

General Linear Modeling in EEGLAB/LIMO

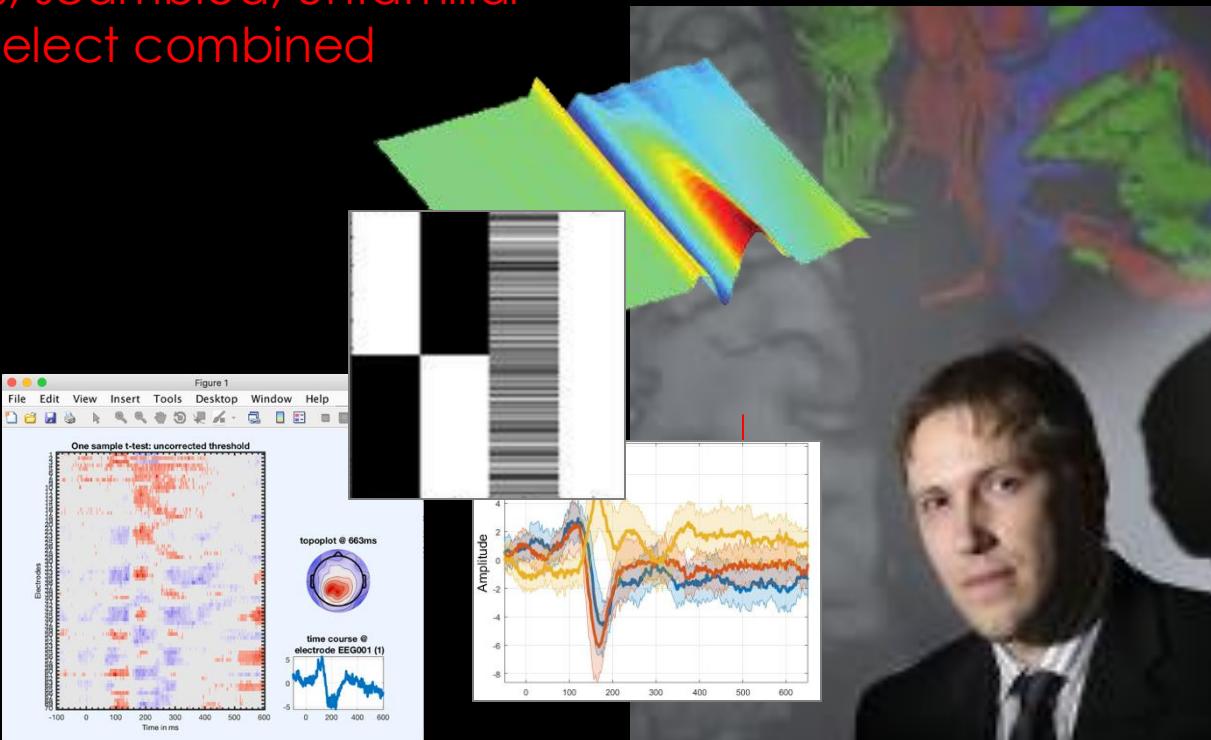
Watch the videos
to be prepared

People cannot reproduce
the video (take screen caps)

Practice

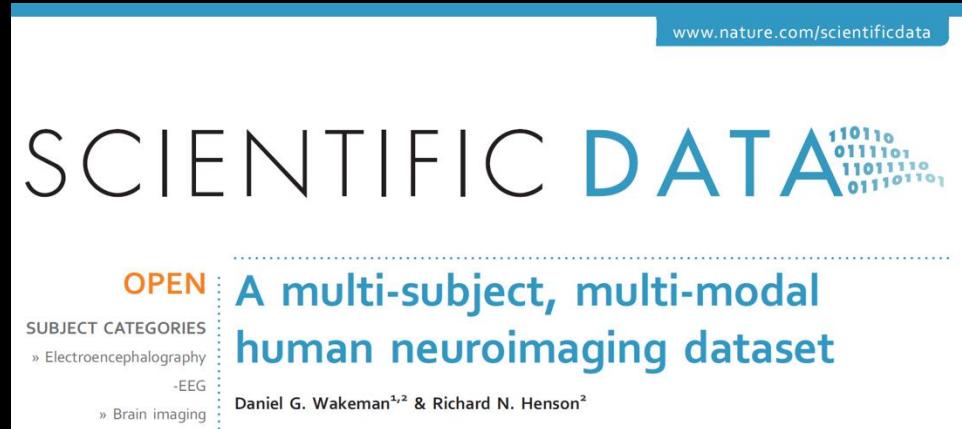
Arnaud Delorme

Video select famous, scrambled, unfamiliar
while in real life we select combined
values



Cyril Pernet

<https://www.nature.com/articles/sdata20151>

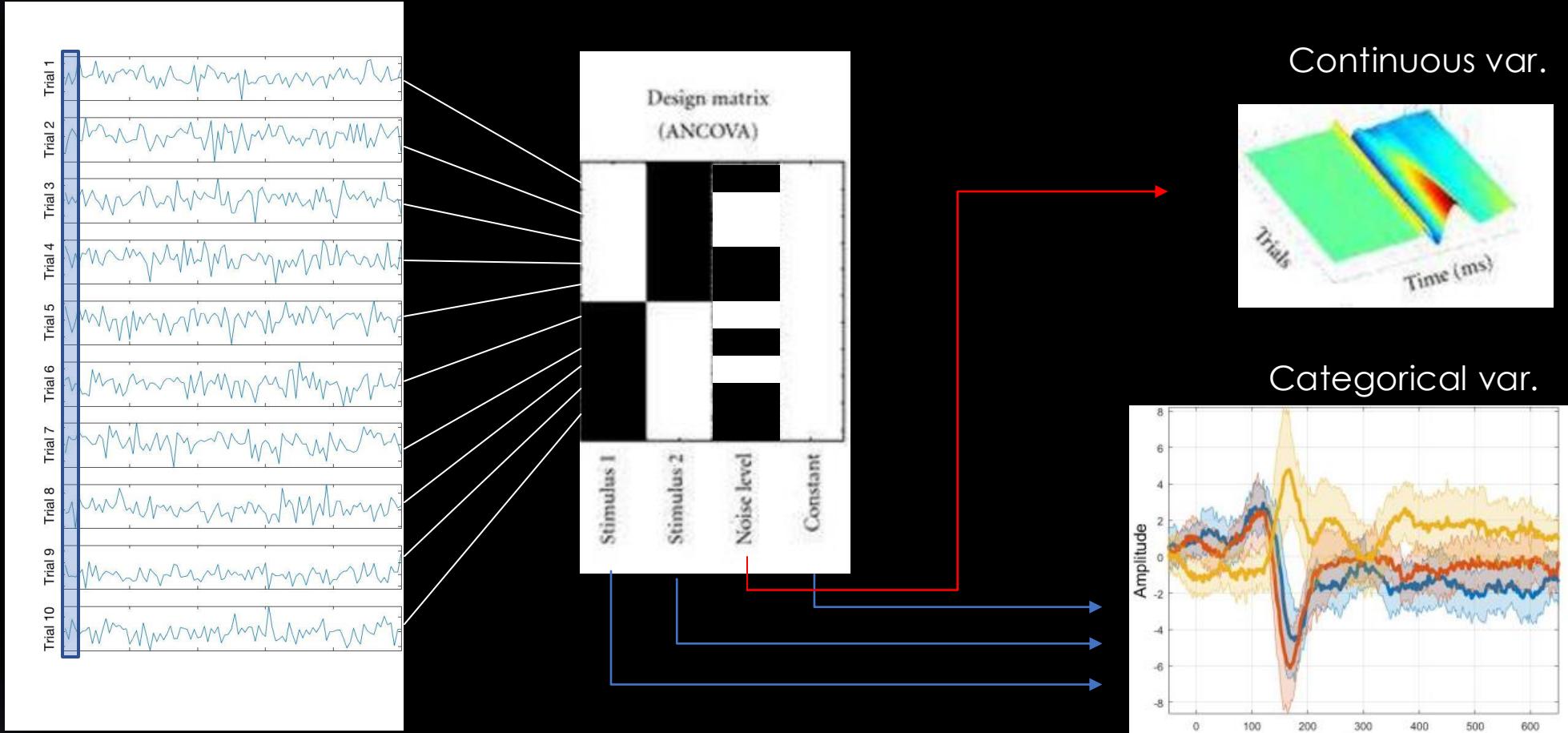


Download the data at
<https://openneuro.org/datasets/ds002718>

The screenshot shows the OpenNEURO dataset page for "Face processing EEG dataset for EEGLAB". It lists versions: Draft (2020-05-26), 1.0.0 (2020-04-21), 1.0.1 (2020-04-23), and 1.0.2 (2020-05-26). The main content includes the dataset title, upload details, download links, and a BIDS Validation section indicating "Valid". The Dataset File Tree shows a hierarchical structure of files including CHANGES, dataset_description.json, participants.json, participants.tsv, README, code, stimuli, and multiple subject folders (sub-002, sub-003, sub-004, sub-005, sub-006, sub-007).

Linear Modeling of EEG data: level 1

Electrode 1



GLM: ordinary least square (OLS) versus weighted least square (WLS)

Significance: bootstrap trials to get confidence interval of beta parameters

The Data



- ▶ 3 types of stimuli: Famous faces, Non-famous faces, Scrambled faces
- ▶ 3 levels of repetition: 1st time, 2nd time (right after), 3rd time (delayed)

We need the conditions computed per subject (1st level) and then do the repeated measure ANOVA to test main effects and interactions.

