

fMRI Analysis Techniques

NIAG

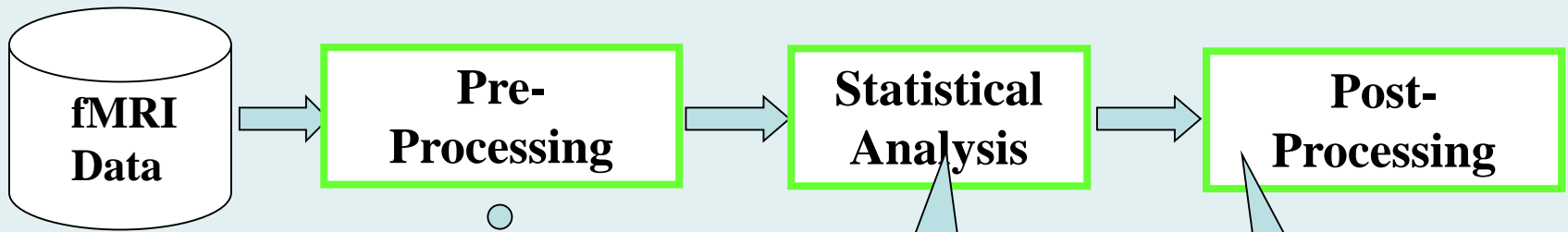
Tehran University of Medical Science

Outline

- An overview on the whole procedure of data analysis in fMRI
- General fMRI Preprocessing steps
- statistical analysis Methods (GLM & ICA)



Overview



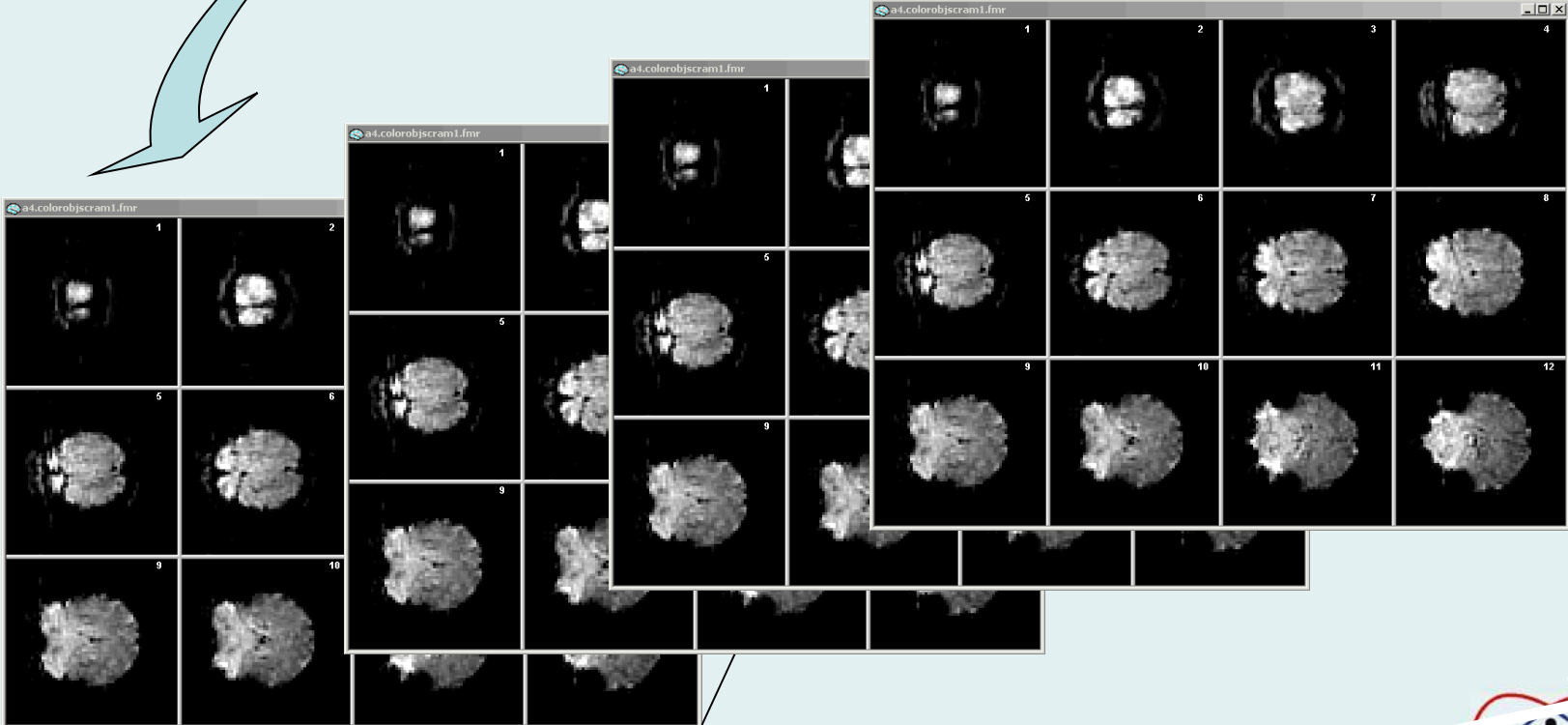
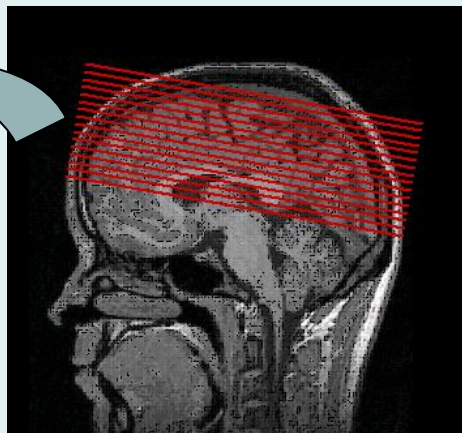
Correction of non-task related variability

Deriving the statistical map of activity
Parametric (Model Based)
Non-parametric (Exploratory)

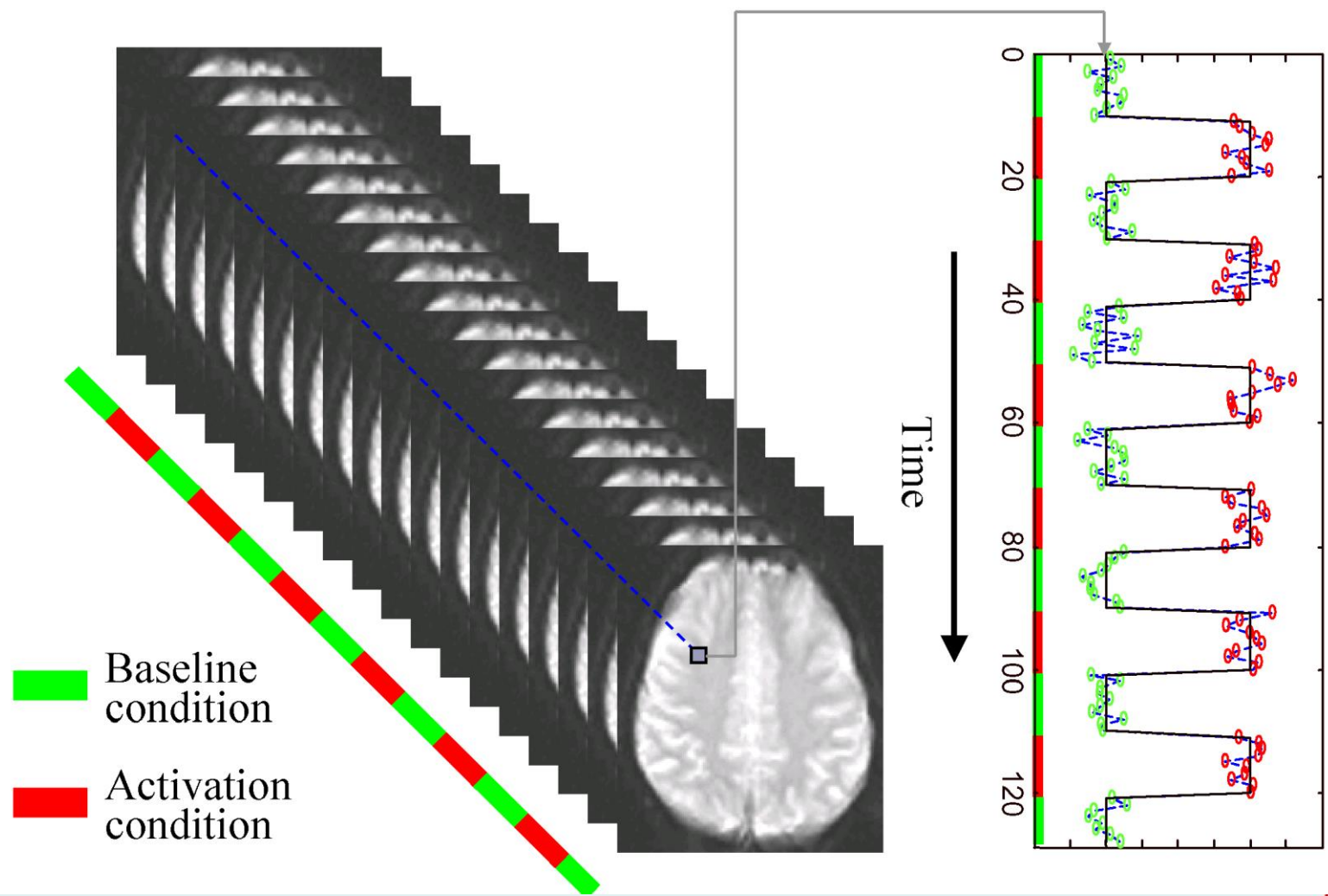
Registration to standard Atlas
Visualization
Interpretation



How is a fMRI Data?



fMRI time series



■ Baseline condition
■ Activation condition



fMRI Image Preprocessing



What is preprocessing?

Correcting for non-task-related variability in experimental data

- Usually done without consideration of experimental design; thus, pre-analysis



Preprocessing Steps

- Data Transfer, and format Conversion
- Removal of first few volumes
- Motion Correction
- Slice Timing Correction
- Spatial Smoothing
- Filtering (Spatial & Temporal)
- Global intensity normalization



Data Format exchange

- **Conventional formats in different MRI system:**
 - ANALYZE
 - ❖ A file with extension *.hdr containing header information
 - ❖ A file with extension *.img containing image raw data
 - NIFTI
 - ❖ Usually a *.nii file containing 4D information
 - AFNI
 - ❖ Header information in *.HEAD
 - ❖ Image raw data in *.BRIK



Removal of first few volumes

- Usually image volumes corresponding to first 10-12 seconds are removed from the sequence
- This is to make sure that spin system is in the steady state condition in acquired images
- Thus unwanted changes in image weight are removed

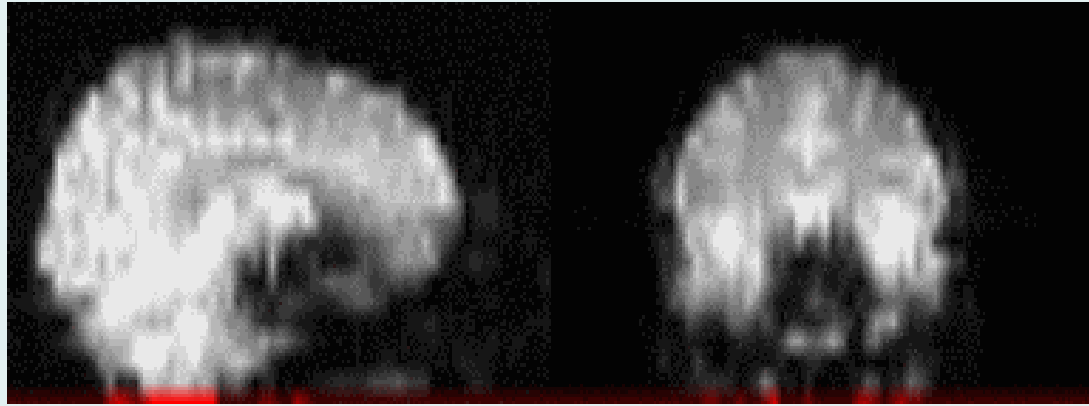


Head Motion Correction:

- Even with padding around the
- head people move in the scanner
- Motion correction realigns all images to a common reference
- 3-12 parameters: 3 translation, 3 rotation, Skewness
- Minimization of some cost function



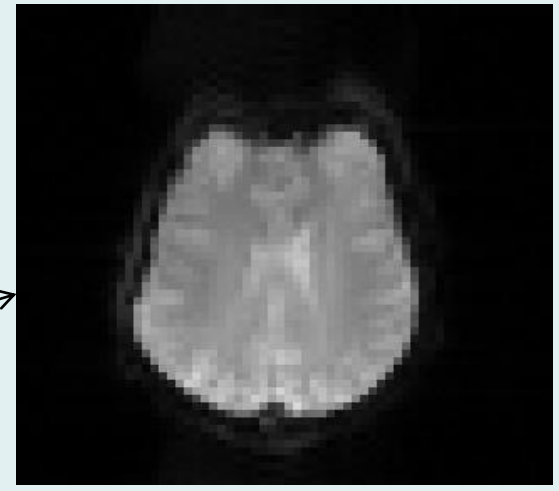
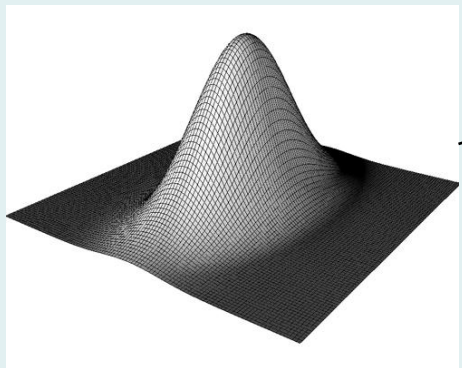
Slice Timing Correction



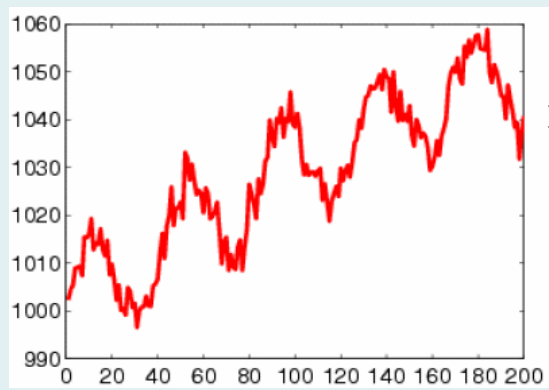
- Each slice is scanned at a slightly different time
- Slice timing correction shifts the *data* (each voxel's time series) as if whole volume was acquired at exactly the same time

Spatial & Temporal Filtering:

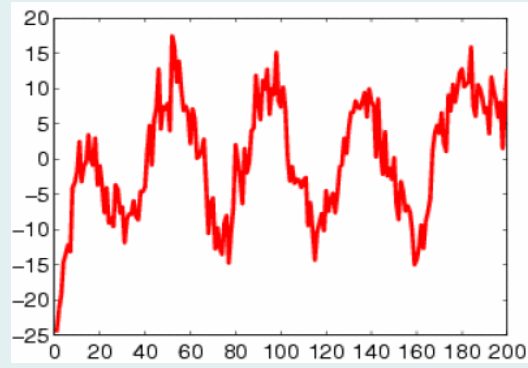
Spatial Filtering



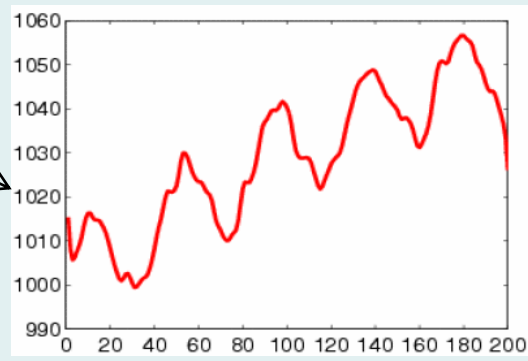
Temporal filtering



HP



LP



Global intensity normalization

Why:

- variation in the intensity between different volumes
- How: compute 4D average and make it the same across subjects/sessions



Statistical Analysis: General Linear Model



Outline

Aim:

- detecting the status of each voxel (active rest)?

Method:

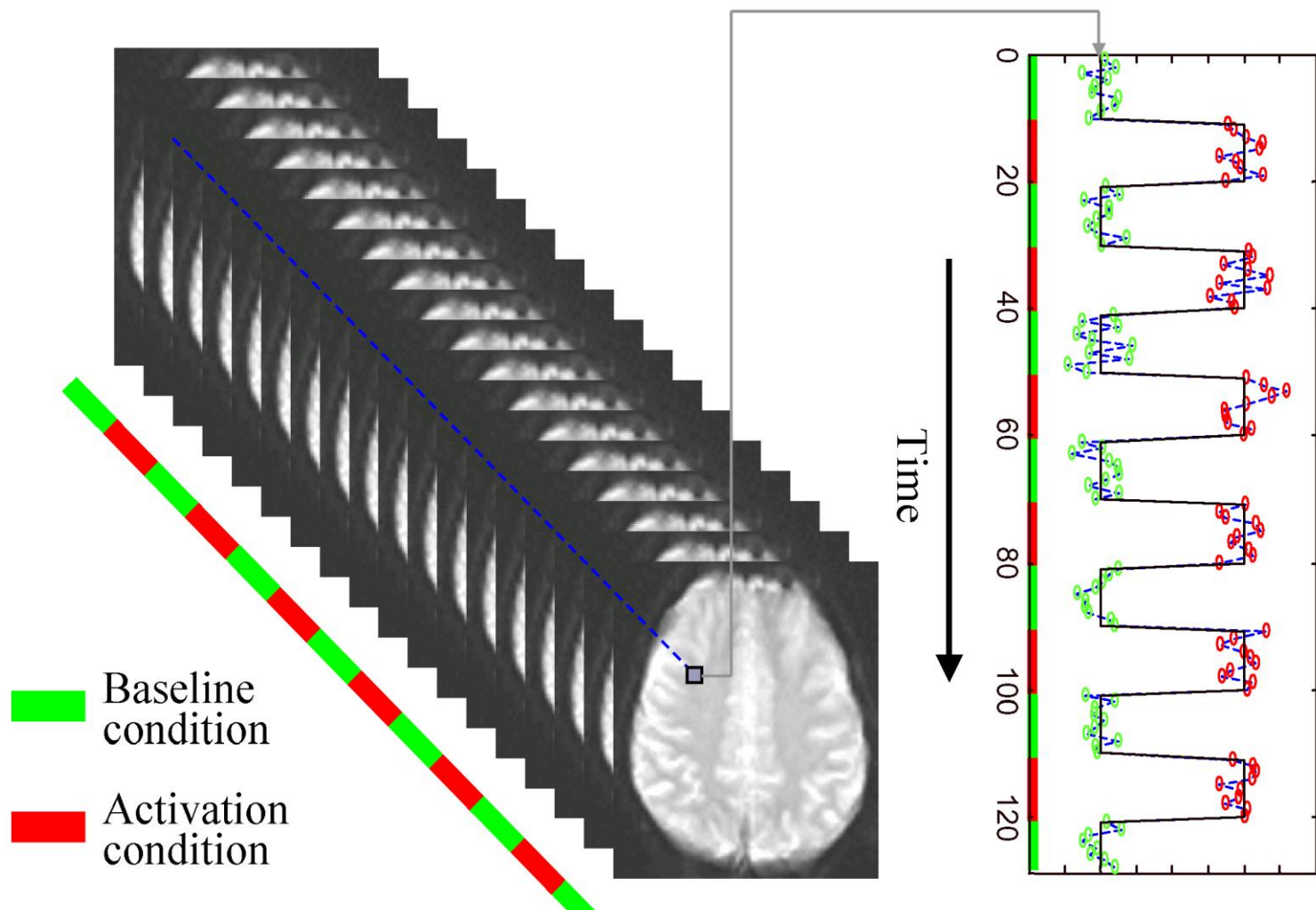
- General Linear Model (GLM)

Statistical Inference:

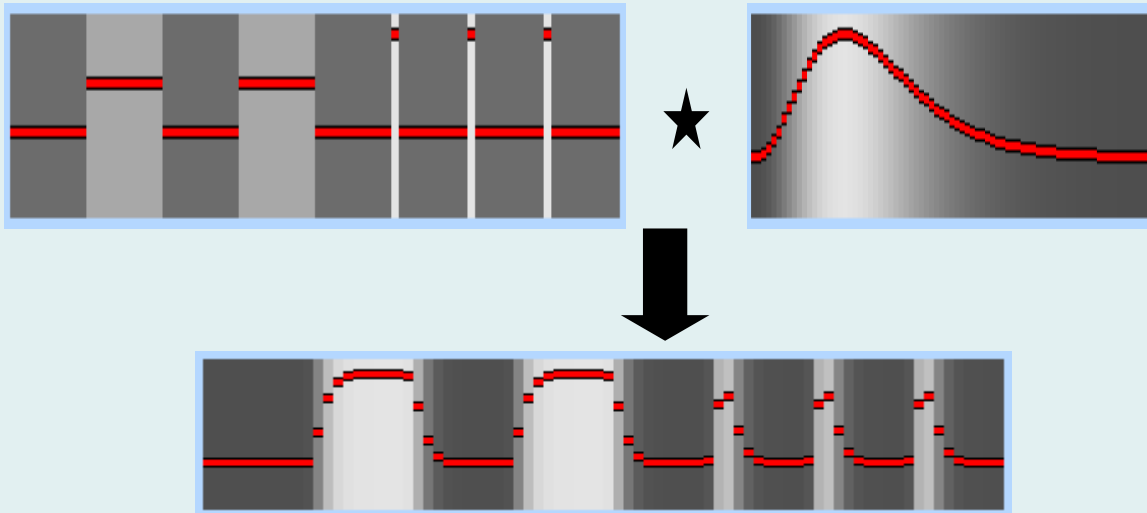
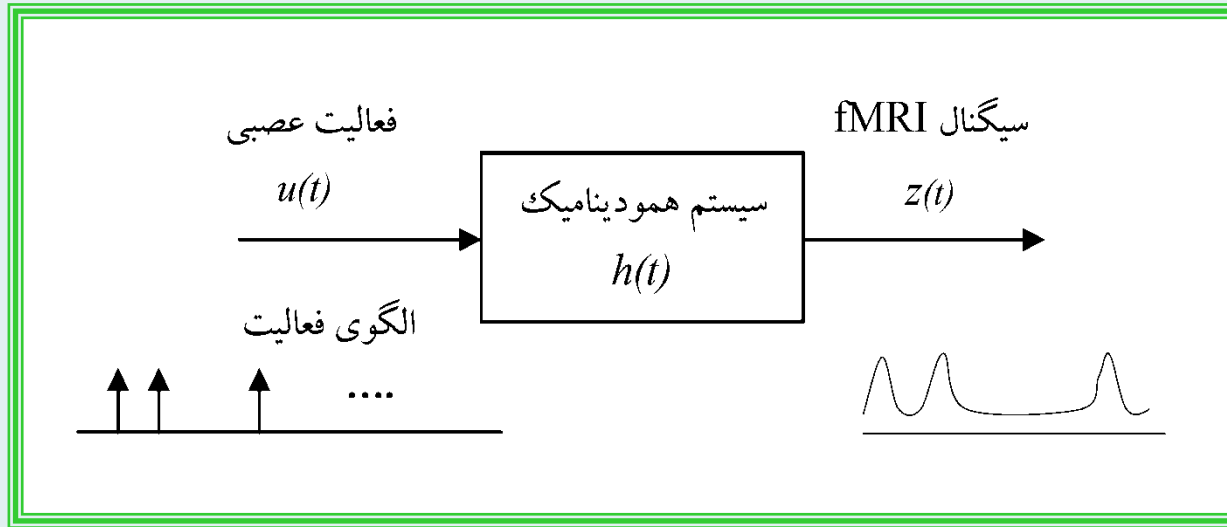
- Finding the Brain Activity and thresholding



