

GLM Analysis using FSL (FMRI Software Library) using (FEAT) (fMRI Expert Analysis Tools)

Some slides from Dr. Scott Huettel &

Chris Petty and FSL website: <http://www.fmrib.ox.ac.uk/fsl/>

Data Analysis: Main Components

- **Within-subjects**
 - *Preprocessing*: removal/minimization of task-independent variability
 - General linear model
 - Model specification: creating and evaluating a model for brain function
 - Model evaluation: testing specific hypotheses
- **Across-subjects**
 - Aggregation of data to increase experimental power
 - Inter-group comparisons
 - Testing of parametric effects

FMRIB Software Library (FSL)

- Created by researchers at the FMRIB in Oxford
- Comprises many tools for analysis of:
 - fMRI data
 - Structural MRI data
 - Diffusion Tensor Imaging data
- Runs natively on Linux/Unix or Mac
 - Runs on Windows with virtual machine (vm-ware)
- Can be run via GUI or via scripts
- Citation
 - S.M. Smith, et al. Advances in functional and structural MR image analysis and implementation as FSL. NeuroImage, 23(S1):208-219, 2004

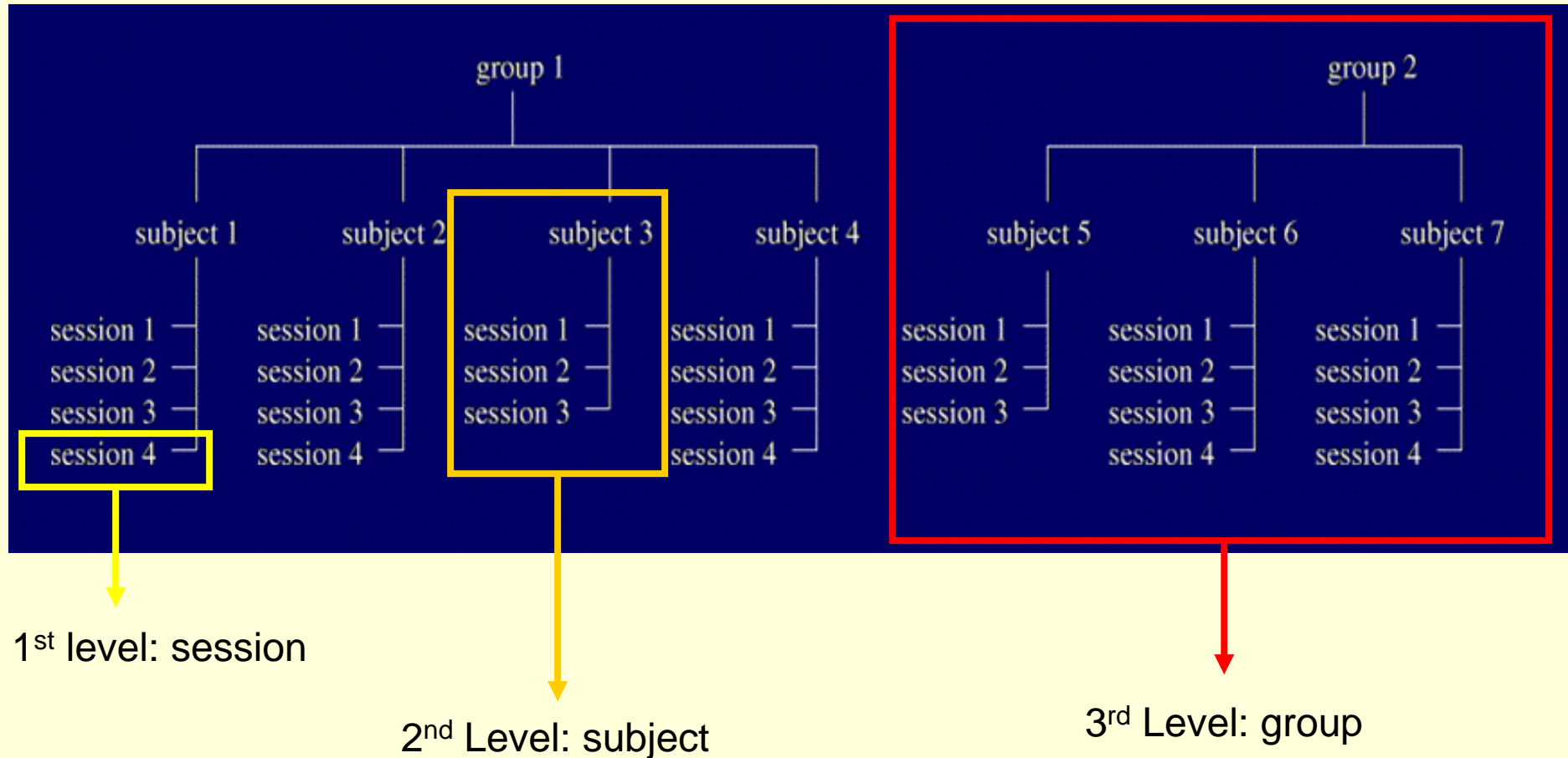
FMRI Software Library (FSL)

- FEAT: fMRI analysis (subject and group levels), preprocessing
 - MCFLIRT: motion correction
 - FLIRT: registration
 - BET: Brain Extraction Tool
- MELODIC: independent components analysis, for model-free analyses and noise removal
- FSL View: displaying data

Preparing your data for FSL

- Convert functional and anatomical data into correct format (nifti)
- Generate orientation matrix for registration
- Generate “3 column files” for behavior
 - Text files with three columns: (1) When did something happen, (2) how long did it take, (3) how much should it be weighted

Levels of FSL analysis



GLM model

		Variables							
		Conditions				Responses			
		x_1	x_2	...	x_p	y_1	y_2	...	y_q
Observations	1	—	—	...	—	—	—	...	—
	2	—	—	...	—	—	—	...	—
	...								
	n	—	—	...	—	—	—	...	—