Let's get started

Name
► sub002
► sub003
► sub004
► sub005
► sub006
► sub007
► sub008
► sub009
► sub010
► sub011
► sub012
► sub013
► sub014
► sub015
► sub016
► sub017
► sub018
► sub019

- Download the data
<https://openneuro.org/datasets/ds002718>
- Open Matlab
- Start EEGLAB

```

% start EEGLAB
clear
[ALLEEG, EEG, CURRENTSET, ALLCOM] = eeglab;

% import BIDS
filepath      = 'XXX\WakemanHenson_Faces\eeg';
[STUDY, ALLEEG] = pop_importbids(filepath, 'bidsevent','on','bidschanloc','on', ...
    'studyName','Face_detection');
ALLEEG = pop_select( ALLEEG, 'nochannel',{ 'EEG061','EEG062','EEG063','EEG064'} );
CURRENTSTUDY = 1; EEG = ALLEEG; CURRENTSET = [1:length(EEG)];

% Remove bad channels
EEG = pop_clean_rawdata( EEG,'FlatlineCriterion',5,'ChannelCriterion',0.8, ...
    'LineNoiseCriterion',4,'Highpass',[0.25 0.75] ,...
    'BurstCriterion','off','WindowCriterion','off','BurstRejection','off',...
    'Distance','Euclidian','WindowCriterionTolerances','off' );

% Rereference using average reference
EEG = pop_reref( EEG,[],'interpchan',[]);

% Run ICA and flag artifactual components using IClabel
EEG = pop_runica(EEG,s, 'icatype','runica','concatcond','on','options',{'pca',-1});
EEG = pop_iclabel(EEG,s,'default');
EEG = pop_icflag(EEG(s),[NaN NaN;0.8 1;0.8 1;NaN NaN;NaN NaN;NaN NaN;NaN NaN]);
EEG = pop_subcomp(EEG(s), []); % remove bad components

% clear data using ASR - just the bad epochs
EEG = pop_clean_rawdata( EEG,'FlatlineCriterion','off','ChannelCriterion','off',...
    'LineNoiseCriterion','off','Highpass','off','BurstCriterion',20, ...
    'WindowCriterion',0.25,'BurstRejection','on','Distance','Euclidian',...
    'WindowCriterionTolerances',[-Inf 7] );

% Extract data epochs (no baseline removed)
EEG    = pop_epoch( EEG,{ 'famous_new','famous_second_early','famous_second_late',...
    'scrambled_new','scrambled_second_early','scrambled_second_late','unfamiliar_new',...
    'unfamiliar_second_early','unfamiliar_second_late'},[-0.5 1] , 'epochinfo','yes');
EEG    = pop_saveset(EEG, 'savemode', 'resave');
ALLEEG = EEG;

% Create study design
STUDY  = std_checkset(STUDY, ALLEEG);
STUDY  = std_makedesign(STUDY, EEG, 1, 'name','Faces','delfiles','off',...
    'defaultdesign','off','variable1','type','values1',{} );
eeglab redraw

```

} Import BIDS dataset/remove unwanted channels

} Clean data lightly

} Re-reference

} Run/Remove ICA

} Clean data aggressively

} Extract epochs

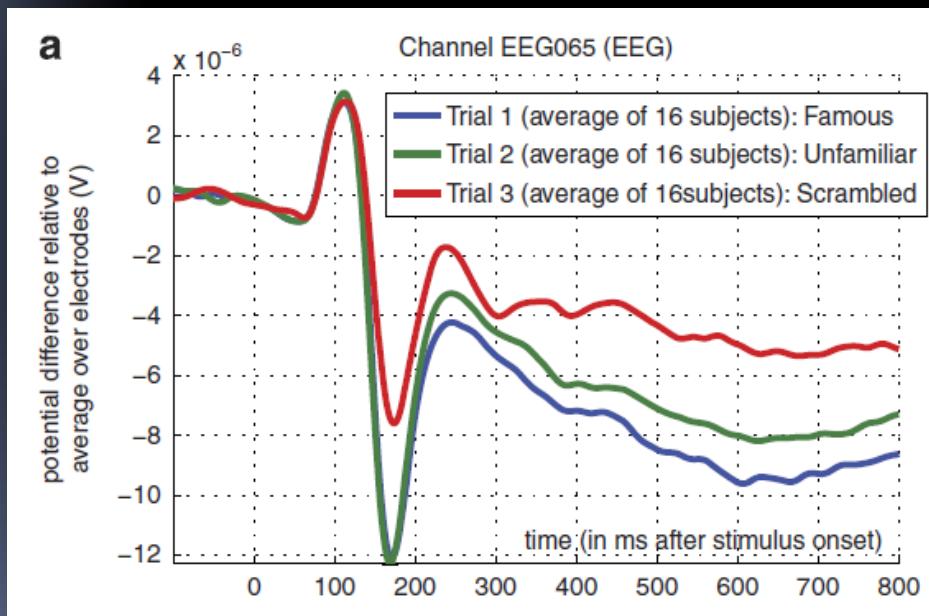
} Create STUDY and STUDY design

Watch movie
Session_5.2_movie_1.mp4

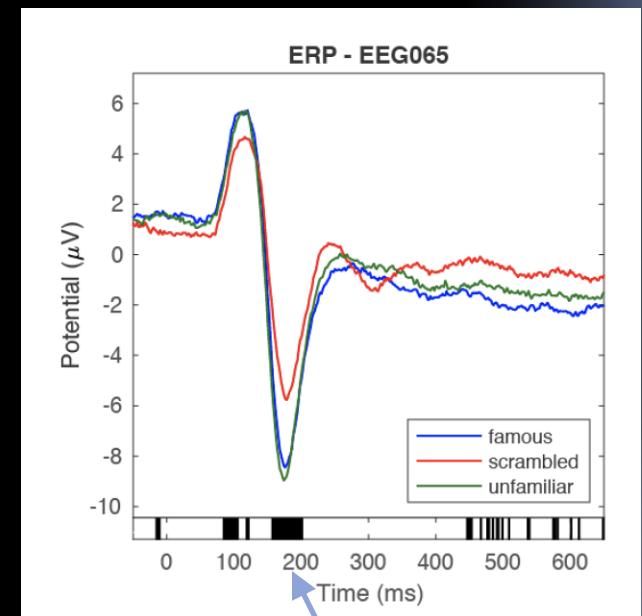
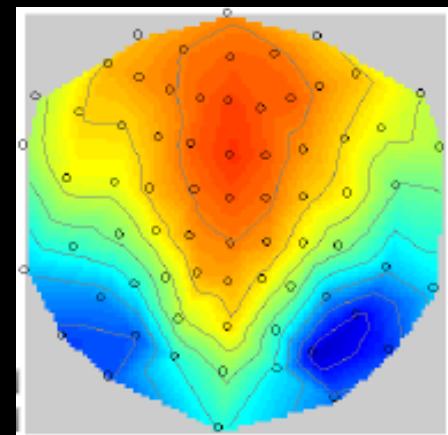
What are we going to do?

