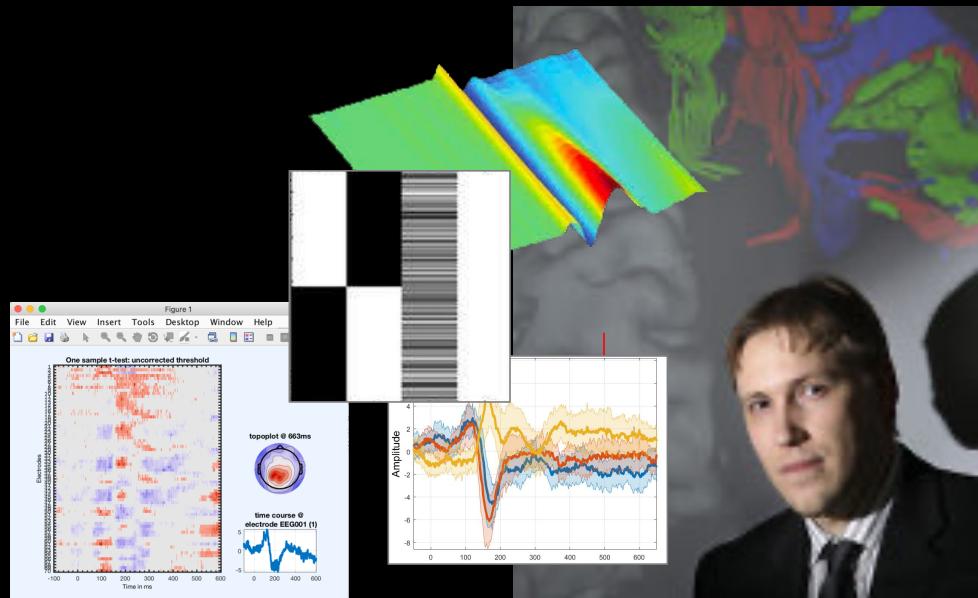


# General Linear Modeling in EEGLAB/LIMO

## Practice

Arnaud Delorme



Cyril Pernet

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

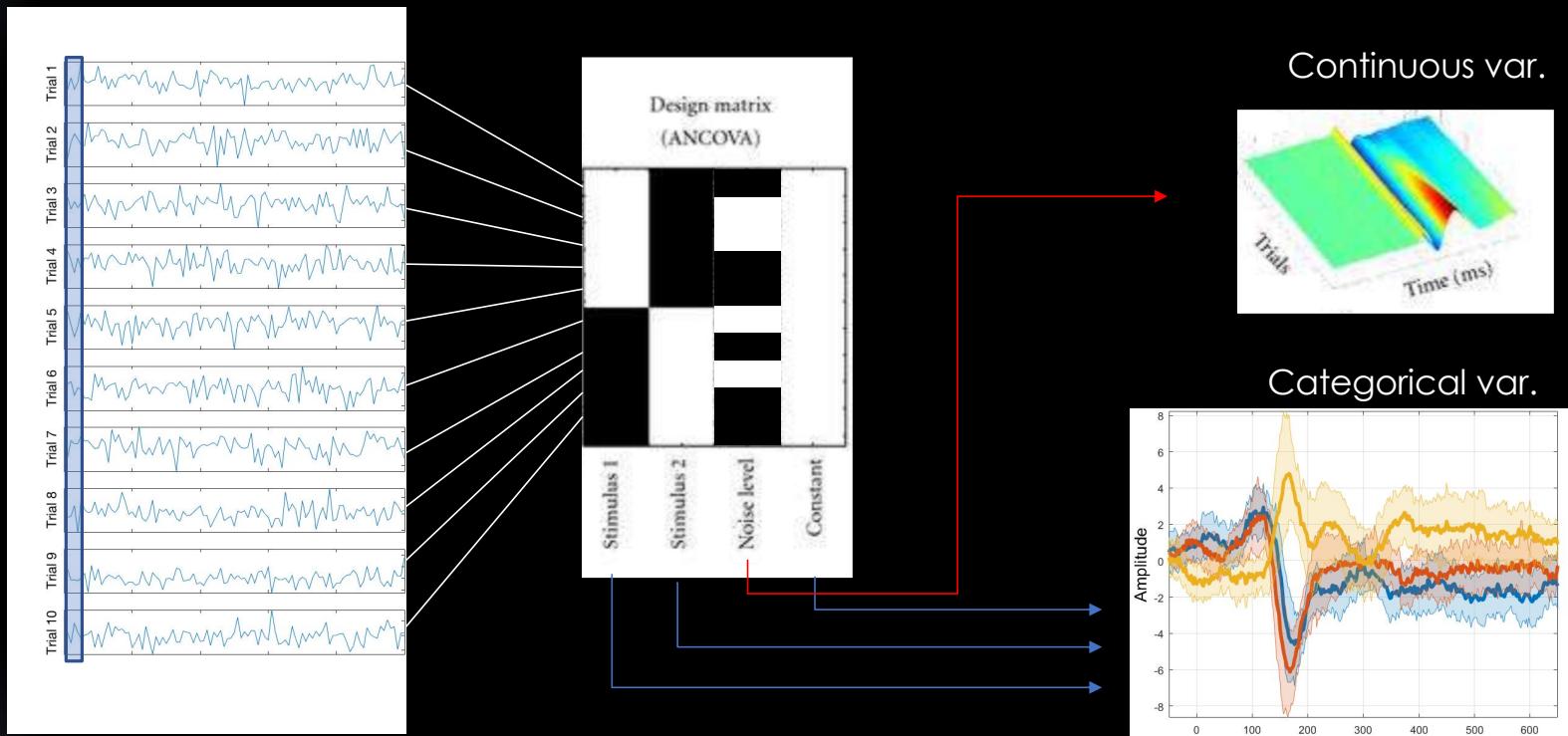
The screenshot shows the header of a scientific article. At the top right is the URL [www.nature.com/scientificdata/](http://www.nature.com/scientificdata/). Below it is the title "SCIENTIFIC DATA" in large, bold letters. To the right of the title is a binary sequence graphic: 1101110, 0111101, 11011110, 011101101. Underneath the title, the word "OPEN" is displayed in a red box. To the left of the title, under "SUBJECT CATEGORIES", are "Electroencephalography", "-EEG", and "Brain imaging". The main title of the article is "A multi-subject, multi-modal human neuroimaging dataset". Below the title are the authors' names: Daniel G. Wakeman<sup>1,2</sup> & Richard N. Henson<sup>2</sup>.

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

The screenshot shows the OpenNeuro dataset page for "Face processing EEG dataset for EEGLAB". The top navigation bar includes links for "MY DASHBOARD", "PUBLIC DASHBOARD", "SUPPORT", "FAQ", and "UPLOAD DATASET". On the left, there's a sidebar titled "Versions" showing four 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 area displays the dataset details: "Face processing EEG dataset for EEGLAB", "BIDS Validation" (Valid), and a "Dataset File Tree" showing the directory structure of the dataset files.