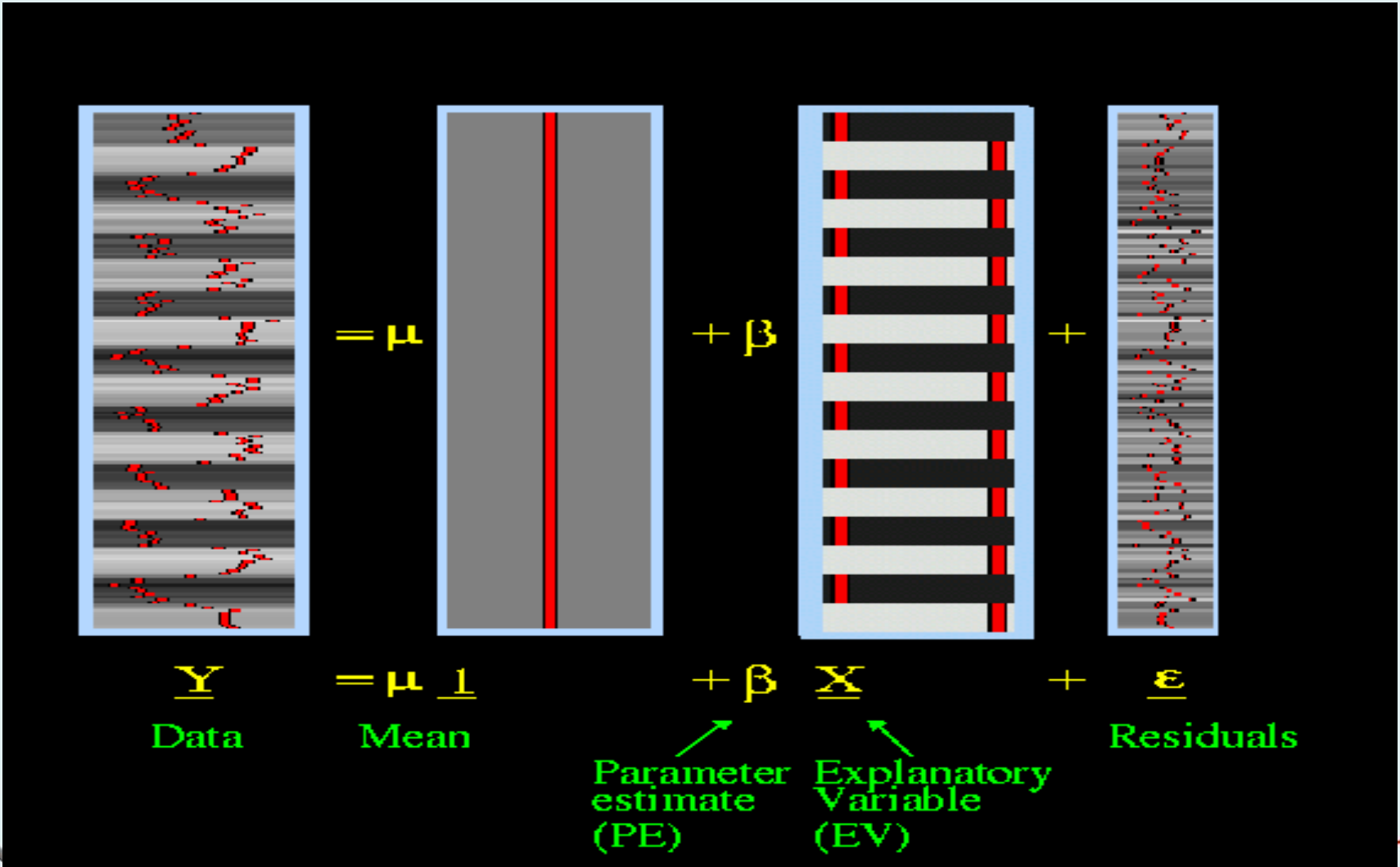
Basic rule for detecting the status of each voxel (active rest)?



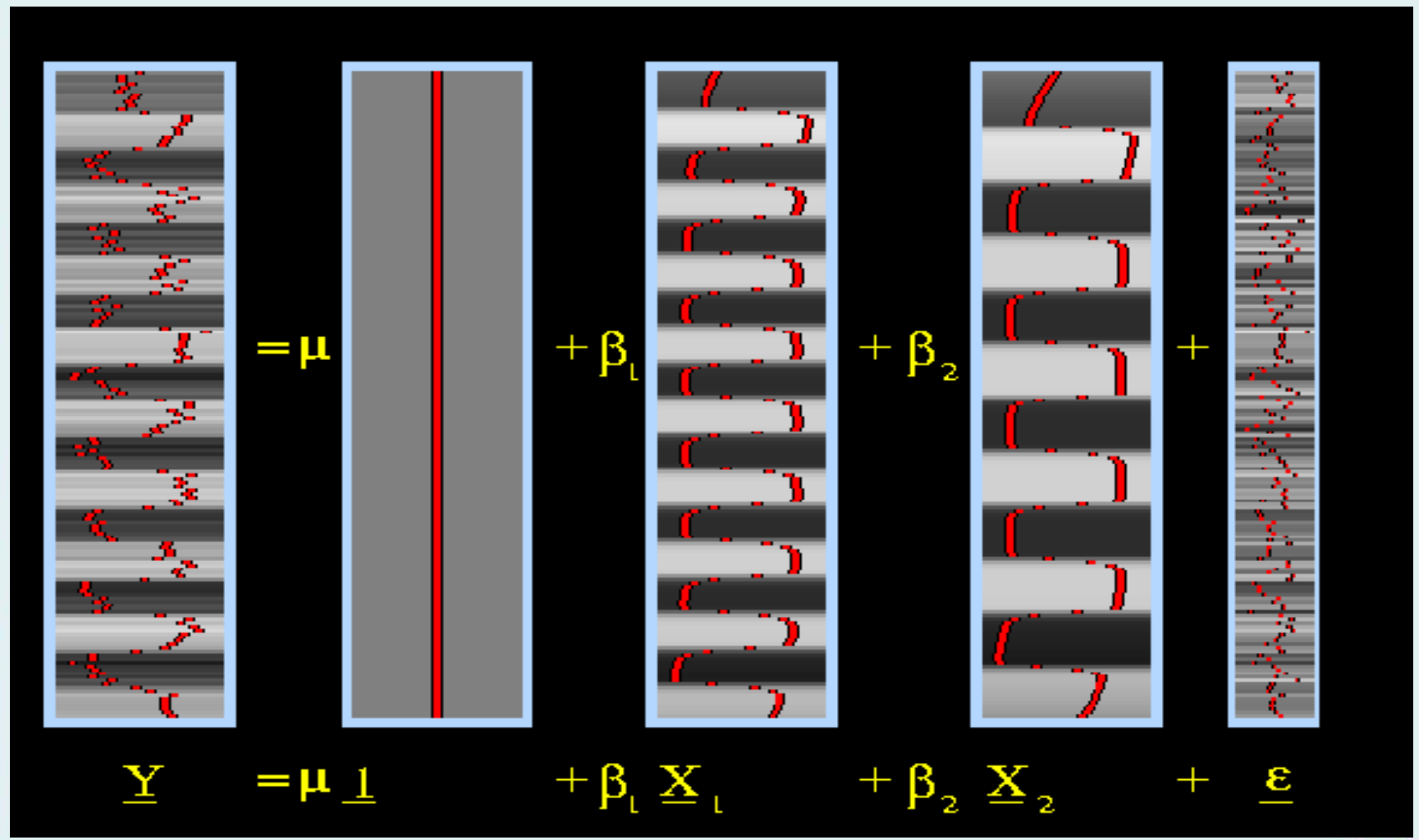
The Role of Hemodynamic Response Function (HRF)



GLM: General Linear Model (A simple view)



GLM: General Linear Model



Finding the Brain Activity

$$\begin{aligned}
 \underline{Y} &= \cancel{\mu \underline{1}} + \beta_1 \underline{X}_1 + \beta_2 \underline{X}_2 + \underline{\varepsilon} \\
 &= \begin{pmatrix} \underline{X}_1 & \underline{X}_2 \end{pmatrix} \begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix} + \underline{\varepsilon} \\
 &= \underline{X} \underline{\beta} + \underline{\varepsilon}
 \end{aligned}$$

Explanatory Variable (EV)

Mean

Design Matrix

Parameter estimates (PEs)

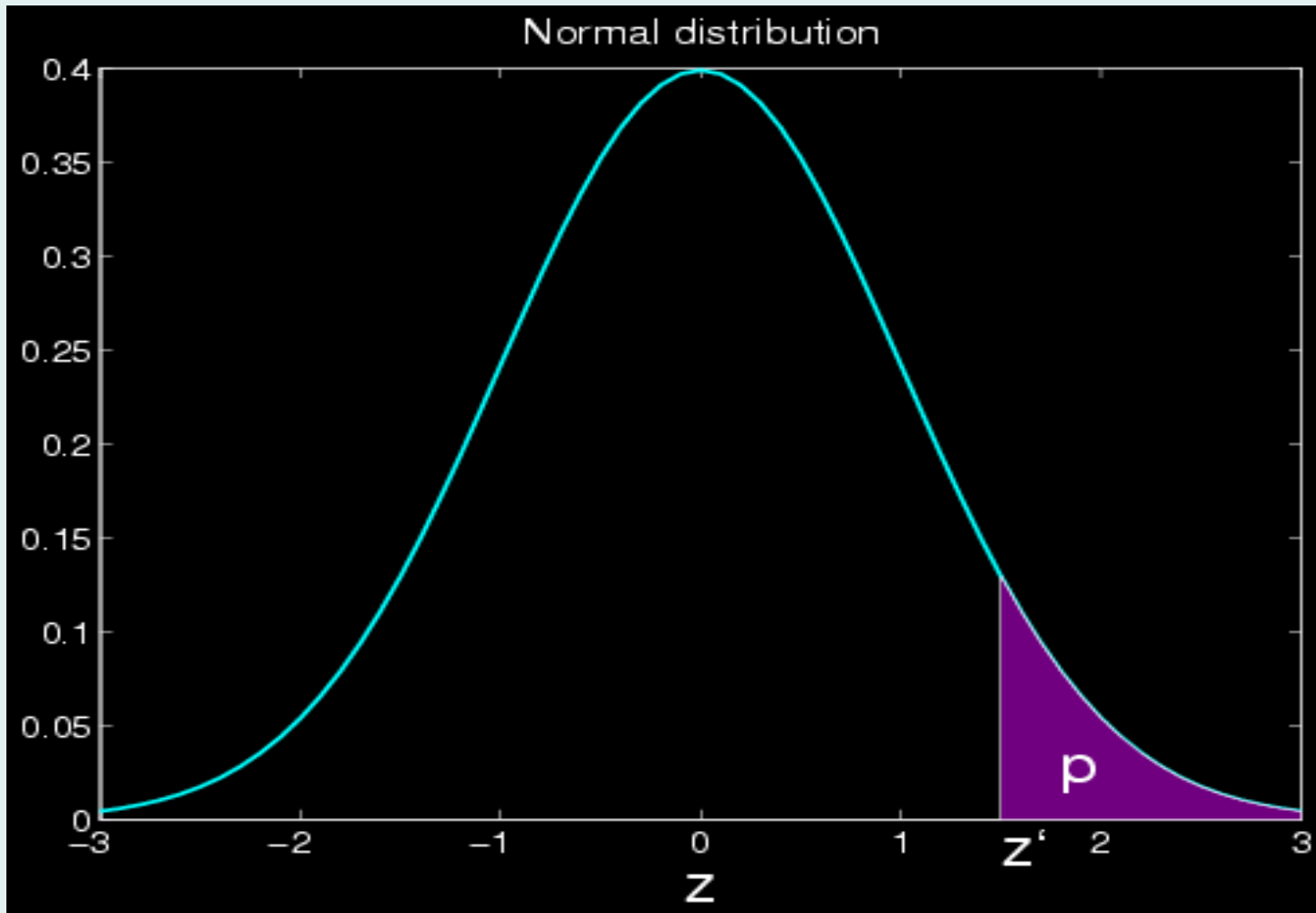
$$\hat{\beta} = (X^T X)^{-1} X^T Y$$

Contrast = [1 0]
 COPE = 1 × β₁ + 0 × β₂
 T-stat = COPE / STD(COPE)



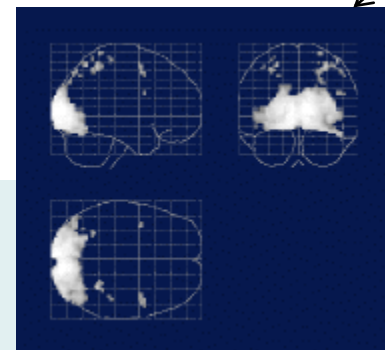
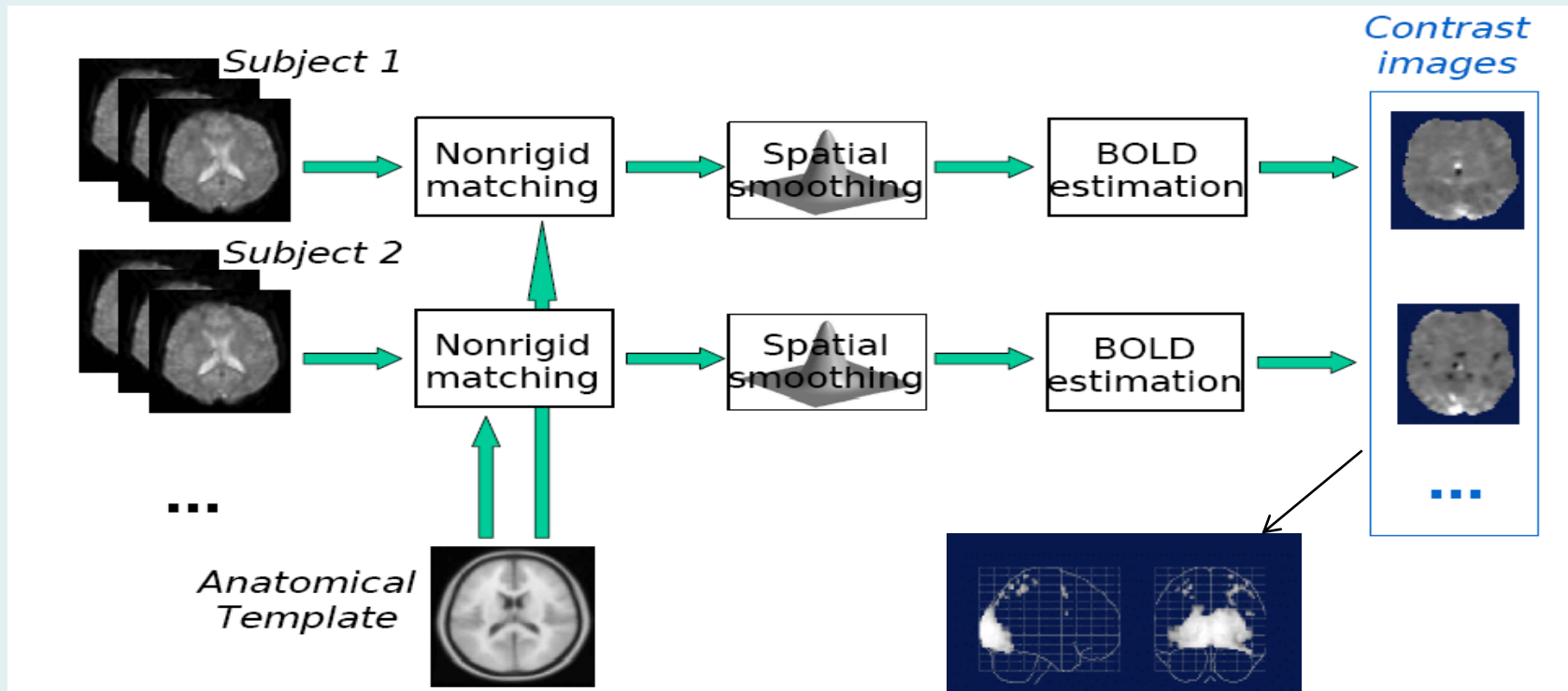
Thresholding

- Z (Gaussian Distributed), t (T-student), or F (Fisher) statistics



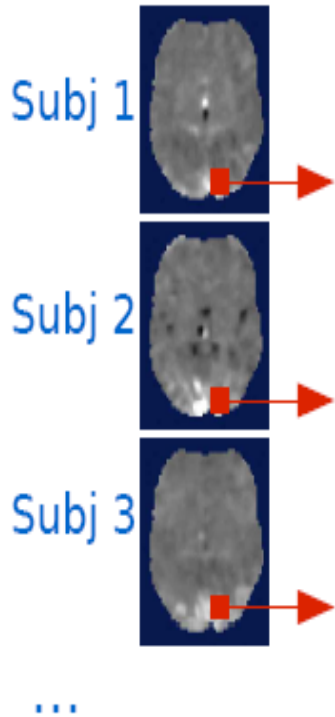
Group fMRI Analysis:

- Which brain regions are consistently activated in a population in response to a given contrast of experimental conditions?

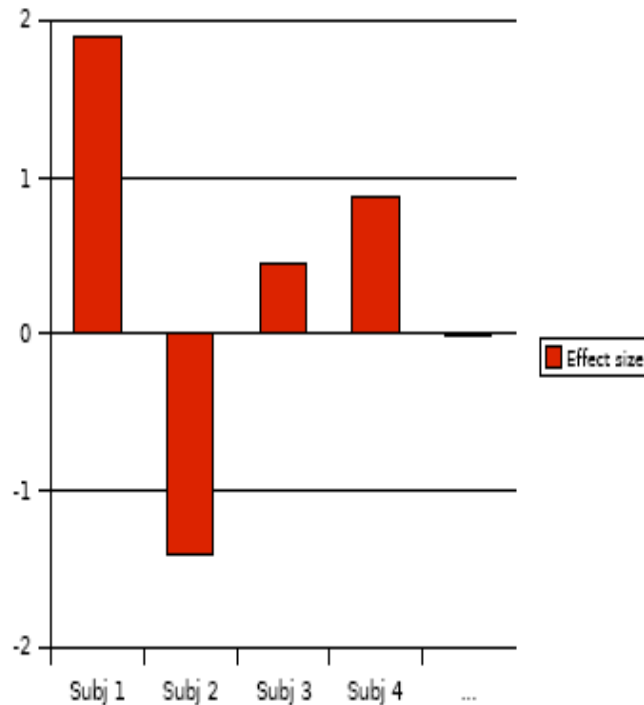


Group fMRI Analysis:

Contrast images



Observations in one voxel



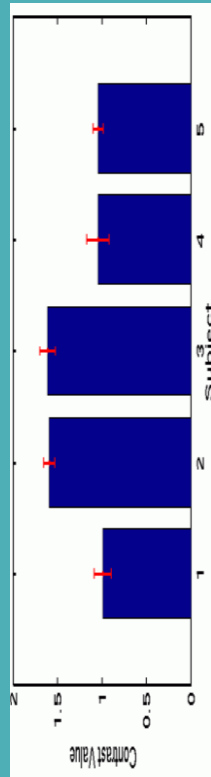
Significant mean effect?

Higher Level GLM Analysis

One-Sample Group Mean (OSGM)

$$y = X * \beta$$

Observations
(Low-Level Contrasts)



Data from
one voxel

$$= \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \beta_G$$

Design Matrix
(Regressors)

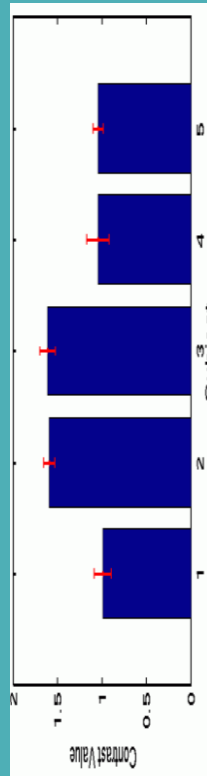
Vector of
Regression
Coefficients
("Betas")

Contrast Matrix:
 $C = [1]$
Contrast = $C * \beta = \beta_G$

Two Groups GLM Analysis

$$y = X * \beta$$

Observations
(Low-Level Contrasts)



Data from
one voxel

$$= \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 1 & 0 \\ 0 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \beta_{G1} \\ \beta_{G2} \end{bmatrix}$$

Contrasts: Two Groups GLM Analysis

1. Does Group 1 by itself differ from 0?

$$C = [1 \ 0], \text{ Contrast} = C * \beta = \beta_{G1}$$

2. Does Group 2 by itself differ from 0?

$$C = [0 \ 1], \text{ Contrast} = C * \beta = \beta_{G2}$$

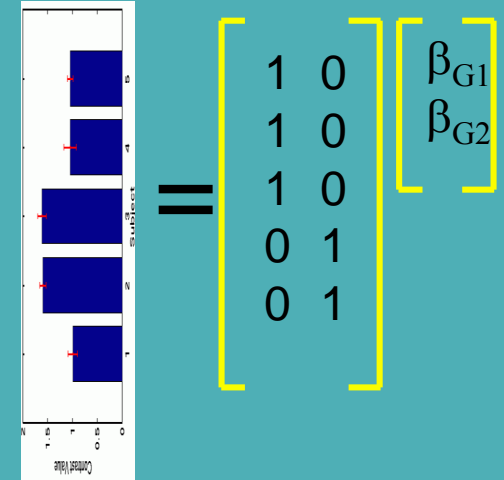
3. Does Group 1 differ from Group 2?

$$C = [1 \ -1], \text{ Contrast} = C * \beta = \beta_{G1} - \beta_{G2}$$

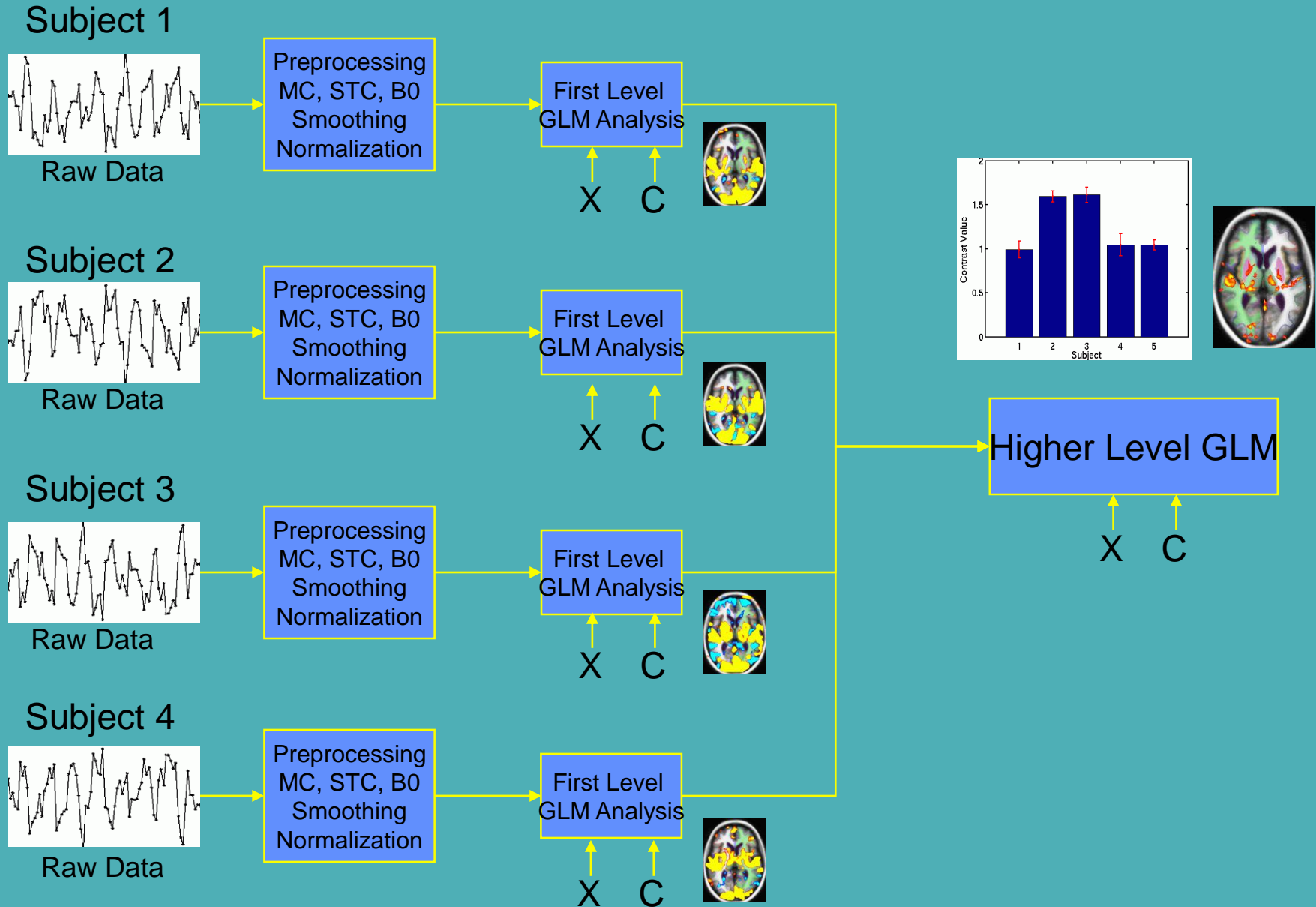
4. Does either Group 1 or Group 2 differ from 0?