- ▶ 1 – Replicate Henson et al. – faces vs. scrambled

EEGLAB

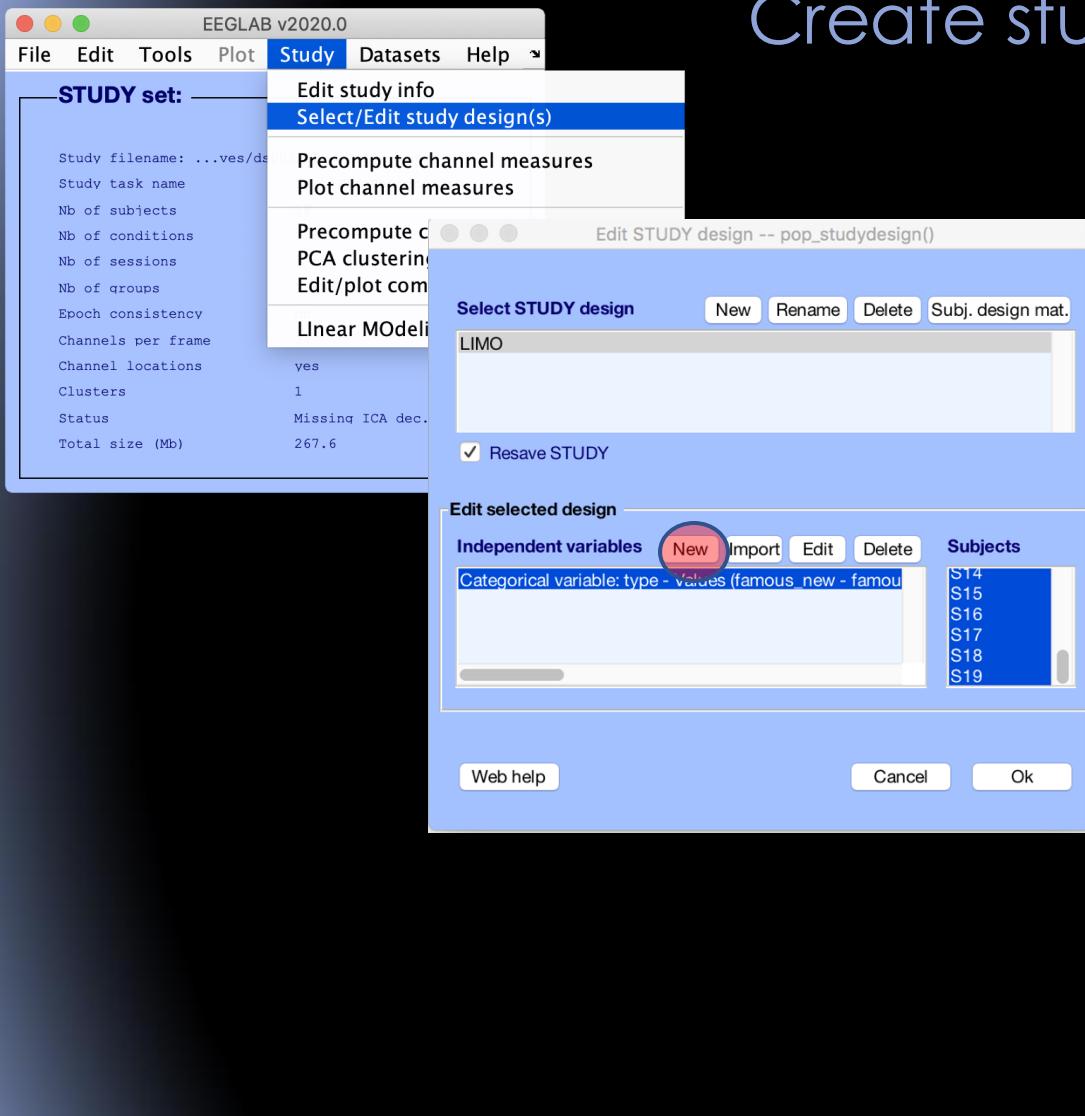


Topography 170 ms



Watch movie
Session_5.2_movie_2.mp4

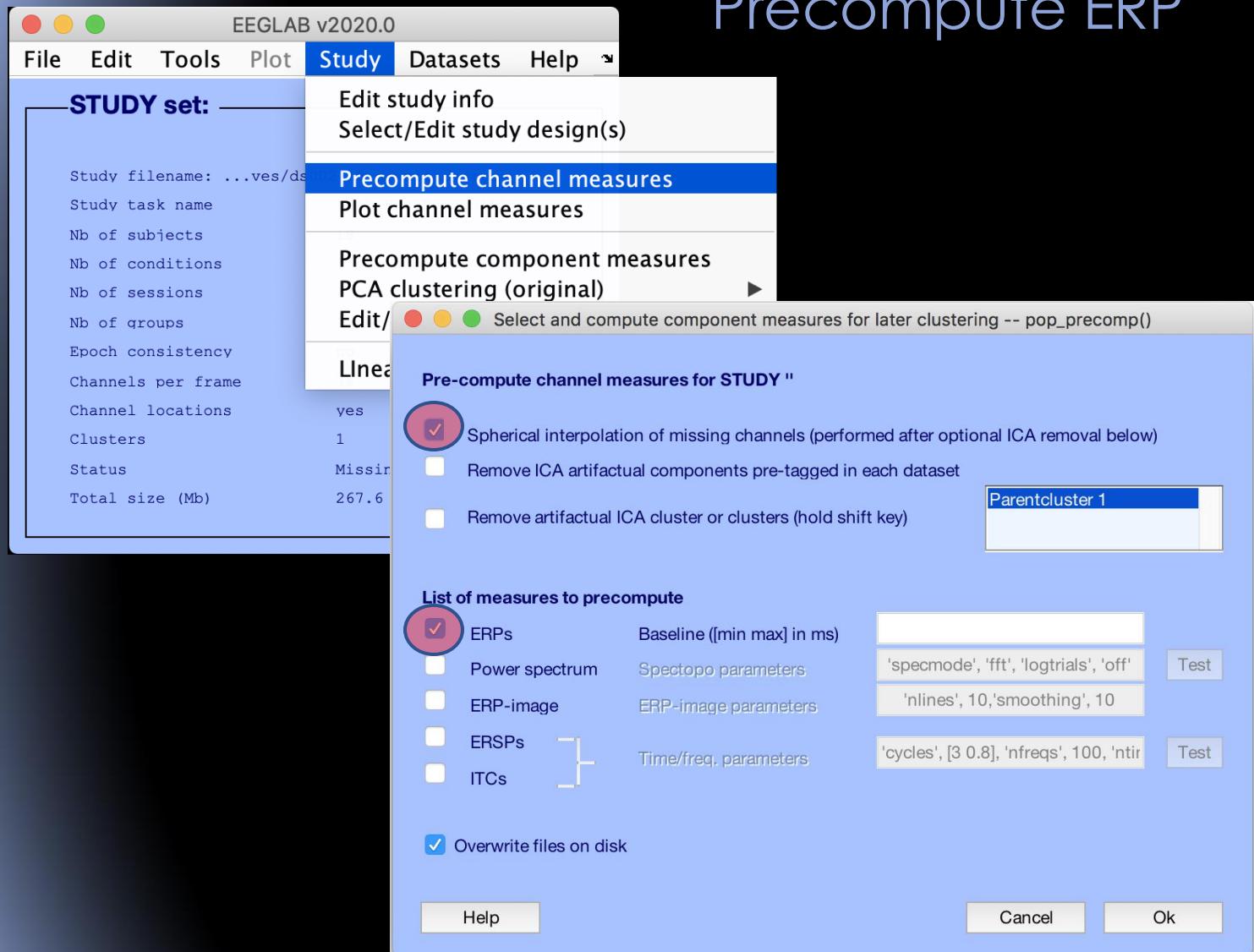
Create study designs



Here, we pick the 'type'
and select all 9 conditions

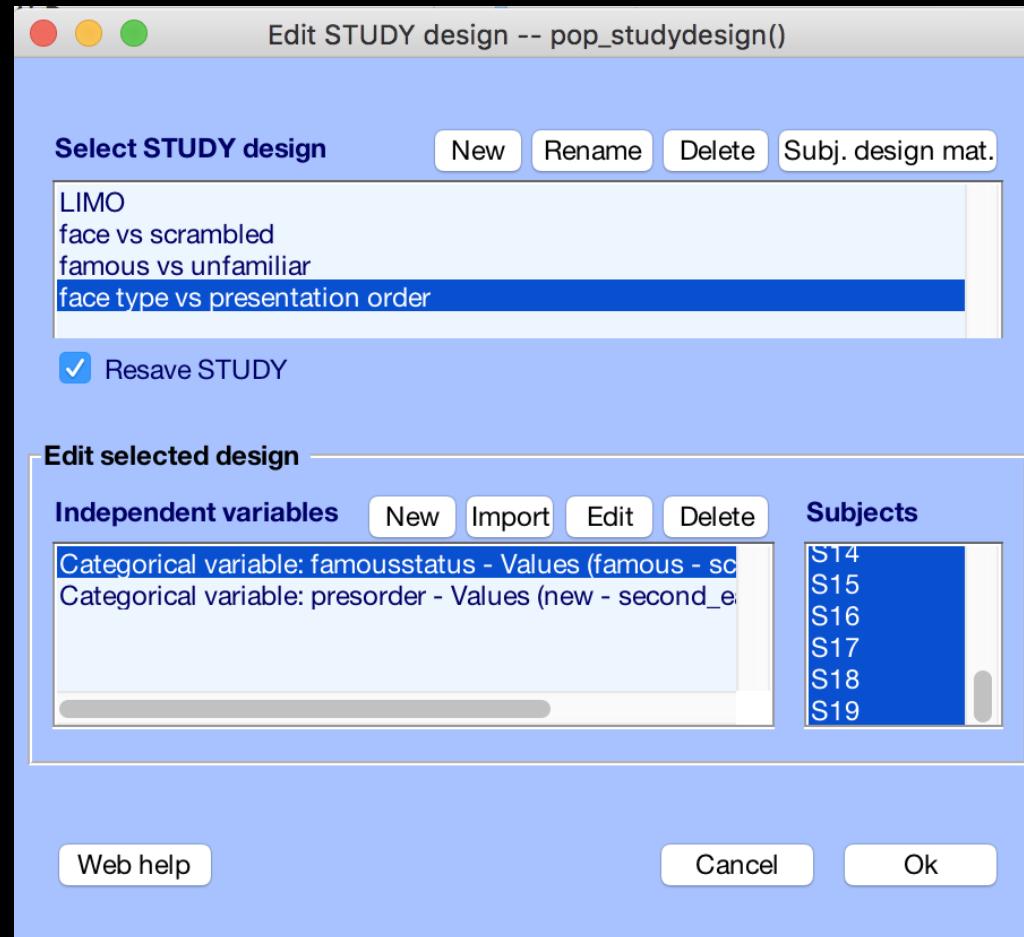
The new STUDY interface allows all sorts of designs. By default, you model each and any condition / covariate (the more complete the model the better)

Precompute ERP



Precompute channel measures (single trials)
with or without interpolation of missing channels