# 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

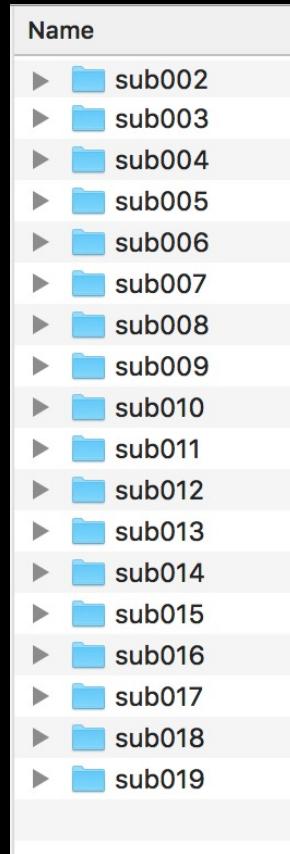
# The Data



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

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

# Let's get started



- 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

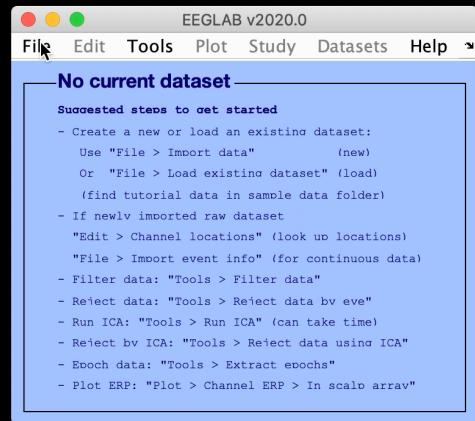
% clear data using ASR - just the bad epochs
EEG = pop_clean_rawdata( EEG,'FlatlineCriterion','off','ChannelCriterion','off',...
'LineNoiseCriterion','off','Highpass','off','BurstCriterion',20, ...
'WindowCriterion',0.25,'BurstRejection','on','Distance','Euclidian',...
'WindowCriterionTolerances',[-Inf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

