

Robust Linear Modelling of EEG data: the LIMO EEG plug-in

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Copy the LIMOEEG folder and move limo_egg to plugins

Overview

- What is LIMMO EEG for?
- Plug-in overview
- Theory and practice
 - 1st level robust GLM
 - 2nd level robust stats
 - Design brainstorming

Accounting for within and between subjects variance

WHAT IS IT FOR?

Full 'brain' analysis

- Traditionally compute averages per condition and do your statistics on peaks
 - Peaks are NOT necessarily important events (they are likely to mark the end of a process)
 - Increase type 1 FWER by choosing electrodes
 - Averages don't account for trial variability
 - Fixed effect can be biased
 - Design flexibility

Hierarchical Linear Model

1st level analysis:

GLM: $Y = X\beta + \epsilon$

→ 1 β per column of X
(= within subject effects)

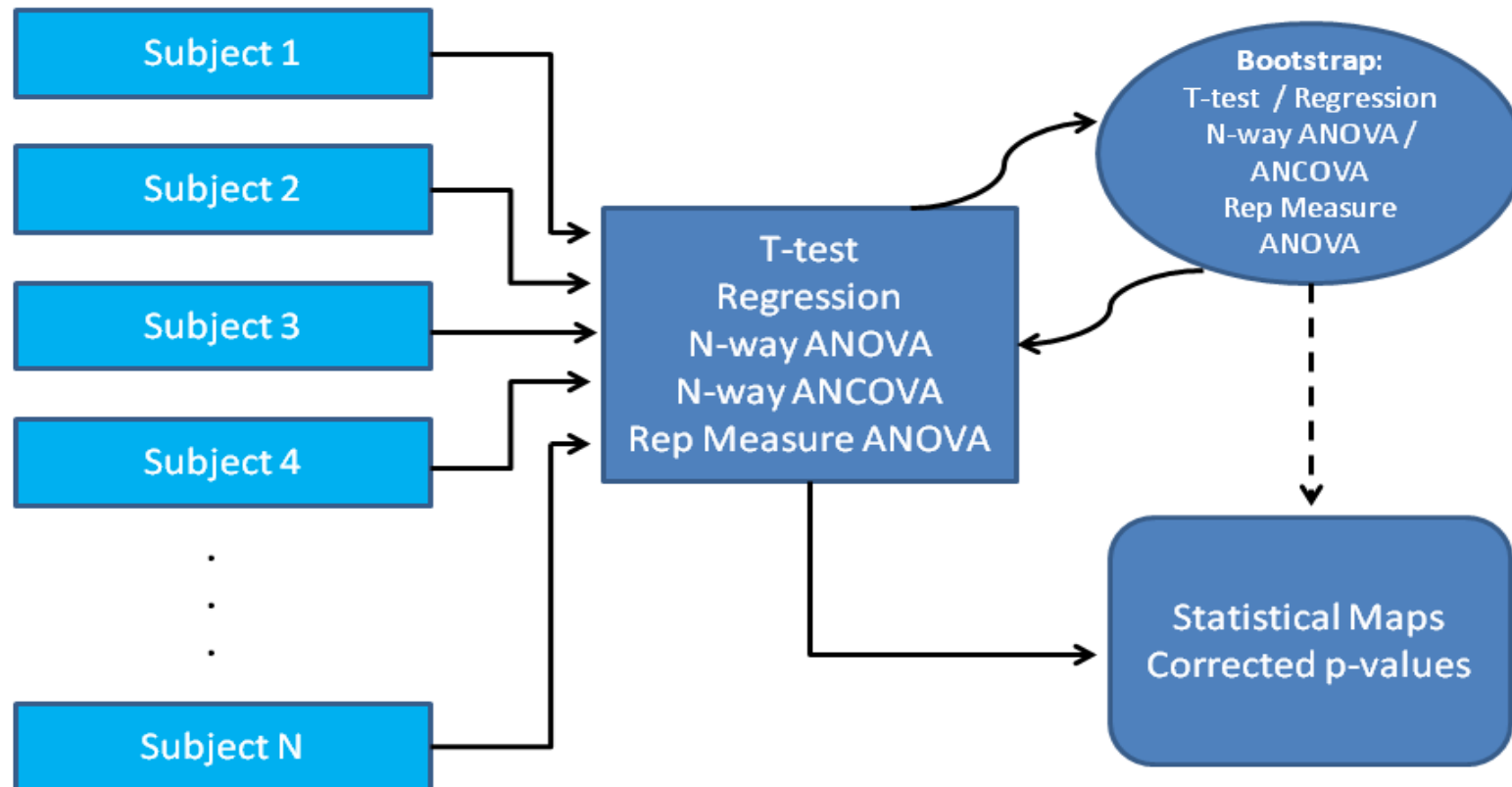
2nd level analysis:

Robust stats (Yuen t-tests, robust GLM, robust Hotelling T²)

Multiple Comparison

Correction:

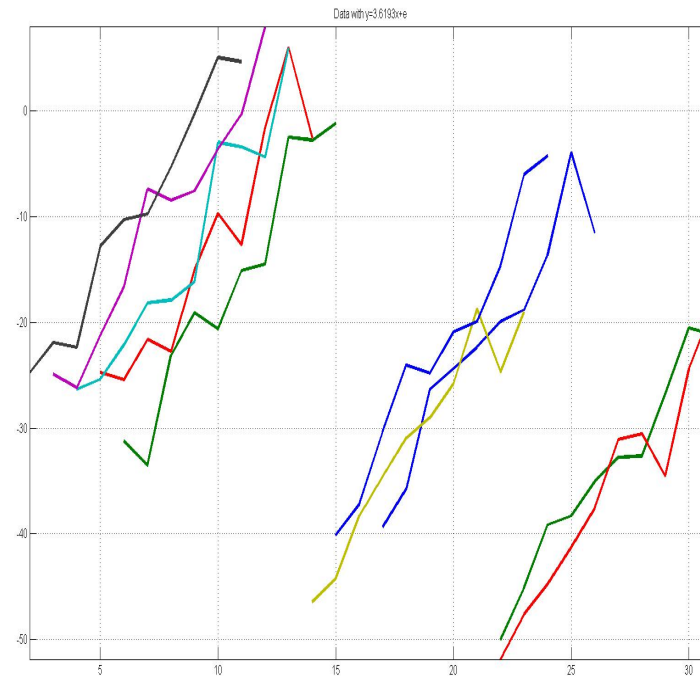
Max, Cluster-Mass, TFCE



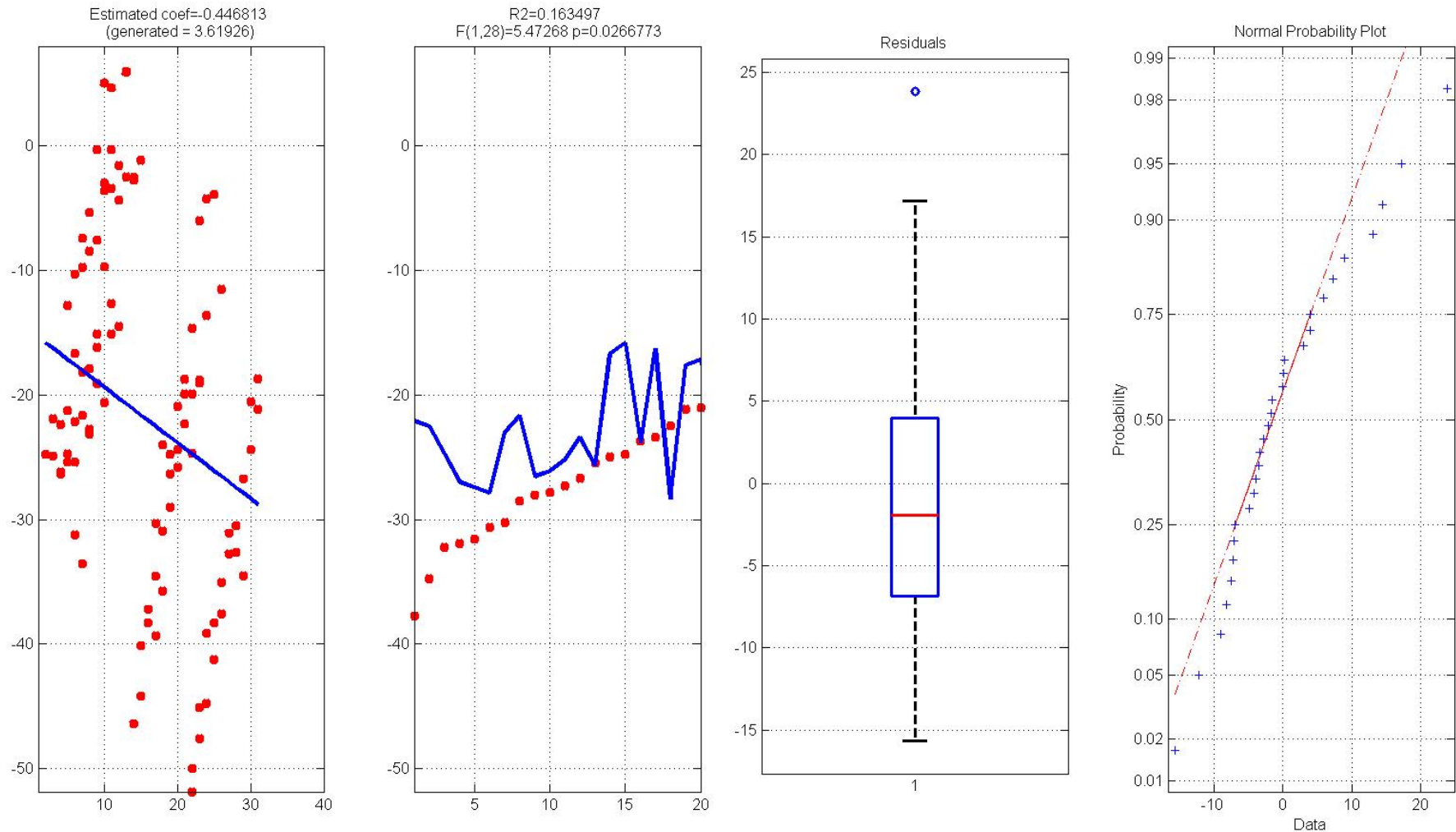
Random Effect Model

Model the data with fixed effects (the experimental conditions) and a random effect (subjects are allowed to have different overall values – considering subjects as a random variable)

Example: present stimuli from intensity -5 units to +5 units around the subject perceptual threshold and measure RT
→ Plot the data per intensity

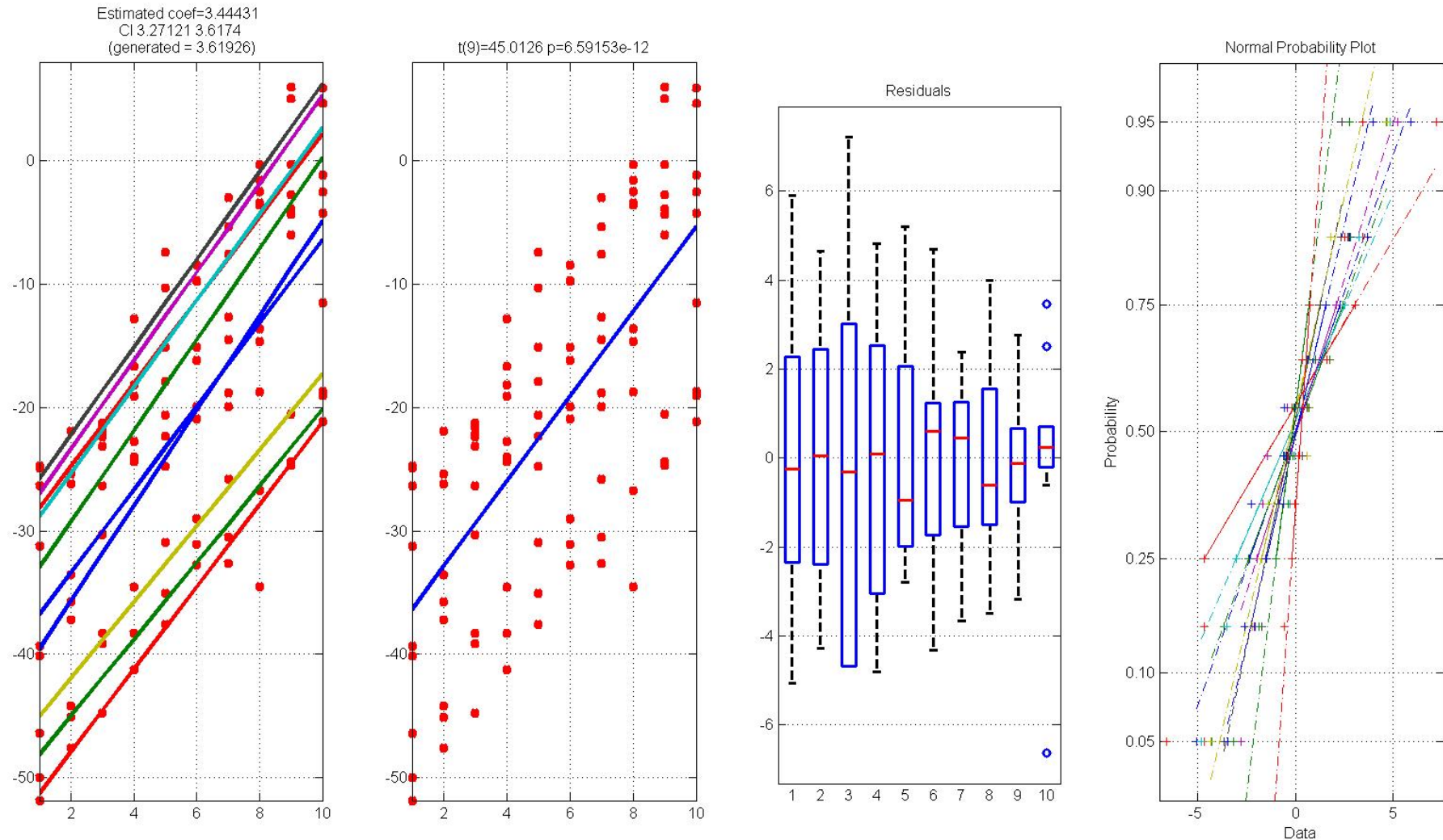


Random Effect Model



Fixed effect = average across subjects \rightarrow negative correlation?

Random Effect Model



Fixed effect = average across subjects → negative correlation?

Mixed effect = effect per subject with variable (random) offsets

Design considerations

- ALLOWS YOU TO ANALYZE ANY (PAIRED / UNPAIRED) DESIGNS

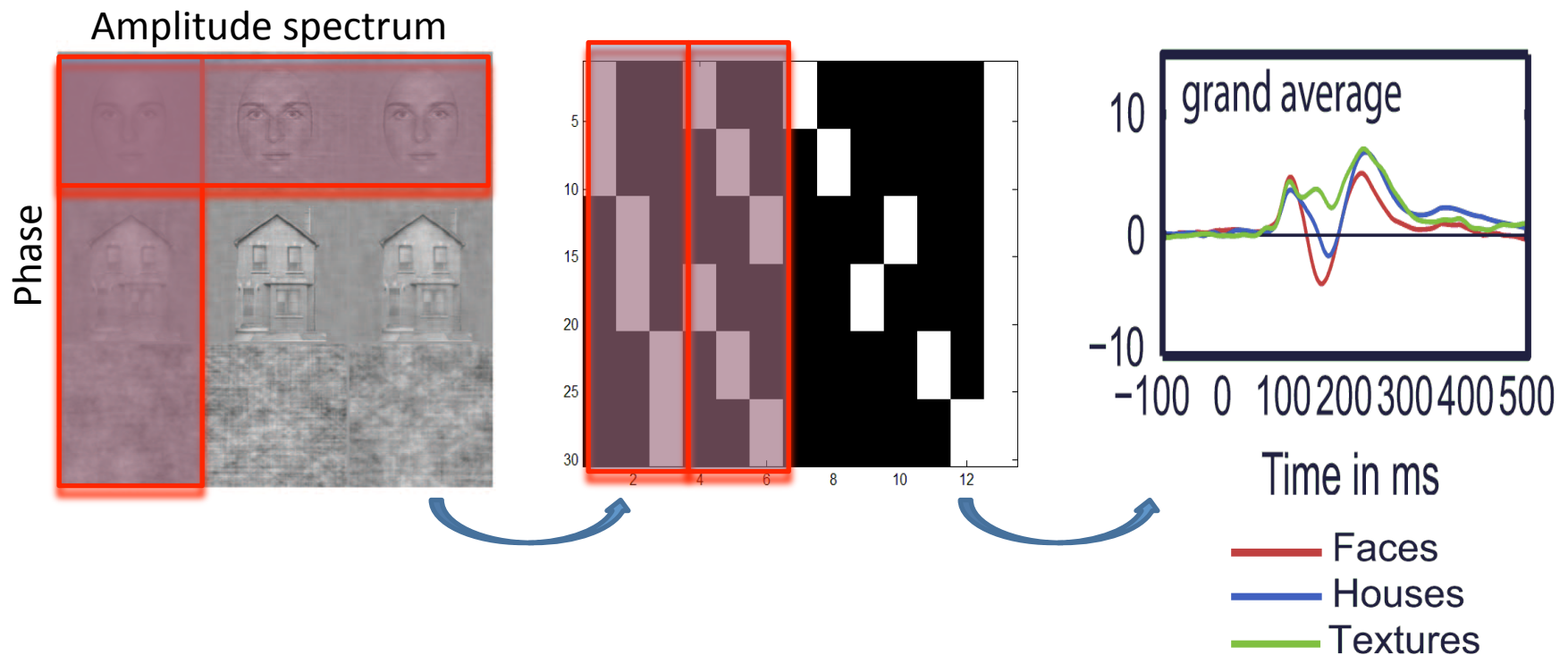
Illustration with a set of studies looking at the effect of stimulus phase information