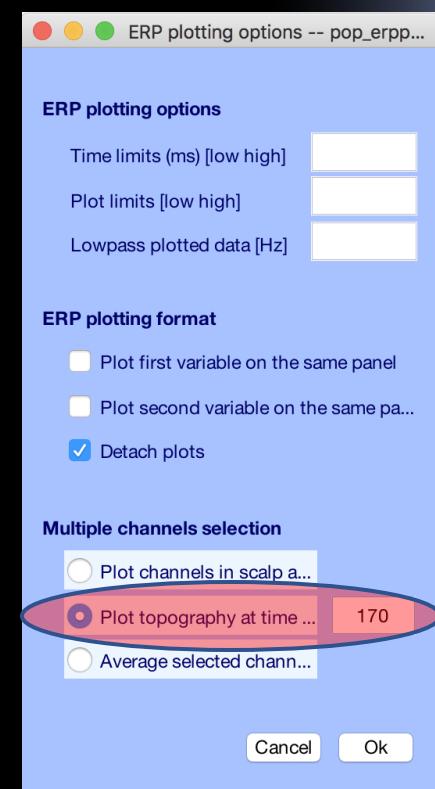
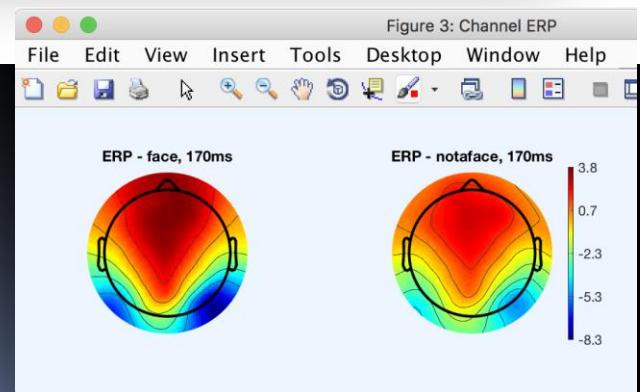
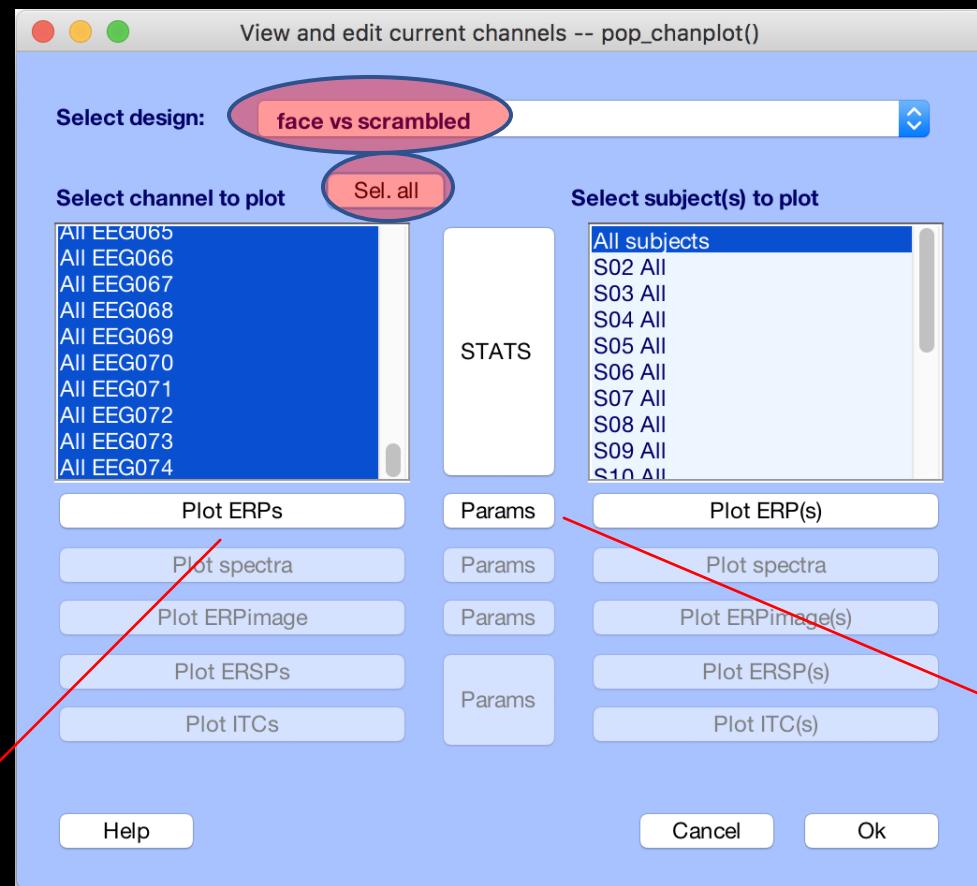
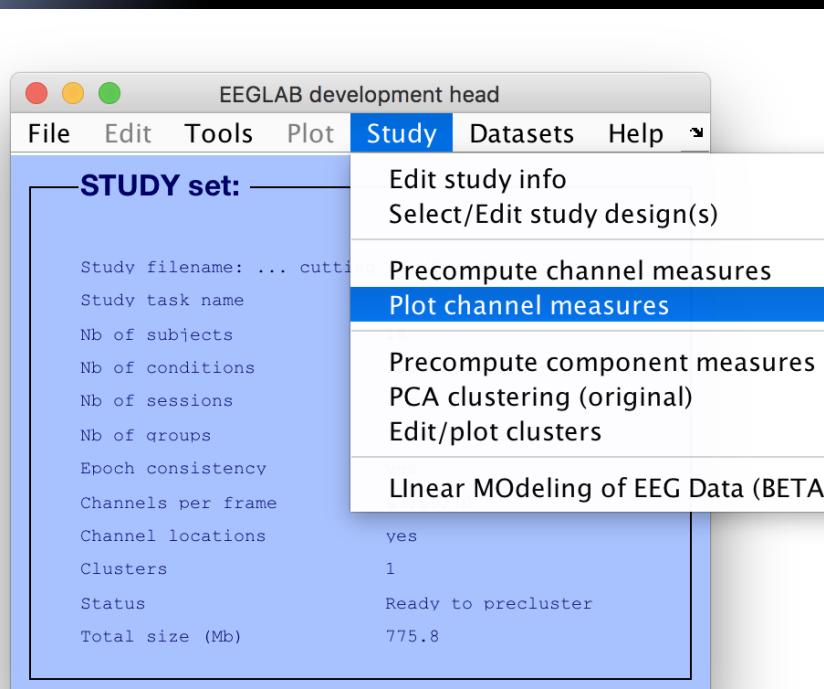
Create other study designs



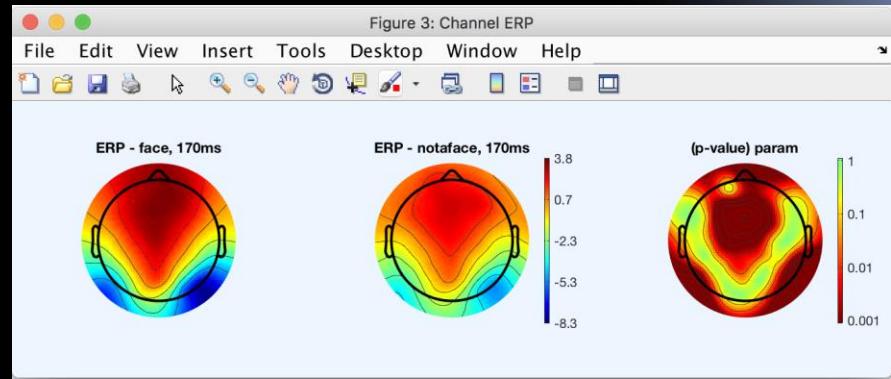
Select/edit study design

The new STUDY interface allows all sorts of designs. By default, you model each and any condition / covariate (the more complete the model the better)