- The rows (observations) may refer to **observations at different times**, to different subjects, or to different brain locations
- The columns (**x_j variables**) **describe the condition in which the observation was made** (it may describe particular **response condition**, groups, or subjects)
- **y_k are observed measurements, for example activations measured different locations.**
- The basic goal of the analysis is to find a way to describe the y_k as **functions of the x_j s**

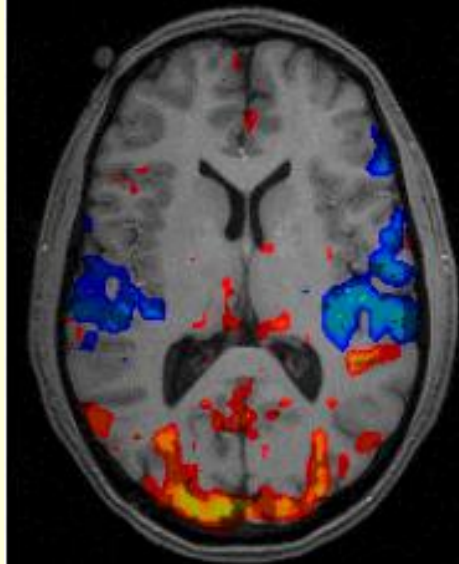
t	Stim	x_0	x_1	x_2	y_1	y_2	y_3	y_4
1	0	1	0	0	9.45	13.25	11.23	16.48
2	0	1	0	0	9.86	10.26	11.13	13.62
3	0	1	0	0	10.17	13.90	11.74	15.13
4	1	1	1	0	12.97	11.76	10.97	16.63
5	1	1	1	0	11.31	13.83	10.65	16.42
6	1	1	1	0	12.70	10.96	10.12	17.85
7	0	1	0	0	11.38	12.95	11.15	13.65
8	0	1	0	0	10.29	12.12	11.56	15.96
9	0	1	0	0	11.82	10.29	12.73	14.27
10	2	1	0	1	10.27	12.45	14.15	19.39
11	2	1	0	1	11.54	13.25	14.33	18.49
12	2	1	0	1	8.93	8.93	14.32	16.73
13	0	1	0	0	11.01	11.69	10.40	17.31
14	0	1	0	0	8.92	11.52	10.87	14.62
15	0	1	0	0	11.04	12.85	11.09	14.00
16	2	1	0	1	9.45	11.65	15.50	17.54
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots

Data from two stimuli presented at different times. Responses are recorded from four locations.

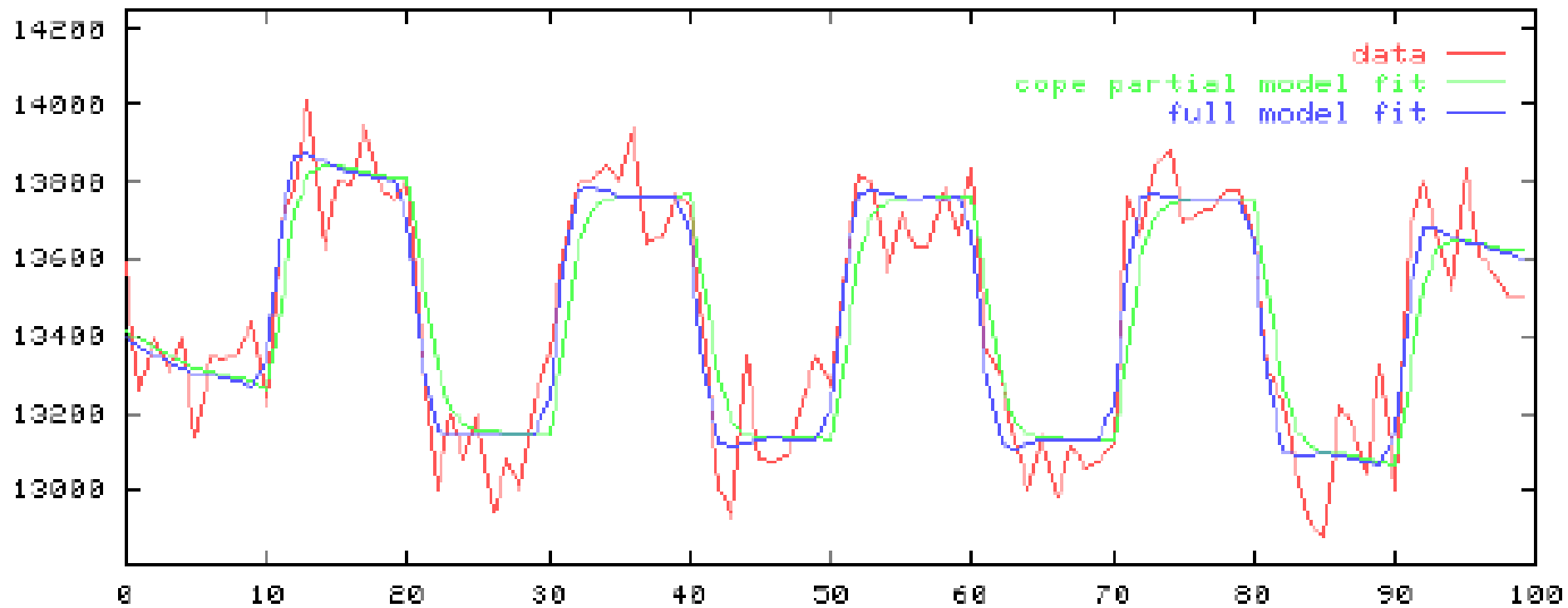
GLM concept

$$Y_j = x_{j1}\beta_1 + \dots + x_{jl}\beta_l + \dots + x_{jL}\beta_L + \epsilon_j$$

