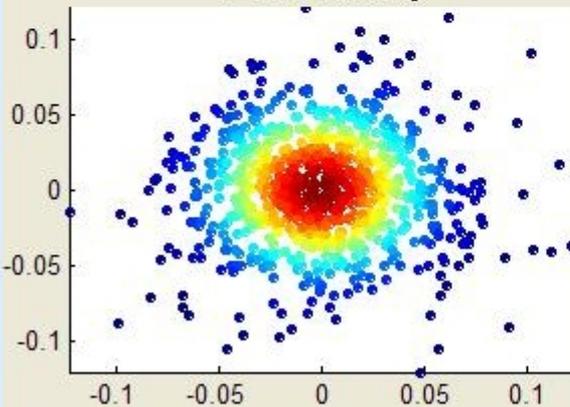
<http://www.math.umn.edu/~wittman/mani/>

<Student Version> : mani

**Original Manifold**



**Embedding**



Running LTSA.  
LTSA complete: 1.031s  
Embedding data written to matrix "maniY"

**Manifold**

Matrix:

File name:

Color Vector:

Examples:

# Points =  Sigma =

**Parameters**

Target Dimension  $d =$

Nearest Neighbors  $K =$

Sigma =  Alpha =

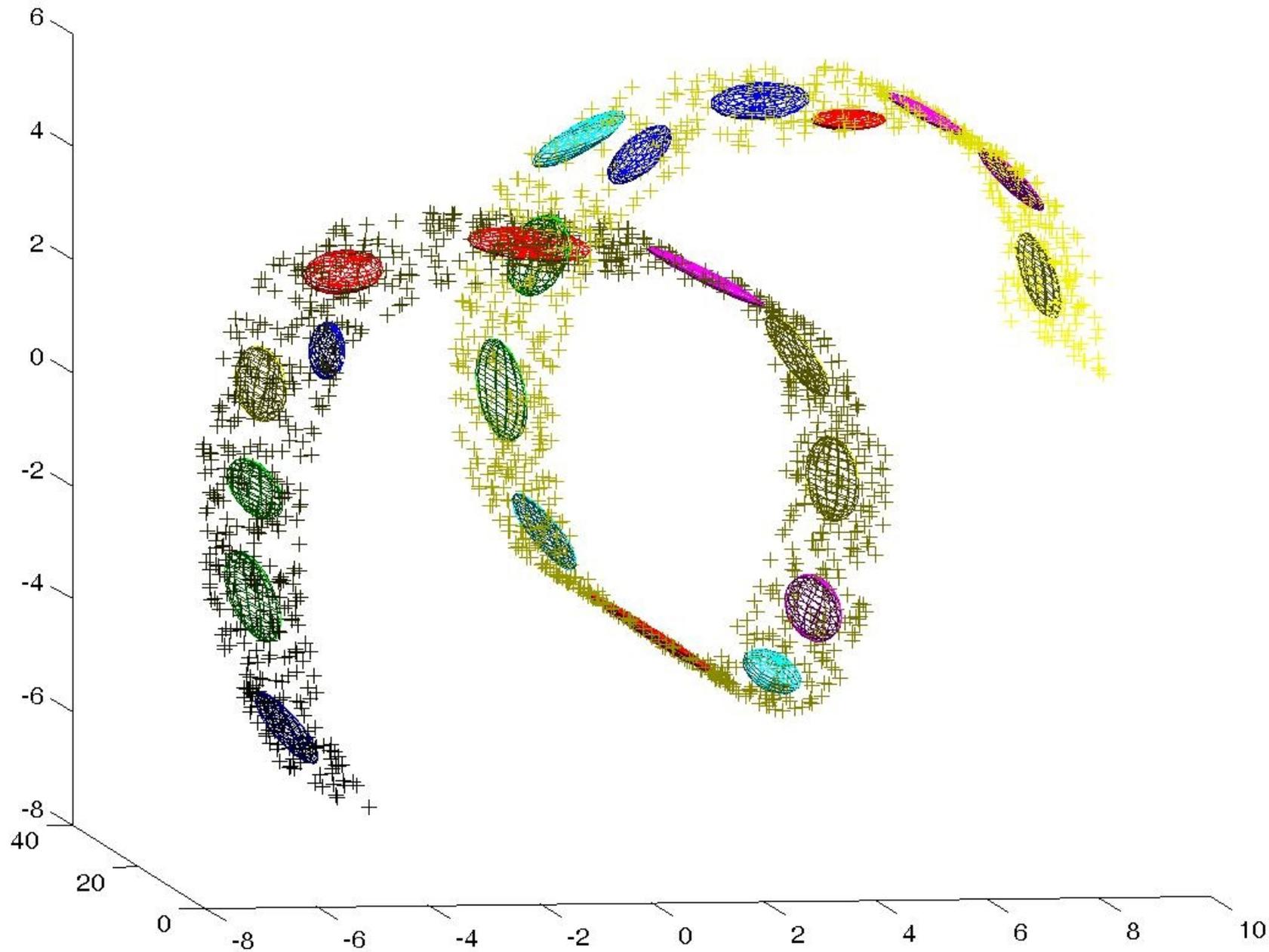
**Algorithms**

MDS	PCA	ISOMAP
LLE	Hessian LLE	Laplacian
Diffusion Map	LTSA	Run All 8

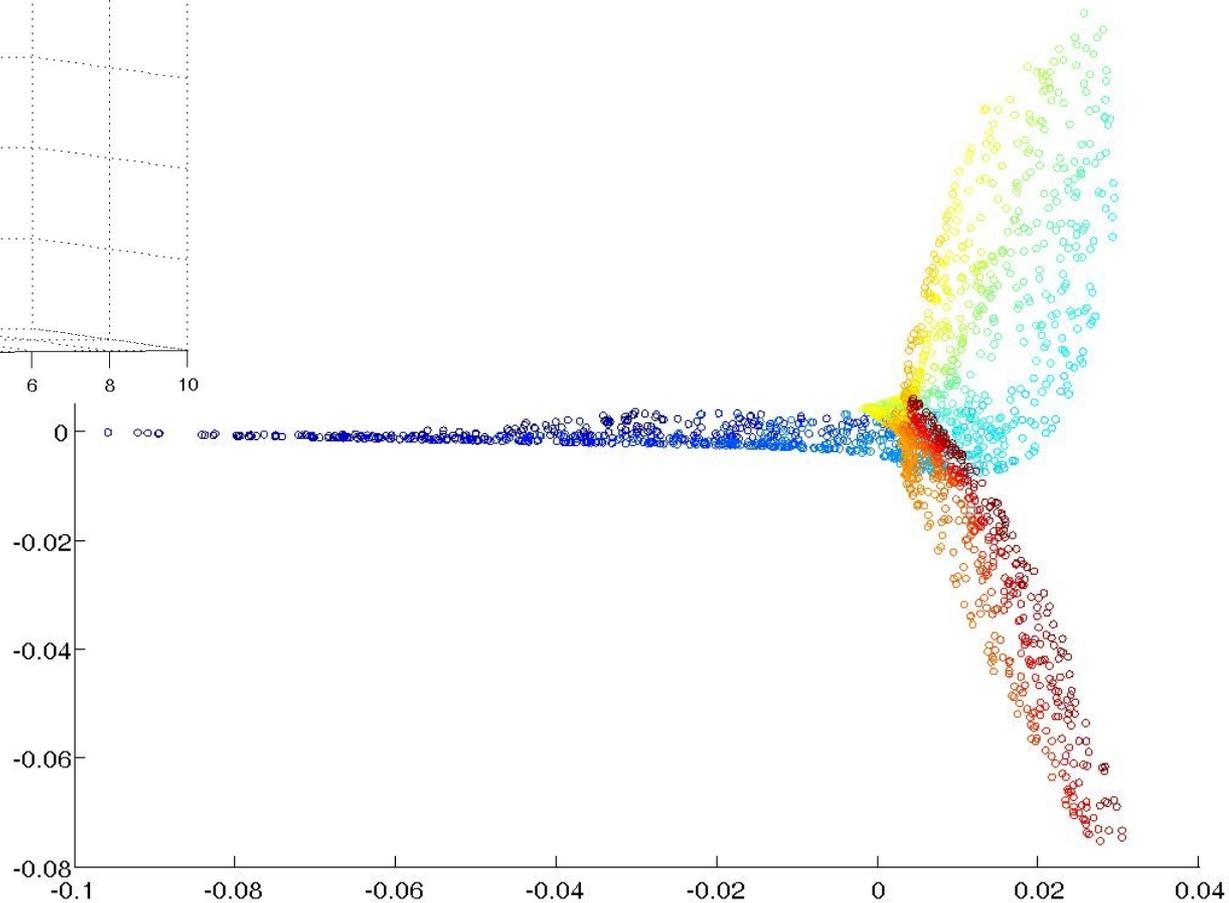
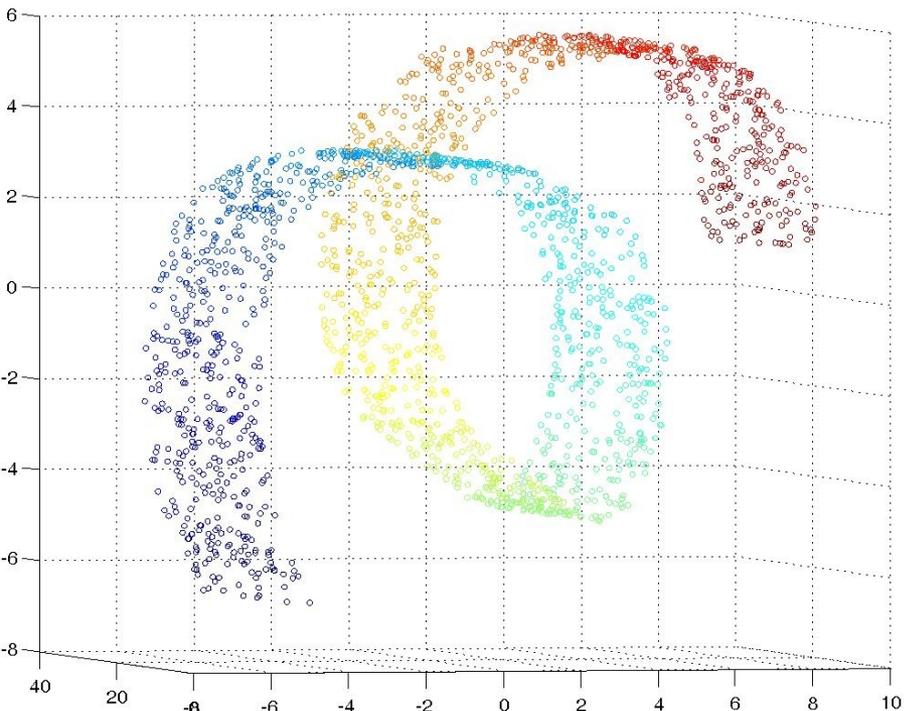
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# Manifold learning

- The algorithms usually assume iid sample.
  - ***Can we exploit time correlations???***
    - E.g. using Hidden Markov Models in the Variational Bayesian Mixture Factor Analyzer model (vbMFA, Beal & Ghahramani)...
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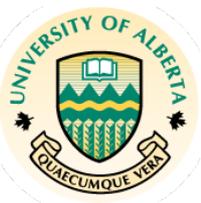
# LTSA Failure



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# Thanks for the attention!