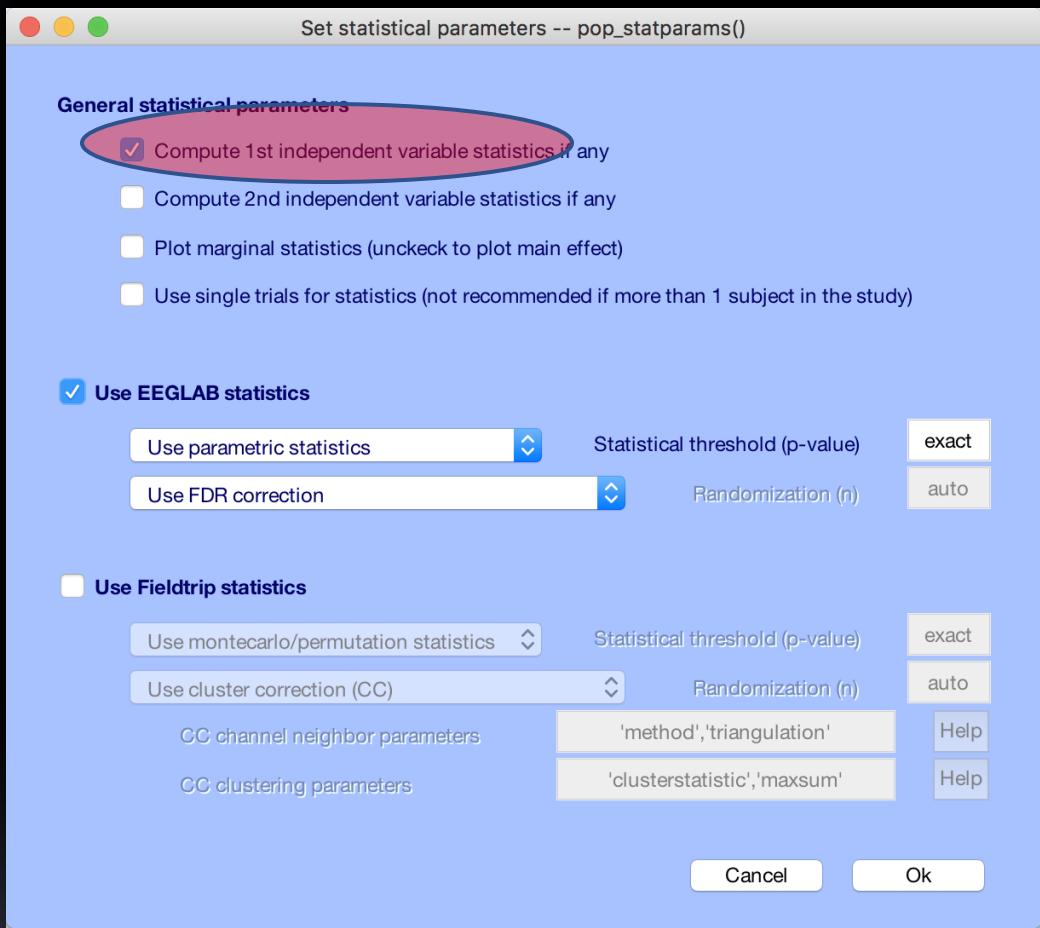
Standard EEGLAB statistics



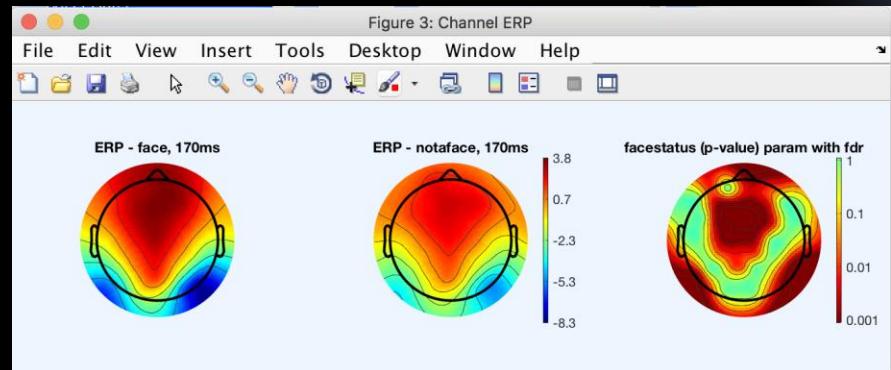
Uncorrected



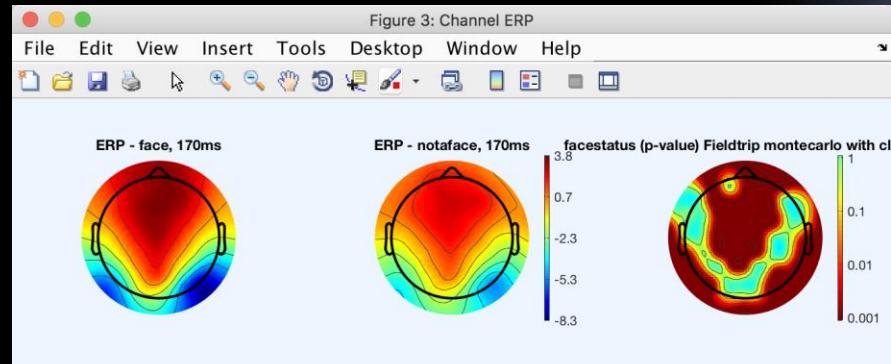
Standard EEGLAB statistics

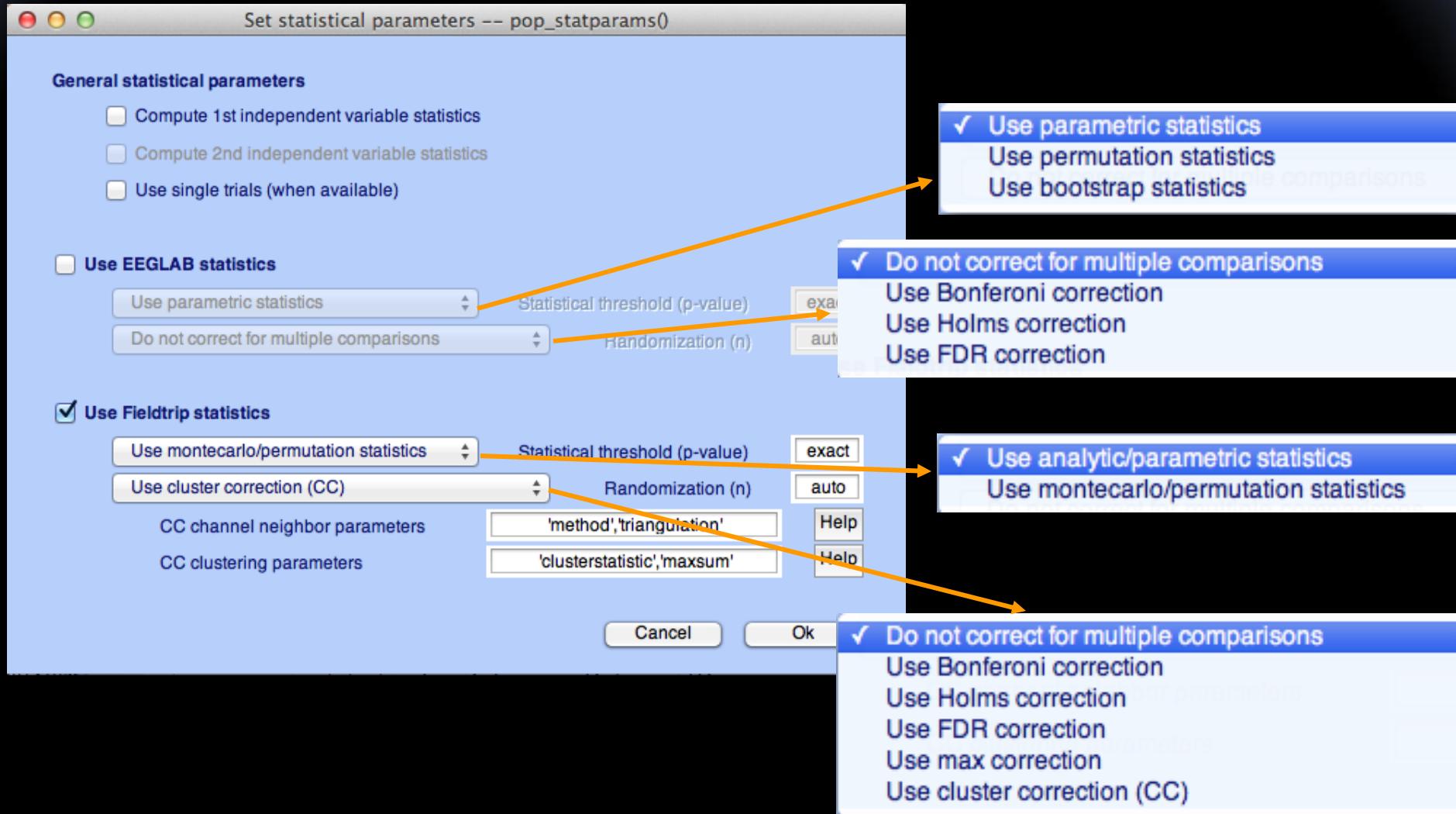


FDR corrected



Cluster corrected



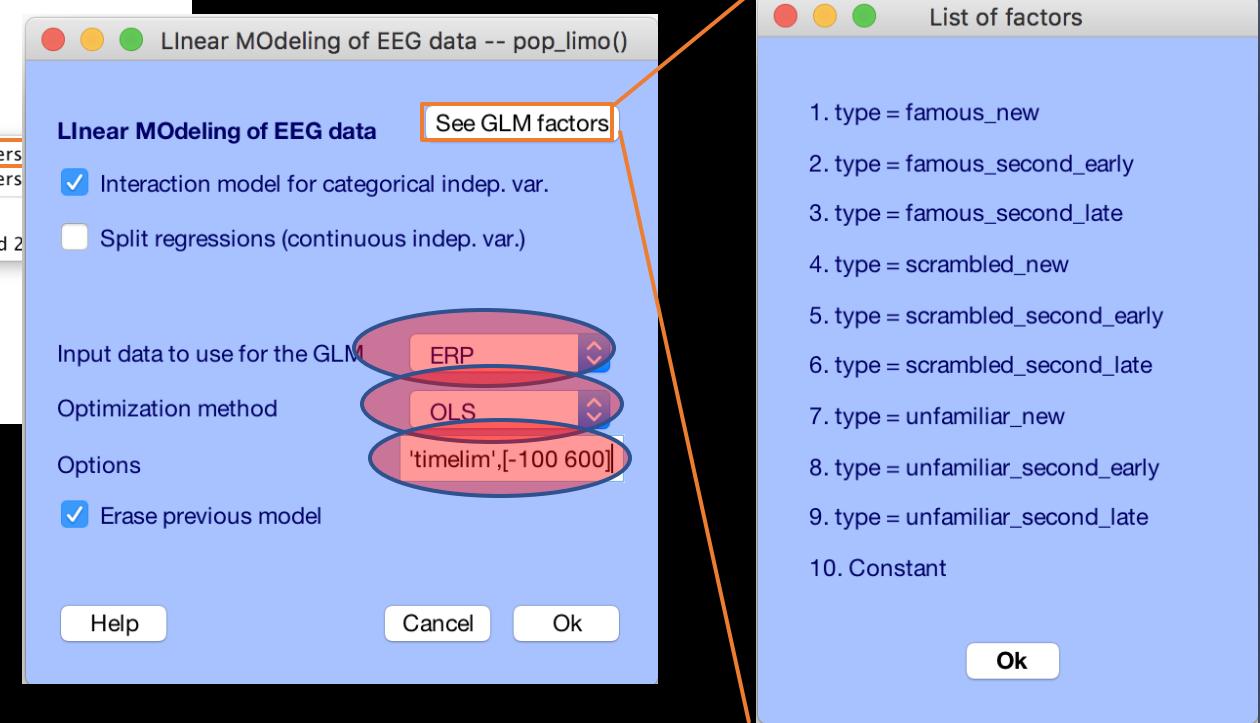
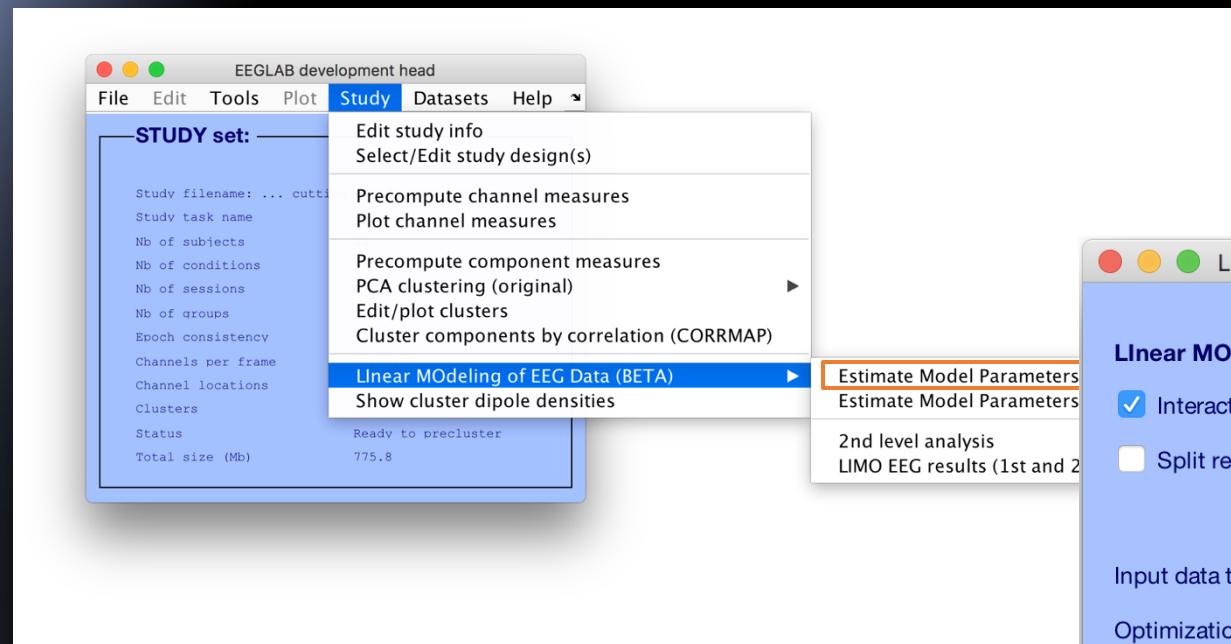