- GLM explains the activation measure (response variable) Y_j in terms of a linear combination of different stimuli (EV: explanatory variables) plus error term.
- Y_j is activation in a particular voxel at **observation (time) j**
- X_{jl} is explanatory variable (stimulus) for **various stimuli, groups, or subjects $l=1, \dots, L$**
- β_l is the Parameter Estimate (fitted value for each stimulus)
- ϵ_j is error for each observation



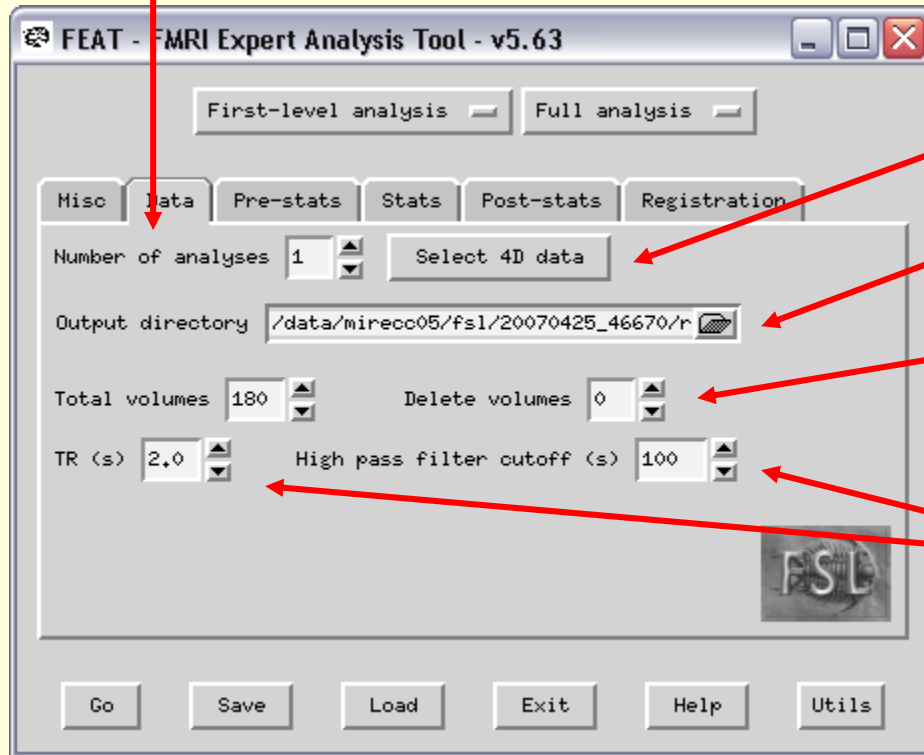
Contrast 1 [1 0] : max Z stat of 13.779 at voxel (32,7,7)



1st Level: Data Parameters

How many runs:

-Typically do runs individually



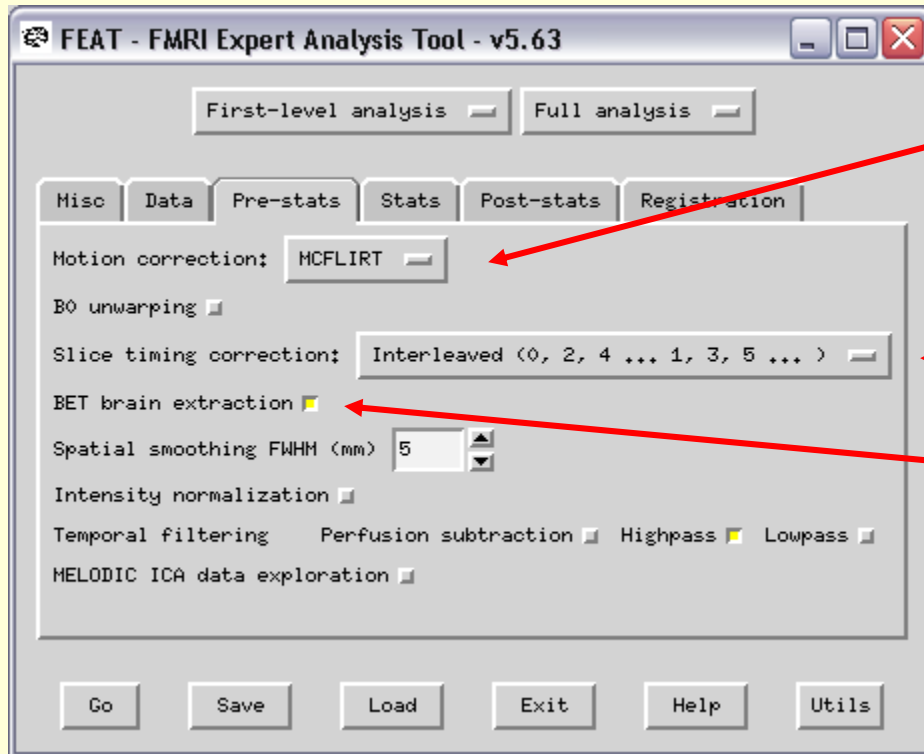
Select the 4-d data for your run

Where will the data be saved?

Are disdaqs already thrown out?