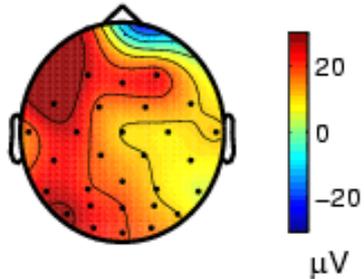
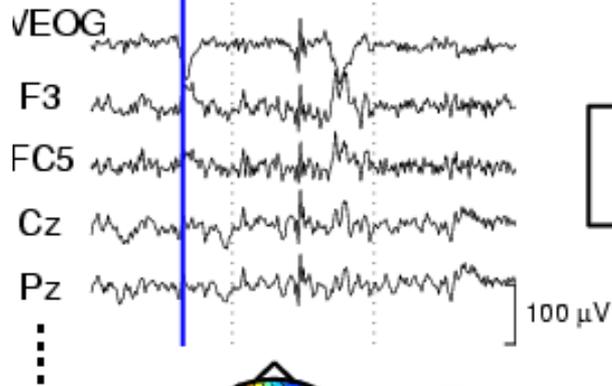
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# ICA decomposition

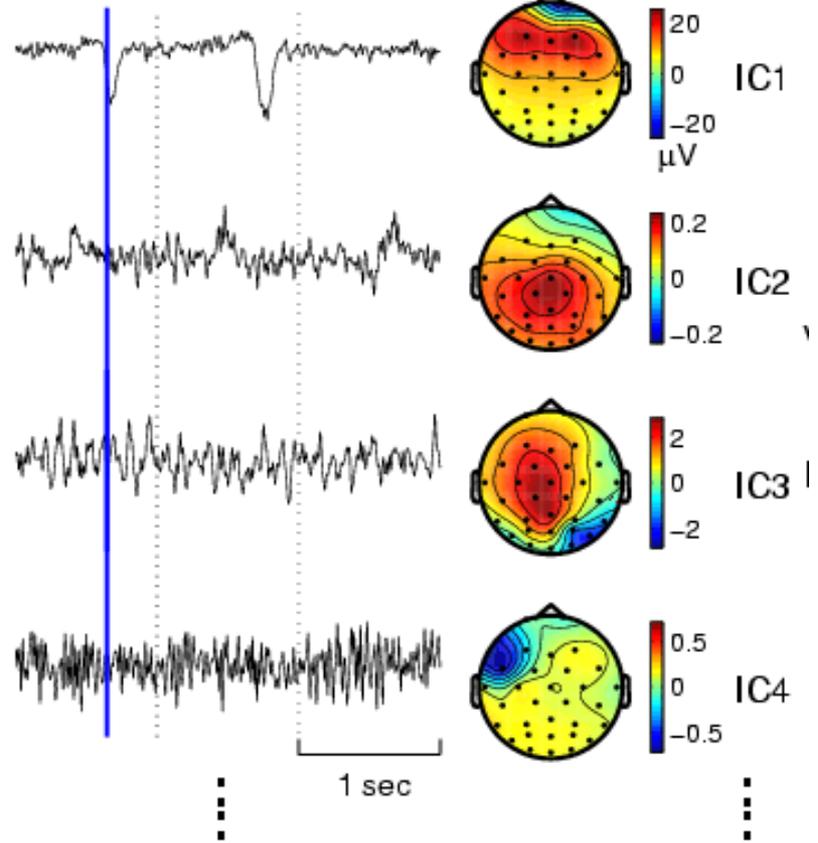


EEG Scalp Channels



unmixing  
( $W$ )