C has two rows: F-test (vs t-test)
Concatenation of contrasts #1 and #2

$$C = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$



Higher Level fMRI Analysis Overview



Fixed versus Random Effect Analysis

- Fixed effect Analysis
 - all subjects treated as a single subject (fixed effect)
 - Huge areas of activation
 - Not generalizable beyond sample.
- Random Effect Analysis
 - Model Subjects as a Random Effect
 - Variance comes from a single source
 - Generalizable
- Mixed Effect Analysis
 - Model Subjects as a Random Effect and var comes from both first level and subjects
 - More complicated to compute

Fixed effect vs. Random effect

Do we want to make an inference about

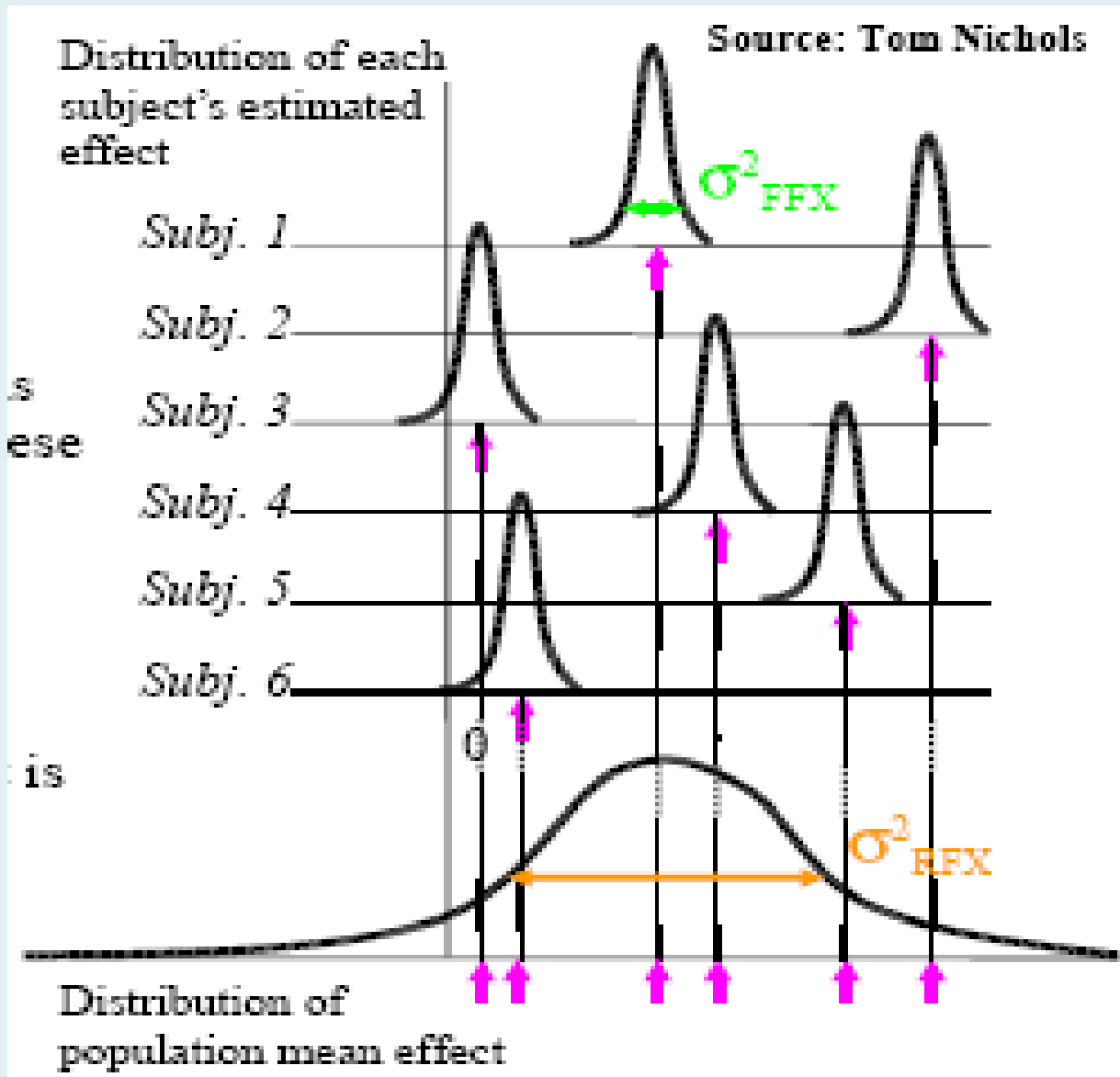
- The particular group we sampled? (Fixed Effect)
- The population they were sampled from? (Random Effect)

Two effect exist:

- Intra-subject variability:
there is some uncertainty on each subject's estimated effect due to "noise" in the fMRI data
- Inter-subject variability:
different subjects may have intrinsically different effects



Fixed effect vs. Random effect



Statistical Analysis: Independent Component Analysis



Different Kinds of Data Analysis:

Classical Data Analysis

How well does my model
Fit to the data?

Problem → Data →

Model → Analysis

→ Results

- results depend on the model

Exploratory Data Analysis

Is there any interesting thing in
the data?

Problem → Data →

Analysis → Model

→ Results

- can give unexpected results



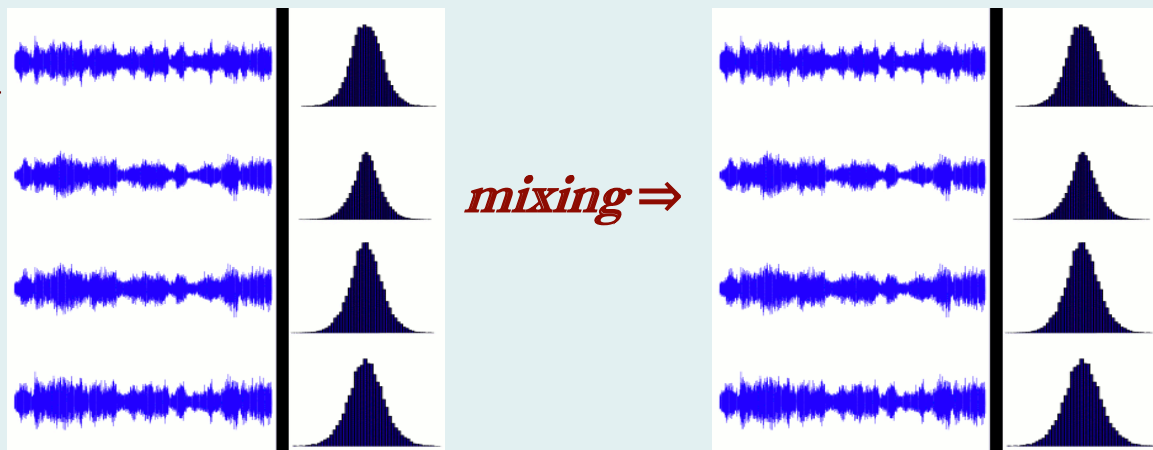
Why We need ICA in fMRI analysis?

- **Experiment:**
suboptimal event timing, non-efficient design, etc.
- **Physiology:**
secondary activation, ill-defined baseline, resting-fluctuations
- **Analysis:**
filtering & sampling artifacts, design misspecification.
- **MR Physics Analysis:**
MR noise, field inhomogeneity ,MR artifacts etc.



ICA Basis:

- **Independency** among the sources (Spatial & Temporal)
- **Non-Gaussianity**

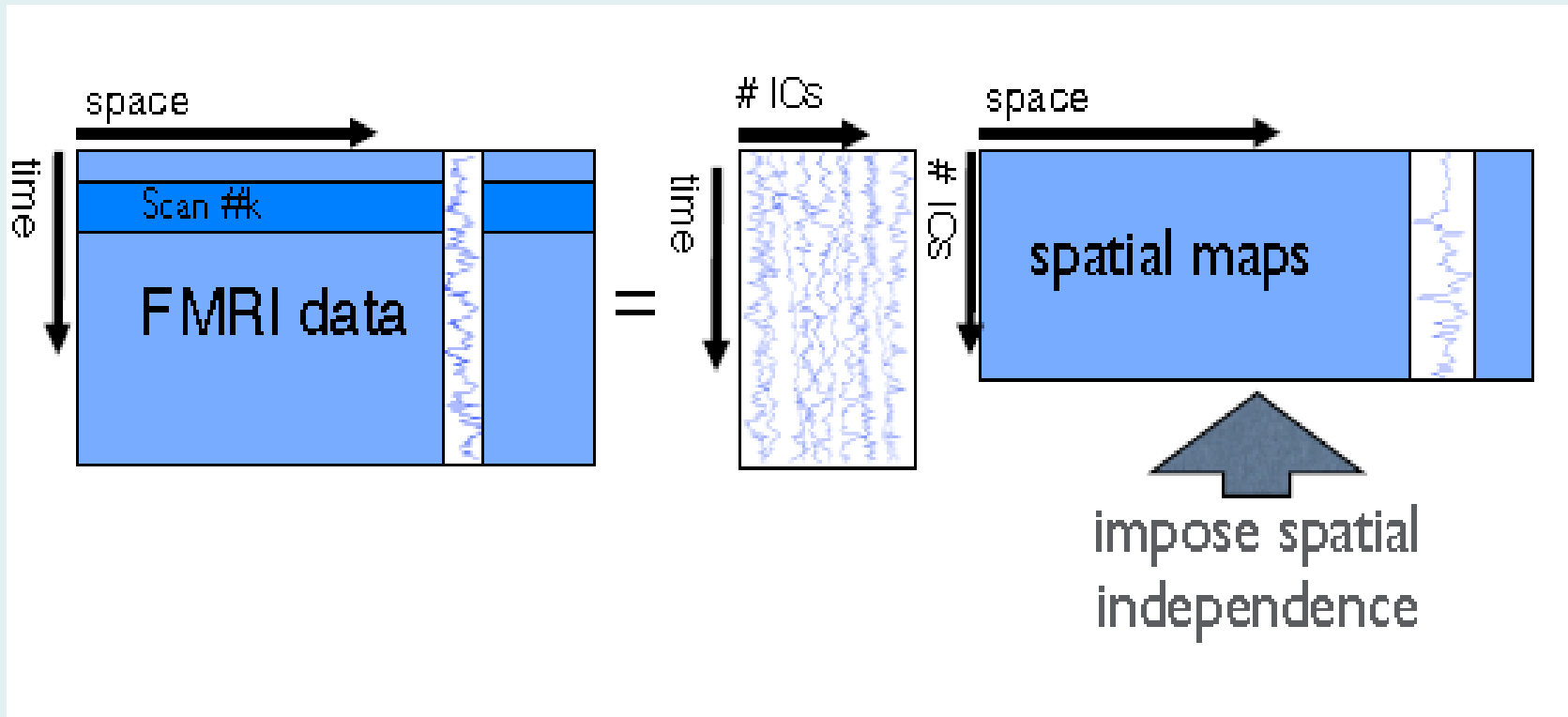


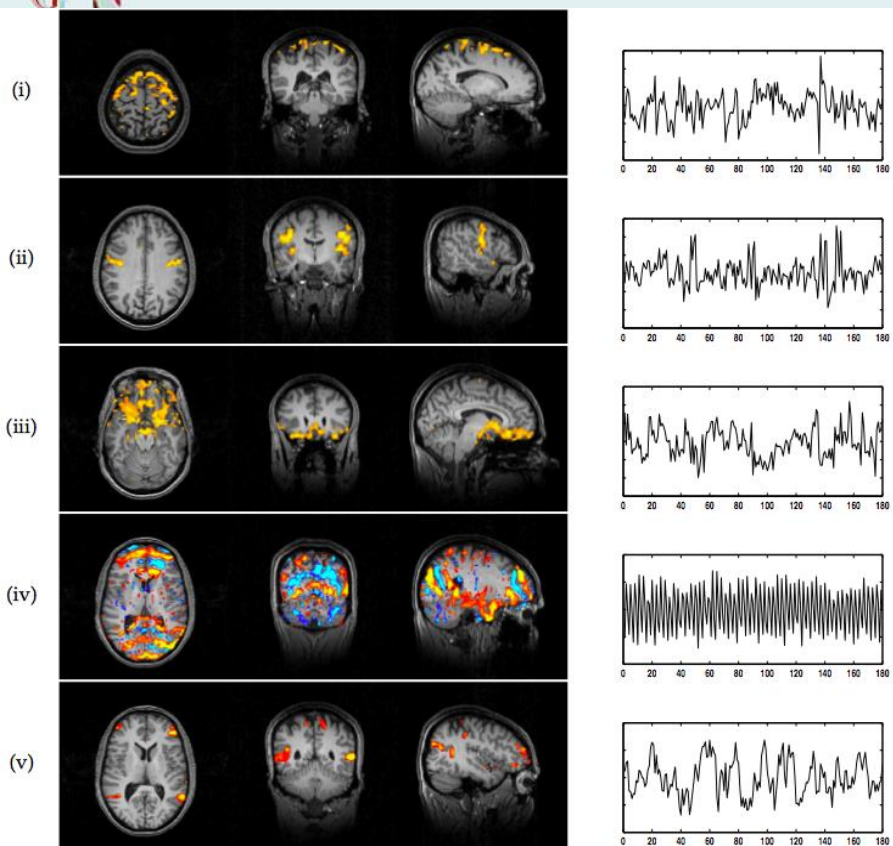
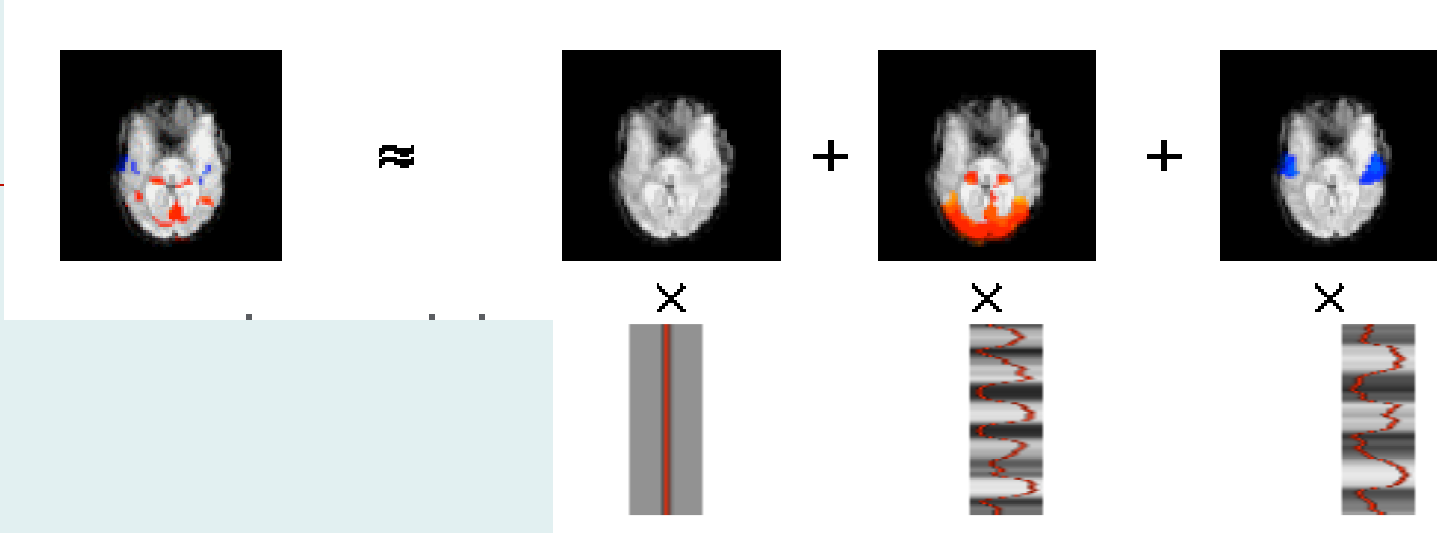
ICA Algorithms:

objective function + optimization technique

- maximum entropy & gradient descent (InfoMax)
- kurtosis or cumulates & gradient descent (Jade)
- negentropy & fixed point iteration (FastICA)

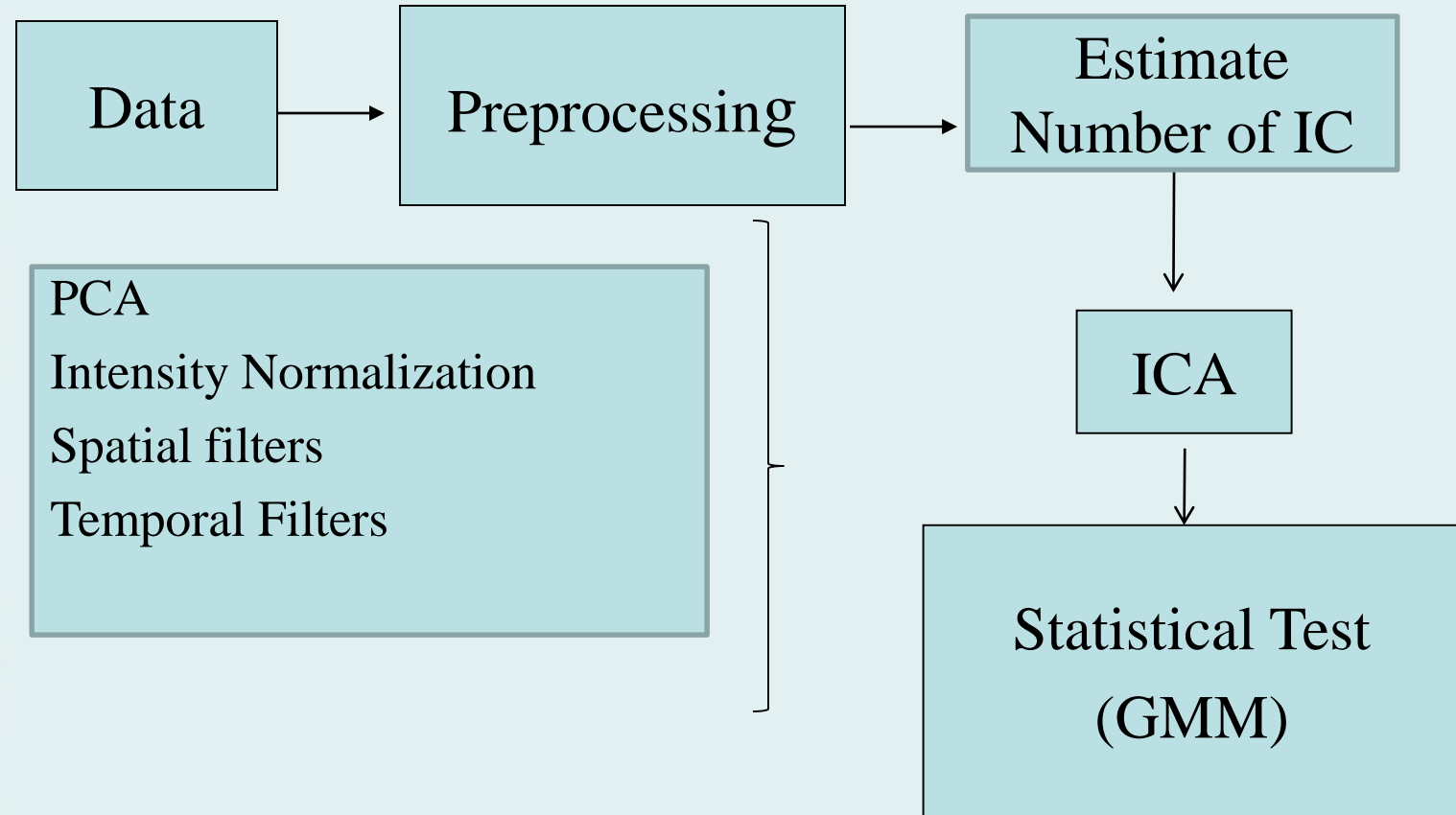
Spatial / Temporal ICA in FMRI





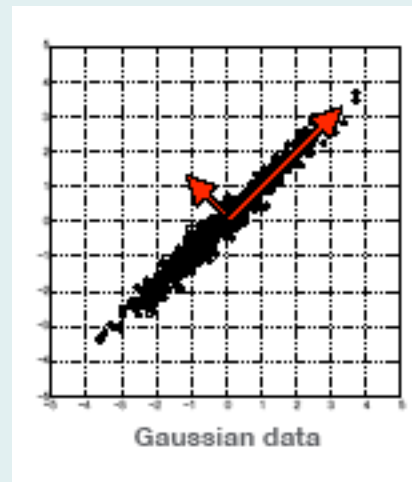
What are components in fMRI?