std_stat() function in EEGLAB

Exercise

1. Load “ds002718processed/Face_detection.study” file
2. Edit STUDY design and delete current variable(s) (menu item *STUDY > Select/Edit STUDY design(s)*) – Do not resave STUDY
3. Create a new indep. Variable design to compare Famous vs. Unfamiliar stimuli
4. Recompute spectrum and ERP (remove labeled ICA comp.) (menu item *STUDY > Precompute channel measures*)
5. Plot ERP for Electrode 65 (menu item *STUDY > Plot channel measures*)
6. Plot scalp topography at 170 ms (ERP) for both conditions
7. Compare using no correction, FDR correction and permutation statistics cluster correction (Fieldtrip – statistics)

Estimate model parameters



Only the current
design is pooled in
the GLM

Estimate Model Parameter

Have generated single trials, specified the model, we now do the stats
→ Restrict 'timelim', [-50 650]

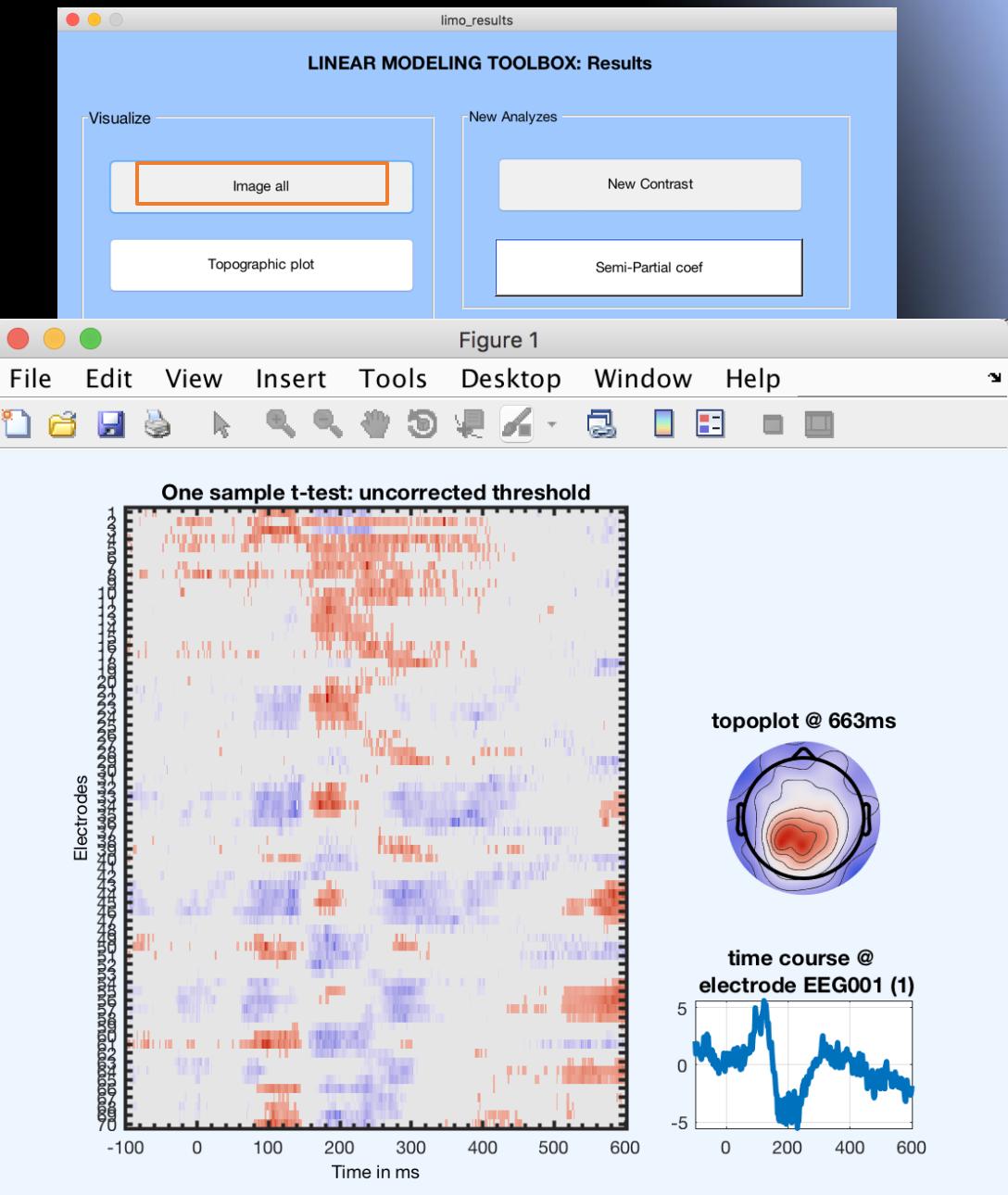
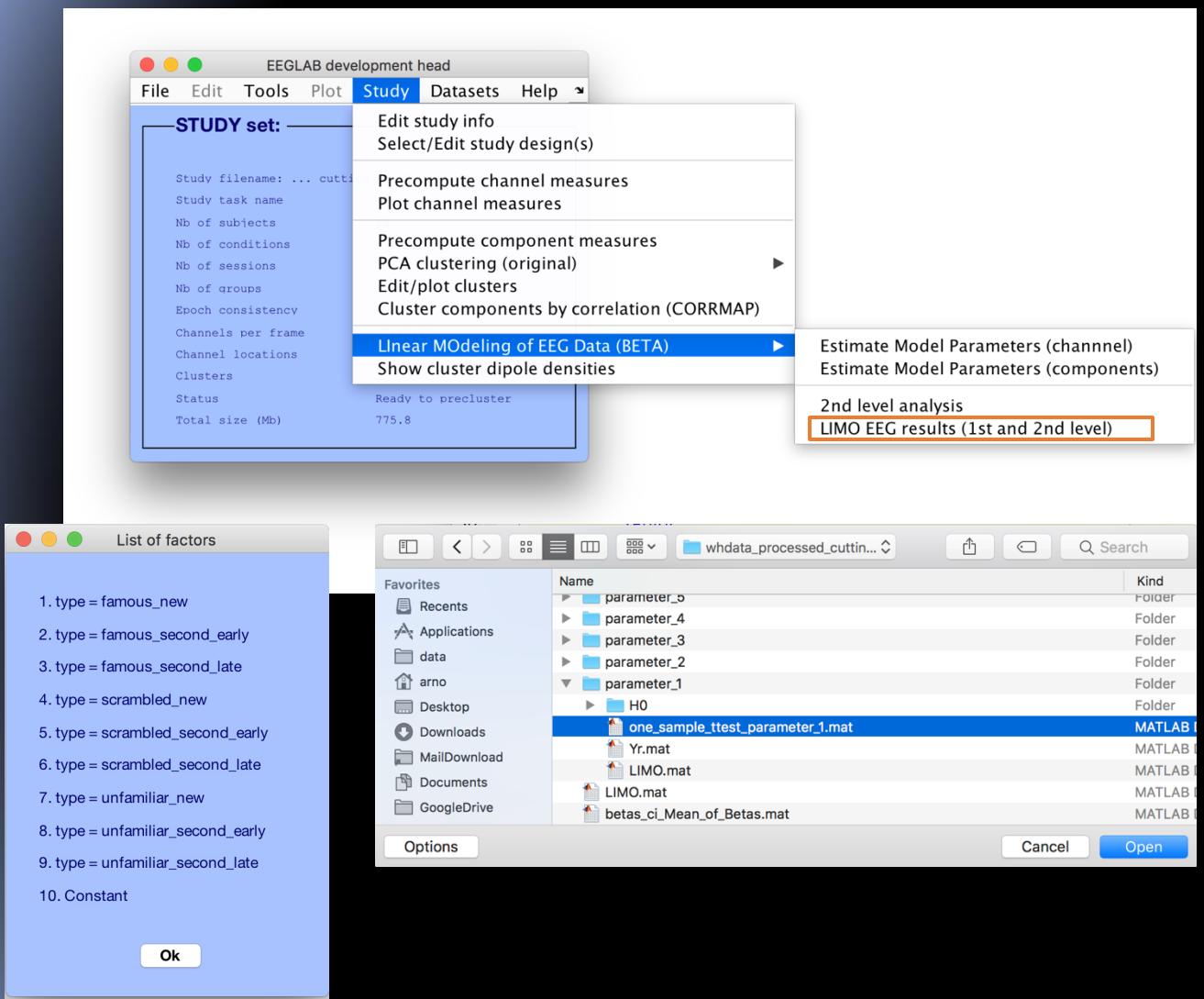
Are Beta significant?