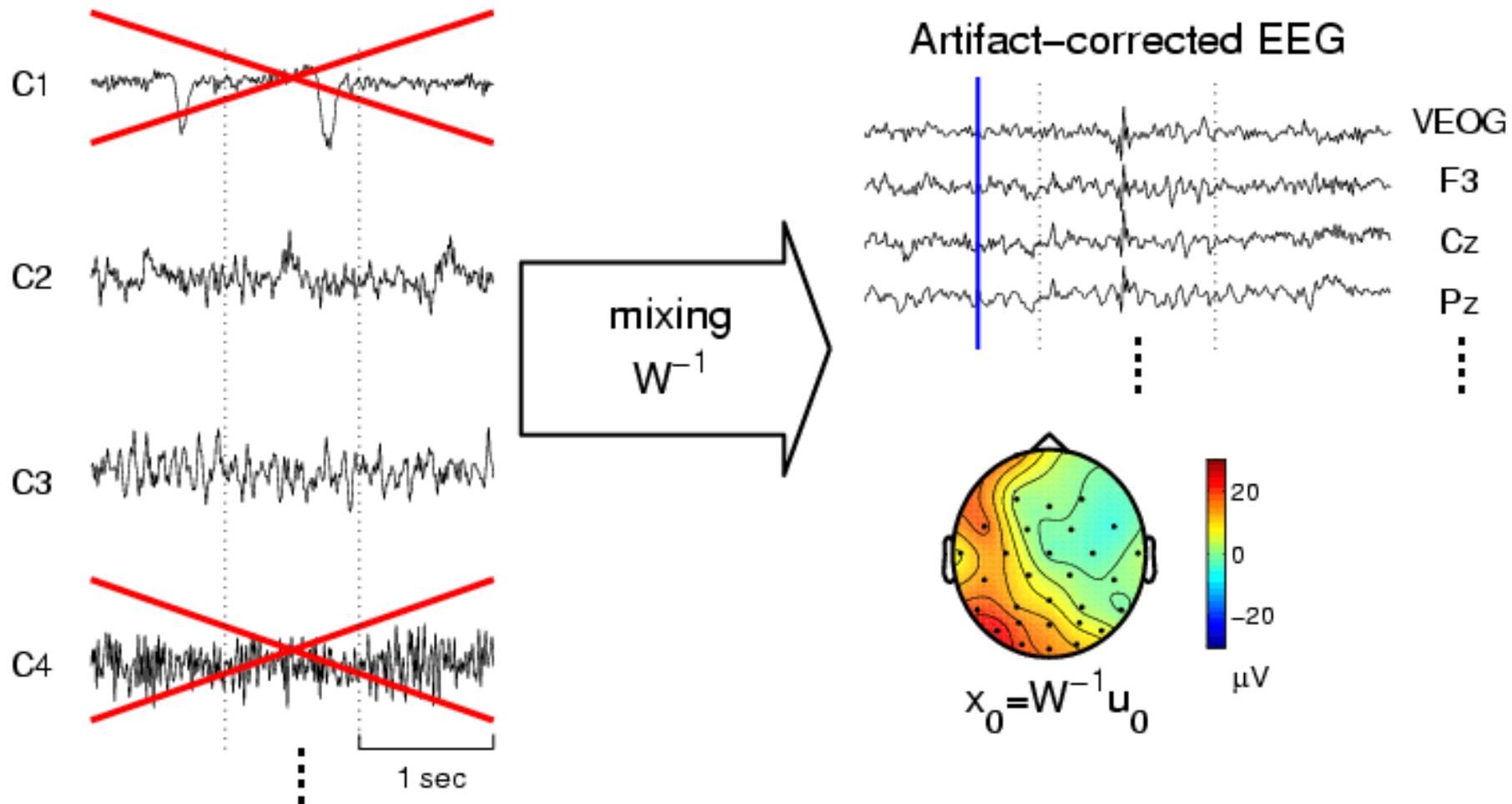
## Independent Components



activations  
( $u=WX$ )

scalp maps  
( $W^{-1}$ )