Self explanatory

1st Level: Preprocessing

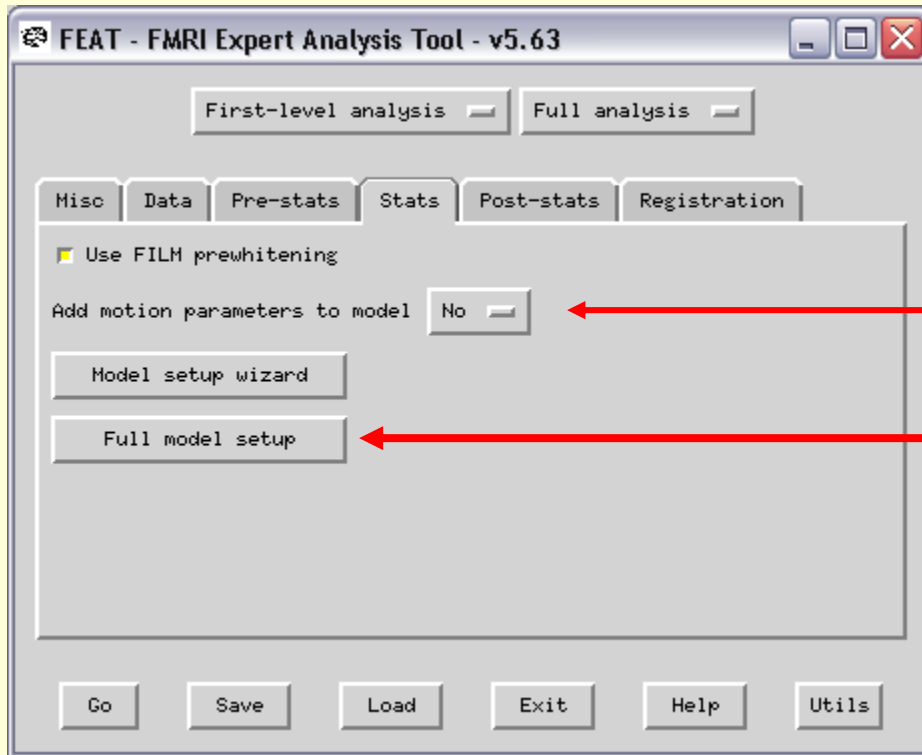


Correct for motion?

Interleaved, ascending, descending, custom order

Brain only voxels

1st Level: Statistics



Want to model motion?

This is the only hard part
(i.e., model specification)

1st Level: General Linear Model

How many explanatory variables?

General Linear Model

EVs | Contrasts & F-tests

Number of original EVs: 9

1 2 3 4 5 6 7 8 9

EV name: memoran

Basic shape: Custom (3 column format)

Filename: /data/wirecc05/fsl/20070425_46

Convolution: Gamma

Phase (s): 0

Stddev (s): 3

Mean lag (s): 6

Orthogonalise

Add temporal derivative

Apply temporal filtering

Select your 3 column file

Shape of your correlation waveform

1st Level: Contrasts

How many zmaps? (each condition + contrasts)

General Linear Model

EVs | Contrasts & F-tests

Setup contrasts & F-tests for

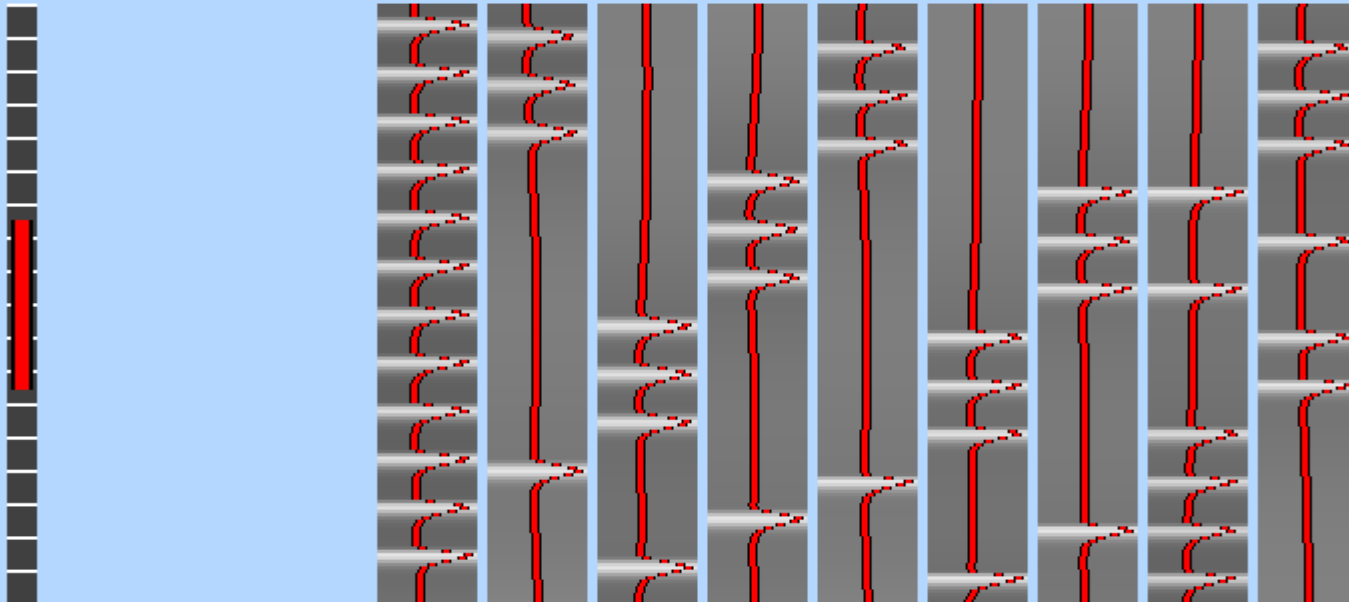
Contrasts F-tests

Paste	Title	EV1	EV2	EV3	EV4	EV5	EV6	EV7	EV8	EV9	F1
OC1	memoranda	1	0	0	0	0	0	0	0	0	
OC2	emo_dist	0	1	0	0	0	0	0	0	0	
OC3	neu_dist	0	0	1	0	0	0	0	0	0	
OC4	scr_dist	0	0	0	1	0	0	0	0	0	
OC5	probe	0	0	0	0	1	1	1	0	0	
OC6	emo_probe	0	0	0	0	1	0	0	0	0	
OC7	neu_probe	0	0	0	0	0	1	0	0	0	
OC8	scr_probe	0	0	0	0	0	0	1	0	0	
OC9	probe_corr	0	0	0	0	0	0	0	1	0	

What is a contrast?