[Rousselet, Pernet, Bennet, Sekuler \(2008\). Face phase processing. BMC Neuroscience 9:98](#)

Factorial Designs: $N*N*N*...$

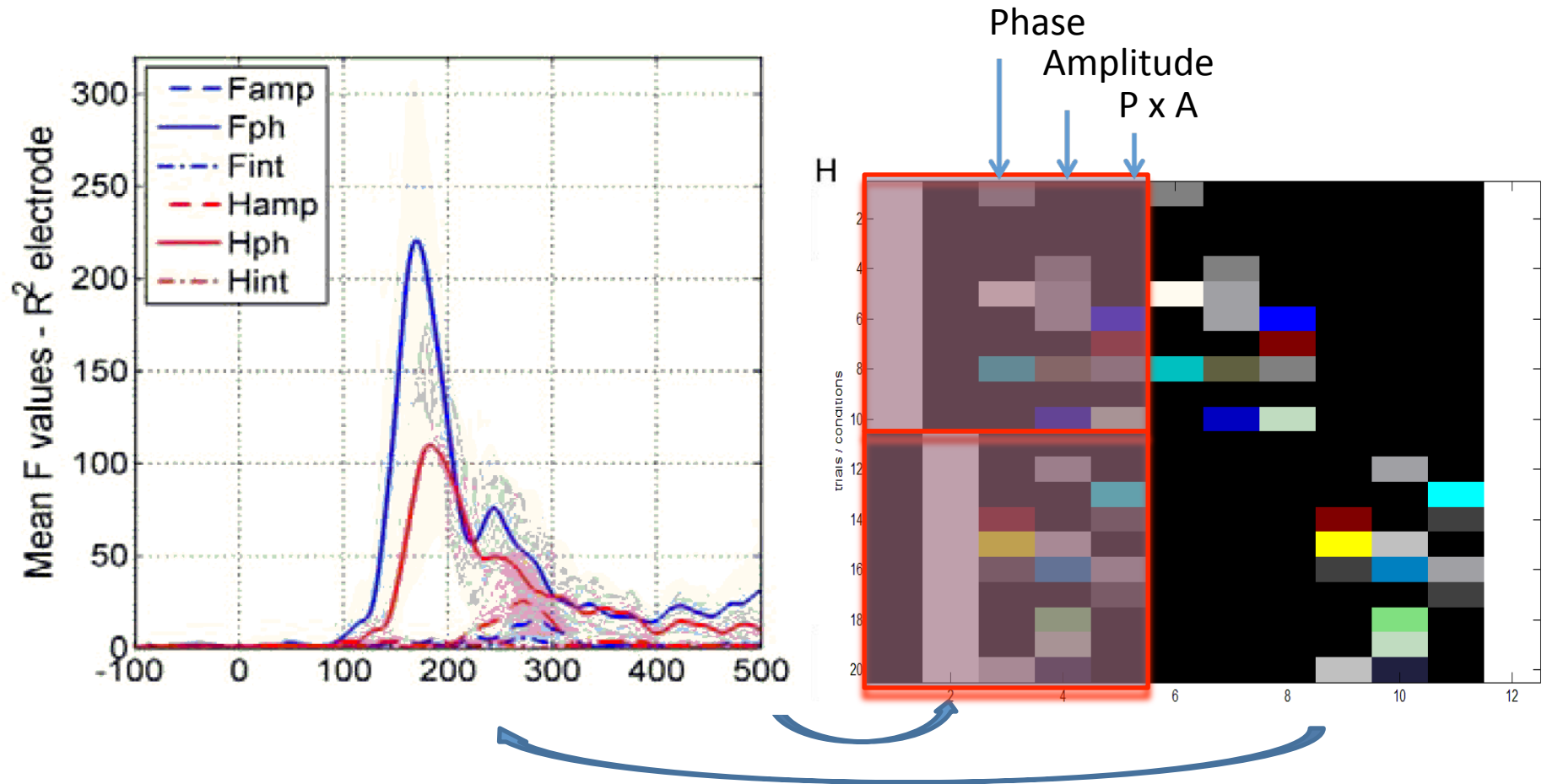
Categorical designs: Group level analyses of course but also Individual analyses with bootstrap



[Bienek, Pernet, Rousselet \(2012\). Phase vs Amplitude Spectrum. Journal of Vision 12\(13\), 1–24](#)

Regression based designs

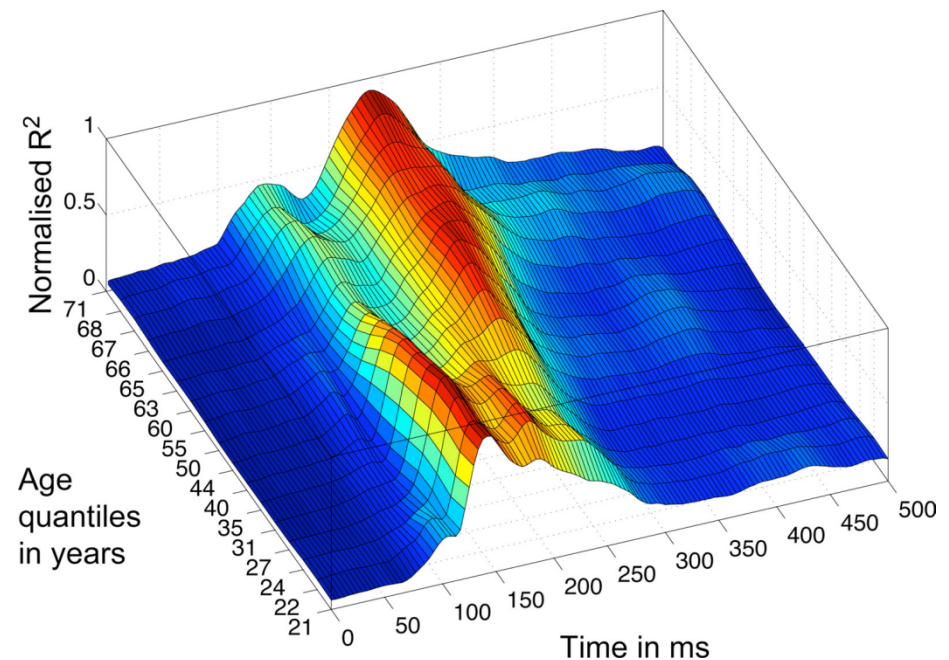
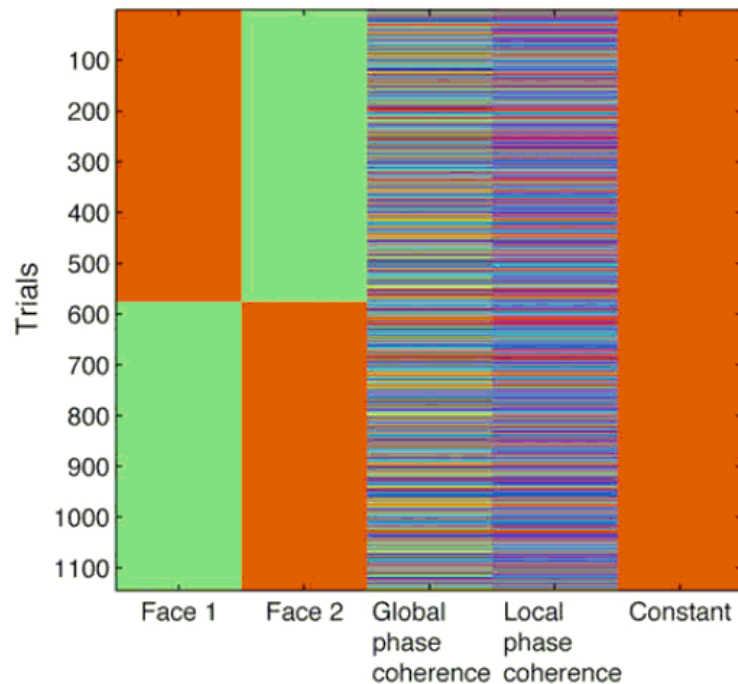
Mixed design: Control of low level physical properties