Chart of fMRI Analysis by ICA:



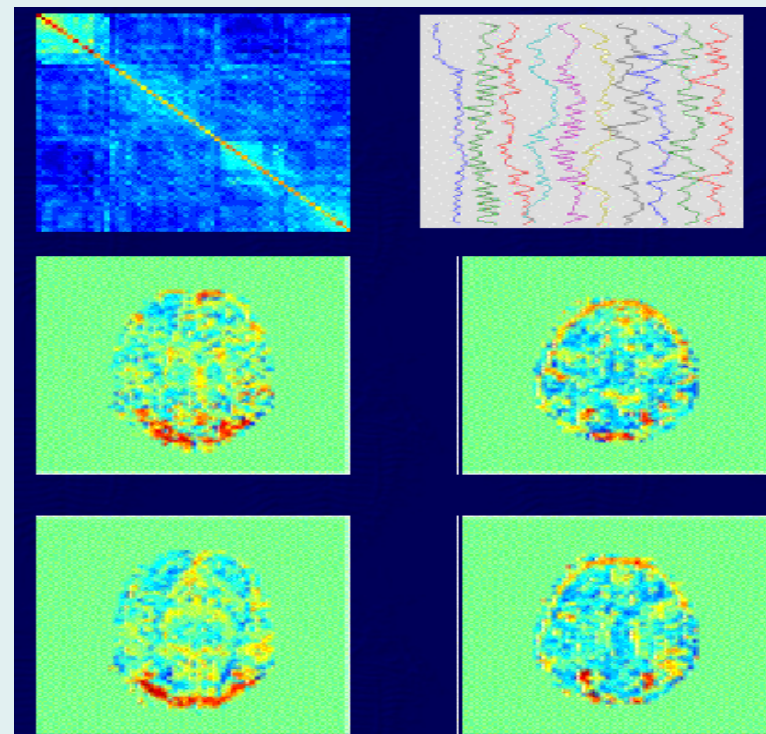
Preprocessing: PCA

- PCA finds projections of maximum amount of variance in Gaussian data.



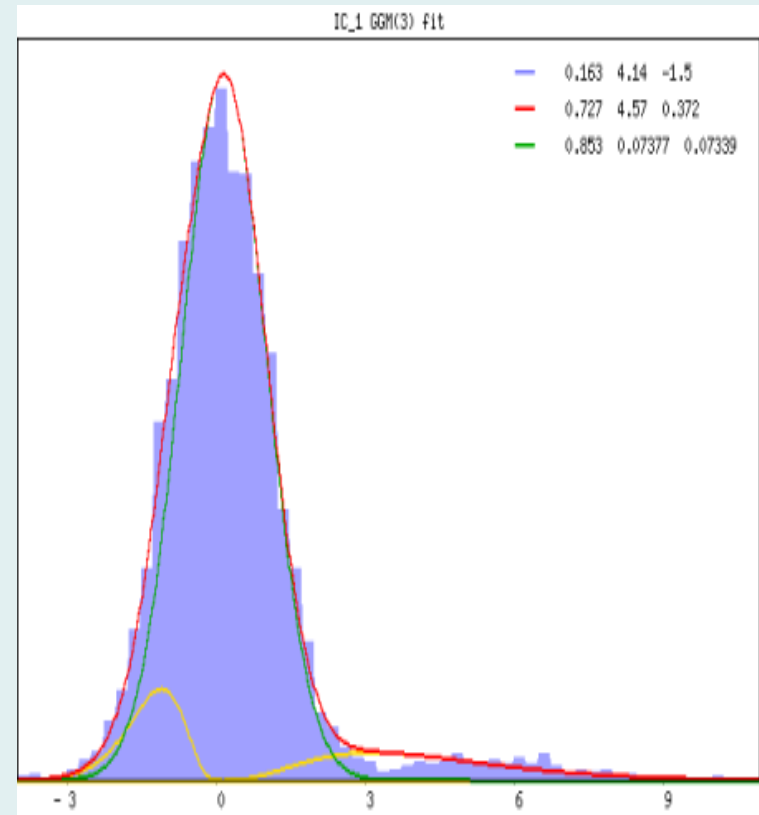
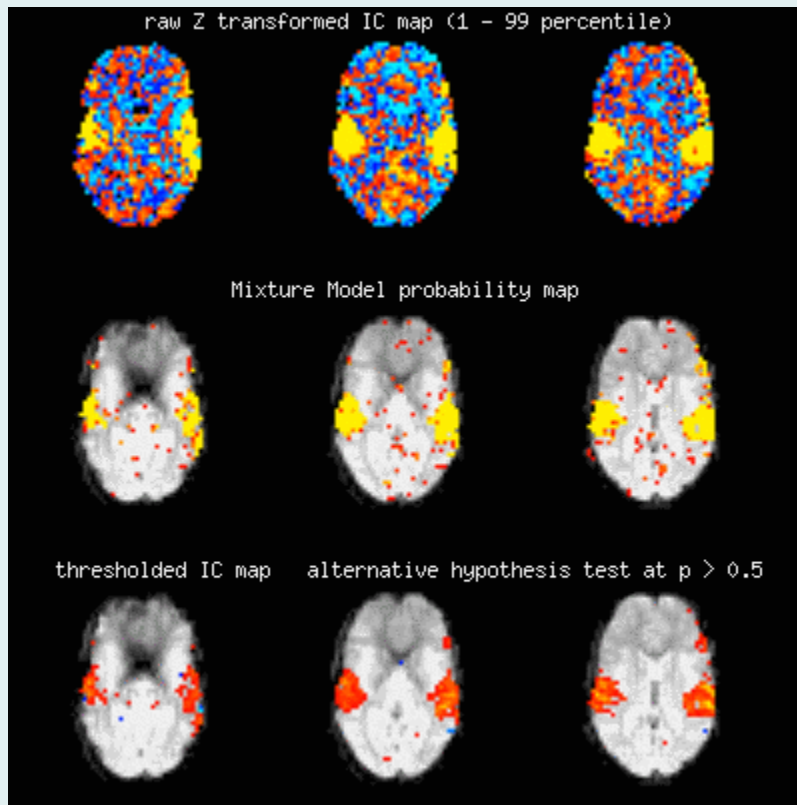
In fMRI data:

- calculate the data covariance matrix
- calculate the full set of Eigenvectors
- calculate the principal components by projecting the data onto the Eigenvectors



Alternative hypothesis test

use a Gaussian / mixture model fitted to the histogram of intensity values (using EM)



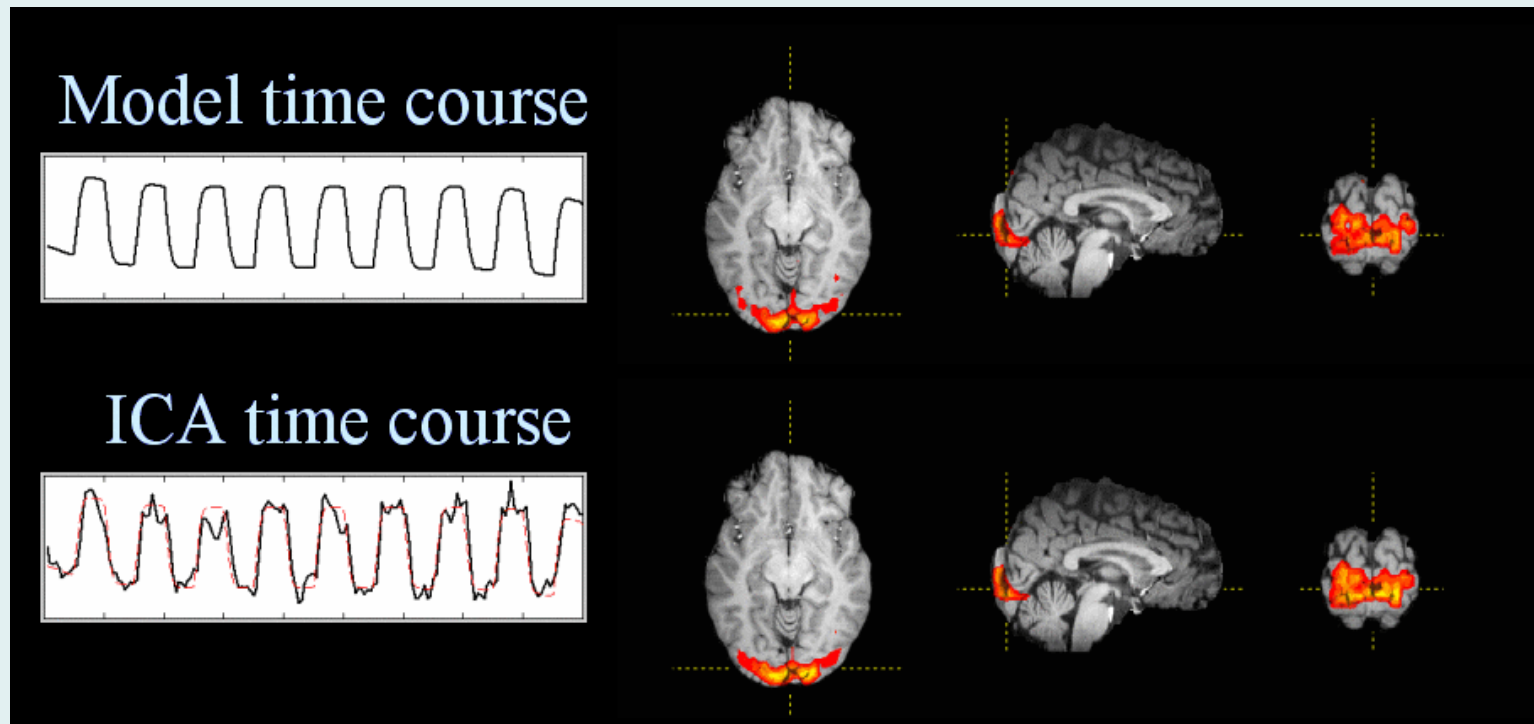
Applications

- MELODIC can be useful to investigate the BOLD for finding areas of 'activation' which respond in a more complex way to an external stimulus
- estimate artifacts in the data
- analyze data for which no model of the BOLD response is available



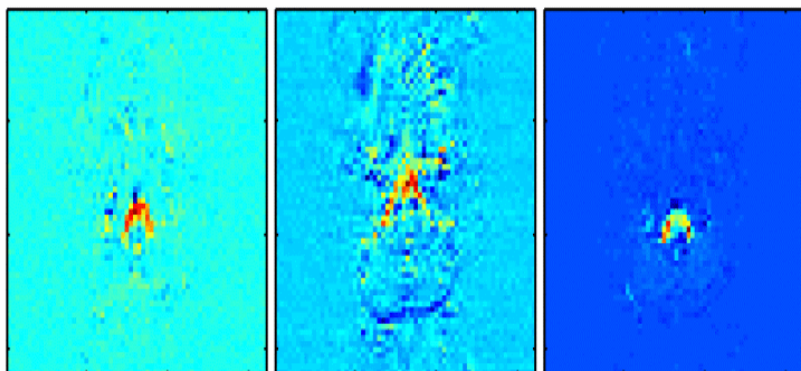
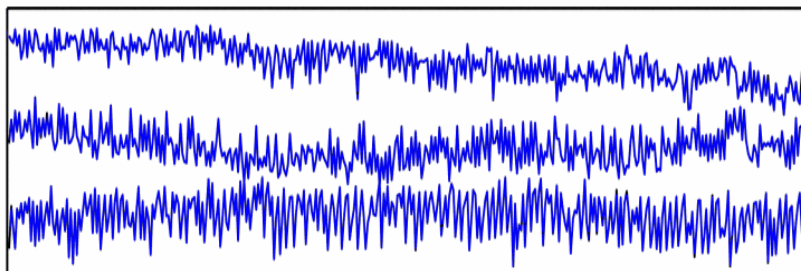
Investigating the BOLD response

- audio-visual stimulation
- 30s on/off reversing checkerboard with colour: GLM results and primary PICA map

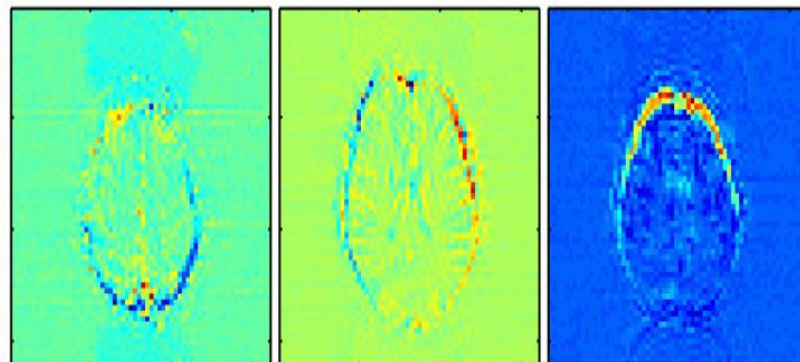
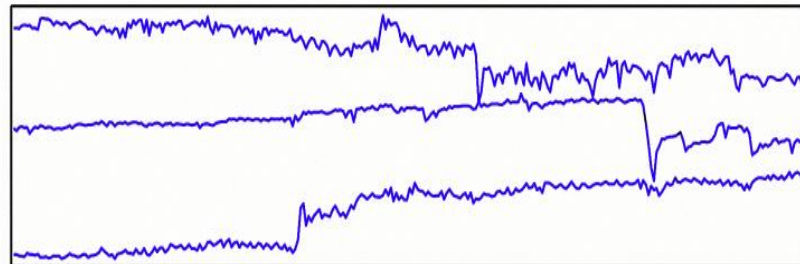


Different kinds of Artifact

high-frequency noise

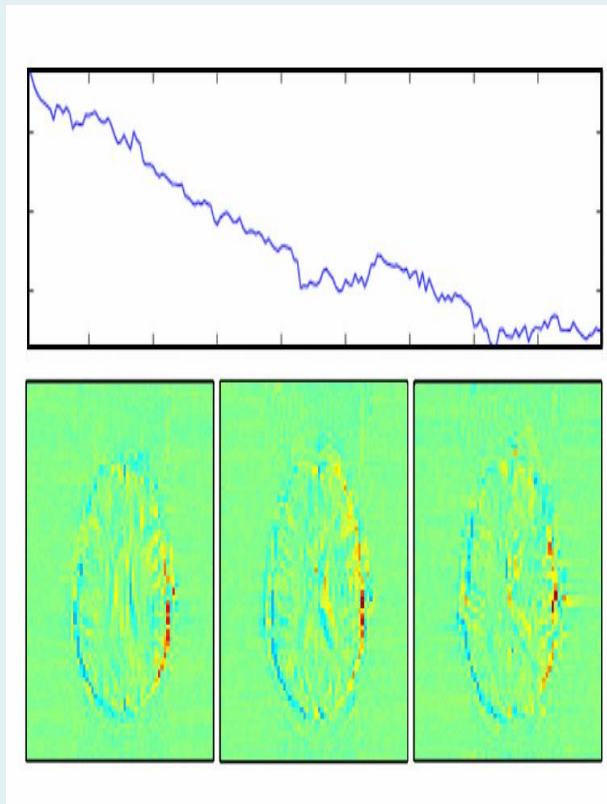


abrupt head motion

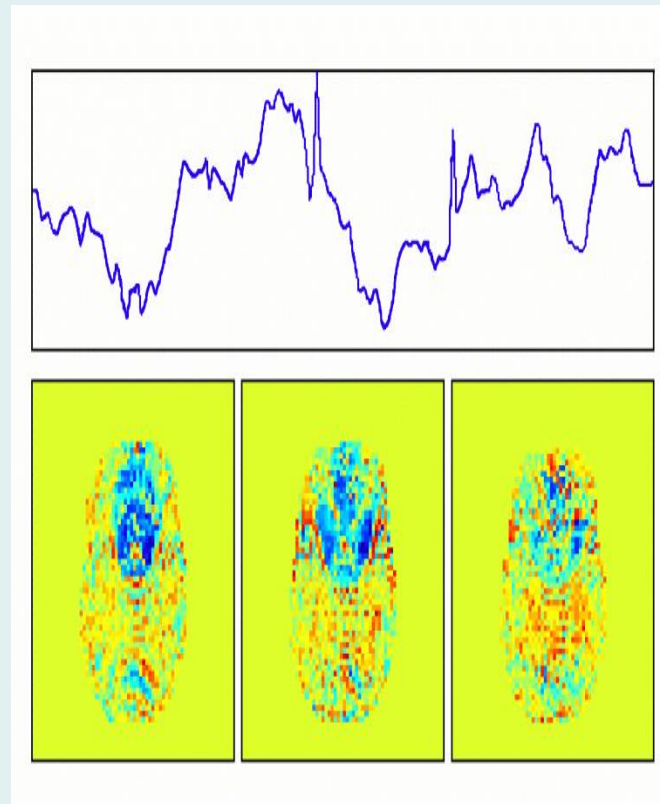


Different kinds of Artifact

slow head motion

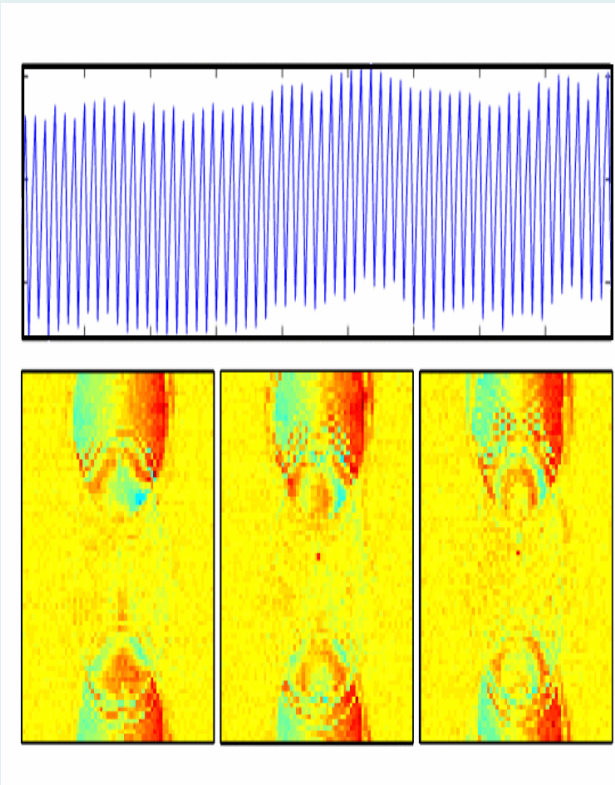


signal loss due to B0 inhomogeneity

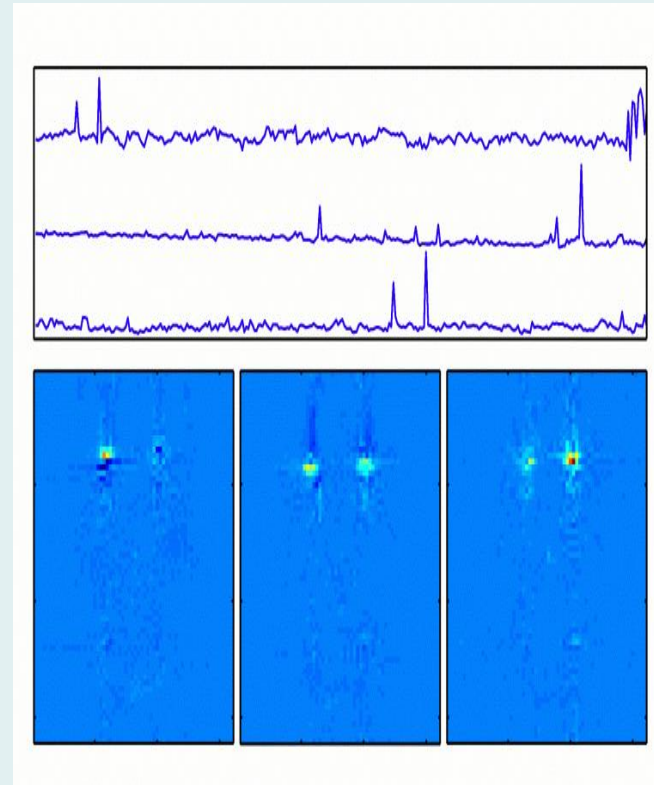


Different kinds of Artifact

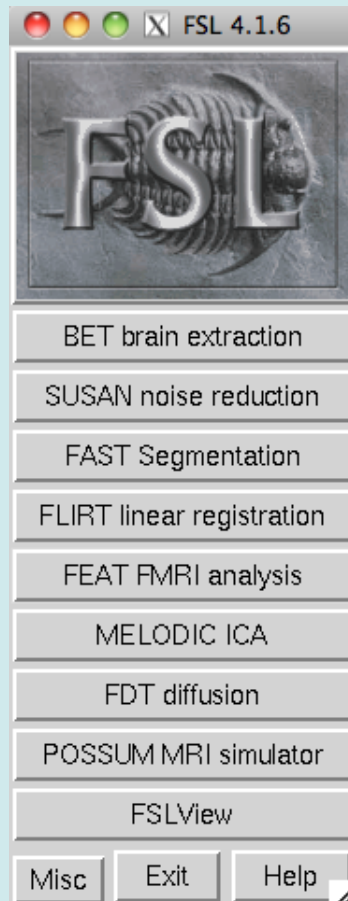
EPI (N/2) ghosting



Eye-related artefacts



fMRI Analysis Processing with FSL



Presented By:
Nafiseh Hasani

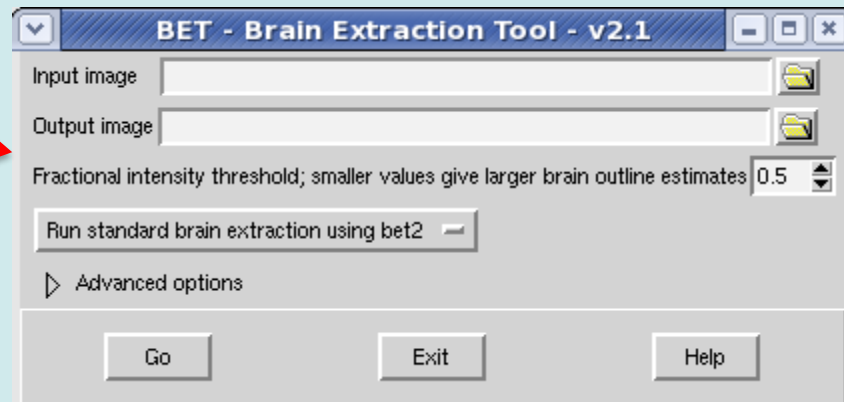
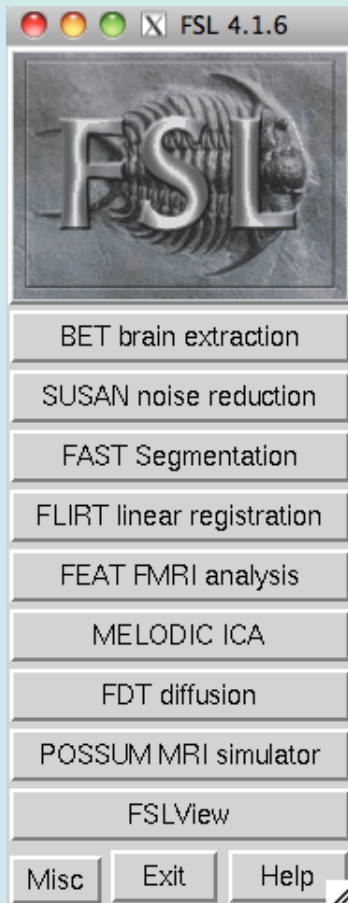
Preparing Data for fMRI Analysis

- ❑ With Dicomworks and DCM2Nii software
convert functional and anatomical data into
correct format (nifti ,4D)