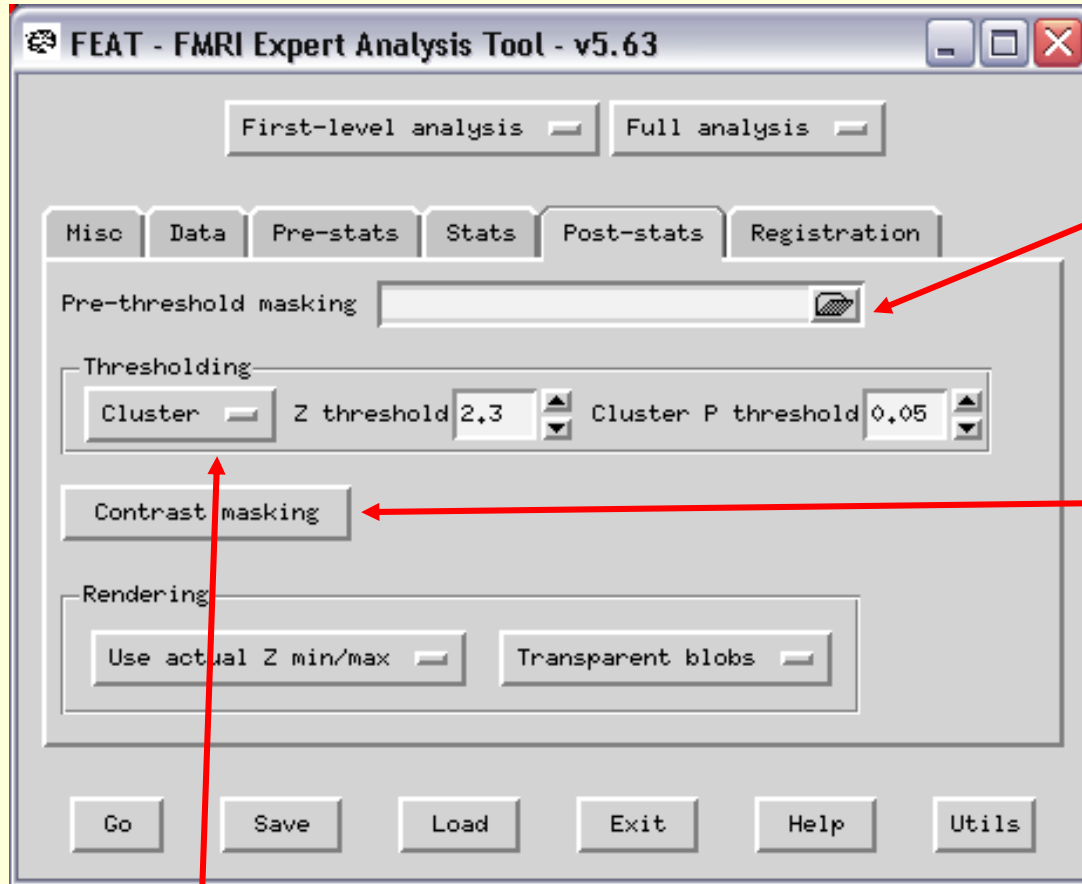
- Key concept in fMRI analysis
 - Remember: fMRI provides *relative* measures
- We contrast terms in our analysis model to evaluate whether they modulate the brain differently
 - Faces > Houses
 - “Tapping right hand” > “Tapping left hand”
- Analogous to subtractive techniques in psychology

Model



	memoranda	emo_dist	neu_dist	scr_dist	emo_probe	neu_probe	scr_probe	probe_corr	probe_incorr	F
C1 memoranda	1	0	0	0	0	0	0	0	0	<input checked="" type="checkbox"/>
C2 emo_dist	0	1	0	0	0	0	0	0	0	<input checked="" type="checkbox"/>
C3 neu_dist	0	0	1	0	0	0	0	0	0	<input checked="" type="checkbox"/>
C4 scr_dist	0	0	0	1	0	0	0	0	0	<input checked="" type="checkbox"/>
C5 probe	0	0	0	0	1	1	1	0	0	<input checked="" type="checkbox"/>
C6 emo_probe	0	0	0	0	1	0	0	0	0	<input checked="" type="checkbox"/>
C7 neu_probe	0	0	0	0	0	1	0	0	0	<input type="checkbox"/>
C8 scr_probe	0	0	0	0	0	0	1	0	0	<input type="checkbox"/>
C9 probe_corr	0	0	0	0	0	0	0	1	0	<input type="checkbox"/>
C10 probe_incorr	0	0	0	0	0	0	0	0	1	<input type="checkbox"/>
C11 emo-d > neu-d	0	1	-1	0	0	0	0	0	0	<input type="checkbox"/>
C12 emo-d > scr-d	0	1	0	-1	0	0	0	0	0	<input type="checkbox"/>
C13 emo-d > neu-d + scr-d	0	2	-1	-1	0	0	0	0	0	<input type="checkbox"/>
C14 neu-d > emo-d	0	-1	1	0	0	0	0	0	0	<input type="checkbox"/>
C15 neu-d > scr-d	0	0	1	-1	0	0	0	0	0	<input type="checkbox"/>
C16 scr-d > neu-d	0	0	-1	1	0	0	0	0	0	<input type="checkbox"/>
C17 scr-d > emo-d	0	-1	0	1	0	0	0	0	0	<input type="checkbox"/>
C18 emo-d > neu-d	0	0	0	0	1	1	0	0	0	<input type="checkbox"/>