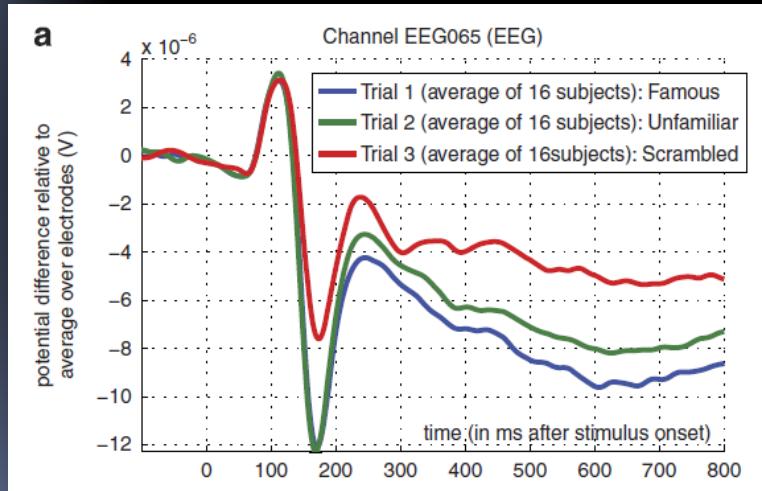
```

- } Import BIDS dataset/remove unwanted channels
- } Clean data lightly
- } Re-reference
- } Run/Remove ICA
- } Clean data aggressively
- } Extract epochs
- } Create STUDY and STUDY design

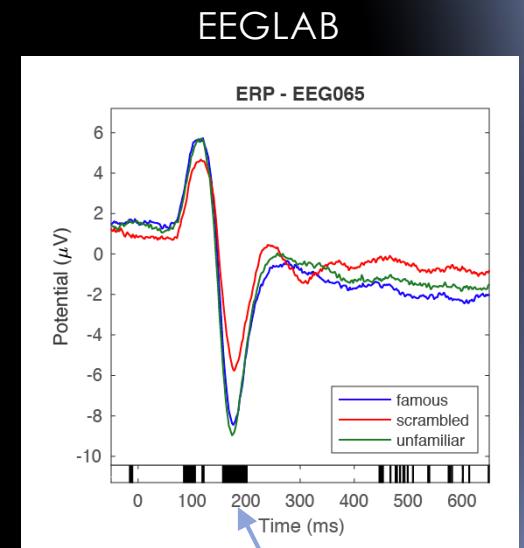
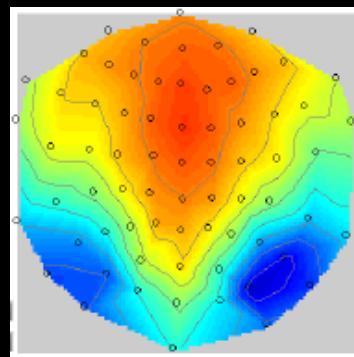


# What are we going to do?

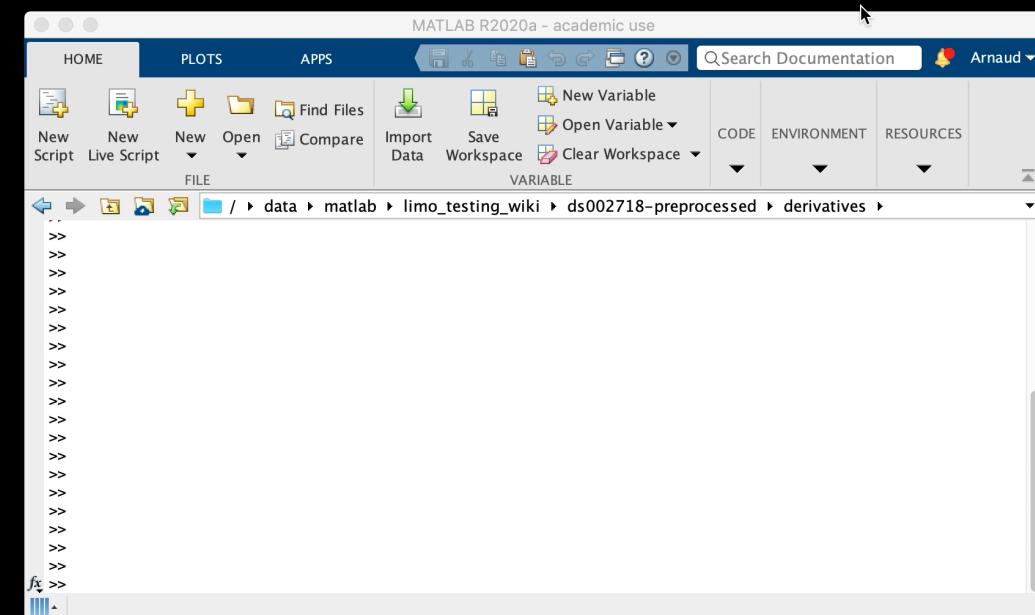
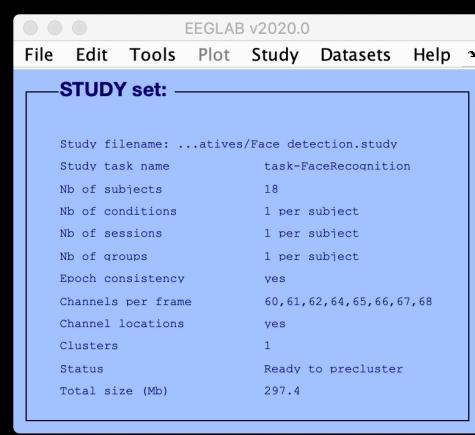
- ▶ 1 – Replicate Henson et al. – faces vs. scrambled



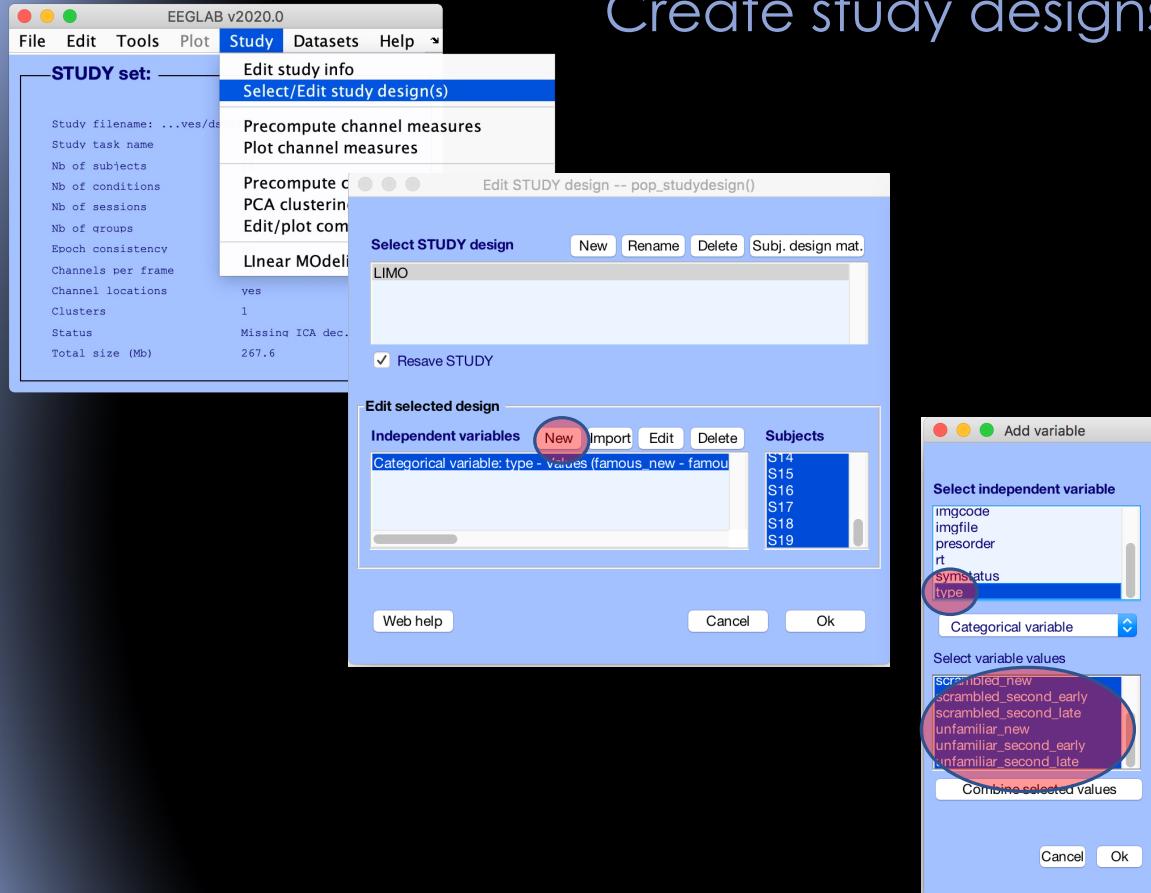
Topography 170 ms



p<0.001