[Bienek, Pernet, Rousselet \(2012\). Phase vs Amplitude Spectrum. Journal of Vision 12\(13\), 1–24](#)

Regression based designs (2 levels)

Parametric designs:
study the effect of stimulus properties within subjects
effect of aging between subjects



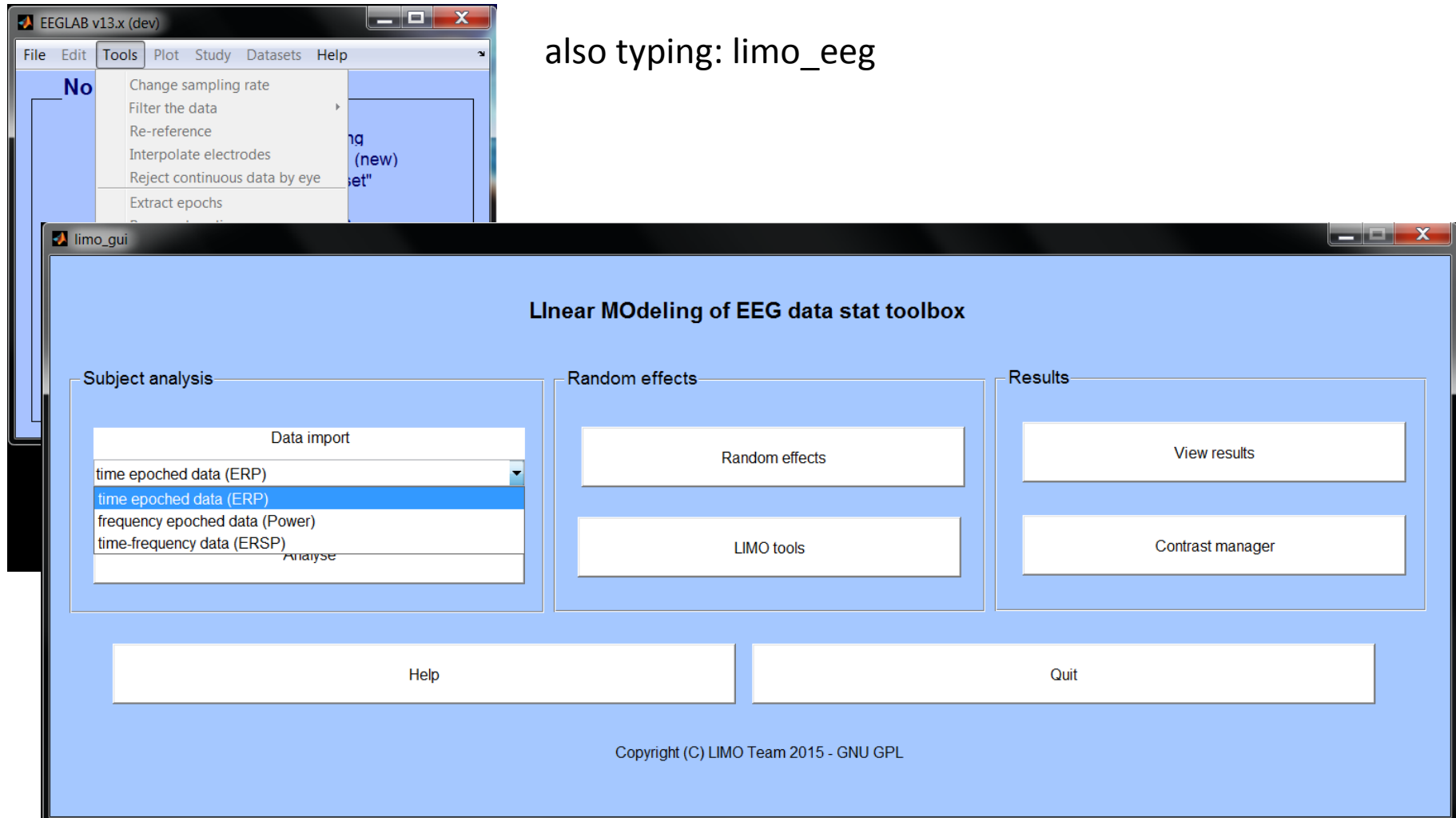
[Rousselet, Gaspar, Pernet, Husk, Bennett, Sekuler \(2010\). Aging and face perception. Front Psy](#)