1st Level: Thresholding



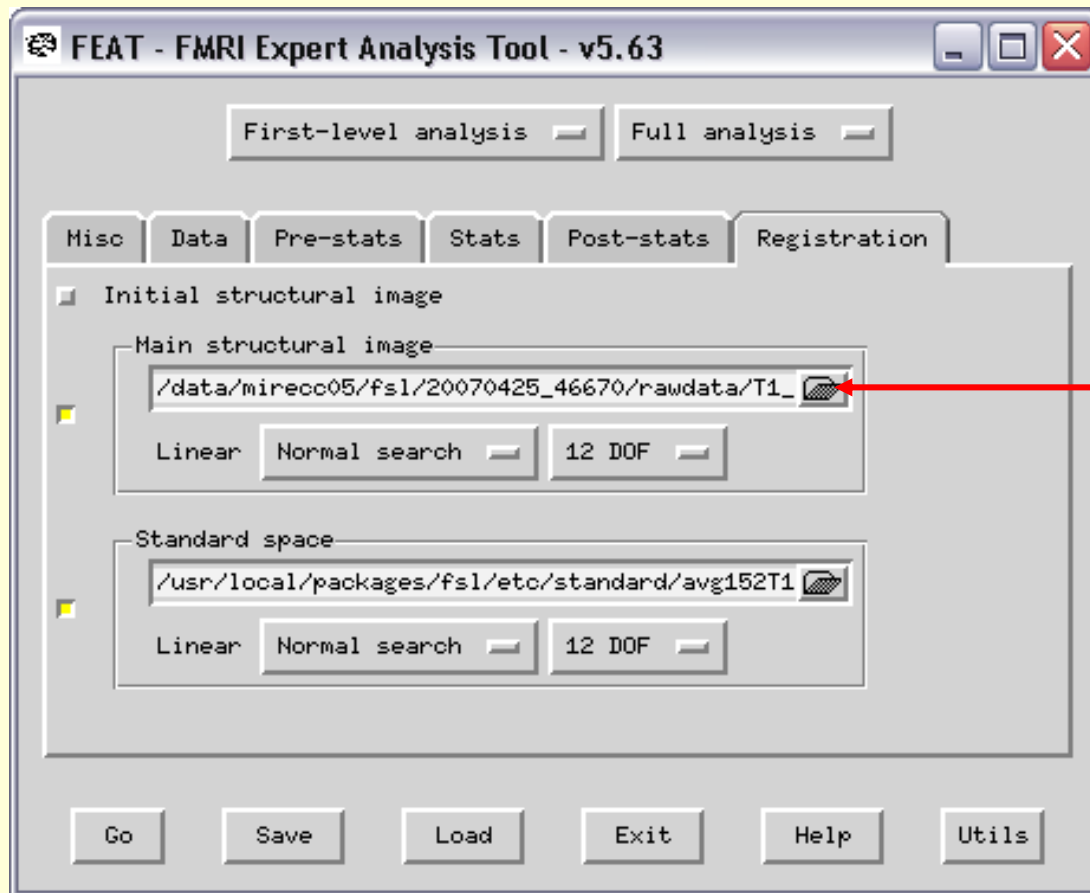
Mask by a bitmap of some specified region

You can mask one cope by another

If **Cluster**: Then each cluster's estimated significance level (from GRF-theory) is compared with the cluster probability threshold.

If **Voxel** thresholding is selected, GRF-theory-based maximum height thresholding is carried out, with thresholding at the level set, using one-tailed testing. This test is less overly-conservative than Bonferroni correction.

1st Level: Registration

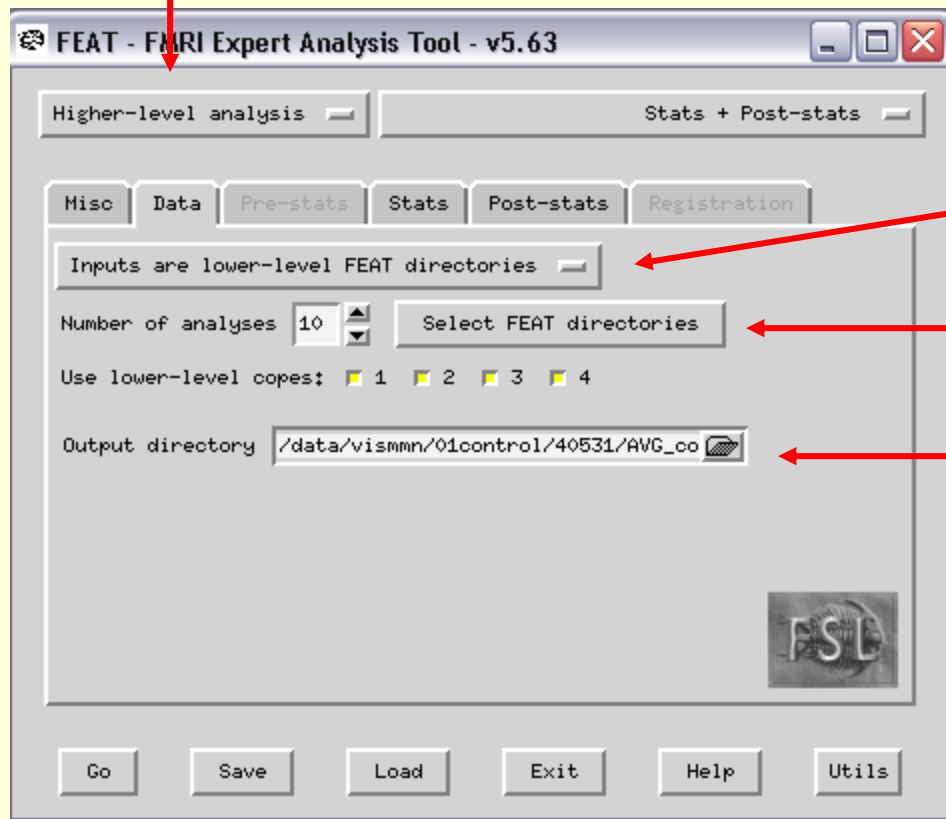


Your skull stripped anatomical

Click **GO** to run FEAT, if you need to insert the orientation matrix, then click **Save**