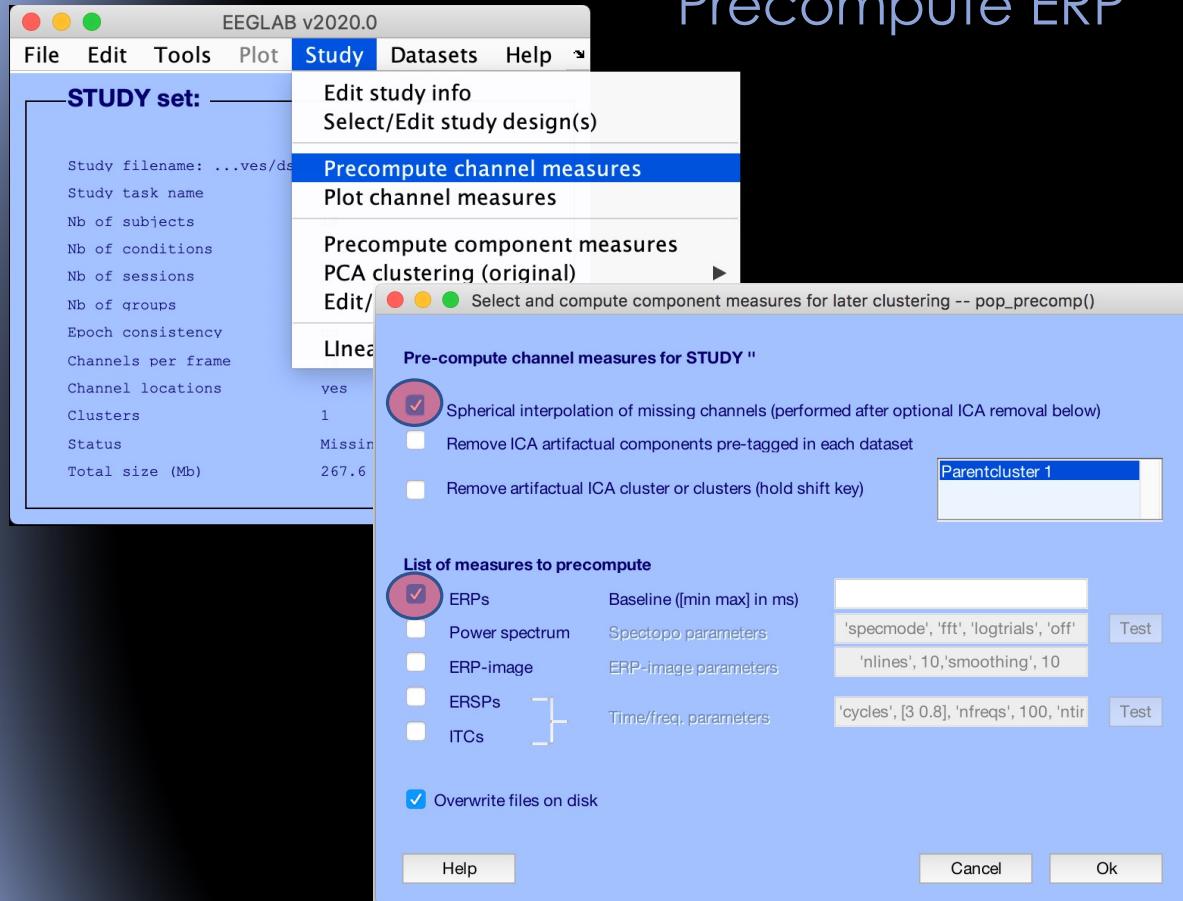
## Create study designs



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

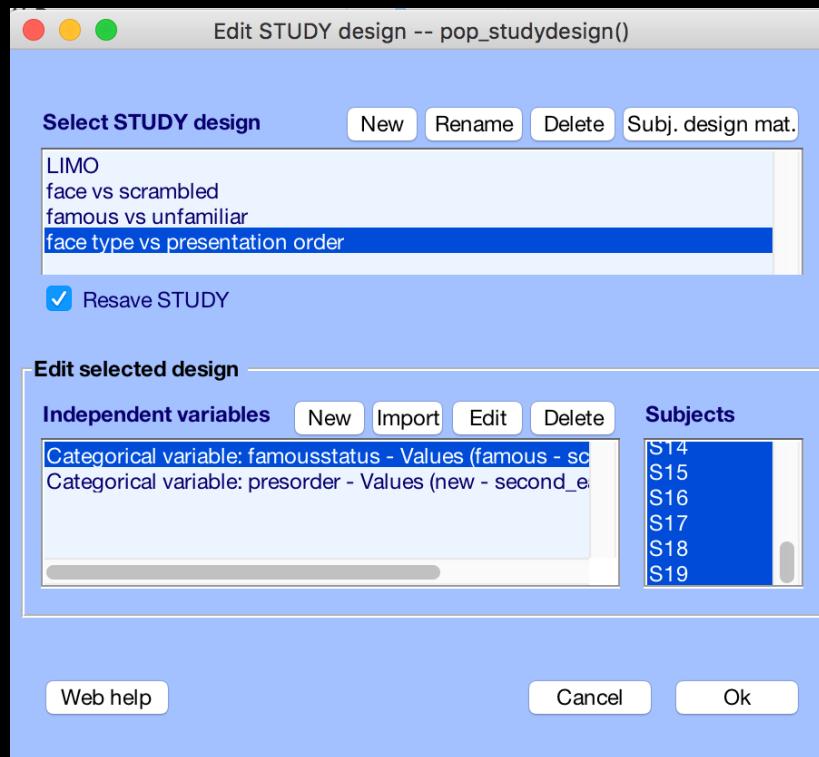
The new STUDY interface allows all sorts of designs. By default, you model each and any condition / covariate (the more complete the model the better)

# Precompute ERP



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

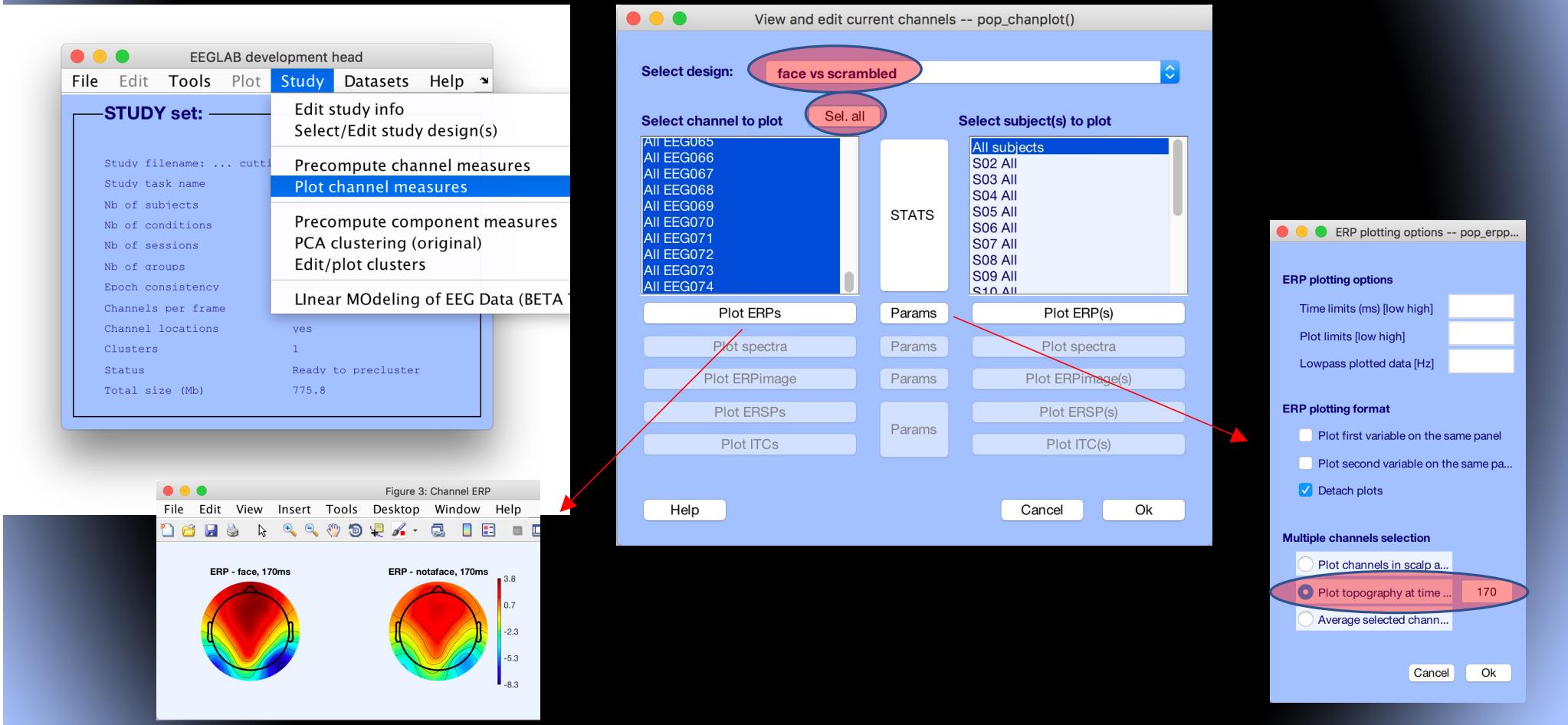
# Create other study designs



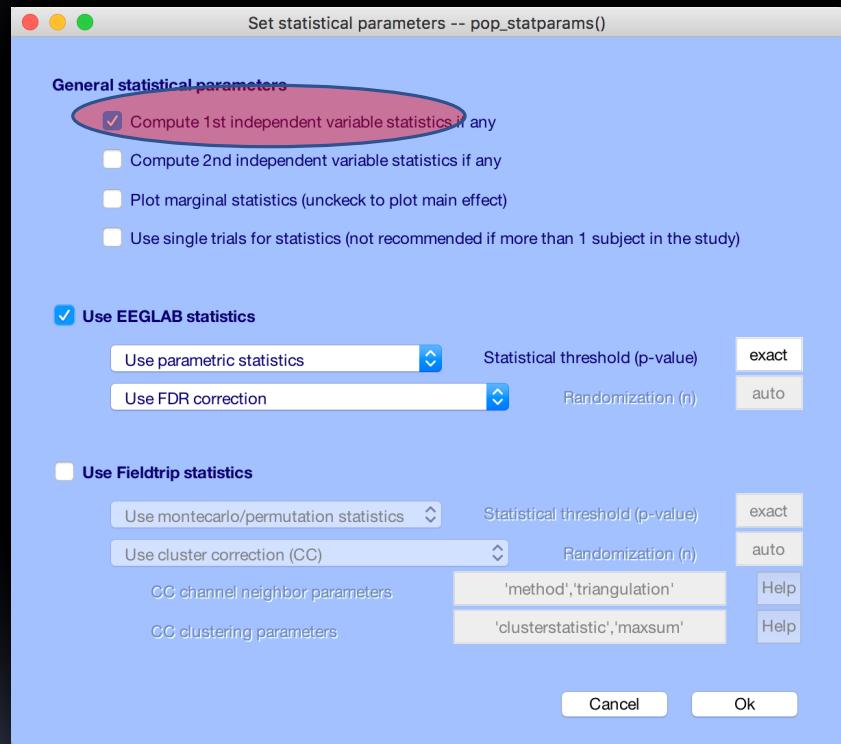
## Select/edit study design

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)