Quick tour of the interface

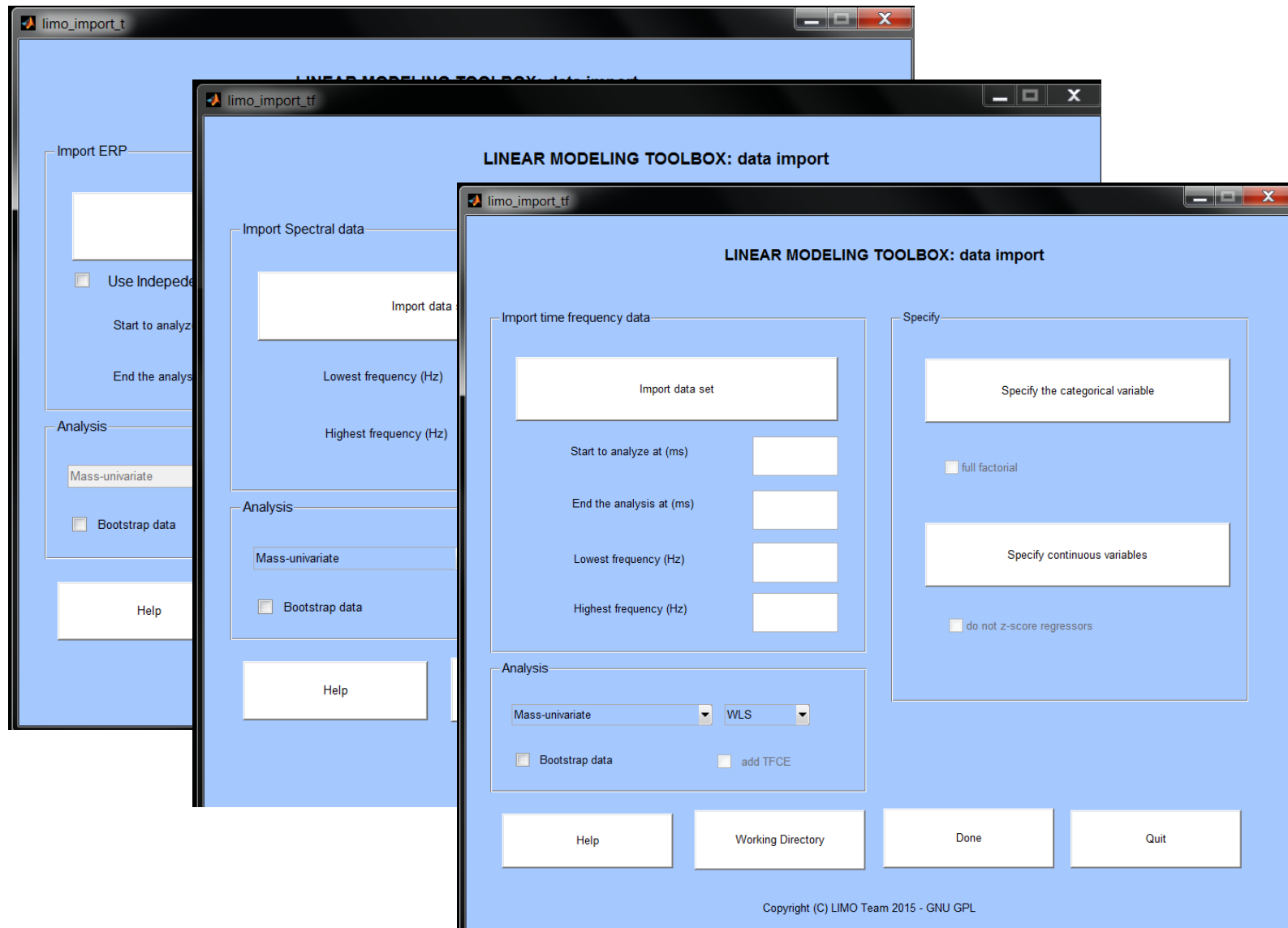
PLUG-IN OVERVIEW

Main GUI

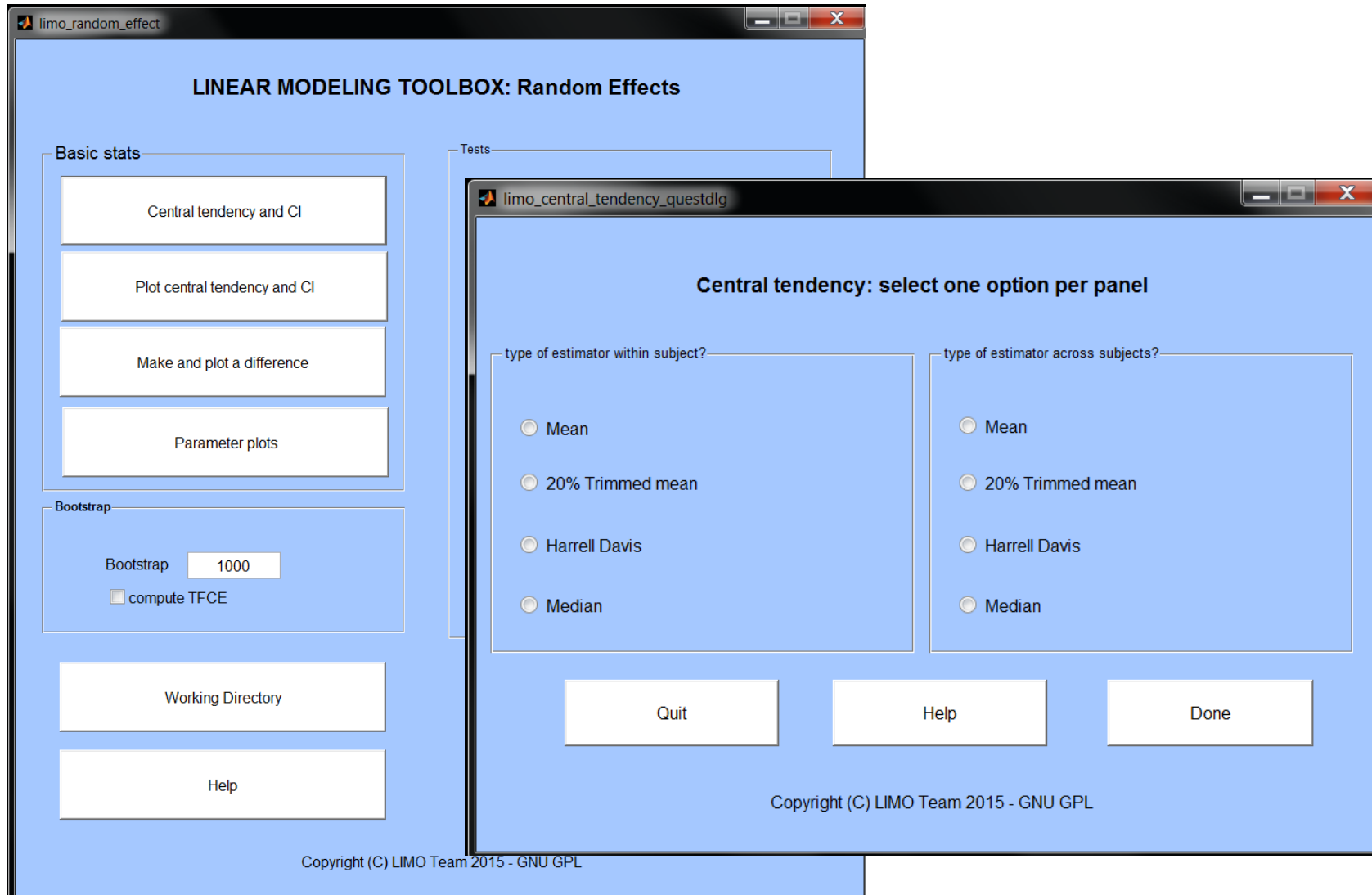
also typing: limo_eeg



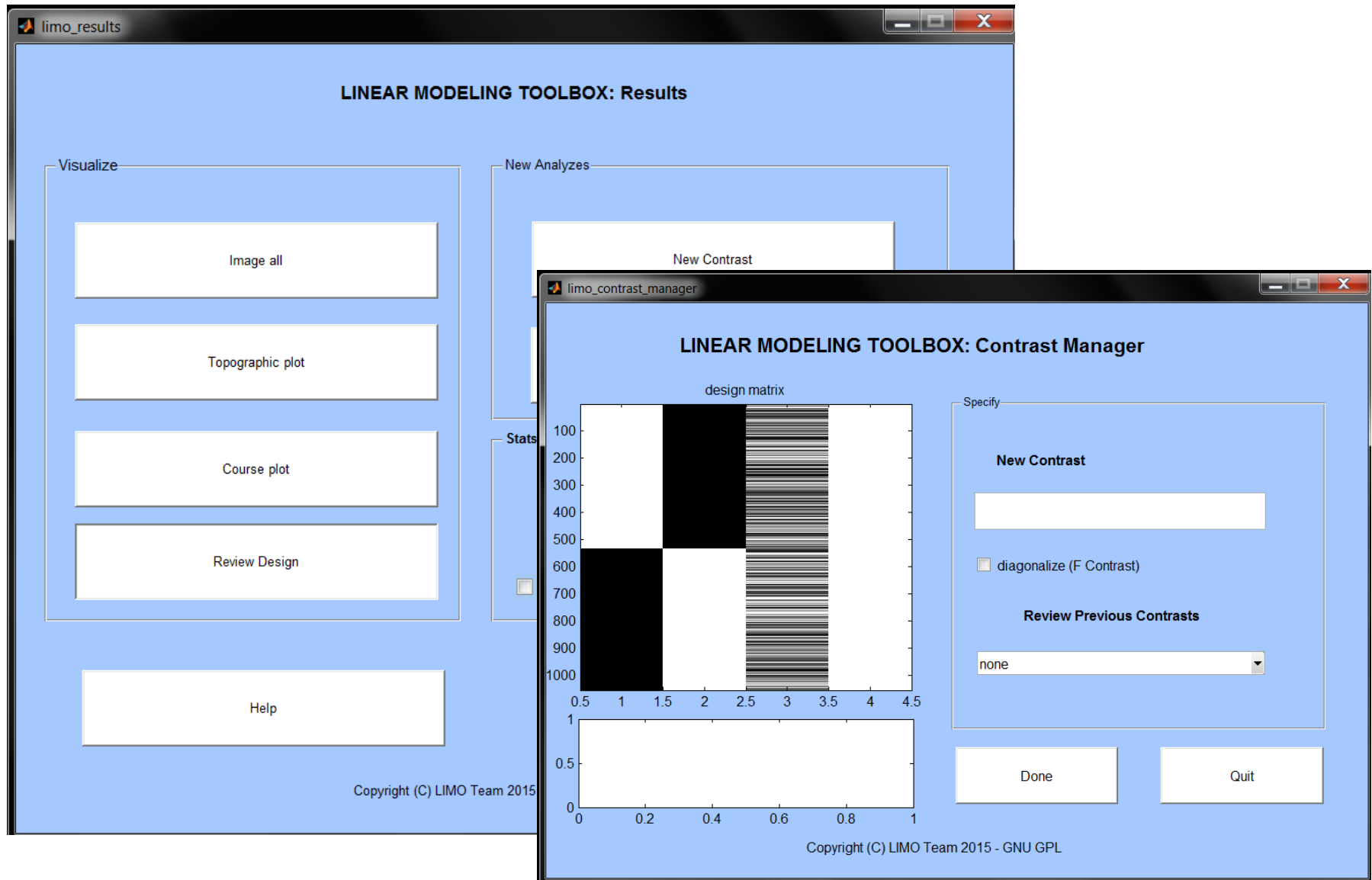
Import (1st level)



Random Effects (2nd level)



Results



GLM, robust stats, GLM, robust stats, GLM, robust stats, GLM, robust stat ..