2nd Level: Within-subject, across-runs

Higher-level analysis – within subject



Inputs are your subject's run directories

of analyses = # of runs

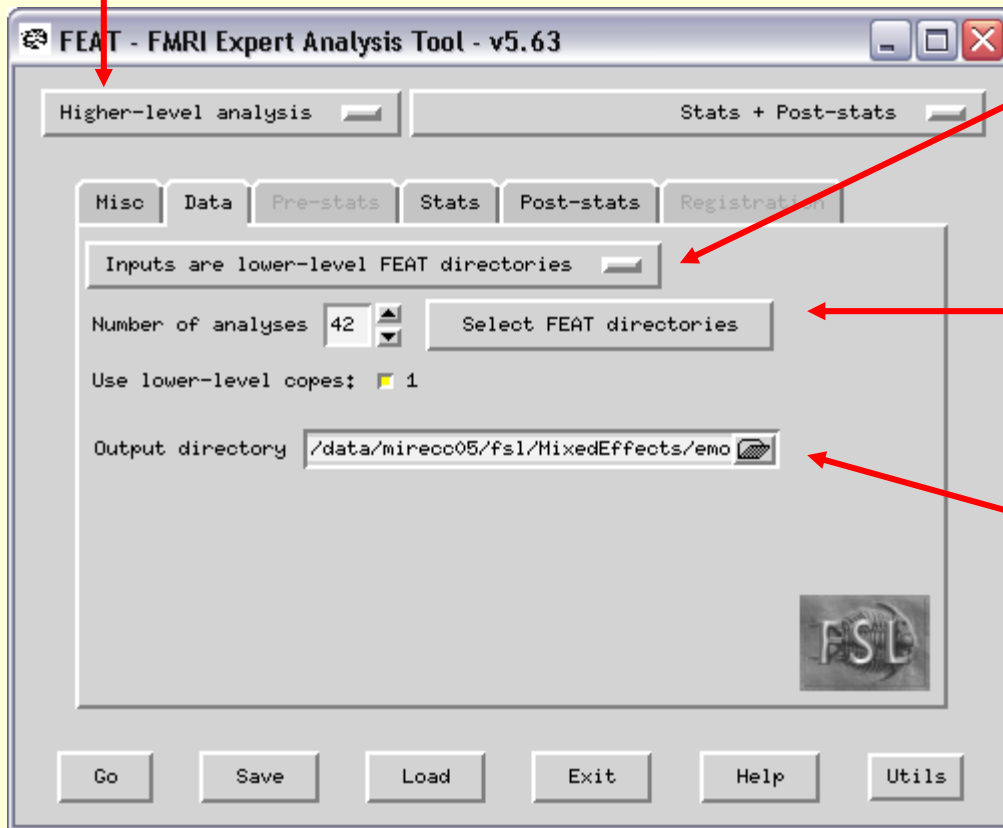
Where do you want your data saved?
-Individual conditions will be saved within
the output directory as separate copes

Error in Group Analysis

- **Fixed-Effects** variance is “the within-session across-time variance” estimated in first level analysis.
- **Random-Effects** variance is “the cross-session variance” in first level parameter estimates.
- In higher level (group analysis) FEAT uses **Mixed Effects** (FLAME= FMRIB Local Analysis of Mixed Effects), which is “the sum of fixed-Effects variance and Random-Effects variance.

3rd Level: Across Subjects

Higher level - Group



Inputs are now cope#.feat directories from the subject averages (2nd Level)

#analyses = #subjects included
- Select /subj#/AVG.gfeat/cope#.feat

Where to save results, what name?

This step needs to be done separately for every condition you are interested in viewing.

Paired Two-group differences: (Two-sample paired T-test)

We have a group of 8 subjects scanned under different conditions, A and B

General Linear Model

EVs Contrasts & F-tests

Number of EVs 8

Number of groups 1

	Group	EV1	EV2	EV3	EV4	EV5	EV6	EV7	EV8	EV9
Input 1	1	1	1	0	0	0	0	0	0	0
Input 2	1	1	0	1	0	0	0	0	0	0
Input 3	1	1	0	0	1	0	0	0	0	0
Input 4	1	1	0	0	0	1	0	0	0	0
Input 5	1	1	0	0	0	0	1	0	0	0
Input 6	1	1	0	0	0	0	0	1	0	0
Input 7	1	1	0	0	0	0	0	0	1	0
Input 8	1	1	0	0	0	0	0	0	0	1
Input 9	1	-1	1	0	0	0	0	0	0	0
Input 10	1	-1	0	1	0	0	0	0	0	0
Input 11	1	-1	0	0	1	0	0	0	0	0
Input 12	1	-1	0	0	0	1	0	0	0	0
Input 13	1	-1	0	0	0	0	1	0	0	0
Input 14	1	-1	0	0	0	0	0	1	0	0
Input 15	1	-1	0	0	0	0	0	0	1	0
Input 16	1	-1	0	0	0	0	0	0	0	1

View design Covariance Done

General Linear Model

EVs Contrasts & F-tests

Contrasts 2 F-tests 0

	Title	EV1	EV2	EV3	EV4	EV5	EV6	EV7	EV8	EV9
C1	A - B	1	0	0	0	0	0	0	0	0
C2	B - A	-1	0	0	0	0	0	0	0	0

EVs

View design Covariance Done

3-Level Analysis: Multi-Session & Multi-Subject

We have 5 subjects each have 3 sessions. Because number of sessions is low, and cross-session variance is required, we can put all subjects into one 2nd-Level analysis.