FSL: Single Subject

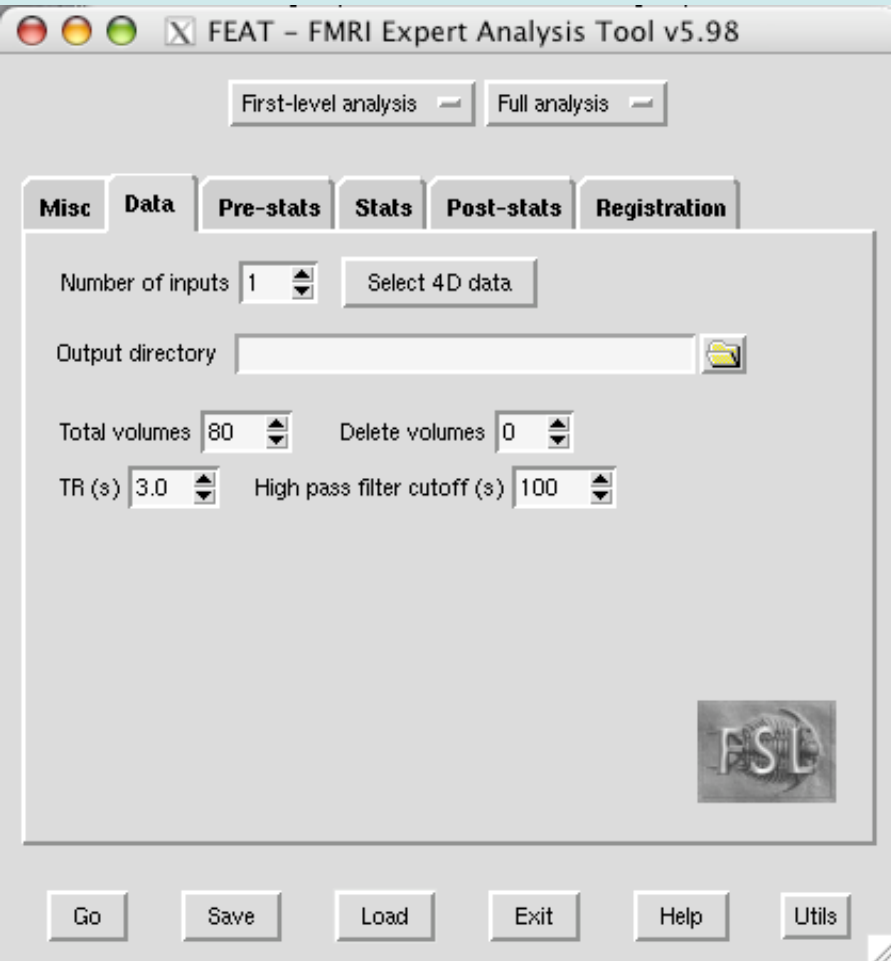
BET

- ❑ All anatomical images have to be run through BET before using FEAT

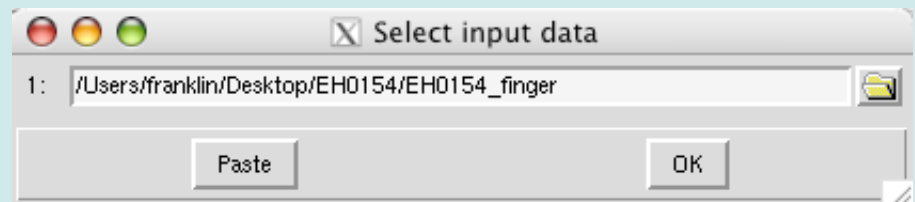


FEAT FMRI Analysis

Data



- Number of analyses
- Total Volumes
- Delete Volumes
- TR
- High Pass Filter Cutoff



FEAT FMRI Analysis

Delete Volumes

- The Number of initial FMRI volumes to delete before any further processing
- Typically two or three volumes
- These volumes are deleted as soon as the analysis is started

FEAT FMRI Analysis

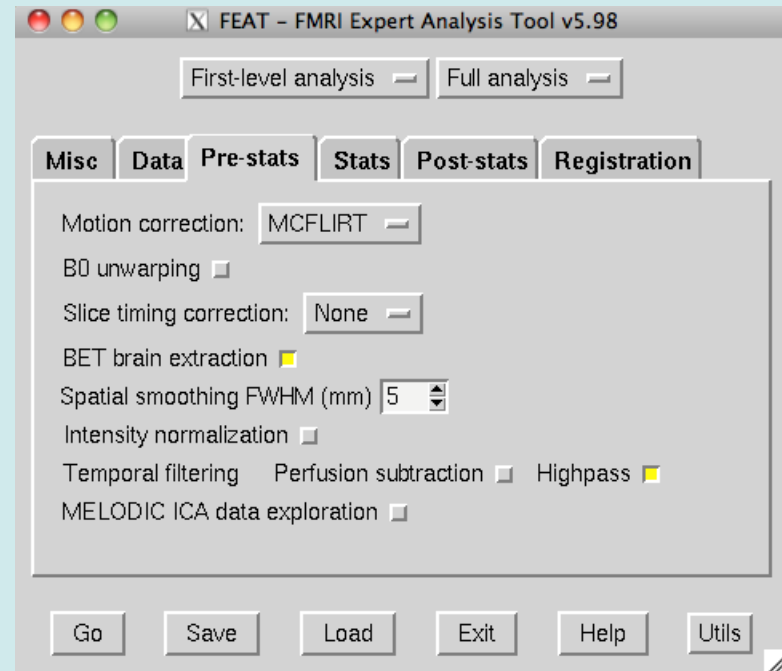
High Pass Filter Cutoff

- ❑ A sensible setting in the case of an rArA or rArB type block design, is the $r+A$ or $r+A+r+B$, total cycle time
- ❑ For event-related designs the rule is not so simple. But in general the cut off can typically be reduced at least to 50s

FEAT FMRI Analysis

Pre-stats

- Motion Correction
- BET Brain Extraction
- Slice timing correction
- Spatial Smoothing FWHM (mm)
- Temporal Filtering



FEAT FMRI Analysis

Slice timing correction

- If slices were acquired from the bottom of the brain to the top select **Regular up**
- If slices were acquired from the top of the brain to the bottom select **Regular down**
- If the slices were acquired with interleaved order (0, 2, 4 ... 1, 3, 5 ...) the choose the **Interleaved** option.

FEAT FMRI Analysis

Spatial Smoothing FWHM

- ❑ This is intended to reduce noise without reducing valid activation; this is successful as long as the underlying activation area is larger than the extent of the smoothing
- ❑ if you are looking for very small activation areas then you should maybe reduce smoothing from the default of 5mm, and if you are looking for larger areas, you can increase it

FEAT FMRI Analysis

Temporal Filtering

- ❑ Highpass temporal filtering : remove low frequency artifacts
- ❑ Lowpass temporal filtering: reduces high frequency noise by Gaussian smoothing ($\sigma=2.8s$), but also reduces the strength of the signal of interest. so is turned off by default.

FEAT FMRI Analysis

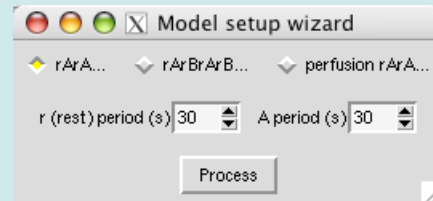
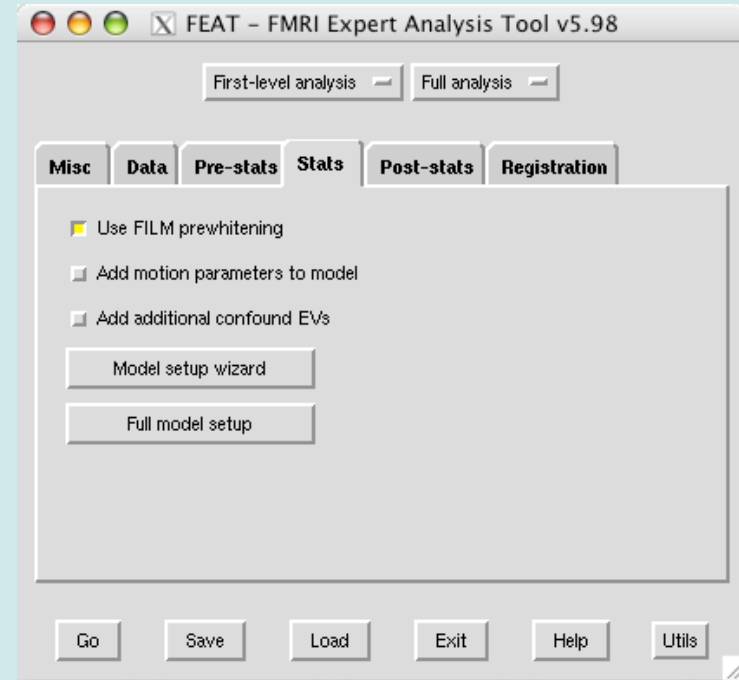
Model Setup:

1. Model Setup Wizard

- Allows you to setup simple experimental designs.

2. Full Model Setup

- You will need to give a text file containing ones and zeros for each explanatory variable.



FEAT FMRI Analysis

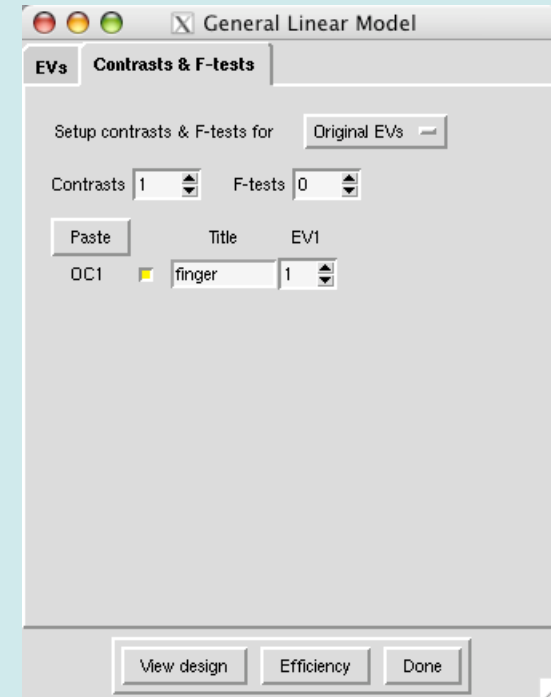
Contrasts

❑ Original EVs- Explanatory Variables

❑ Basic Shape- design matrix Custom (3 Column Format)

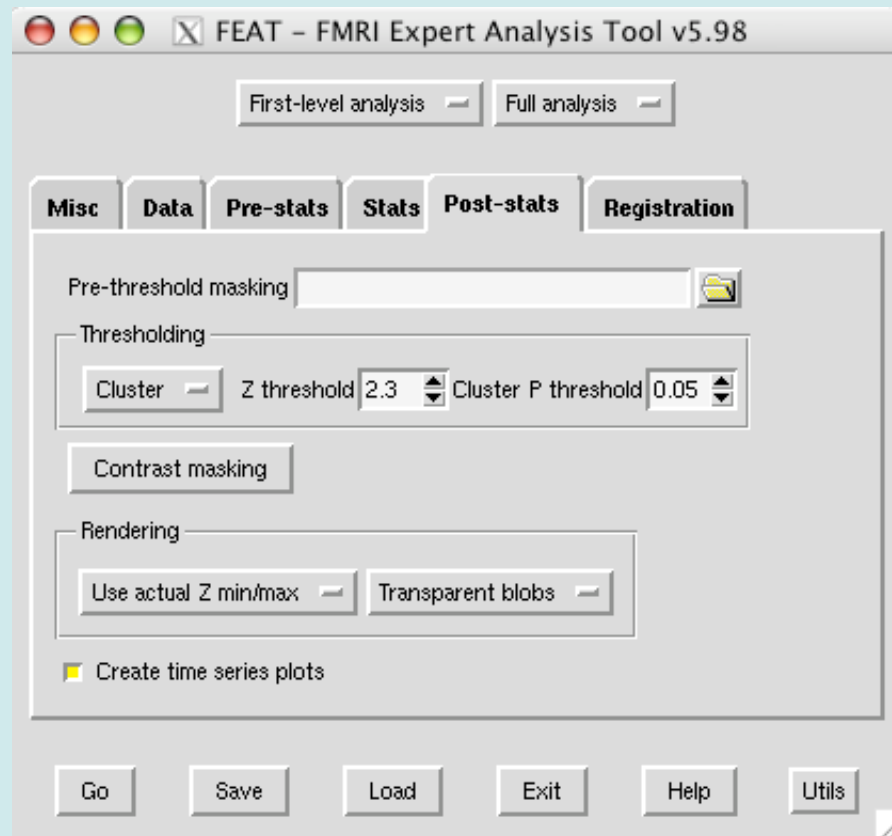
0	30	1
30	30	0
60	30	1
90	30	0
120	30	1
150	30	0
180	30	1
210	30	0

3-Column Text File



FEAT FMRI Analysis

Post-stats



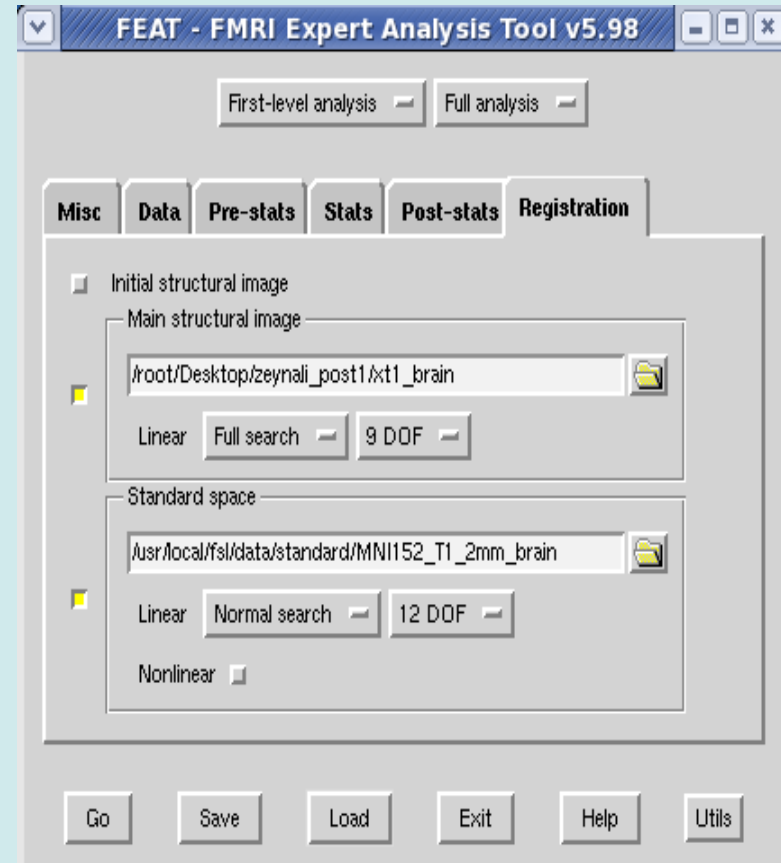
- The data is usually run with the default threshold of $Z > 2.3$.

FEAT FMRI Analysis

Registration

Main Structure – 3D anatomical image.

Standard Space – Should be an image in Talairach space



FEAT FMRI Analysis

Results

- For each analyses ran through FEAT an output directory is created with the extension “.feat”
 - This folder you will contain all statistical images and files.
 - The results can be viewed by clicking on the **report.html**.
 - This will included the Pre-stats, Stats, Post-Stats, Registration, and Log.

FEAT Report

/Users/franklin/Desktop/EH0154/EH0154_finger.feats

Finished at Fri Oct 3 14:34:18 CDT 2008

[Pre-stats](#) - [Stats](#) - [Post-stats](#) - [Registration](#) - [Log](#)



FEAT FMRI Analysis

Results

Pre-stats

Analysis methods

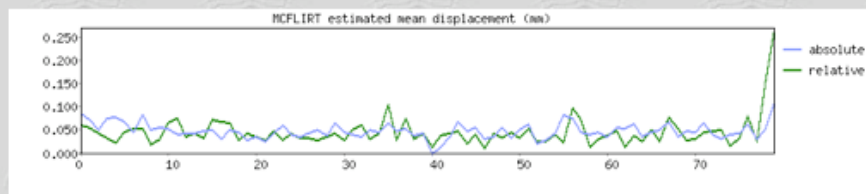
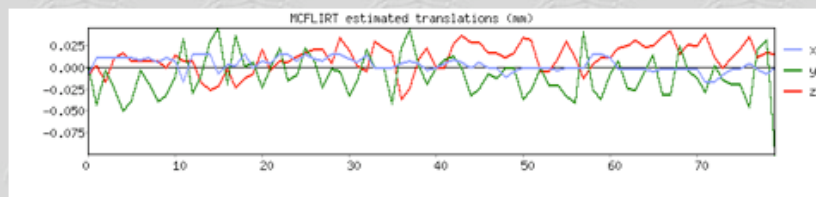
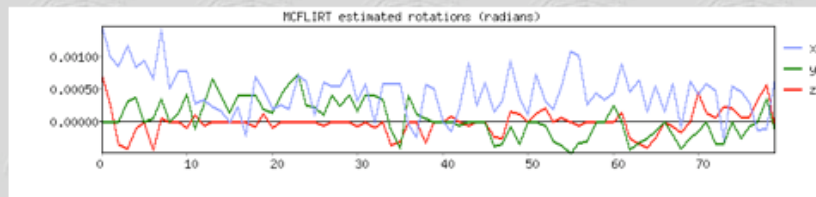
FMRI data processing was carried out using FEAT (FMRI Expert Analysis Tool) Version 5.98, part of FSL (FMRIB's Software Library, www.fmrib.ox.ac.uk/fsl). The following pre-statistics processing was applied; motion correction using MCFLIRT [Jenkinson 2002]; non-brain removal using BET [Smith 2002]; spatial smoothing using a Gaussian kernel of FWHM 5mm; grand-mean intensity normalisation of the entire 4D dataset by a single multiplicative factor; highpass temporal filtering (Gaussian-weighted least-squares straight line fitting, with $\sigma=50.0s$).

References

[Jenkinson 2002] M. Jenkinson and P. Bannister and M. Brady and S. Smith. Improved optimisation for the robust and accurate linear registration and motion correction of brain images. *NeuroImage* 17:2(825-841) 2002.
[Smith 2002] S. Smith. Fast Robust Automated Brain Extraction. *Human Brain Mapping* 17:3(143-155) 2002.

MCFLIRT Motion correction

Mean displacements: absolute=0.04mm, relative=0.04mm



FEAT FMRI Analysis

Results

Stats

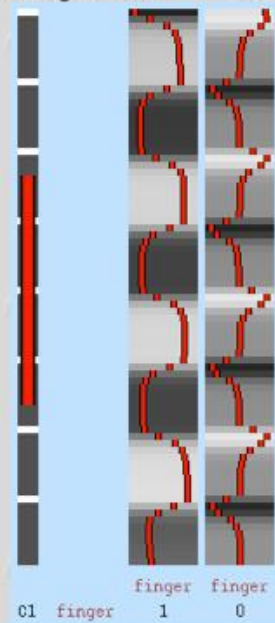
Analysis methods

FMRI data processing was carried out using FEAT (FMRI Expert Analysis Tool) Version 5.98, part of FSL (FMRIB's Software Library, www.fmrib.ox.ac.uk/fsl). Time-series statistical analysis was carried out using FILM with local autocorrelation correction [Woolrich 2001].

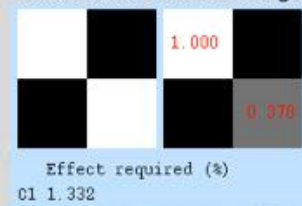
References

[Woolrich 2001] M.W. Woolrich, B.D. Ripley, J.M. Brady and S.M. Smith. Temporal Autocorrelation in Univariate Linear Modelling of FMRI Data. *NeuroImage* 14:6(1370-1386) 2001.

Design matrix



Covariance matrix & design efficiency



FEAT FMRI Analysis Results


Post-stats

Analysis methods

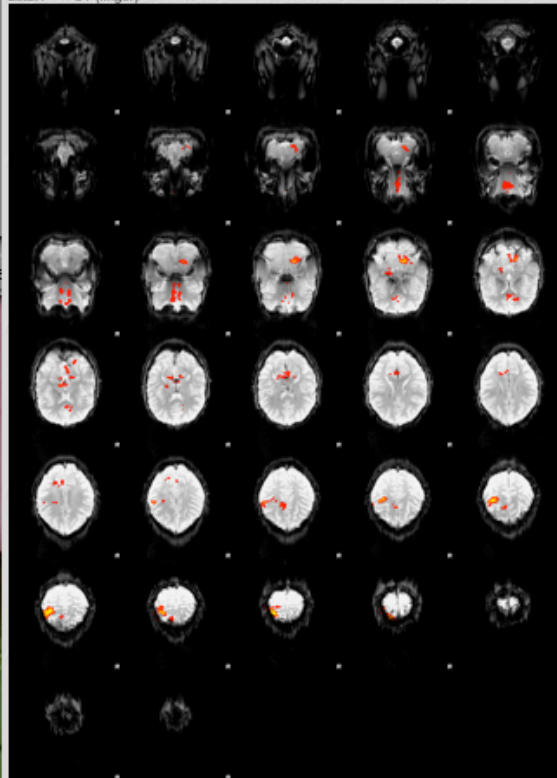
FMRI data processing was carried out using FEAT (FMRI Expert Analysis Tool) Version 5.98, part of FSL (FMRIB's Software Library, www.fmrib.ox.ac.uk/fsl). Z (Gaussianised T/F) statistic images were thresholded using clusters determined by $Z > 2.3$ and a (corrected) cluster significance threshold of $P = 0.05$ [Worsley 2001].

References

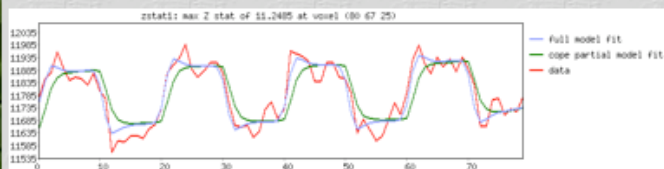
[Worsley 2001] K.J. Worsley. Statistical analysis of activation images. Ch 14, in *Functional MRI: An Introduction to Methods*, eds. P. Jezzard, P.M. Matthews and S.M. Smith. OUP, 2001.

Thresholded activation images 2.3  11.2

zstat1 - C1 (finger)



Time series plots



FEAT FMRI Analysis

Results

Registration

Analysis methods

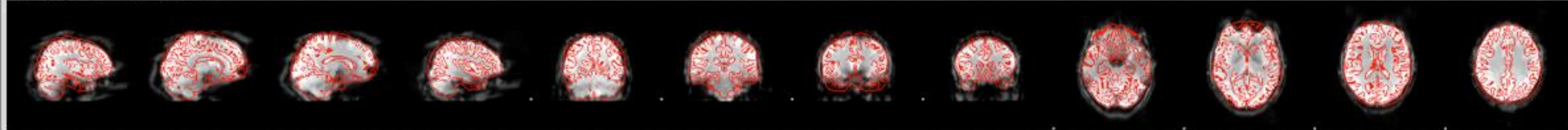
FMRI data processing was carried out using FEAT (FMRI Expert Analysis Tool) Version 5.96, part of FSL (FMRI's Software Library, www.fmrib.ox.ac.uk/fsl). Registration to high resolution structural and/or standard space images was carried out using FLIRT [Jenkinson 2001, 2002].