THEORY AND PRACTICE

Linearity

- Means created by lines
- In maths it refers to equations or functions that satisfy 2 properties: additivity (also called superposition) and homogeneity of degree 1 (also called scaling)
- Additivity $\rightarrow y = x_1 + x_2$ (output y is the sum of inputs x s)
- Scaling $\rightarrow y = \beta x_1$ (output y is proportional to input x)

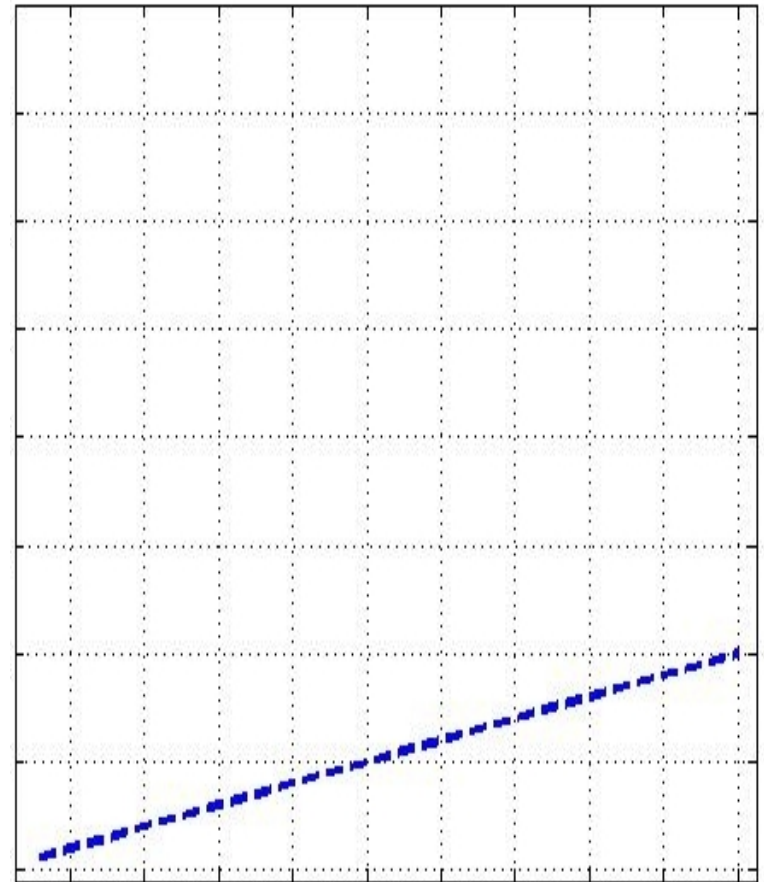
<http://en.wikipedia.org/wiki/Linear>

What is a linear model?

- An equation or a set of equations that models data and which corresponds geometrically to straight lines, planes, hyper-planes and satisfy the properties of additivity and scaling.
- Simple regression: $y = \beta_1 x + \beta_2 + \varepsilon$
- Multiple regression: $y = \beta_1 x_1 + \beta_2 x_2 + \beta_3 + \varepsilon$
- One way ANOVA: $y = u + \alpha_i + \varepsilon$
- Repeated measure ANOVA: $y = u + \alpha_i + \varepsilon$
- ...

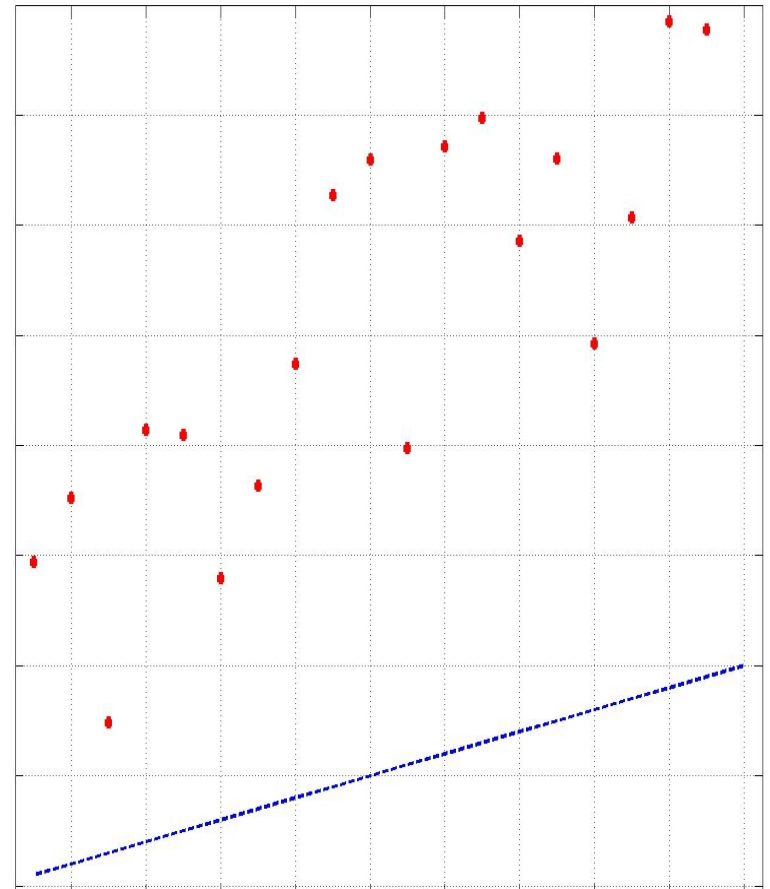
A regression is a linear model

- We have an experimental measure x (e.g. stimulus intensity from 0 to 20)



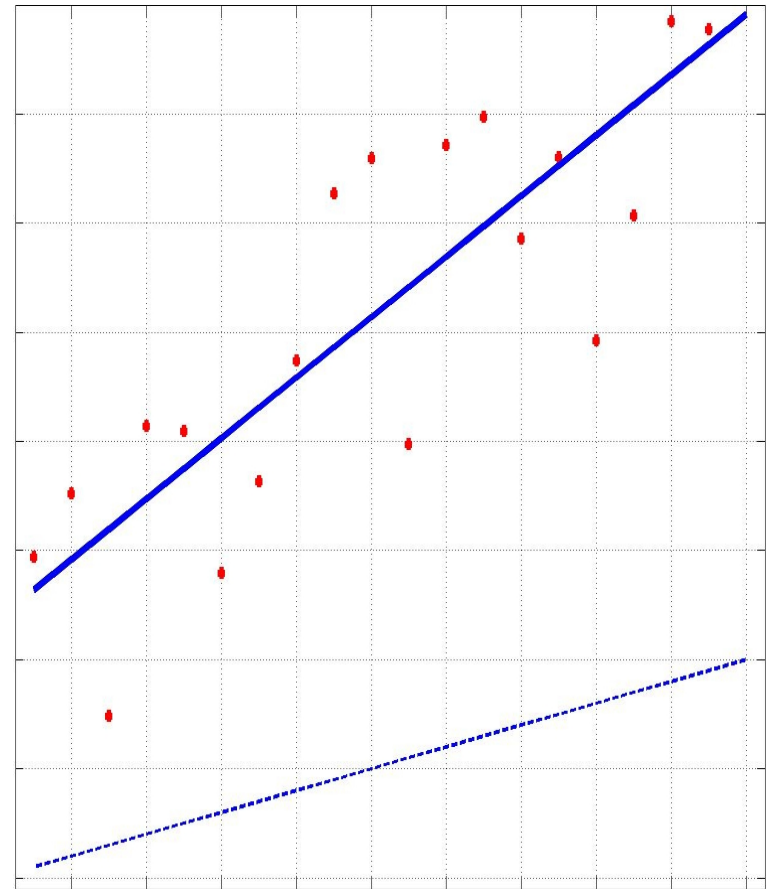
A regression is a linear model

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- We then do the expe and collect data y (e.g. RTs)