General Linear Model

EVs | Contrasts & F-tests

Number of EVs: 5

Number of groups: 1

	Group	EV1	EV2	EV3	EV4	EV5
Input 1	1	1	0	0	0	0
Input 2	1	1	0	0	0	0
Input 3	1	1	0	0	0	0
Input 4	1	0	1	0	0	0
Input 5	1	0	1	0	0	0
Input 6	1	0	1	0	0	0
Input 7	1	0	0	1	0	0
Input 8	1	0	0	1	0	0
Input 9	1	0	0	1	0	0
Input 10	1	0	0	0	1	0
Input 11	1	0	0	0	1	0
Input 12	1	0	0	0	1	0
Input 13	1	0	0	0	0	1
Input 14	1	0	0	0	0	1
Input 15	1	0	0	0	0	1

View design | Covariance | Done

General Linear Model

EVs | Contrasts & F-tests

Contrasts: 5 | F-tests: 0

	Title	EV1	EV2	EV3	EV4	EV5
C1	subject 1	1	0	0	0	0
C2	subject 2	0	1	0	0	0
C3	subject 3	0	0	1	0	0
C4	subject 4	0	0	0	1	0
C5	subject 5	0	0	0	0	1

	EV1	EV2	EV3	EV4	EV5
C1	1	0	0	0	0
C2	0	1	0	0	0
C3	0	0	1	0	0
C4	0	0	0	1	0
C5	0	0	0	0	1

View design | Covariance | Done

F-Tests: With 3 groups of previous subjects, is any group activating on average?

General Linear Model

EVs | Contrasts & F-tests

Number of EVs 3

Number of groups 3

	Group	EV1	EV2	EV3
Input 1	1	1	0	0
Input 2	1	1	0	0
Input 3	1	1	0	0
Input 4	1	1	0	0
Input 5	1	1	0	0
Input 6	2	0	1	0
Input 7	2	0	1	0
Input 8	2	0	1	0
Input 9	2	0	1	0
Input 10	2	0	1	0
Input 11	3	0	0	1
Input 12	3	0	0	1
Input 13	3	0	0	1
Input 14	3	0	0	1
Input 15	3	0	0	1

View design | Covariance | Done

General Linear Model

EVs | Contrasts & F-tests

Contrasts 3 | F-tests 1

	Title	EV1	EV2	EV3	F1
C1	group A	1	0	0	<input checked="" type="checkbox"/>
C2	group B	0	1	0	<input checked="" type="checkbox"/>
C3	group C	0	0	1	<input checked="" type="checkbox"/>

View design | Covariance | Done

3rd Level: Main and Parametric Effects

General Linear Model

EVs Contrasts & F-tests

Number of EVs 2

Paste

	Group	EV1	EV2
Input 1	1	1	-25
Input 2	1	1	38
Input 3	1	1	47
Input 4	1	1	-33
Input 5	1	1	5
Input 6	1	1	16
Input 7	1	1	-42
Input 8	1	1	30
Input 9	1	1	-30
Input 10	1	1	-25
Input 11	1	1	38
Input 12	1	1	43
Input 13	1	1	-12

EV1 represents the main effect across subjects.

EV2 represents some parameter (e.g., a personality test, age, disability scale, etc) that varies across subjects.

General Linear Model

EVs Contrasts & F-tests

Contrasts 2 F-tests 0

	Title	EV1	EV2
C1	group mean	1	0
C2	reaction time	0	1

View design Covariance Done

Other groups

- ANOVA: 1-factor 3-level
 - We have 8 subjects with 1 factor at 3 or 4 levels

General Linear Model

EVs | Contrasts & F-tests

Number of EVs: 4

Number of groups: 1

	Group	EV1	EV2	EV3	EV4
Input 1	1	1	1	0	0
Input 2	1	1	1	0	0
Input 3	1	1	0	1	0
Input 4	1	1	0	1	0
Input 5	1	1	0	0	1
Input 6	1	1	0	0	1
Input 7	1	1	0	0	0
Input 8	1	1	0	0	0

Buttons: View design, Covariance, Done

General Linear Model

EVs | Contrasts & F-tests

Contrasts: 3 | F-tests: 1

	Title	EV1	EV2	EV3	EV4	F1
C1	level1-mean	0	1	0	0	<input checked="" type="checkbox"/>
C2	level2-mean	0	0	1	0	<input checked="" type="checkbox"/>
C3	level3-mean	0	0	0	1	<input checked="" type="checkbox"/>

Buttons: View design, Covariance, Done

Rules for Projects

- Safety
 - There should be no risk to the participant.
 - You must behave professionally and conscientiously while running your subjects.
- Simplicity and Robustness
 - The task must be simple and easily programmed.
 - The design should be simple, ideally blocked with few conditions.

Imaging Cognition II: An Empirical Review of 275 PET and fMRI Studies

A review 275 PET and fMRI studies of:

- Attention (sustained, selective, Stroop, orientation, divided)
- Perception (object, face, space/motion, smell),
- Imagery (object, space/ motion)
- Language (written/spoken word recognition, spoken/ no spoken response)
- Working memory (verbal/numeric, object, spatial, problem solving)
- Semantic memory retrieval (categorization, generation),
- Episodic memory encoding (verbal, object, spatial)
- Episodic memory retrieval (verbal, nonverbal, success, effort, mode, context)
- Priming (perceptual, conceptual)
- Procedural memory (conditioning, motor, and nonmotor skill learning).

New feasible Proposals

- Cortical thickness measurement for cognitive status and neuro-degenerative disease
- Connectivity evaluation in cognitive status and neuro-degenerative disease (eg. Epilepsy)
- Evaluation of Pre-frontal dementia for changes in attention, Language, and memory

Much More Info on FSL

FEAT Expert Guide

<http://www.fmrib.ox.ac.uk/fsl/feat5/detail.html>

FSL Course Slides

<http://www.fmrib.ox.ac.uk/fslcourse/>

More Expert Guidance

<http://fsl.fmrib.ox.ac.uk/fsl/feat5/>