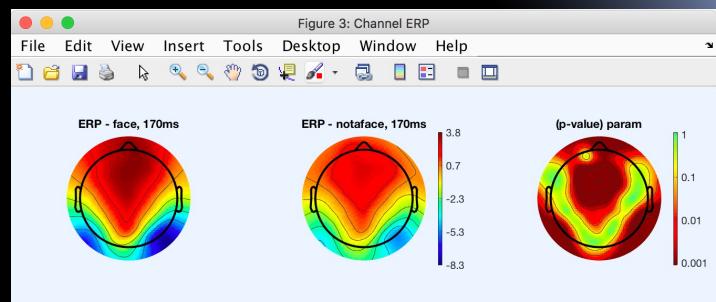
# Standard EEGLAB statistics



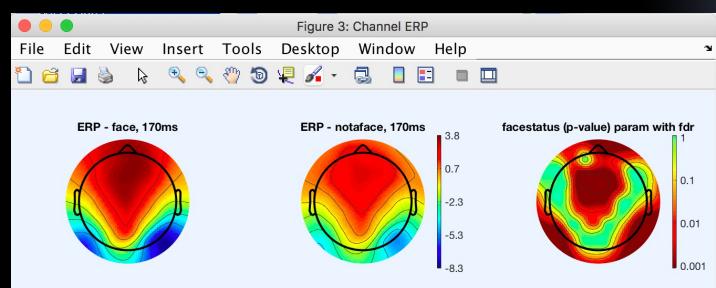
# Standard EEGLAB statistics



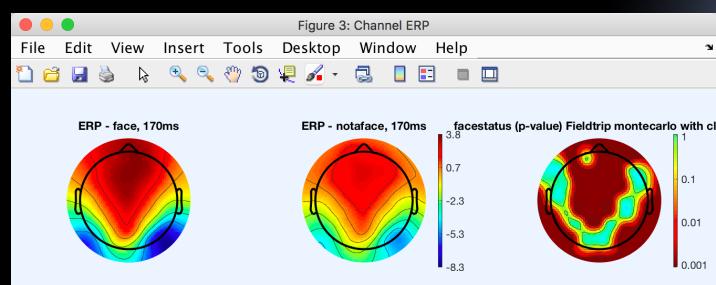
Uncorrected

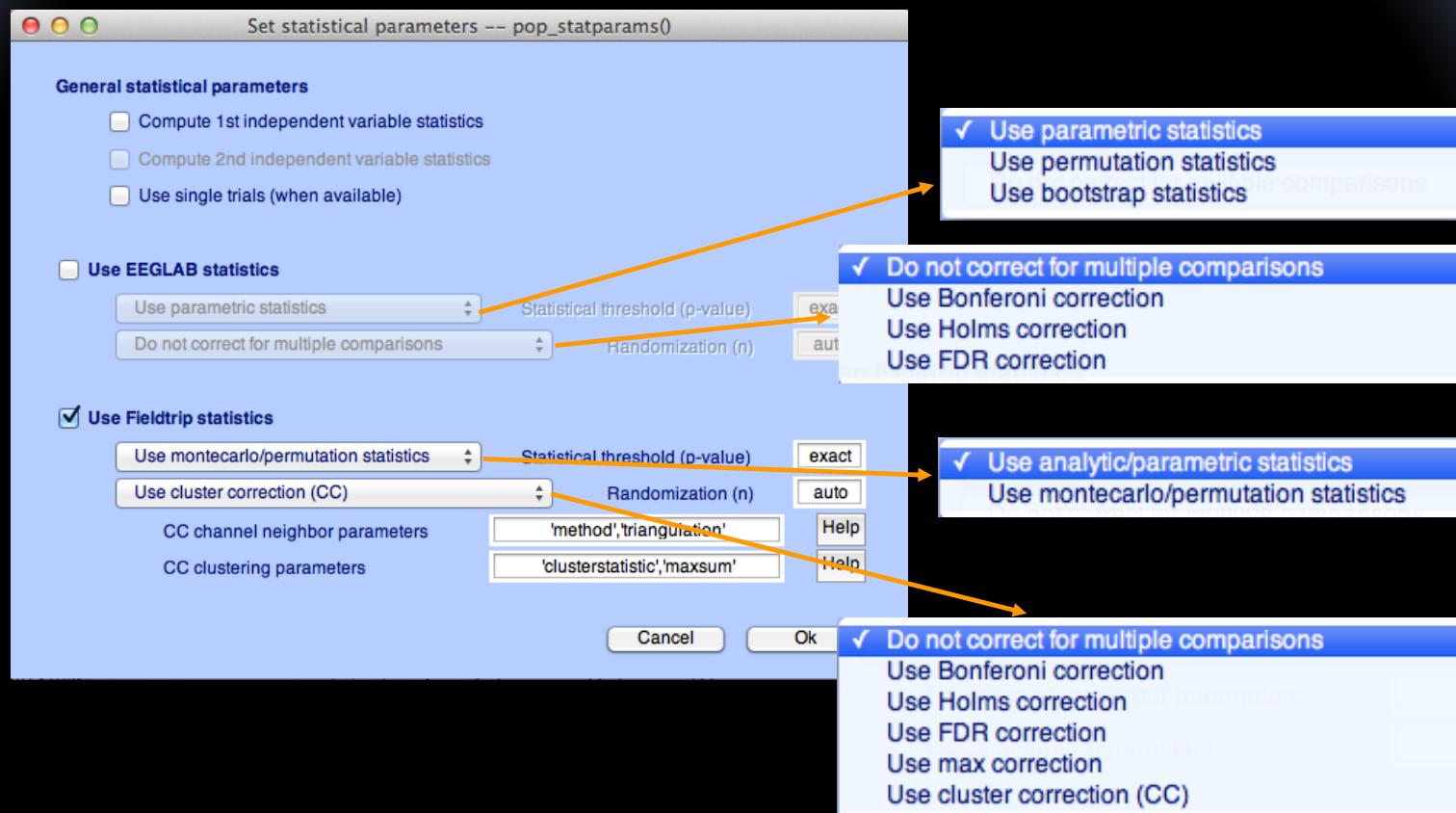


FDR corrected



Cluster corrected

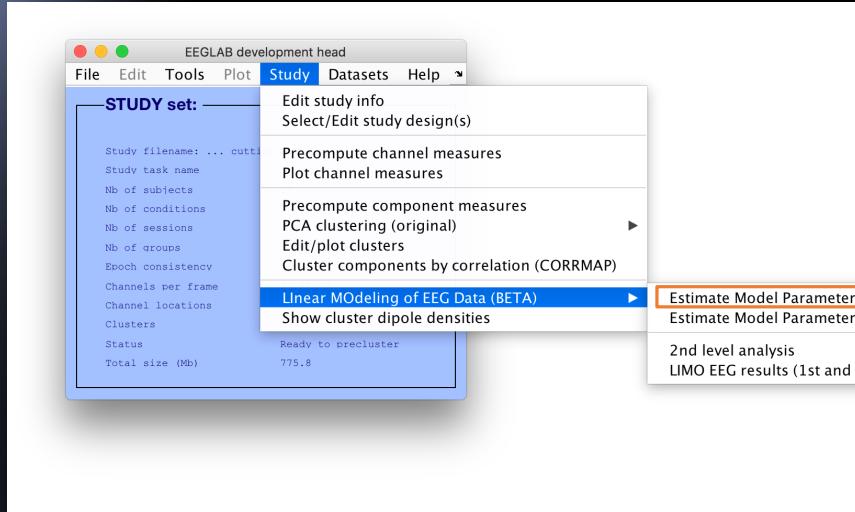




**std\_stat() function in EEGLAB**

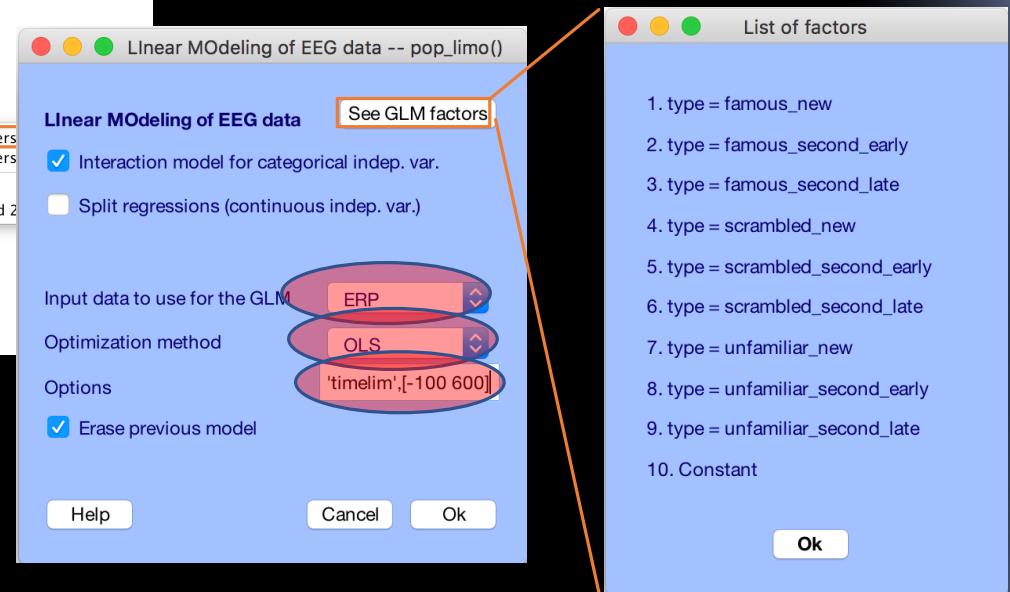
## Exercise

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 Famour 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)