References

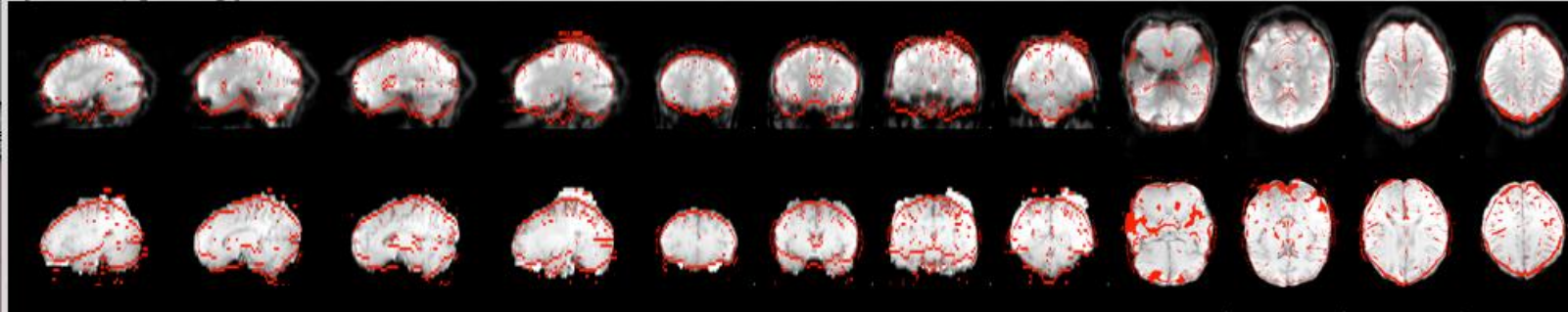
[Jenkinson 2001] M. Jenkinson and S.M. Smith. A Global Optimisation Method for Robust Affine Registration of Brain Images. *Medical Image Analysis* 5:2(143-156) 2001.

[Jenkinson 2002] M. Jenkinson, P. Bannister, M. Brady and S. Smith. Improved Optimisation for the Robust and Accurate Linear Registration and Motion Correction of Brain Images. *NeuroImage* 17:2(825-841) 2002.

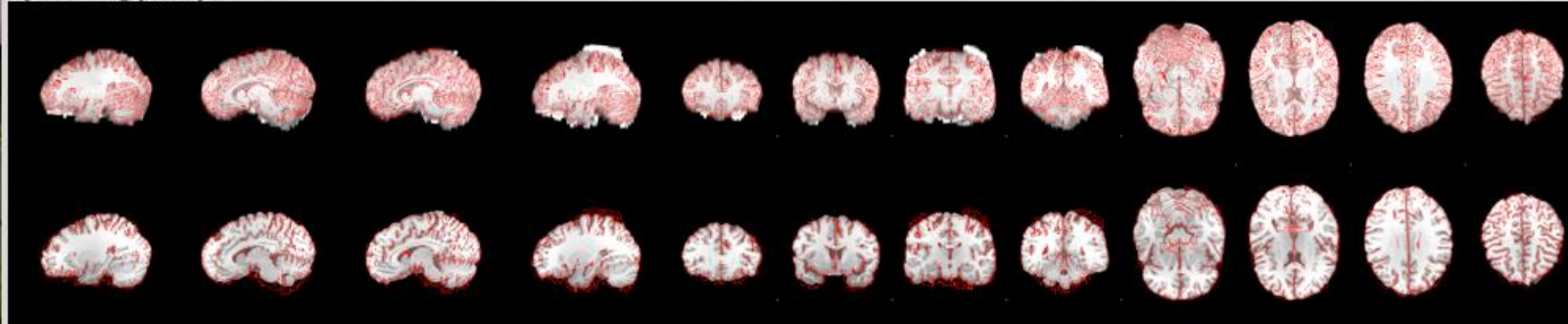
Summary registration, FMRI to standard space



Registration of example_func to initial_highres

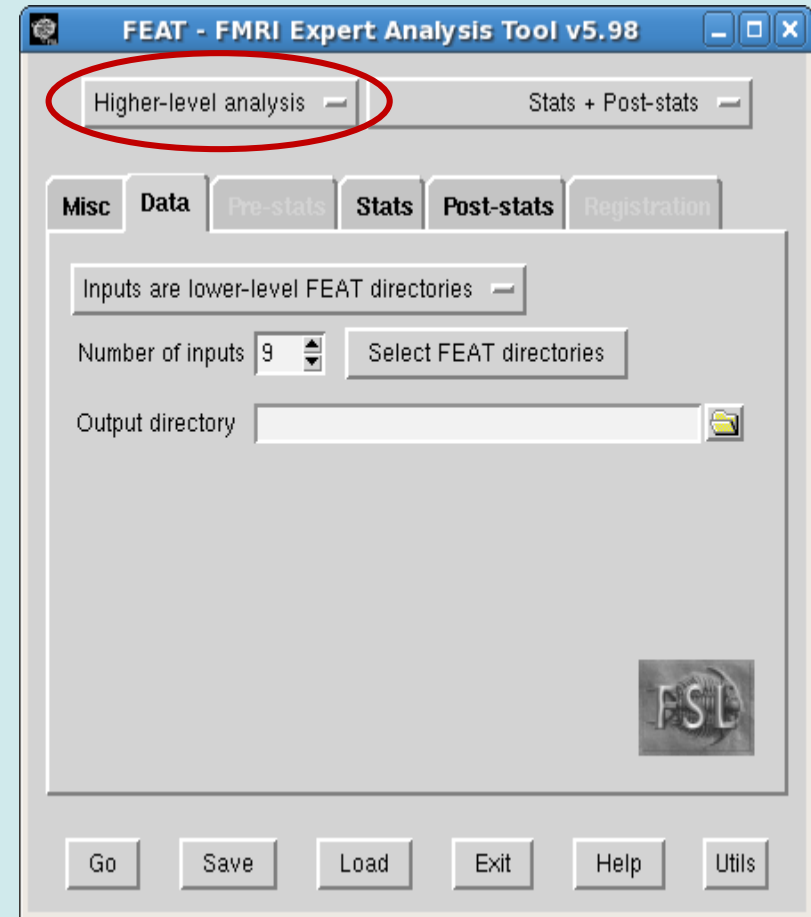


Registration of initial_highres to highres

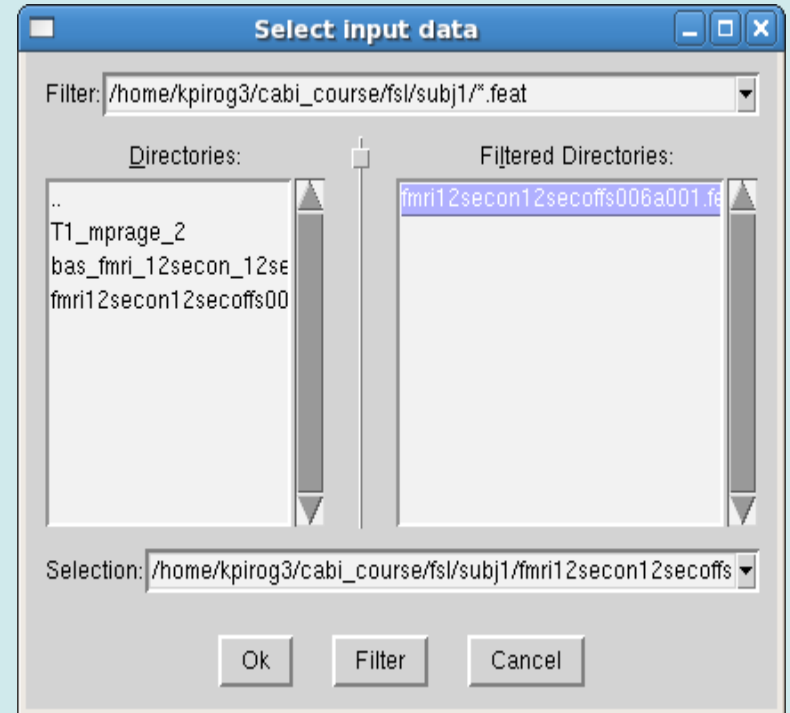
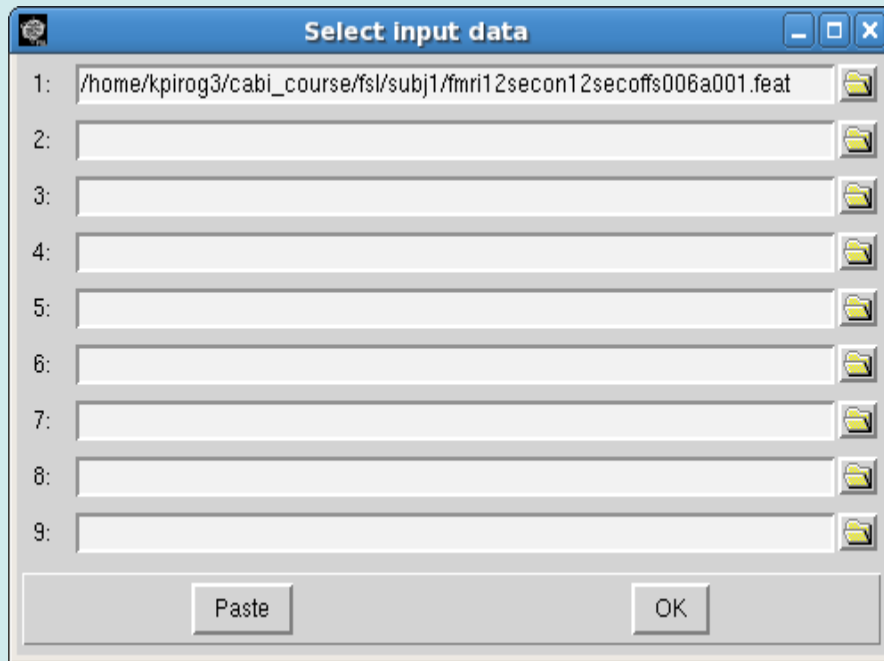


FSL: Group Analysis

- ❑ Set tab at top left to Higher-level analysis
- ❑ Data Tab:
 - Inputs are FEAT directories
 - 9 inputs
 - Output directory should be ok if you launched FSL from the directory I told you to create for yourself!

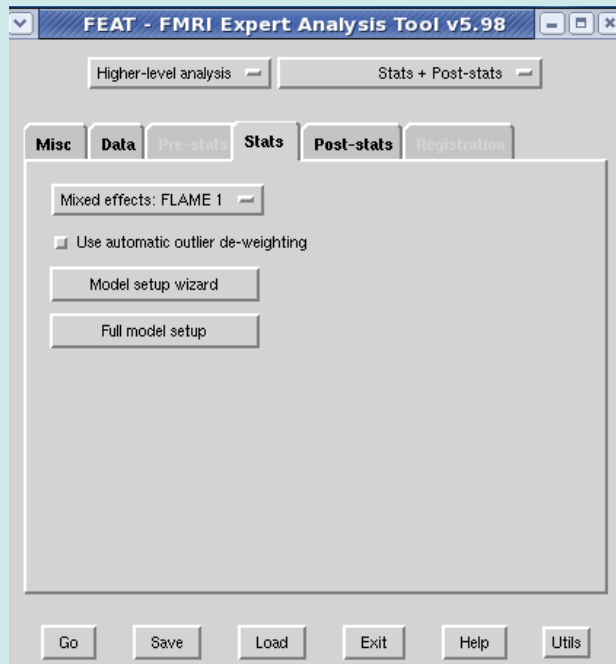


FSL: Group Analysis



- ❑ Click on folders to get to file selection dialog and select the fmri...fe directory for each participant

FSL: Group Analysis



- Fixed effects
- Mixed effects: FLAME1
- Mixed effects: FLAME 1+2
- Mixed effects: Simple OLS

FSL: Group Analysis

Fixed effects

- ❑ FE ignores cross-session/subject variance, reported activation is with respect to the group of sessions or subjects present, and not representative of the wider population

FSL: Group Analysis

Mixed effects: Simple OLS

- ❑ OLS (ordinary least squares) is a fast estimation technique which ignores all lower-level variance estimation and applies a very simple higher-level model. This is the least accurate of the ME options.

FSL: Group Analysis

Mixed effects: FLAME1

- ❑ For the most accurate estimation of higher-level activation you should use FLAME (FMRIB's Local Analysis of Mixed Effects) modeling and estimation. it allows separate modeling of the variance in different subject groups
- ❑ The first stage of FLAME is significantly more accurate than OLS, and nearly as fast

FSL: Group Analysis

Mixed effects: FLAME1 +2

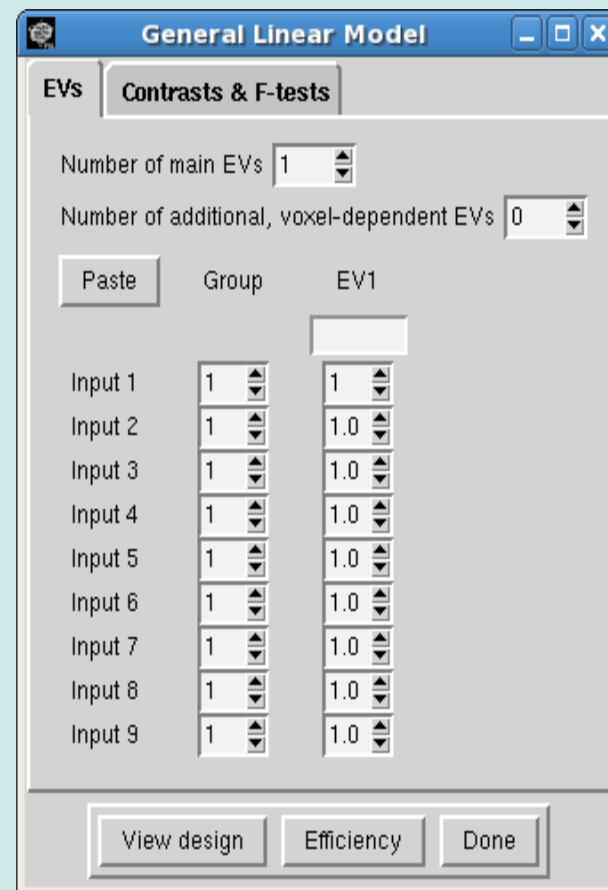
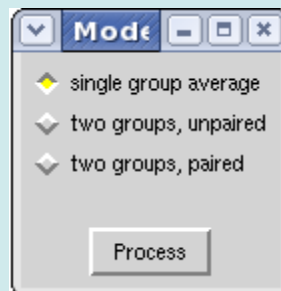
- ❑ The second stage of FLAME increases accuracy slightly over the first stage, but is quite a lot slower (typically 45-200 minutes).
- ❑ **FLAME 1+2** is most significant in a highest-level analysis when you have a small number of subjects (say <10).

FSL: Group Analysis

Model Setup

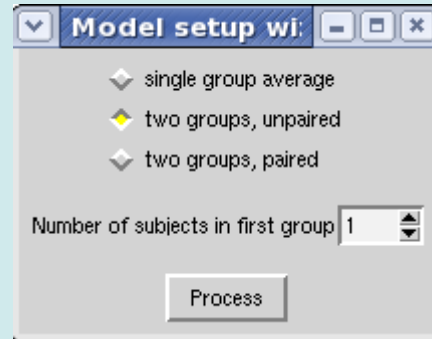
EVs Tab

- Only 1 group of subjects
- Want to see motor activity for all subjects, so set EV to 1 for each input



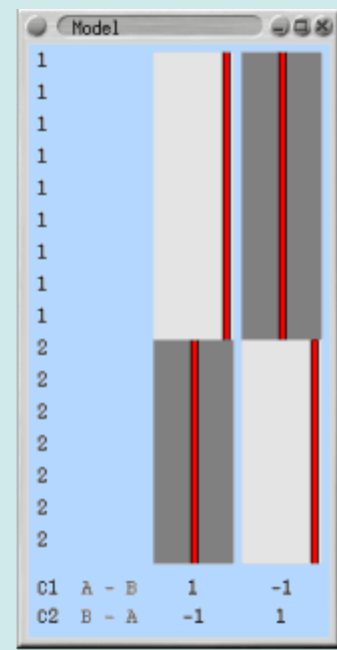
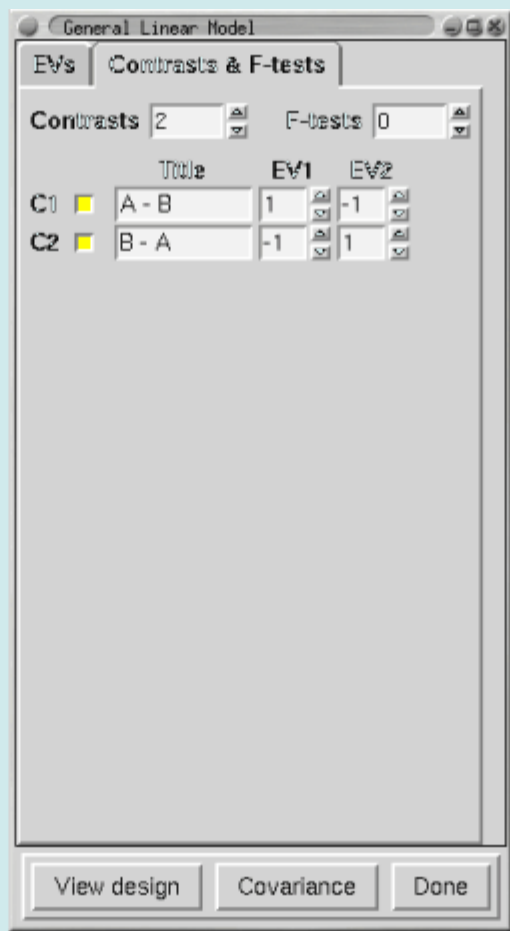
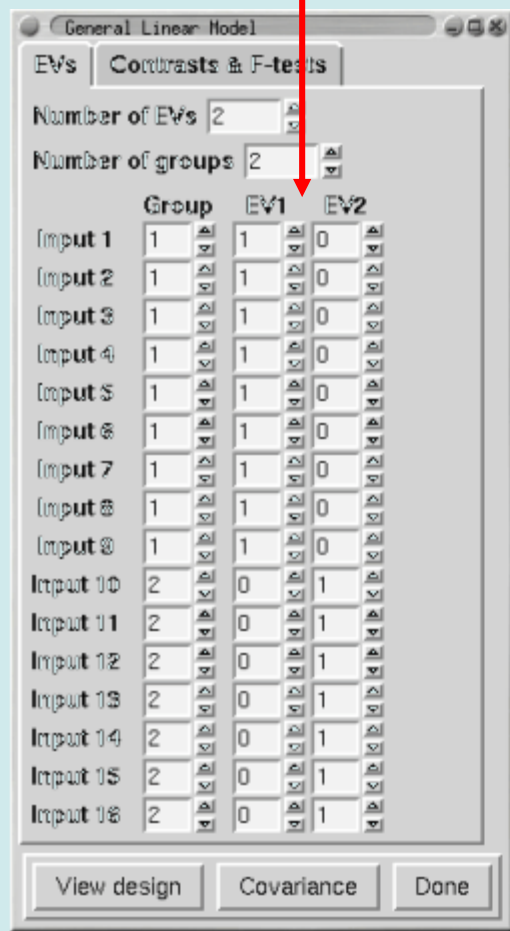
Group Effects

EV1 = include in group1
 EV2 = include in group2



group1

group2



Group Effects

EV1 = include in condition1
 EV2 = include in condition2

General Linear Model

EVs | Contrasts & F-tests

Number of EVs 2

Number of groups 2

	Group	EV1	EV2
Input 1	1	1	0
Input 2	1	1	0
Input 3	1	1	0
Input 4	1	1	0
Input 5	1	1	0
Input 6	1	1	0
Input 7	1	1	0
Input 8	1	1	0
Input 9	1	1	0
Input 10	2	0	1
Input 11	2	0	1
Input 12	2	0	1
Input 13	2	0	1
Input 14	2	0	1
Input 15	2	0	1
Input 16	2	0	1

condition1 (bracketed for inputs 1-9)
 condition2 (bracketed for inputs 10-16)

View design | Covariance | Done

Model

A > B s1

C1	condition A > B	1	0
C2	condition B > A	-1	0

Model

- ◆ single group average
- ◆ two groups, unpaired
- ◆ two groups, paired

Process

General Linear Model

EVs | Contrasts & F-tests

Contrasts 2 | F-tests 0

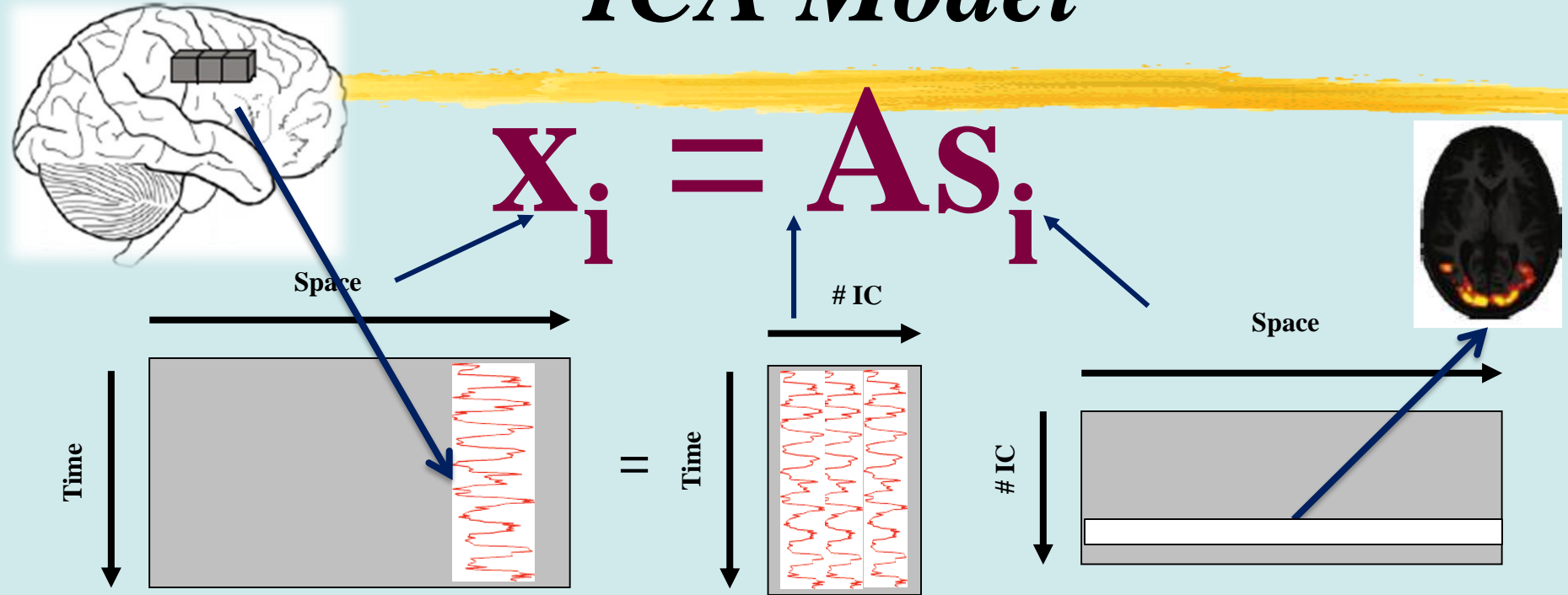
	Title	EV1	EV2
C1	A - B	1	-1
C2	B - A	-1	1