# Summed Projection of Selected Components



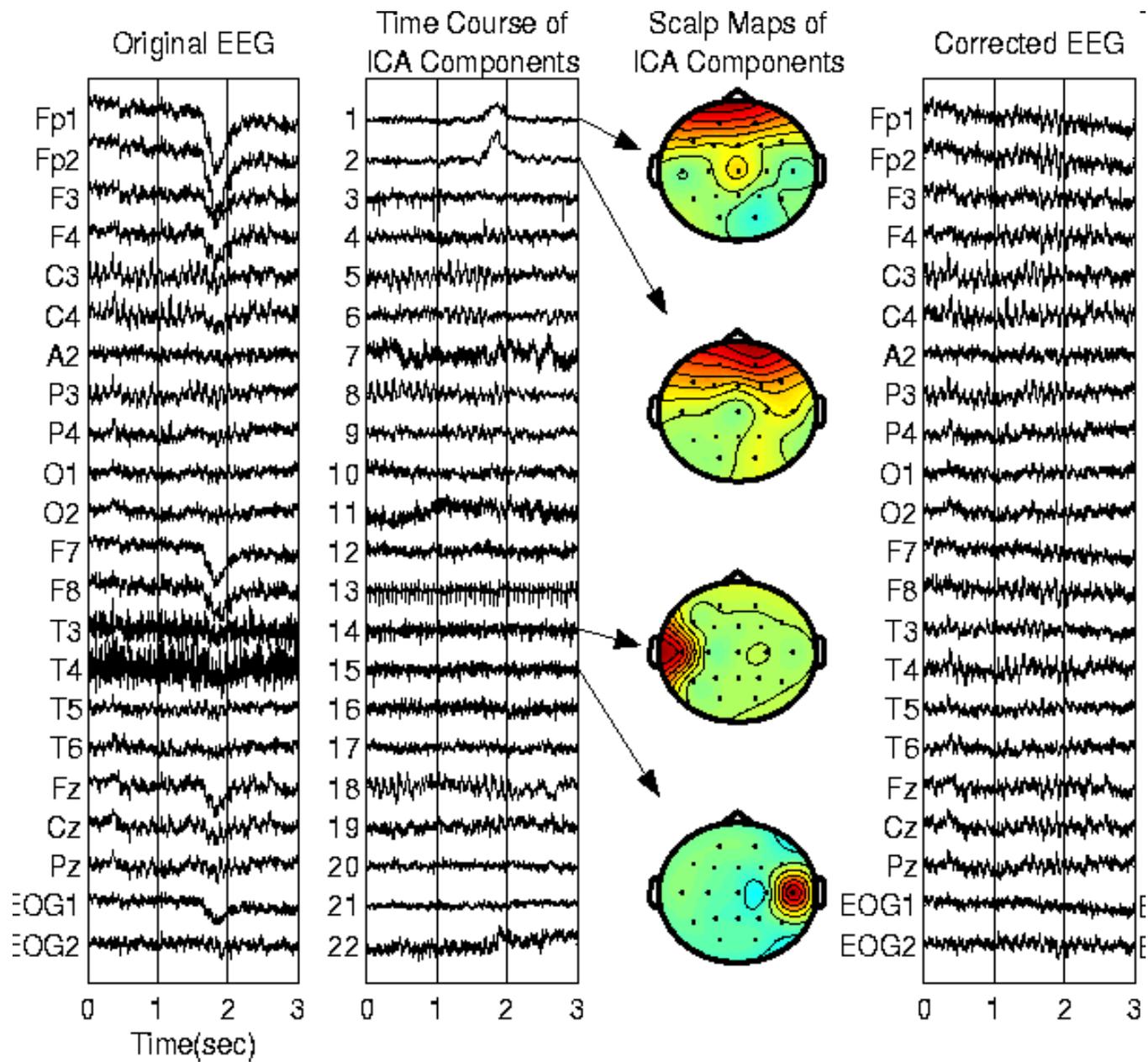


Fig from Jung

# ICA for Microarray data processing

(Schachtner et al, ICA07)