Only the current design is pooled in the GLM

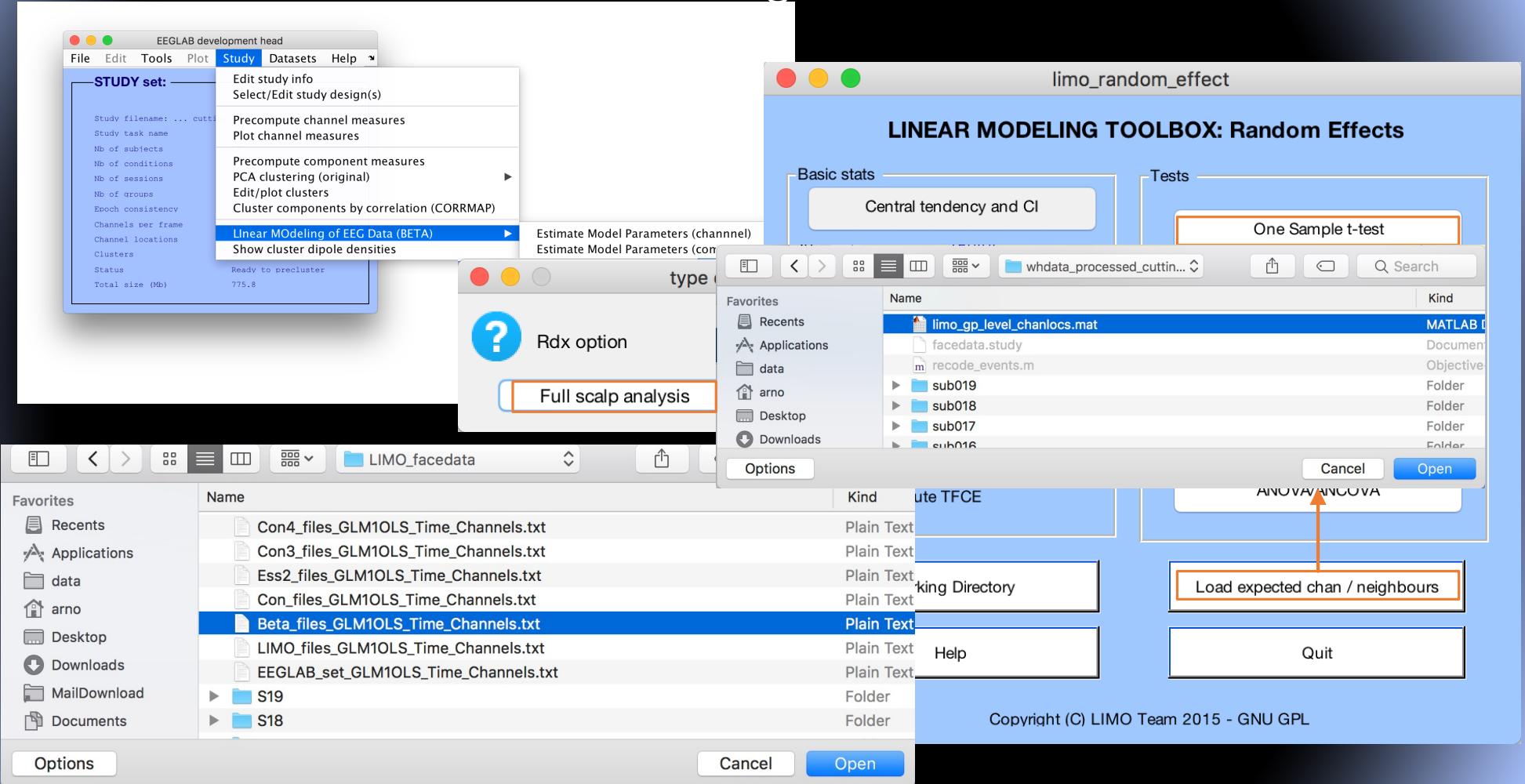
## Estimate model parameters

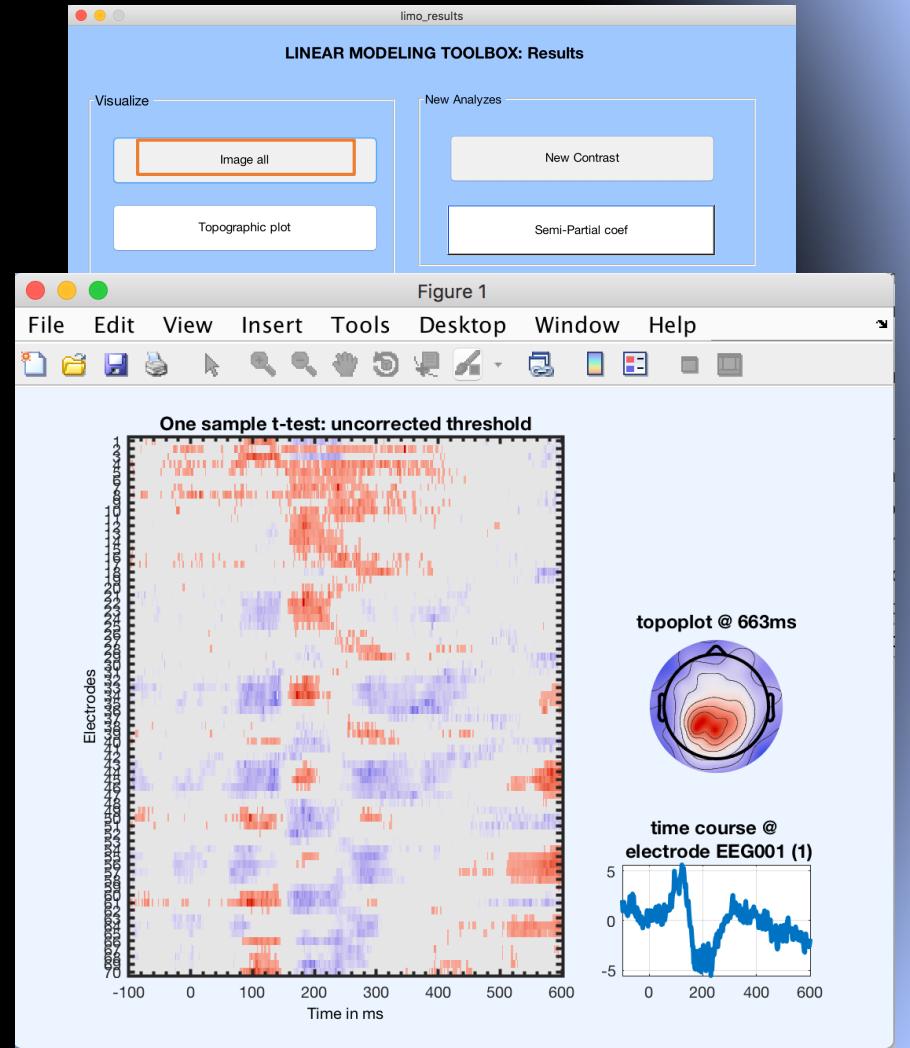
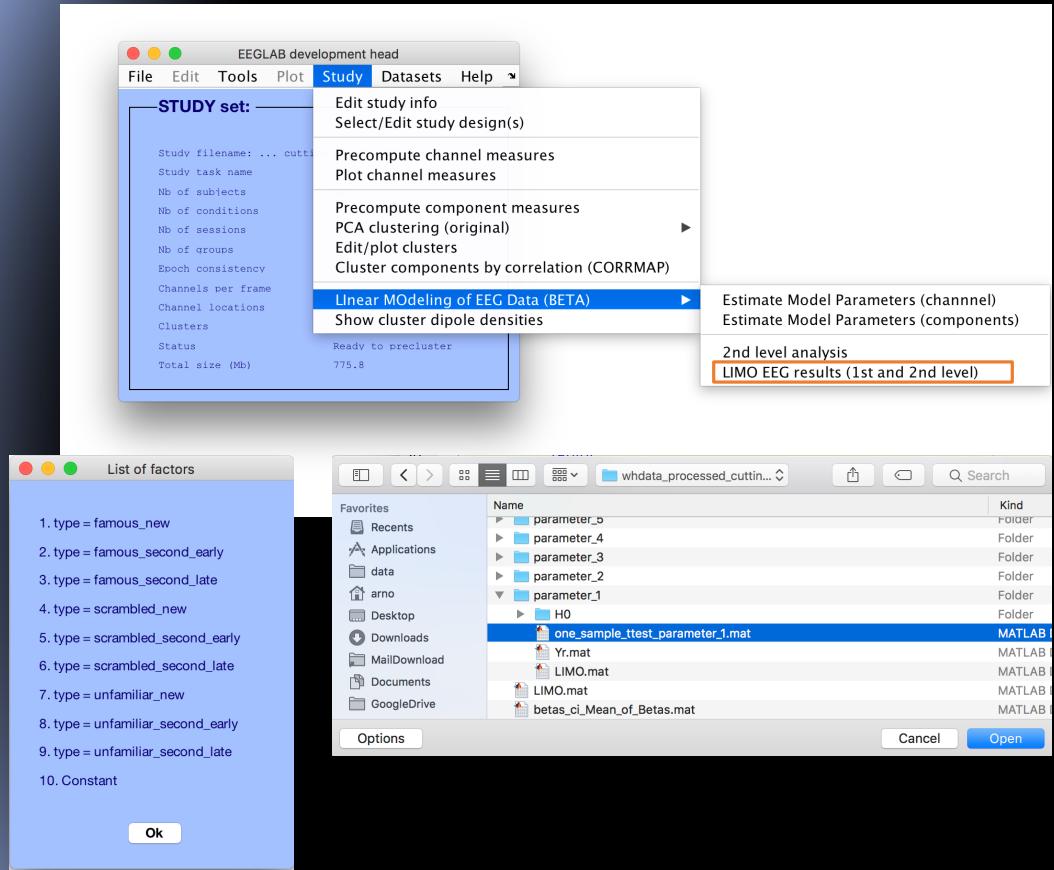


### Estimate Model Parameter

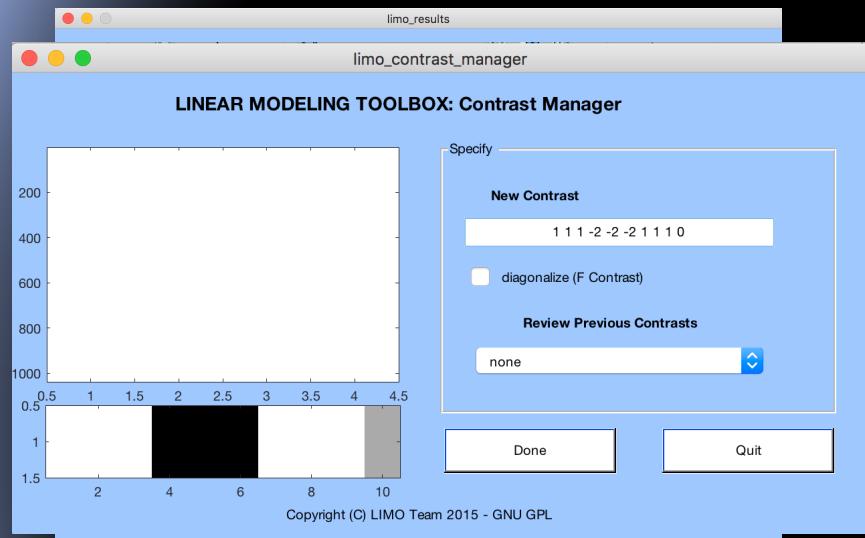
Have generated single trials, specified the model, we now do the stats  
→ Restrict 'timelim' [-50 650]

# Are Beta significant?





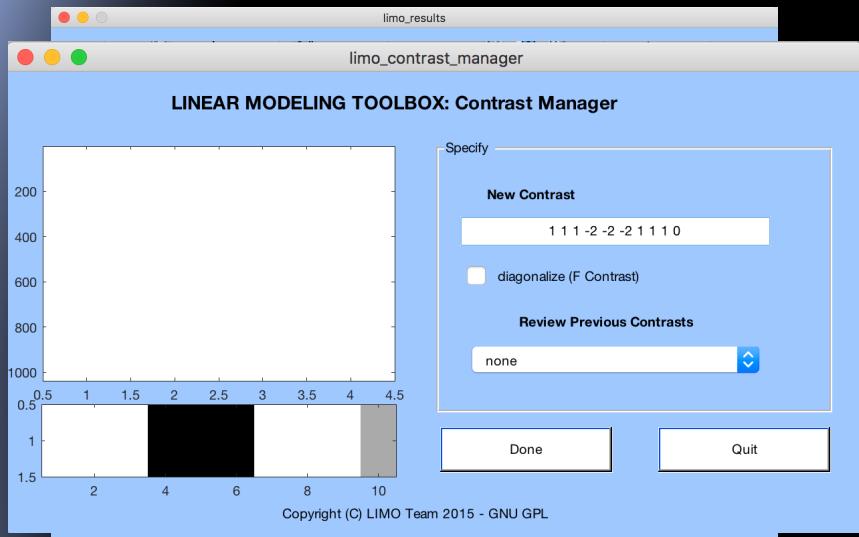
# Grouping betas and differences between conditions



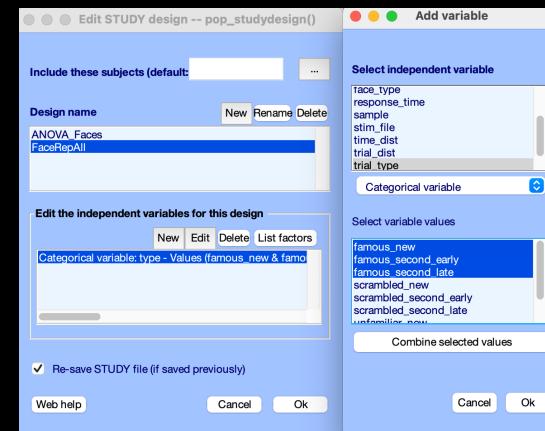