View design | Covariance | Done

Model

C1	A - B	1	-1
C2	B - A	-1	1

ICA Model



FMRI Data

$P \times N$

P= Num. of Scans

N= Num. of Voxels

Columns are Vectors of Observations

Mixing Matrix

$P \times K$

P= Num. of Scans

K= Num. of IC

No Statistical Constraint
(Possibly correlated)

Spatial Maps

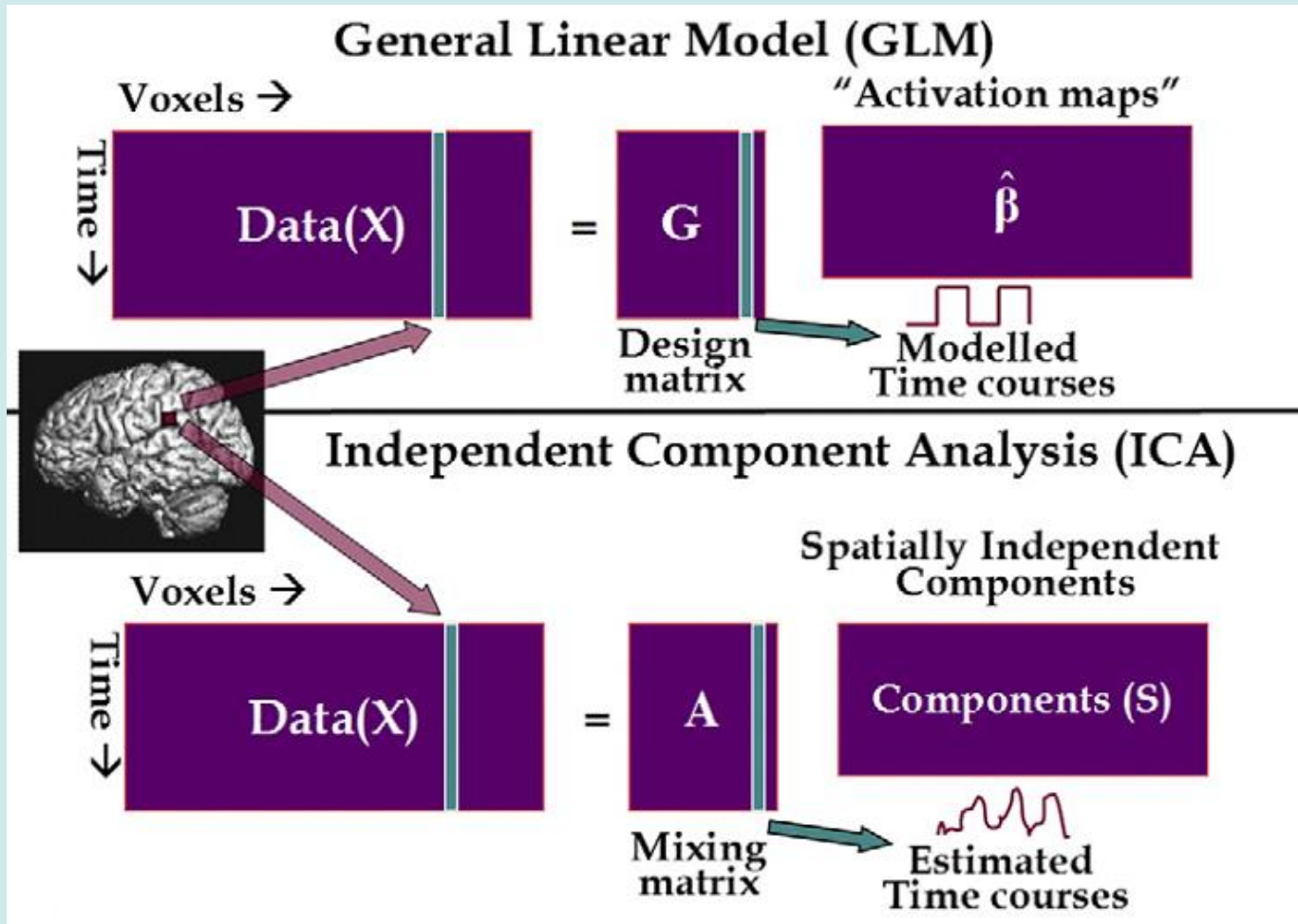
$K \times N$

K= Num. of IC

N=Num. of Voxels

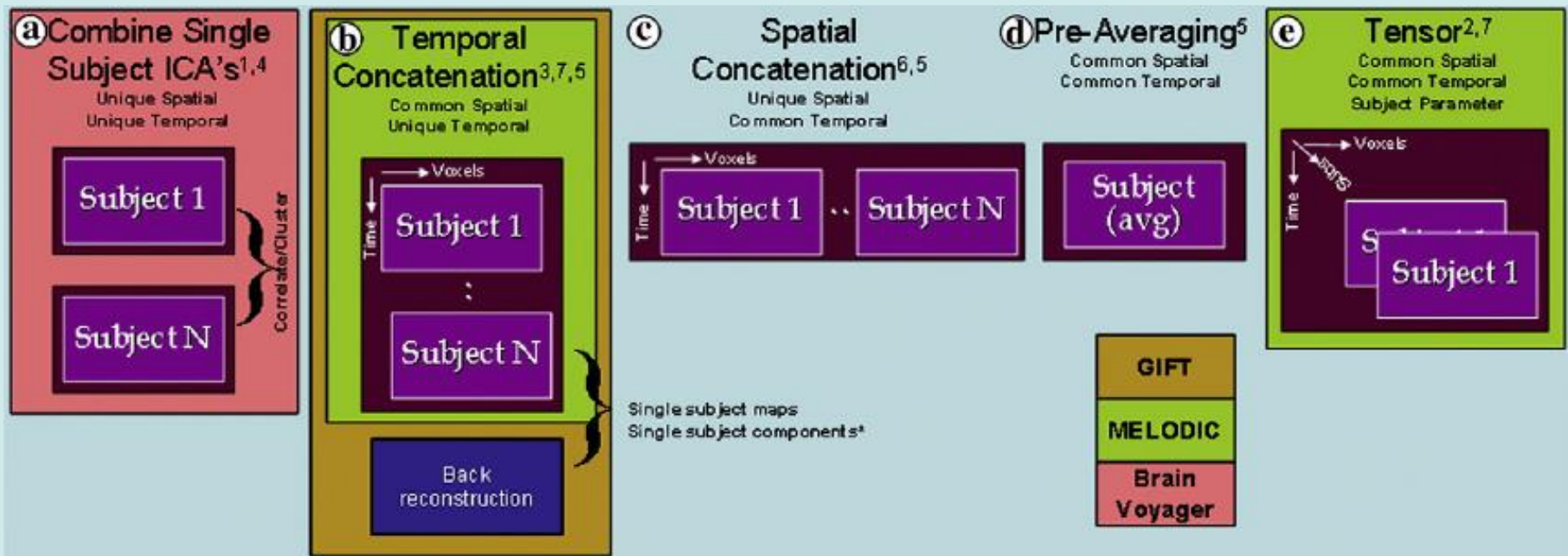
Statistically Independent
Non-Gaussian Sources
(including Noise)

GLM vs. ICA

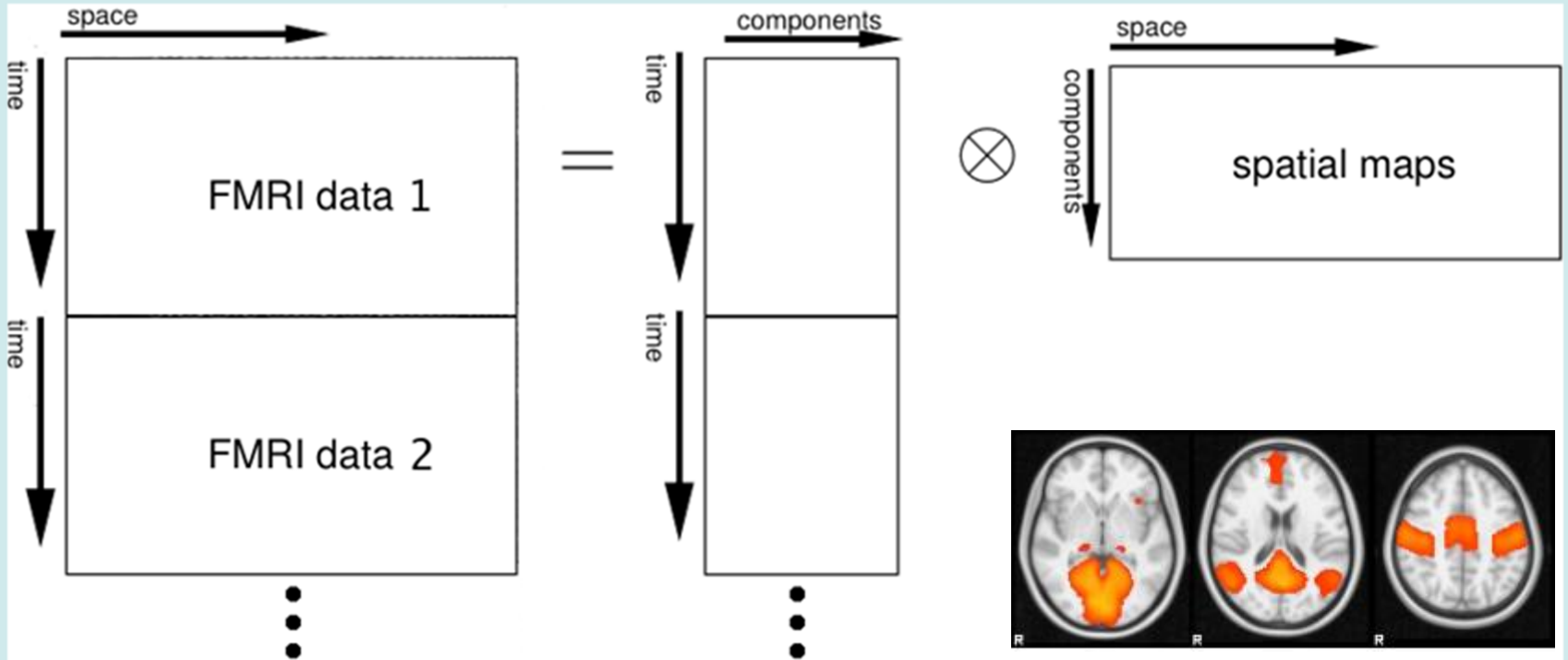


Group ICA models

Each Single subject analysis yields a different set of ICs

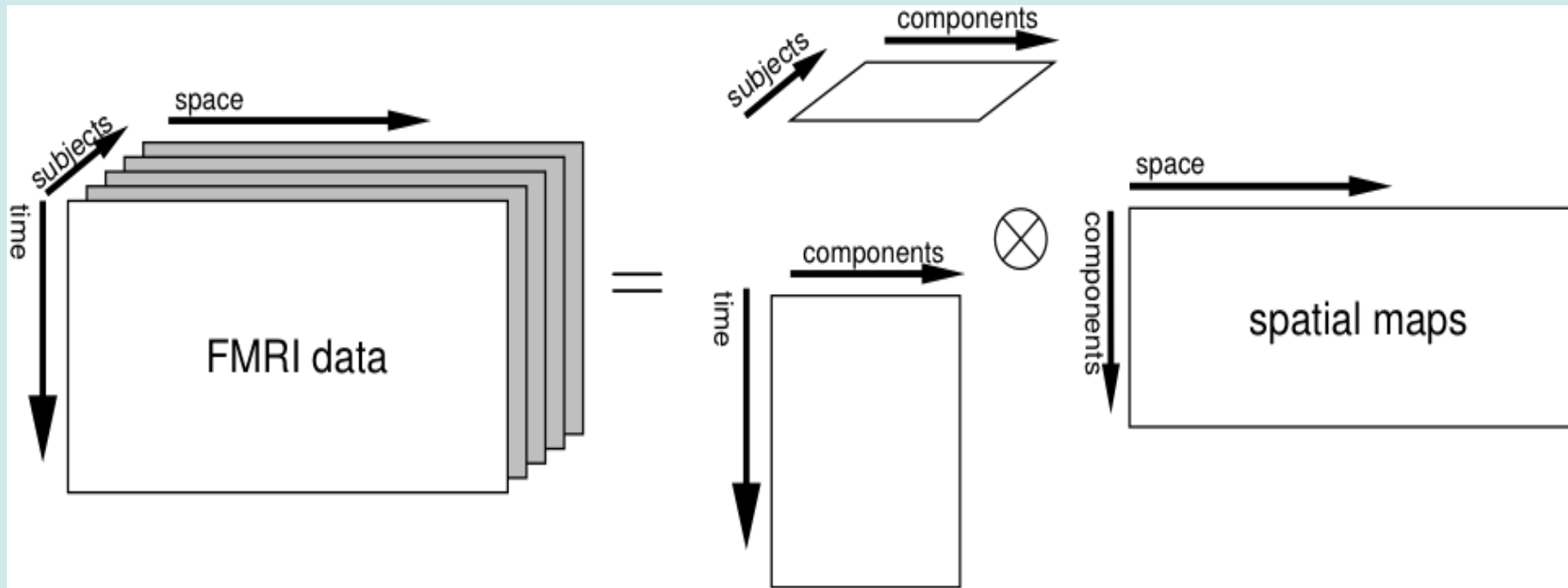


Temporal Concatenation



- ✓ This approach does not assume that the temporal response pattern is the same across the population.

Tensor ICA



- ✓ It is recommended to use this approach for data where the stimulus paradigm is consistent between session/subjects. Tensor-ICA assumes that the temporal response pattern is the same across the population and provides a single decomposition for all original data sets.

Practical



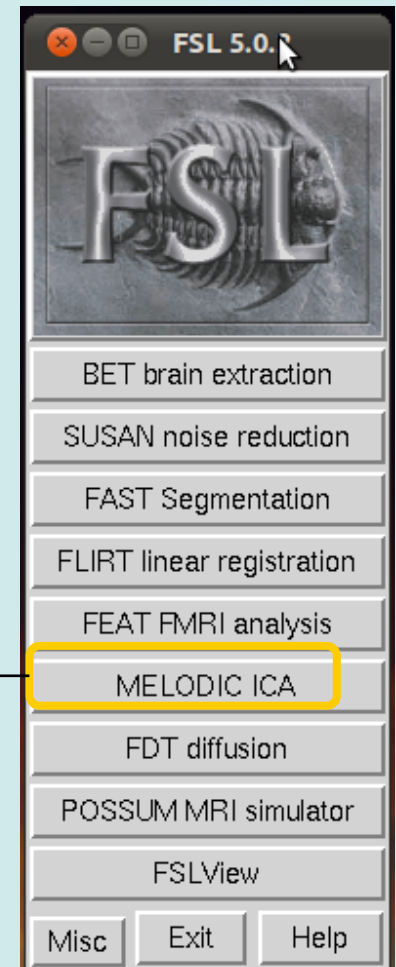
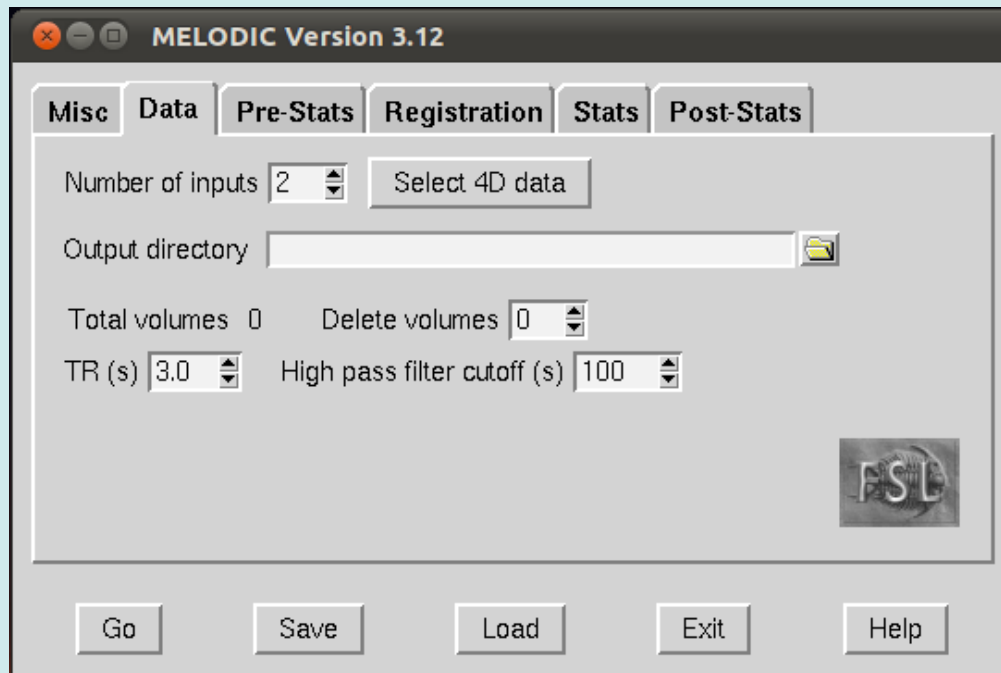
FSL

MELODIC

**Multivariate Exploratory Linear Optimized
Decomposition into Independent Components**

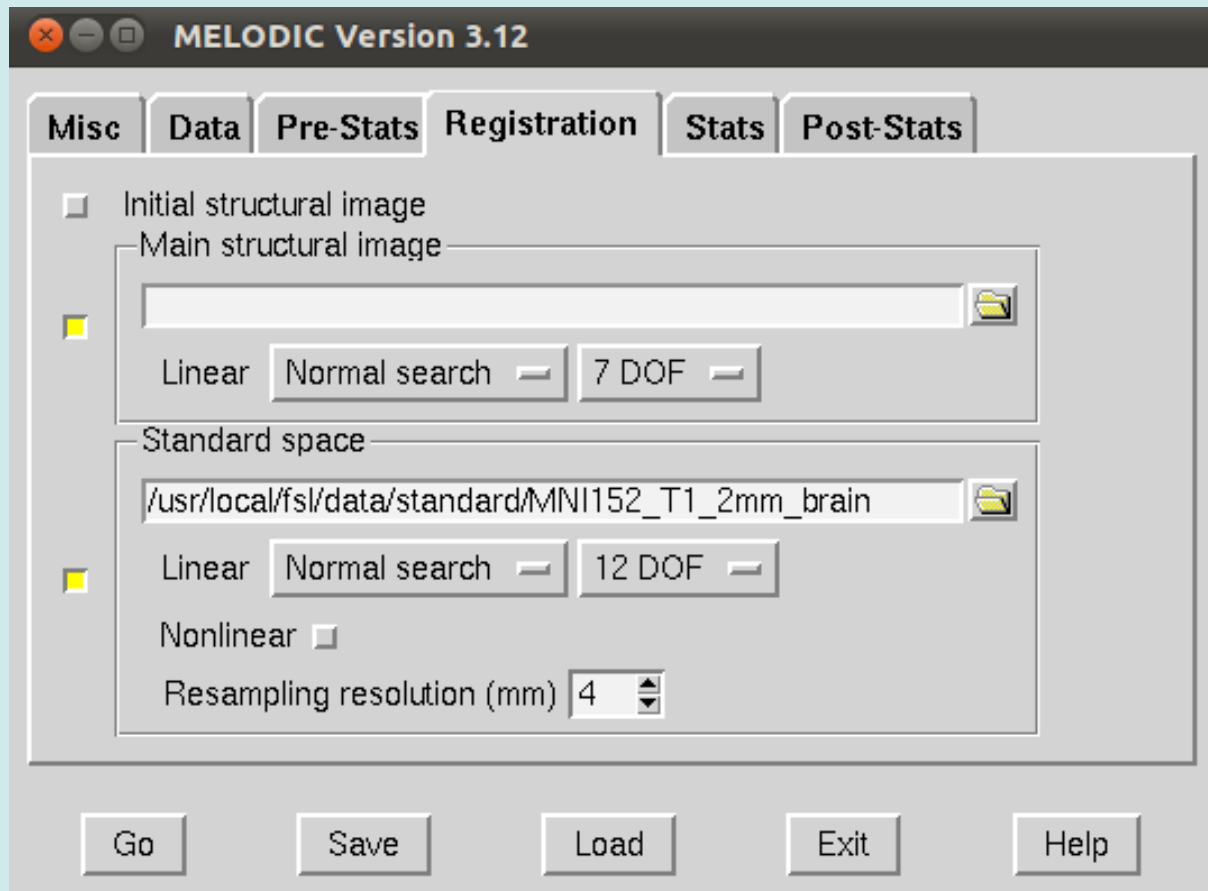
Practical-set

- To call the MELODIC GUI, type **Melodic** in a terminal.