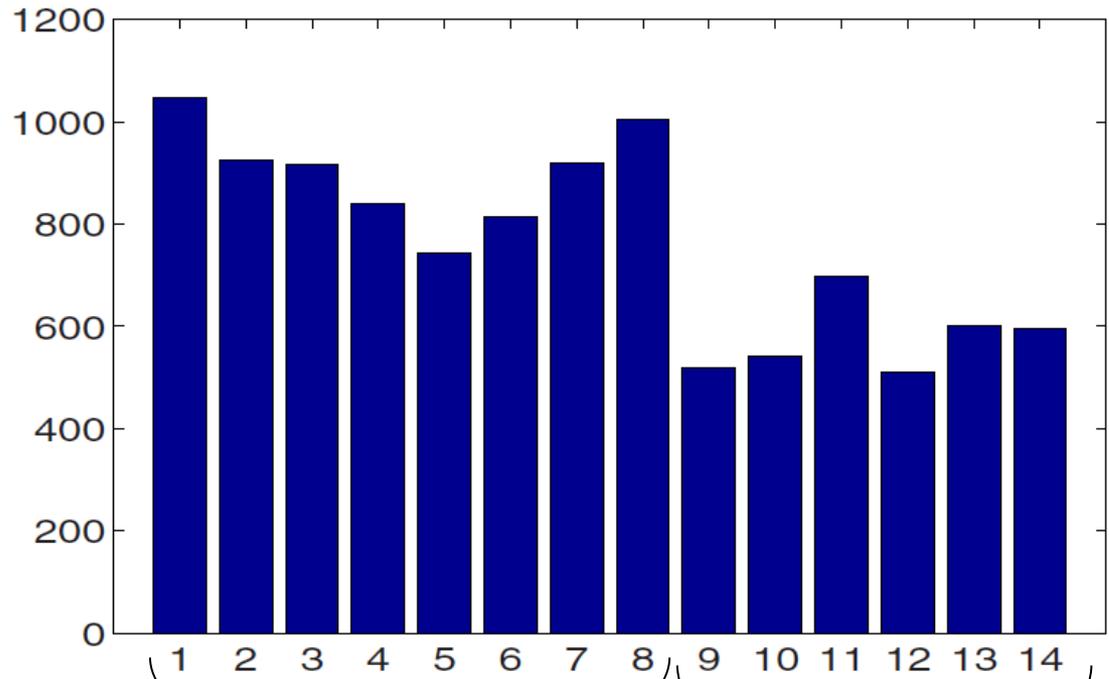
## Brest Cancer Data set

M=14 Experiments

N=22283 genes

2 classes

9<sup>th</sup> column of **A**:



$|\text{Corr}(\mathbf{a}_9, \mathbf{d})|=0.89$ , where  $\mathbf{d}$  is the vector of class labels

Class 1,  
weak metastasis

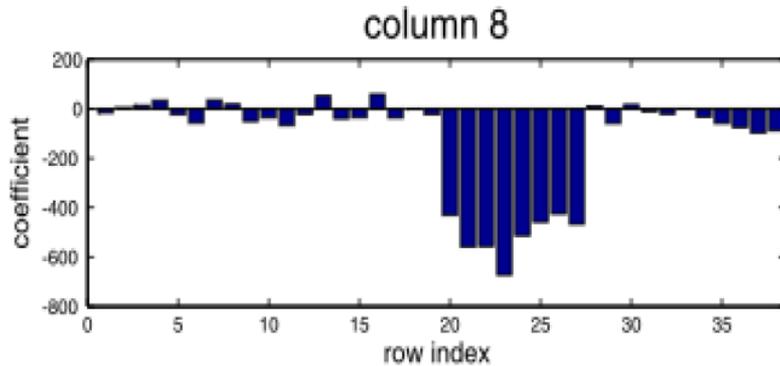
Class 2,  
strong metastasis

# ICA for Microarray data processing (Schachtner et al, ICA07)

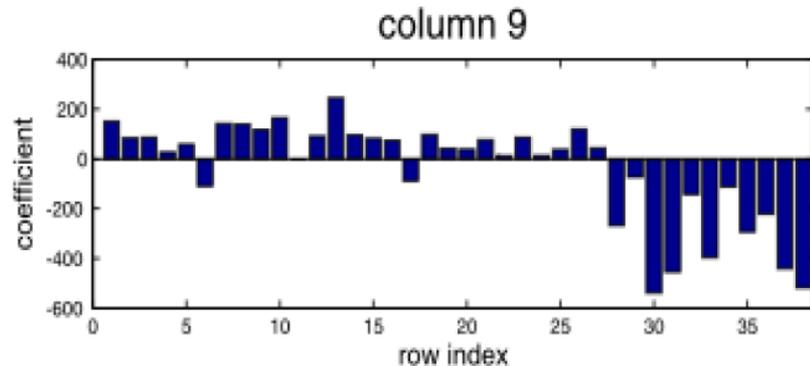
**Leukemia Data set** M=38 Experiments

N=5000 genes

3 classes: ALL-B, ALL-T, AML



ALL-B ALL-T AML



ALL-B ALL-T AML