	Faces vs non-faces	Famous	Scrambled	Unfamiliar	ANOVA (famous/scrambled/unfamiliar)	ANOVA (new/early/late)
1. type = famous_new	1	1	0	0	1 0 0	1 0 0
2. type = famous_second_early	1	1	0	0	1 0 0	0 1 0
3. type = famous_second_late	1	1	0	0	1 0 0	0 0 1
4. type = scrambled_new	-2	0	1	0	0 1 0	1 0 0
5. type =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

At the bottom left is an 'Ok' button.

# Grouping betas and differences between conditions



Contrast were automatically created when we combined variables

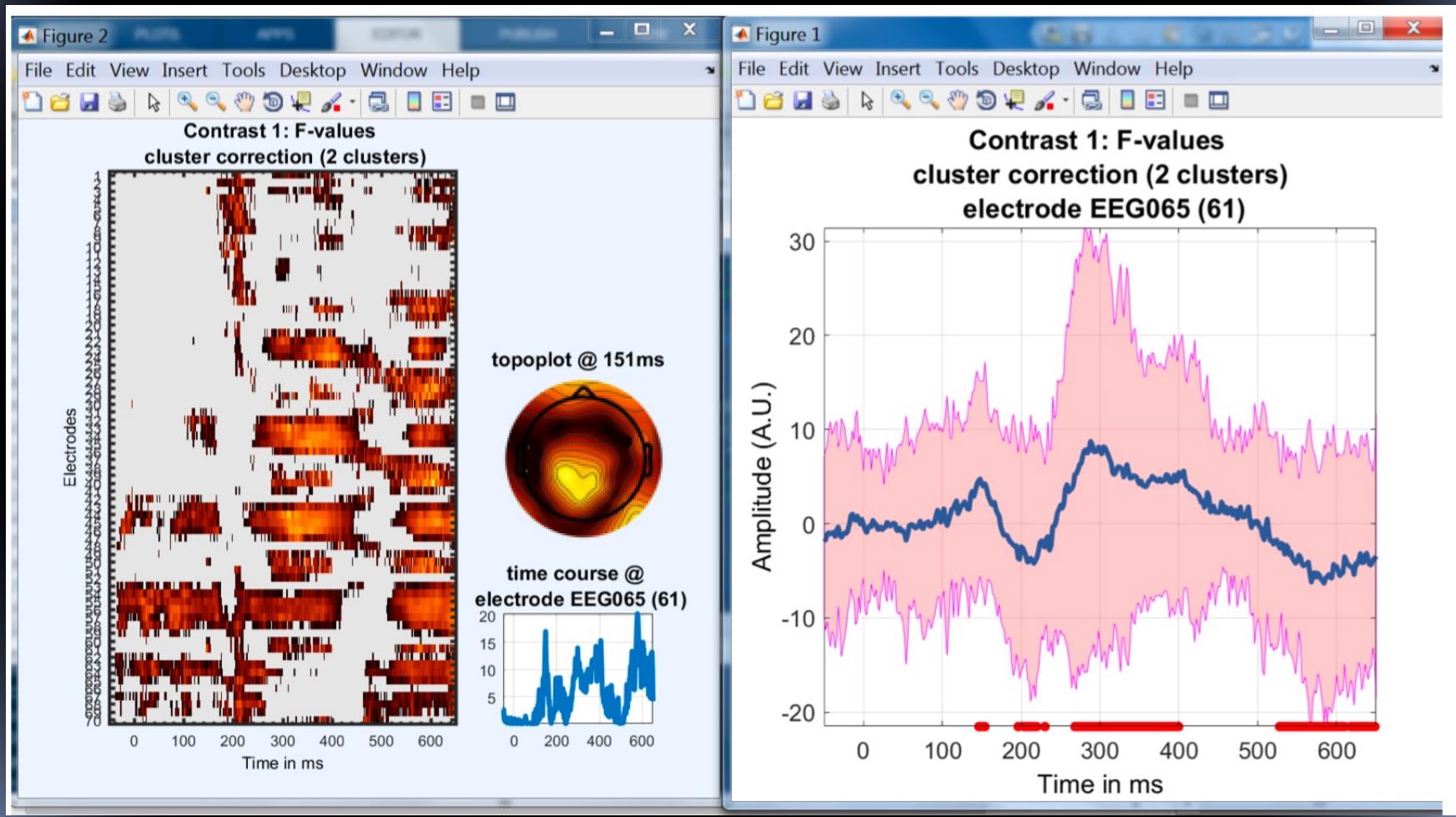


	Faces vs non-faces	Famous	Scrambled	Unfamiliar	
1. type = famous_new		1	0	0	1 0 0 1 0 0
2. type = famous_second_early		1	0	0	1 0 0 0 1 0
3. type = famous_second_late		1	0	0	1 0 0 0 0 1
4. type = scrambled_new		-2	0	1	0 1 0 1 0 0
5. type = scrambled_second_early		-2	0	1	0 1 0 0 1 0
6. type = scrambled_second_late		-2	0	1	0 1 0 0 0 1
7. type = unfamiliar_new		1	0	0	0 0 1 1 0 0
8. type = unfamiliar_second_early		1	0	0	0 0 1 0 1 0
9. type = unfamiliar_second_late		1	0	0	0 0 1 0 0 1
10. Constant		0	0	0	0 0 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 > 2<sup>nd</sup> Level analysis*)