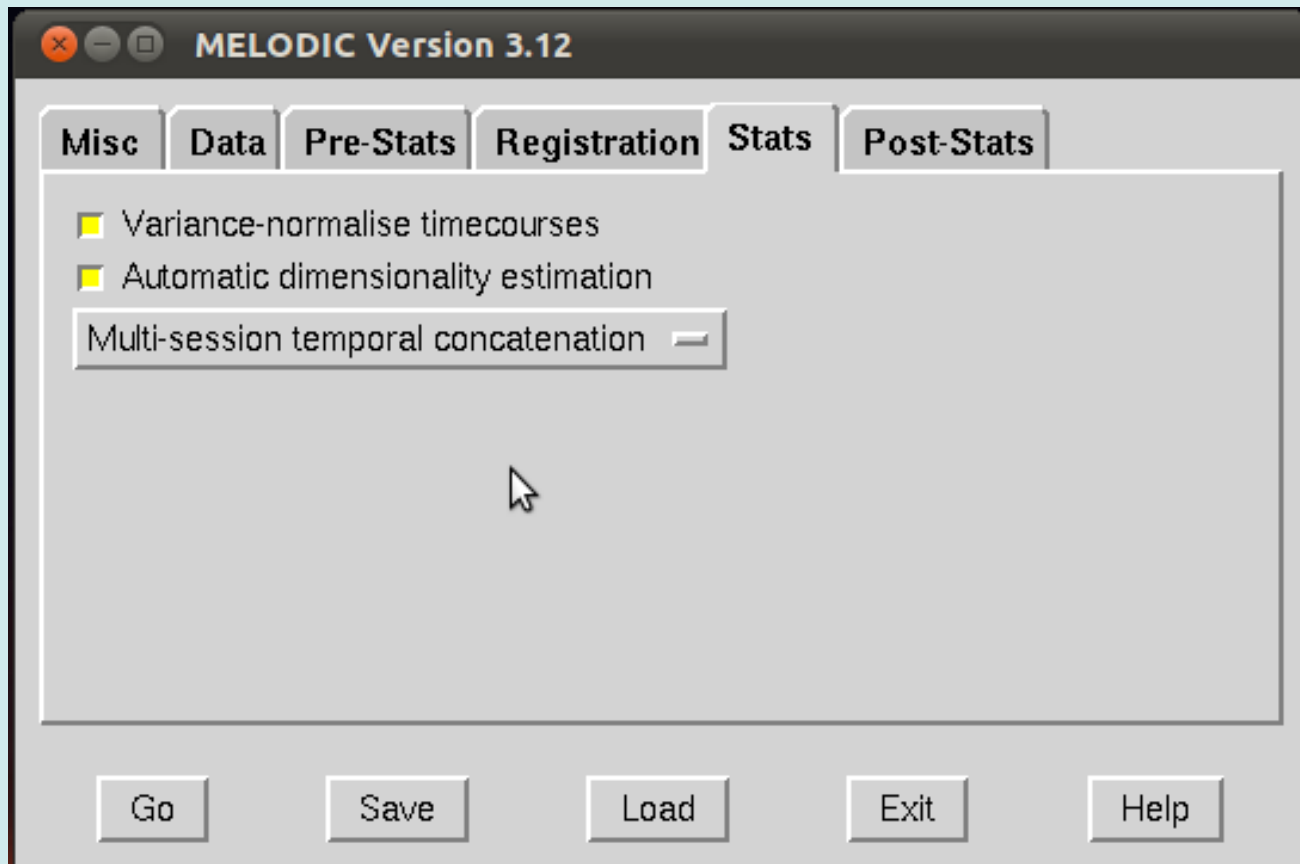


Press the **MELODIC** button ←

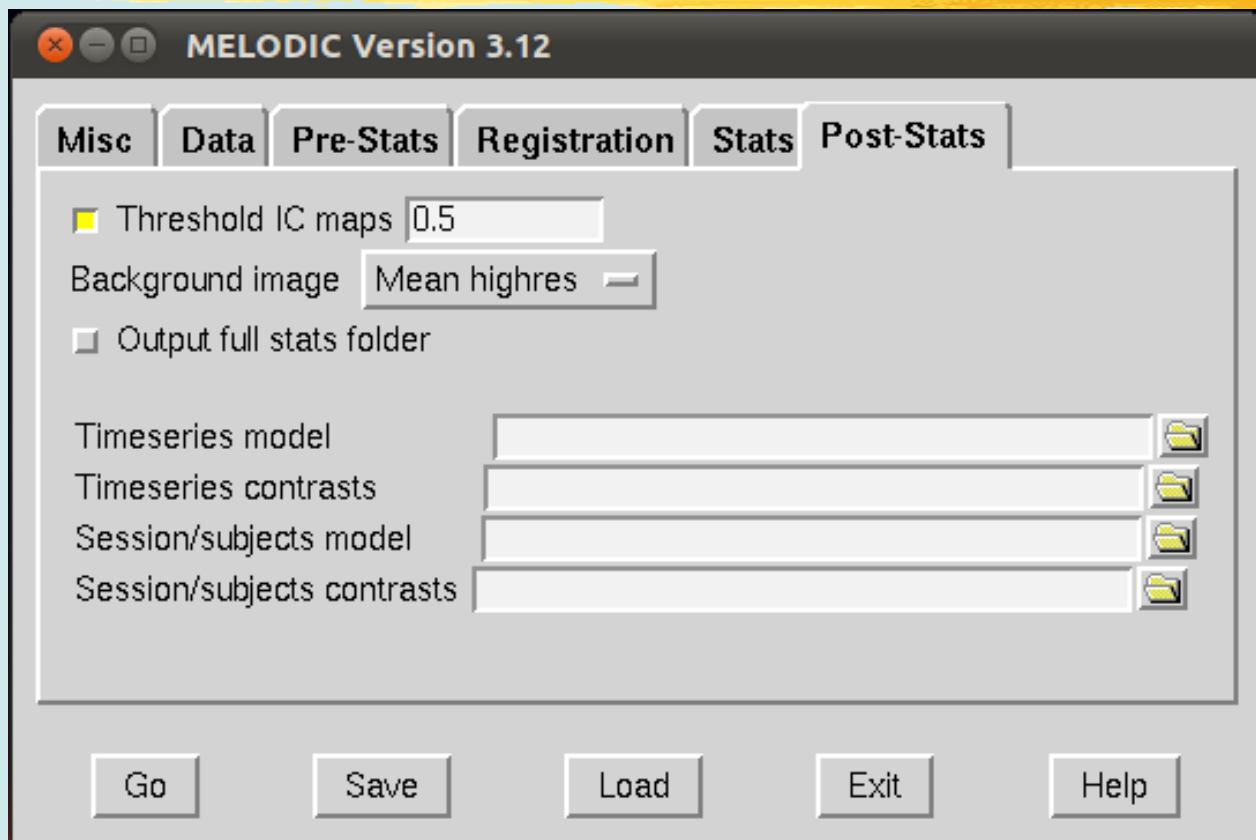
Practical-set



Practical-set



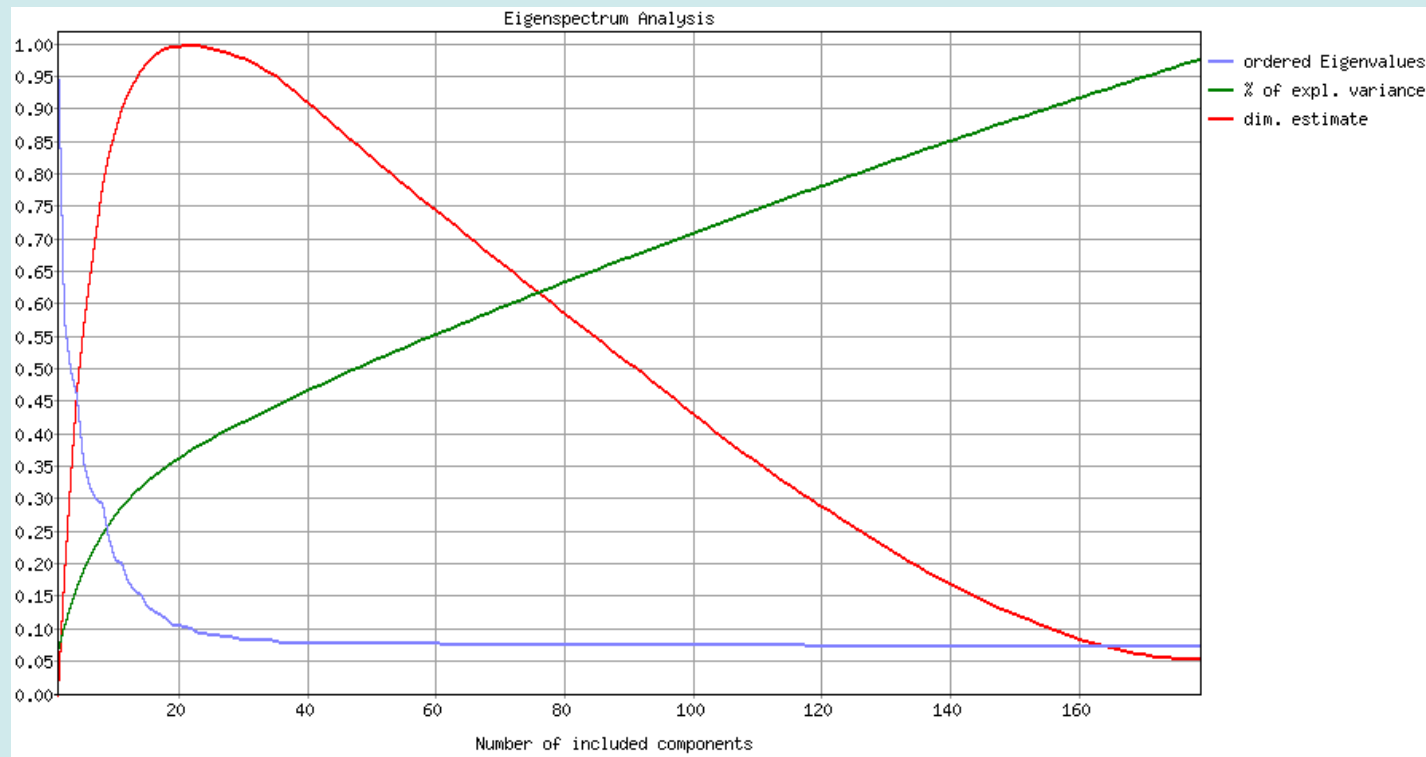
Practical-set



If a temporal design was specified in the **Post-Stats** section then the time series plot will also contain a plot of the total model fit

Practical-output

ICs Estimation



Practical-output

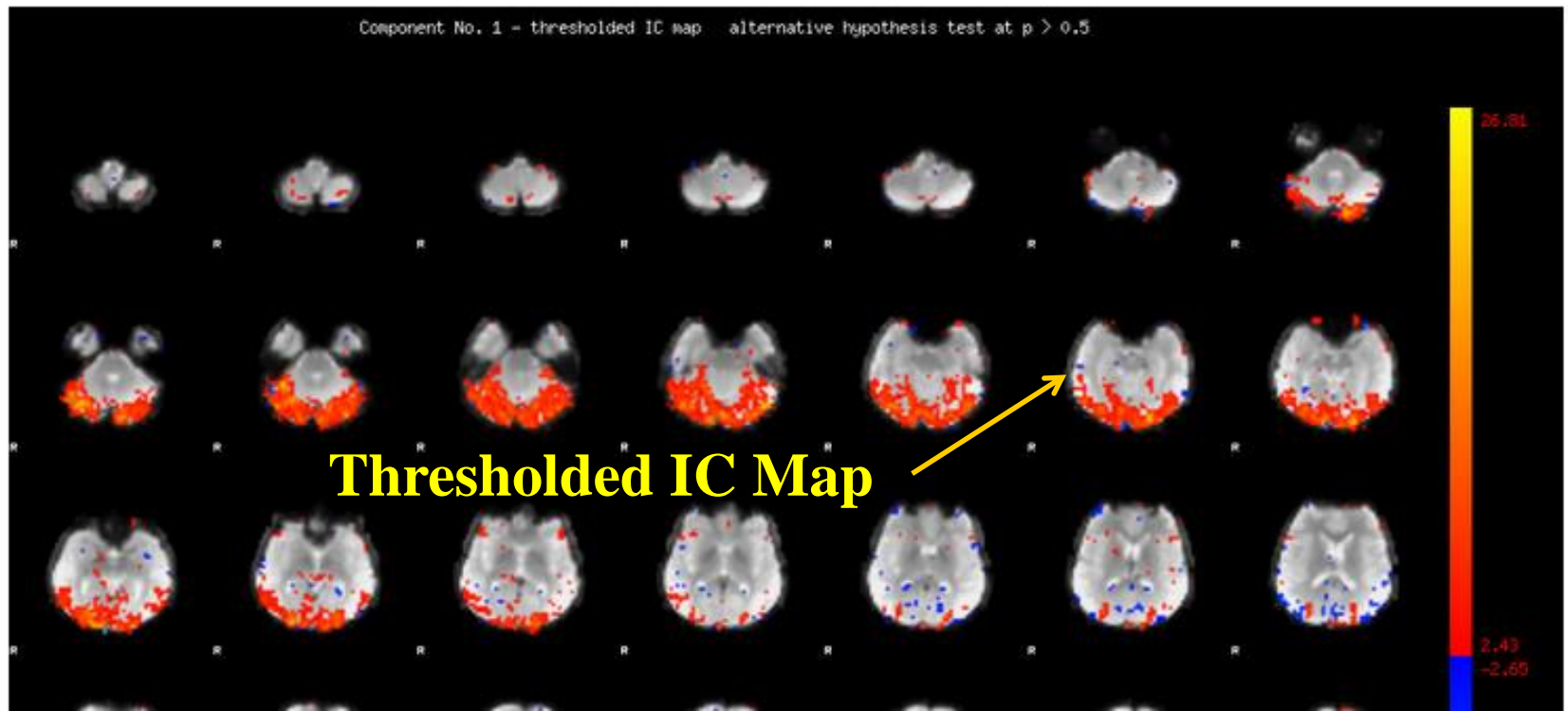
Main - Components: [1](#) [2](#) [3](#) [4](#) [5](#) [6](#) [7](#) [8](#) [9](#) [10](#) [11](#) [12](#) [13](#) [14](#) [15](#) [16](#) [17](#) [18](#) [19](#) [20](#) [21](#) [22](#)

≤-≥

MELODIC Component 1

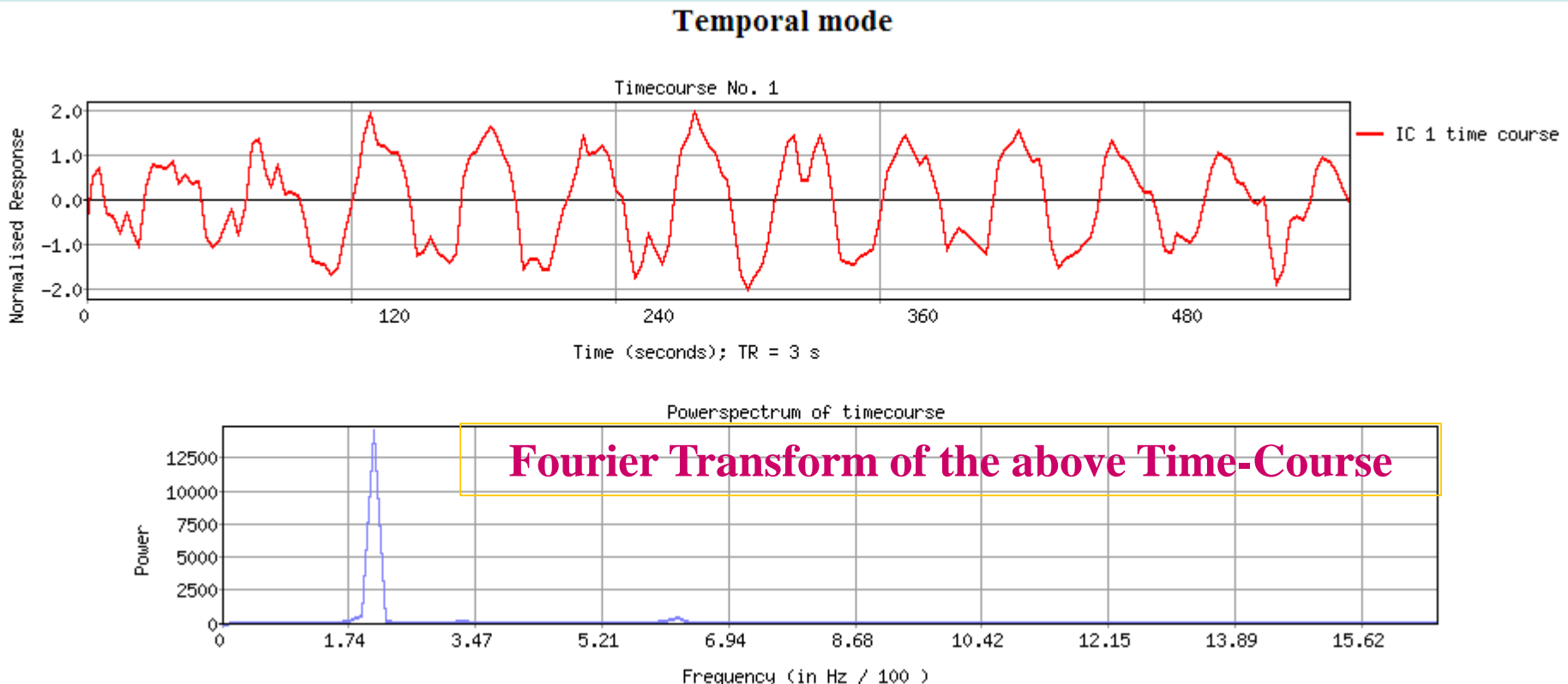
22
Components

8.72 % of explained variance; 3.28 % of total variance



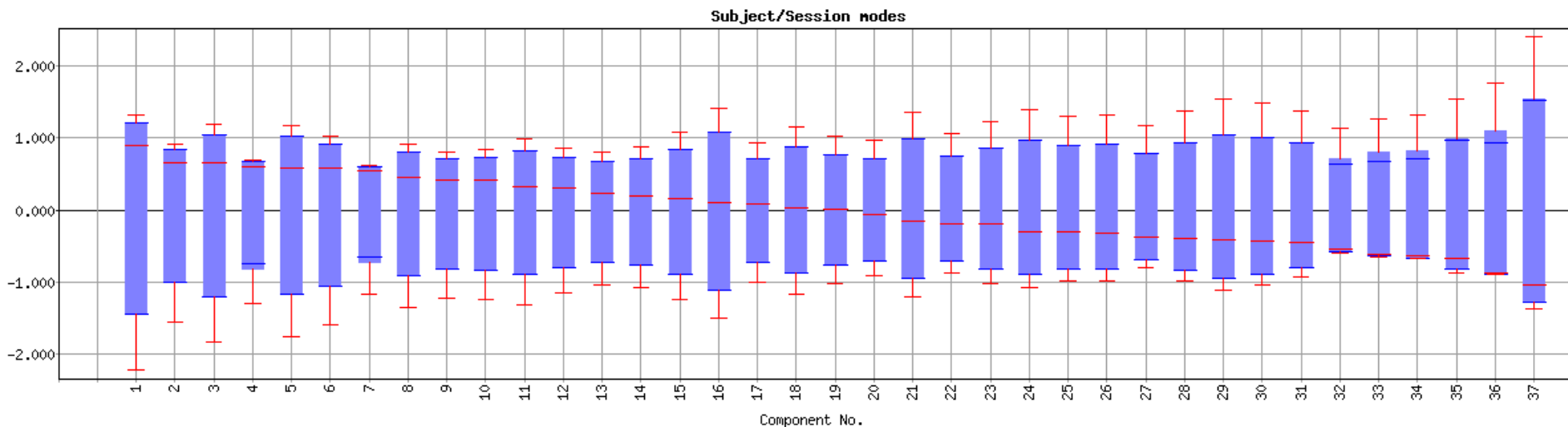
Practical-output

✓ For each component the final mixing matrix `melodic_mix` contains the temporal response of all different data sets concatenated into a single column vector



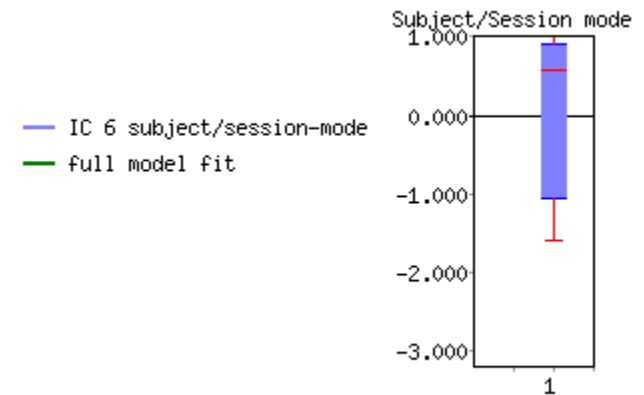
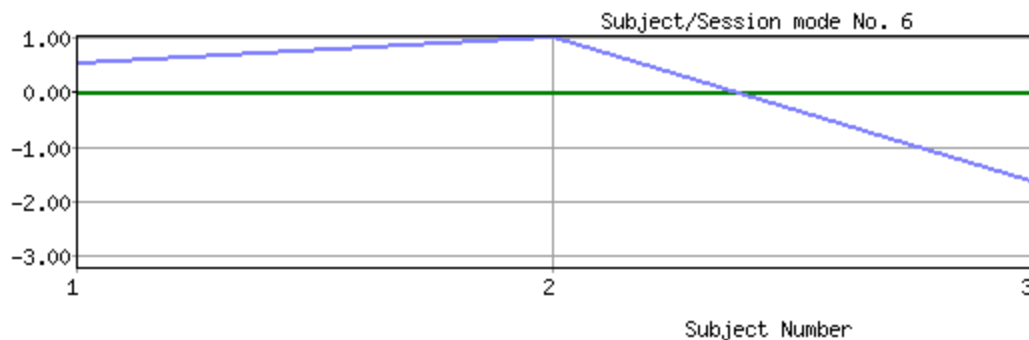
Practical-output

The TICA Subject/Session modes plot contains a series of boxplots, one per estimated component. The boxplots themselves show the distribution of the relative effect sizes per subject for a given component. The components are ordered in decreasing order of median effect size.



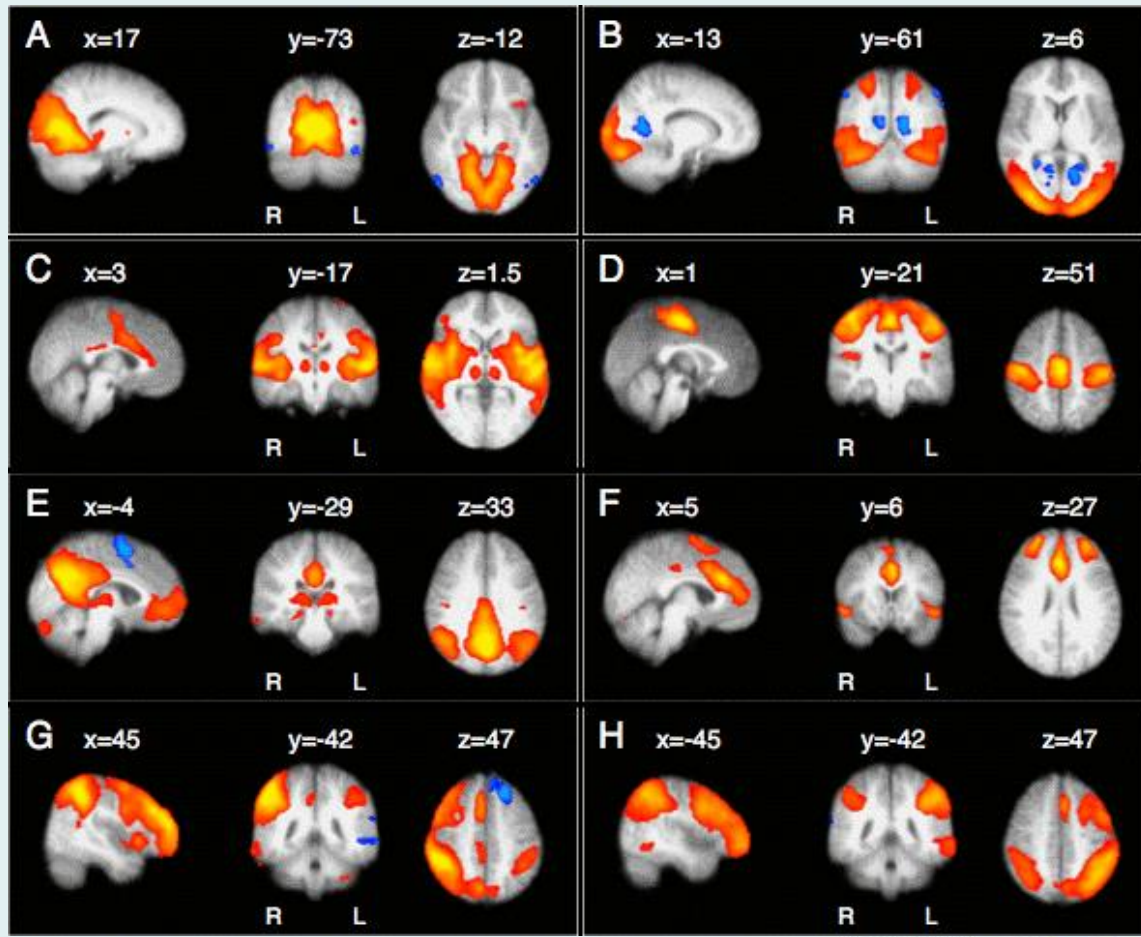
Practical-output

Each component has three main parts: spatial map, time course and subject/session effect sizes.



ICA on resting/null data

- Group study
- (10 subjects under rest): consistent resting fluctuations
- medial (A) visual cortex
- lateral (B) visual cortex
- auditory (C)
- sensori-motor (D)
- visuo-spatial system (E)
- executive control (F)
- dorsal visual stream (G,H)



Thanks for Your Attention