The image displays three overlapping windows:

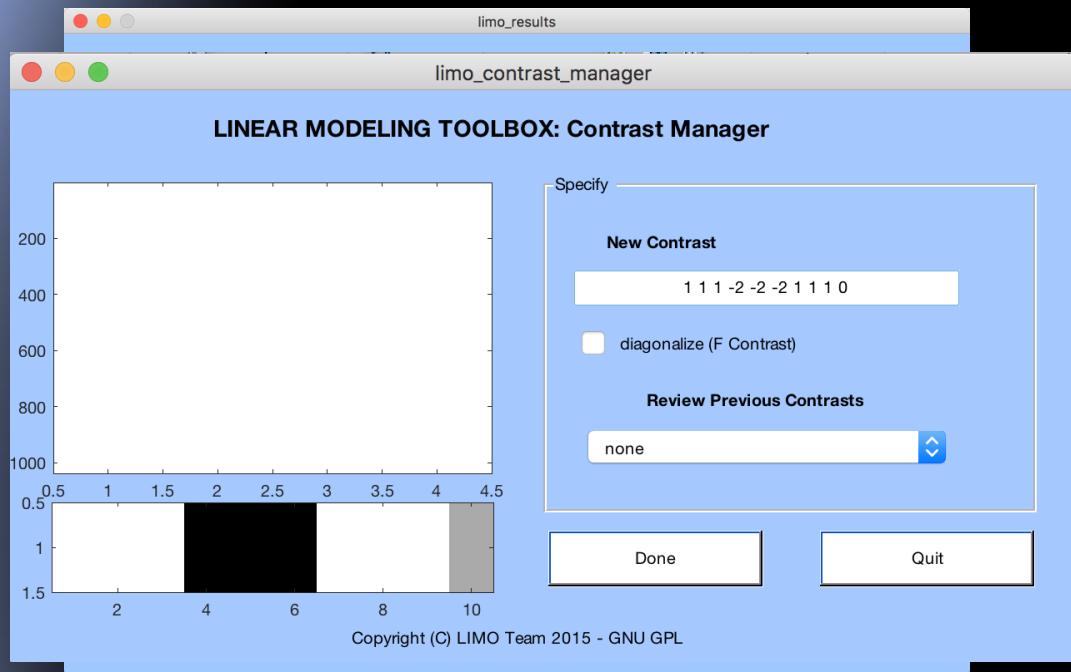
- EEGLAB development head**: A Mac OS X application window. The menu bar includes File, Edit, Tools, Plot, Study, Datasets, and Help. The "Study" menu is open, showing options like "Edit study info", "Select/Edit study design(s)", and "Linear MOdeling of EEG Data (BETA)". The sub-menu for "Linear MOdeling of EEG Data (BETA)" is also open, showing "Estimate Model Parameters (channel)" and "Estimate Model Parameters (com...)".
- LINEAR MODELING TOOLBOX: Random Effects**: A Mac OS X application window titled "limo_random_effect". It contains two main sections: "Basic stats" (with "Central tendency and CI") and "Tests" (with "One Sample t-test").
- File Browser**: A standard Mac OS X file browser window titled "type". It shows a list of files and folders in a directory named "LIMO_facedata". One file, "Beta_files,GLM1OLS_Time_Channels.txt", is selected and highlighted with a blue border. The "Kind" column indicates it is a "Plain Text" file.

Annotations highlight specific features:

- A blue box surrounds the "Beta_files,GLM1OLS_Time_Channels.txt" file in the file browser.
- An orange box surrounds the "One Sample t-test" button in the "Tests" section of the toolbox window.
- An orange arrow points from the "Load expected chan / neighbours" button in the toolbox window up towards the "One Sample t-test" button.



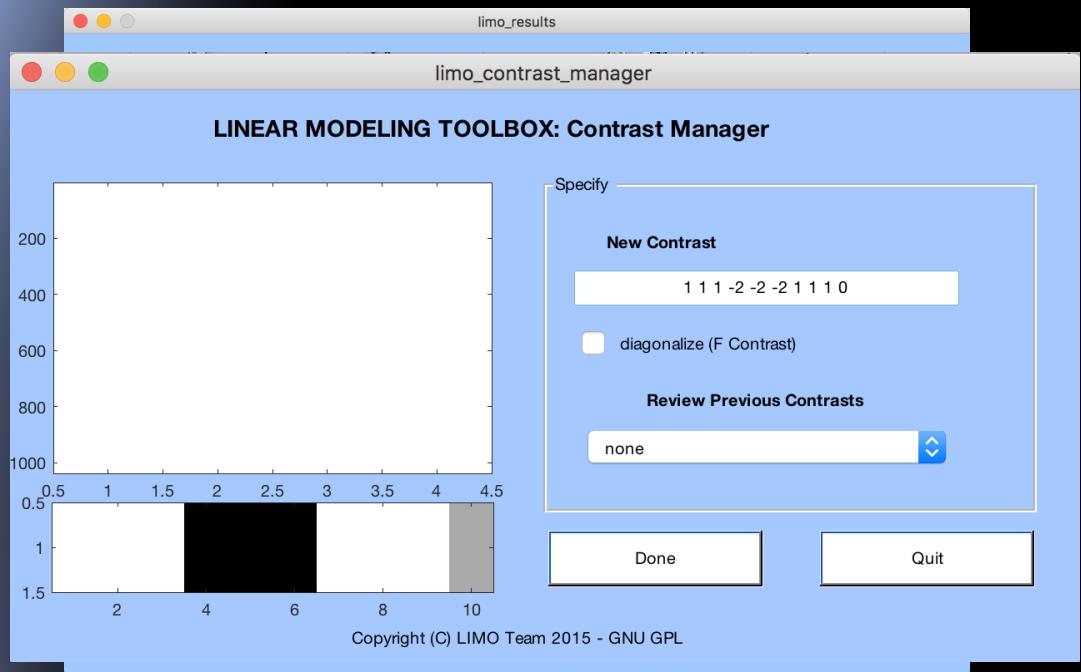
Grouping betas and differences between conditions



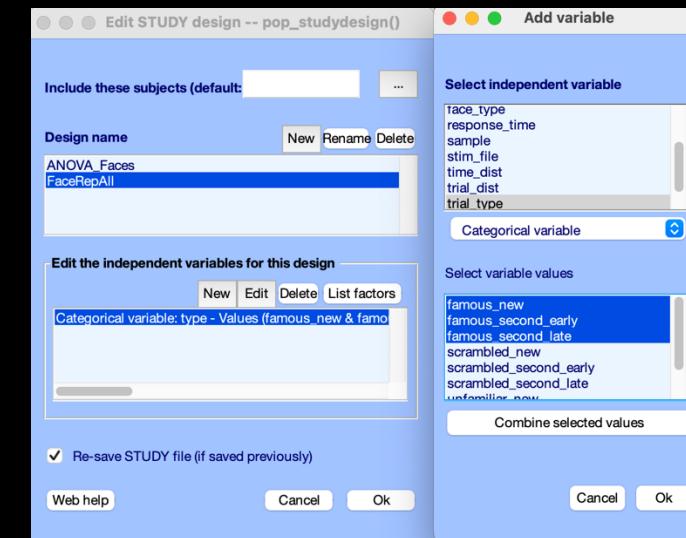
The figure shows a screenshot of the LIMO List of factors dialog. The table lists 10 factors and their corresponding beta coefficients across four conditions: Faces vs non-faces, Famous, Scrambled, Unfamiliar, ANOVA (famous/scrambled/unfamiliar), and ANOVA (new/early/late). Red lines connect the factor names to their respective beta values. The "Constant" row has a value of 0 for all conditions.