A regression is a linear model

- We have an experimental measure x (e.g. stimulus intensity from 0 to 20)
- We then do the expe and collect data y (e.g. RTs)
- Model: $y = \beta_1 x + \beta_2$
- Do some maths / run a software to find β_1 and β_2
- $\hat{y} = 2.7x + 23.6$



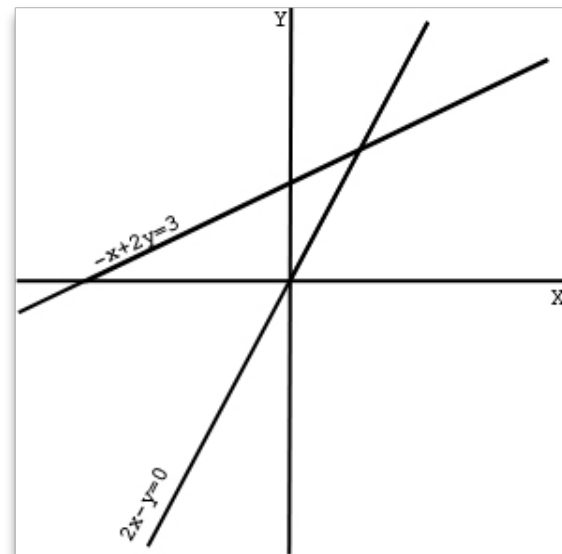
Linear algebra for regression

- Linear algebra has to do with solving linear systems, i.e. a set of linear equations
- For instance we have observations (y) for a stimulus characterized by its properties x_1 and x_2 such as $y = x_1 \beta_1 + x_2 \beta_2$

$$2\beta_1 - \beta_2 = 0$$

$$-\beta_1 + 2\beta_2 = 3$$

$$\beta_1 = 1 ; \beta_2 = 2$$



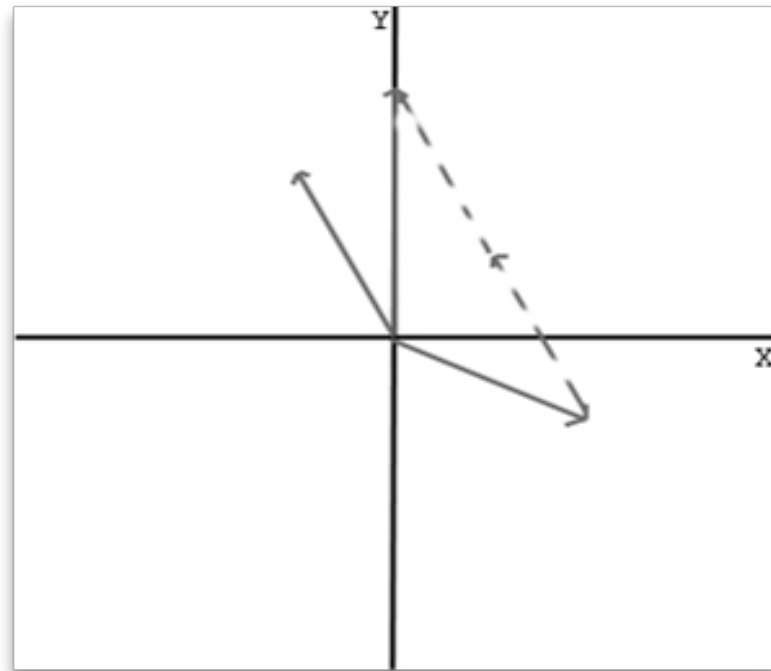
Linear algebra for regression

- With matrices, we change the perspective and try to combine columns instead of rows, i.e. we look for the coefficients with allow the linear combination of vectors

$$\begin{aligned}2\beta_1 - \beta_2 &= 0 \\ -\beta_1 + 2\beta_2 &= 3\end{aligned}$$

$$\begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \end{bmatrix}$$

$$\beta_1 = 1 ; \beta_2 = 2$$



Linear algebra for ANOVA

- In text books we have $y = u + x_i + \varepsilon$, that is to say the data (e.g. RT) = a constant term (grand mean u) + the effect of a treatment (x_i) and the error term (ε)
- In a regression x_i takes several values like e.g. [1:20]
- In an ANOVA x_i is designed to represent groups using 1 and 0

Linear algebra for ANOVA

Y	Gp
8	1
9	1
7	1
5	2
7	2
3	2
3	3
4	3
1	3
6	4
4	4
9	4

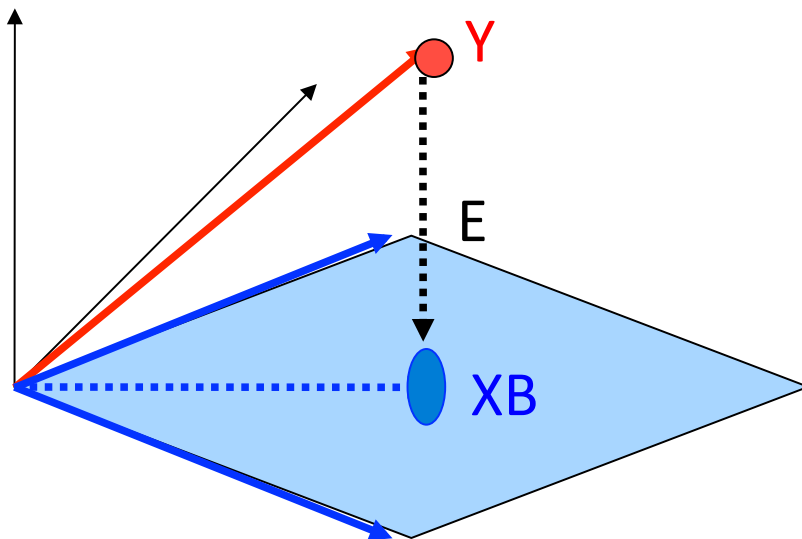
$$\begin{aligned}
 y(1..3)1 &= 1x1 + 0x2 + 0x3 + 0x4 + c + e11 \\
 y(1..3)2 &= 0x1 + 1x2 + 0x3 + 0x4 + c + e12 \\
 y(1..3)3 &= 0x1 + 0x2 + 1x3 + 0x4 + c + e13 \\
 y(1..3)4 &= 0x1 + 0x2 + 0x3 + 1x4 + c + e14
 \end{aligned}$$

$$\begin{pmatrix} 8 \\ 9 \\ 7 \\ 5 \\ 7 \\ 3 \\ 3 \\ 4 \\ 1 \\ 6 \\ 4 \\ 9 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \end{pmatrix} * \begin{pmatrix} \beta_1 \\ \beta_2 \\ \beta_3 \\ \beta_4 \\ c \end{pmatrix} + \begin{pmatrix} e11 \\ e12 \\ e13 \\ e14 \end{pmatrix}$$

→ This is like the multiple regression except that we have ones and zeros instead of 'real' values so we can solve the same way

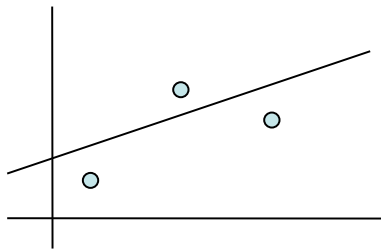
Linear Algebra, geometry and Statistics

- Y = 3 observations X = 2 regressors
- $Y = XB + E \rightarrow B = \text{inv}(X'X)X'Y \rightarrow \hat{Y} = XB$