	Faces vs non-faces	Famous	Scrambled	Unfamiliar	ANOVA (famous/scrambled/unfamiliar)	ANOVA (new/early/late)
1. type = famous_new	1	1	0	0	1	0
2. type = famous_second_early	1	1	0	0	1	0
3. type = famous_second_late	1	1	0	0	1	0
4. type = scrambled_new	-2	0	1	0	0	1
5. type = scrambled_second_early	-2	0	1	0	0	1
6. type = scrambled_second_late	-2	0	1	0	0	1
7. type = unfamiliar_new	1	0	0	1	0	1
8. type = unfamiliar_second_early	1	0	0	1	0	1
9. type = unfamiliar_second_late	1	0	0	1	0	1
10. Constant	0	0	0	0	0	0

Grouping betas and differences between conditions



Contrast were automatically created when we combined variables

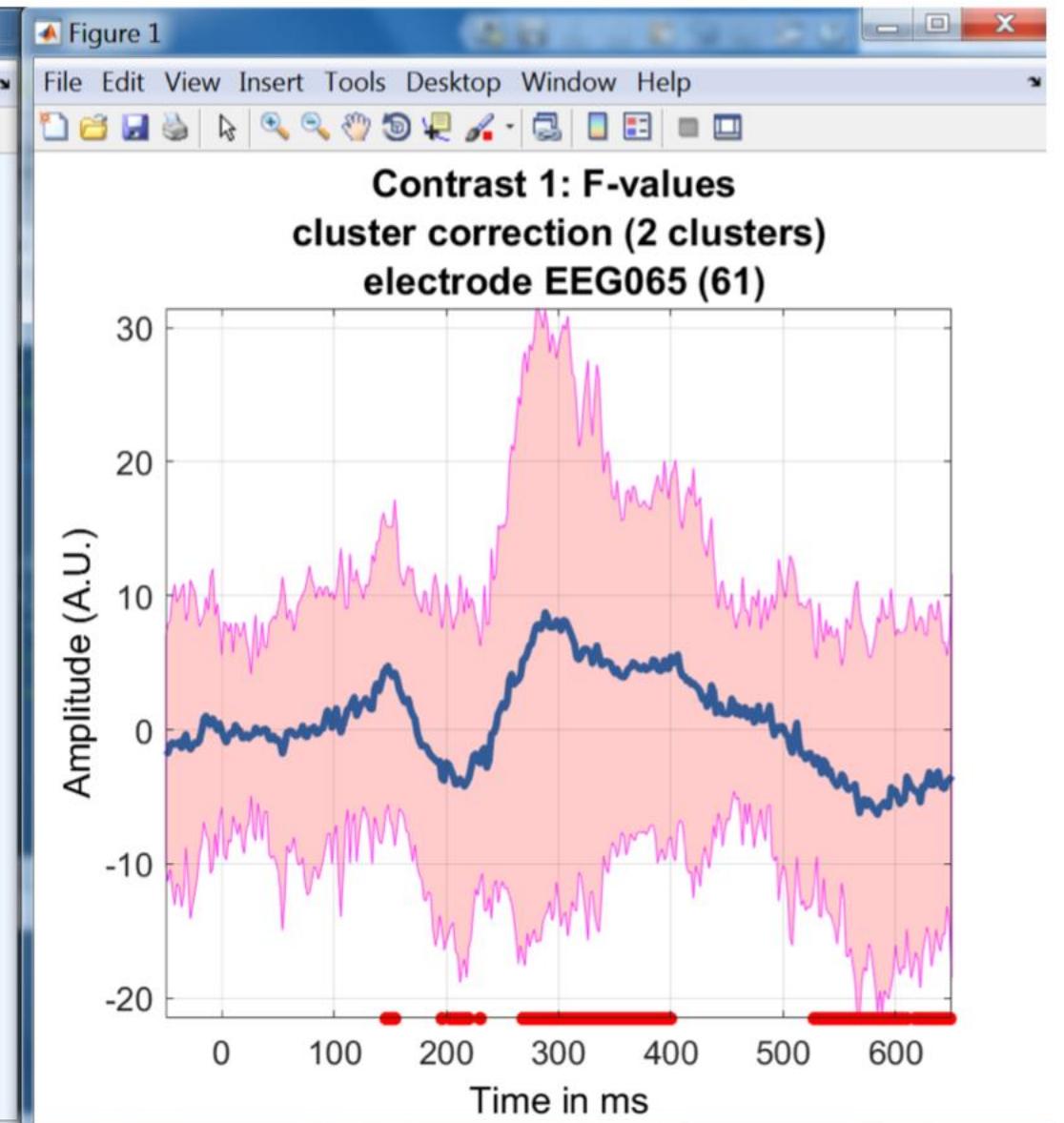
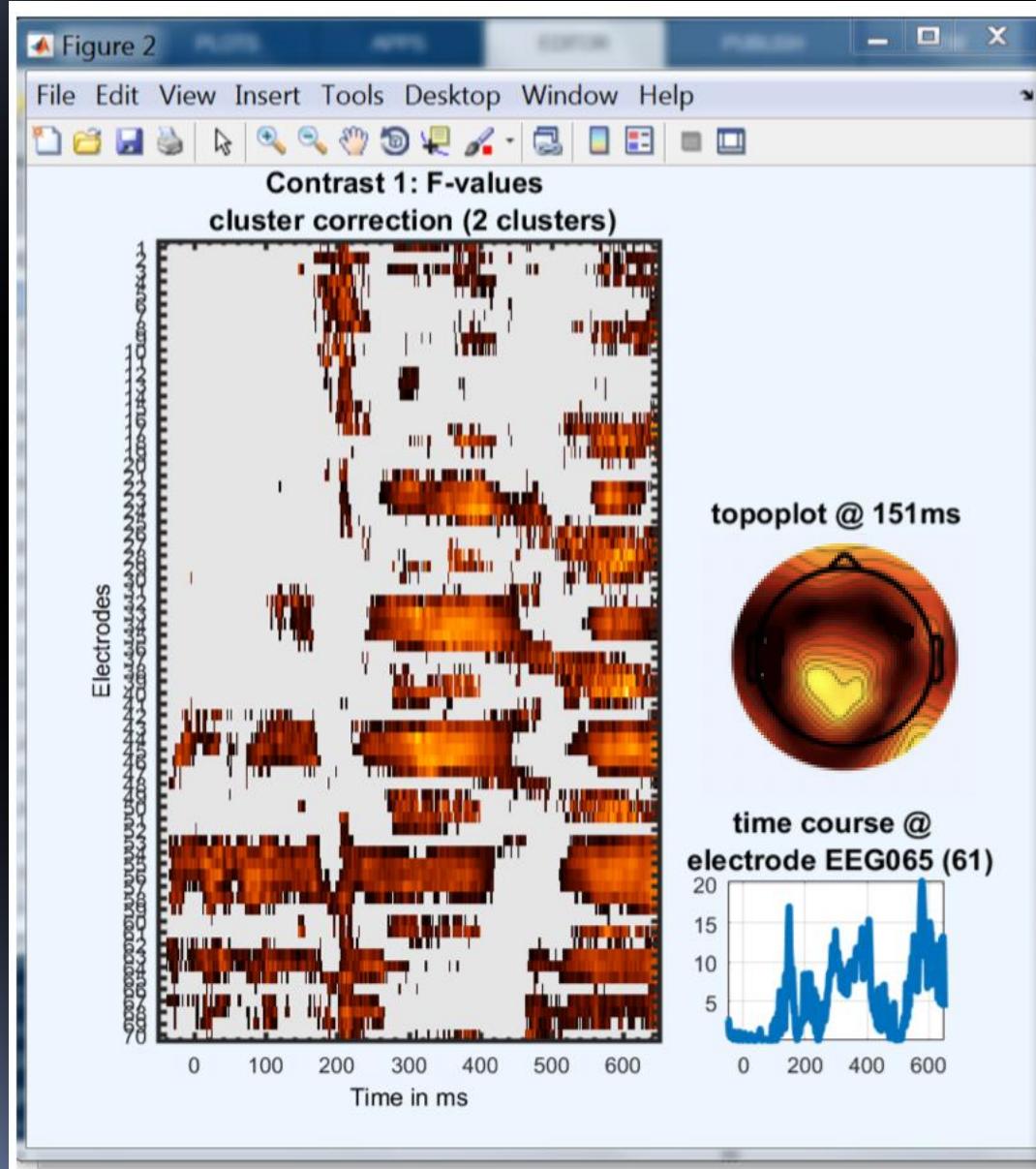


	Faces vs non-faces	Famous	Scrambled	Unfamiliar
1. type = famous_new	1	1	0	0
2. type = famous_second_early	1	1	0	0
3. type = famous_second_late	1	1	0	0
4. type = scrambled_new	-2	0	1	0
5. type = scrambled_second_early	-2	0	1	0
6. type = scrambled_second_late	-2	0	1	0
7. type = unfamiliar_new	1	0	0	1
8. type = unfamiliar_second_early	1	0	0	1
9. type = unfamiliar_second_late	1	0	0	1
10. Constant	0	0	0	0

ANOVA (famous/scrambled/unfamiliar)

ANOVA (new/early/late)

ANOVA (famous/scrambled/unfamiliar)



Exercise

1. Install LIMO plugin
2. Reload “ds002718processed/Face_detection.study” file
3. NOT NEEDED HERE -- Edit STUDY and create design (menu item *STUDY > Select/Edit STUDY design(s)*)
4. NOT NEEDED HERE -- Recompute ERP (remove labeled ICA comp.) (menu item *STUDY > Precompute channel measures*)
5. Compute LIMO for all subjects (Menu *STUDY > Linear Modeling of EEG data > Estimate Model Parameters (channel)*)
6. Compute and plot group level measures (Menu *STUDY > Linear Modeling of EEG data > 2nd Level analysis*)