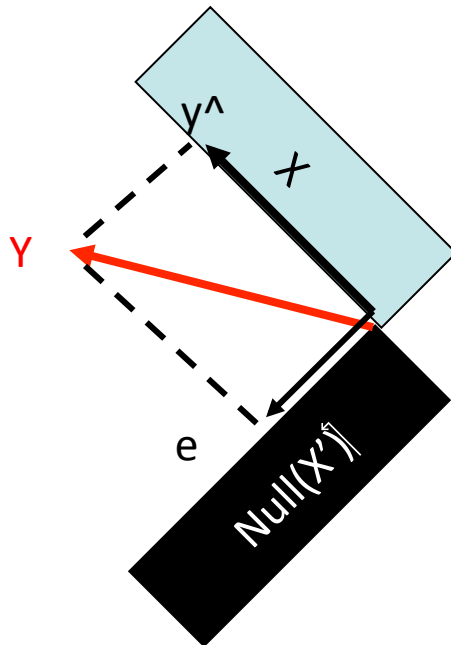
SS total = variance in Y
SS effect = variance in XB
SS error = variance in E
 $R^2 = \text{SS effect} / \text{SS total}$
 $F = \text{SS effect/df} / \text{SS error/dfe}$

Linear Algebra, geometry and Statistics



$$y = \beta x + c$$

Projecting the points on the line at perpendicular angles minimizes the distance²



$$Y = \hat{y} + e$$

$$P = X \text{inv}(X'X) X'$$

$$\hat{y} = PY$$

$$e = (I - P)Y$$

An 'effect' is defined by which part of X to test (i.e. project on a subspace)

$$R0 = I - (X0 * \text{pinv}(X0));$$

$$P = R0 - R;$$

$$\text{Effect} = (B' * X' * P * X * B);$$

Linear Algebra, geometry and Statistics

- Projections are great because we can now constrain \hat{Y} to move along any combinations of the columns of X
- Say you now want to contrast gp1 vs gp2 in a ANOVA with 3 gp, do $C = [1 \ -1 \ 0 \ 0]$
- Compute B so we have XB based on the full model X then using $P(C(X))$ we project \hat{Y} onto the constrained model (think doing a multiple regression gives different coef than multiple simple regression \rightarrow project on different spaces)



UH-OH...LOOKS
LIKE MATH
ANXIETY...

GET
THE
INTEGRAL
SIGN.



Let's analyse one subject

- **Design:** 2 faces (cond1/cond2) + a continuous variable related to the phase information in the stimulus space (\sim noise)
- LIMBO EEG – 1st level analysis
= make a parameter file per condition (like we would for erp)

Let's analyse one subject

The image shows the MATLAB R2013a environment with the LIMO GUI open. The Command Window displays the following text:

```
>> limo_eeg
LIMO_EEG was primarily designed by Cyril Pernet and Guillaume Rousselet,
with the contribution of Andrew Stewart, Nicolas Chauveau, Carl Gaspar,
Luisa Frei, Ignacio Suay Mas and Marianne Latinus.
These authors are thereafter referred as the LIMO Team

LIMO_EEG Copyright (C) 2015 LIMO TEAM
This program comes with ABSOLUTELY NO WARRANTY.
This is free software, and you are welcome to redistribute it under certain conditions

LIMO EEG Ref:
Pernet, C.R., Chauveau, N.,
LIMO EEG: a toolbox for hierarchical
Computational Intelligence and
```

The LIMO GUI, titled "Linear Modeling of EEG data stat toolbox", is divided into three main sections:

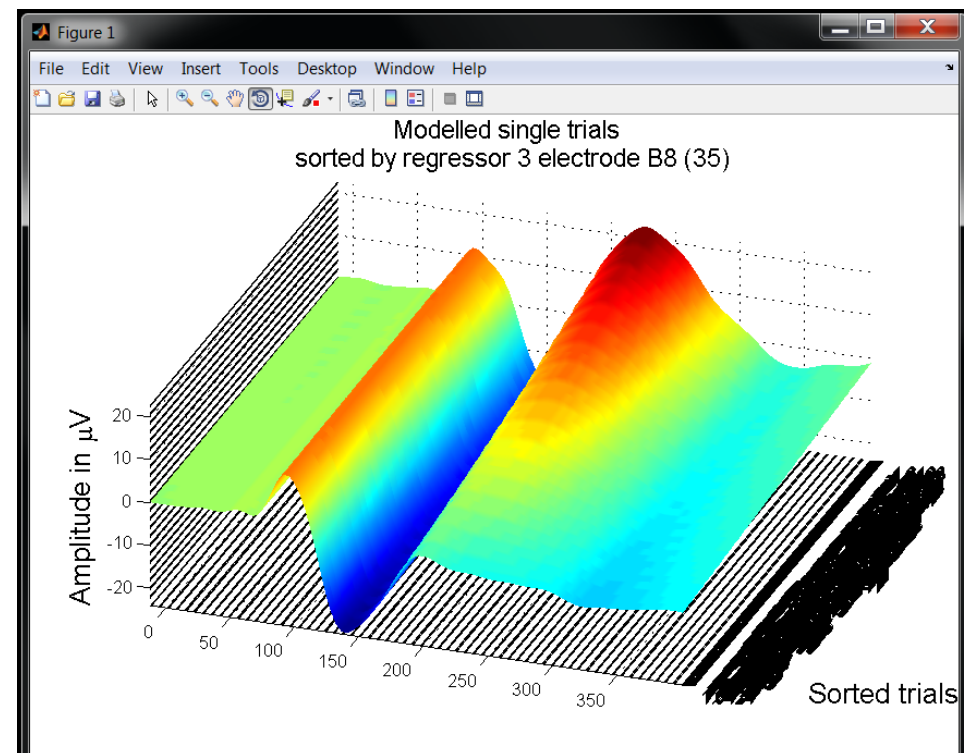
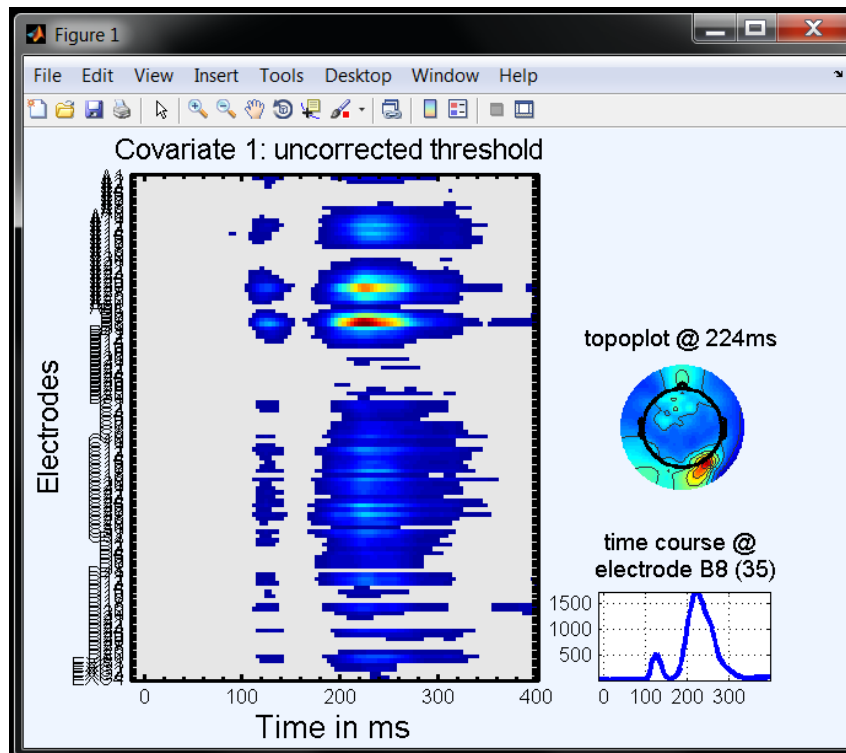
- Subject analysis:** Includes a "Data import" dropdown menu (currently set to "time epoched data (ERP)") and an "Analyse" button.
- Random effects:** Includes a "Random effects" button and a "LIMO tools" button.
- Results:** Includes a "View results" button and a "Contrast manager" button.

At the bottom of the GUI, there are "Help" and "Quit" buttons, and a copyright notice: "Copyright (C) LIMO Team 2015 - GNU GPL".

Name	Date Modified
done_already	27/05/2015 17:18
categorical_variable.txt	06/05/2013 14:50
continuous_variables.txt	06/05/2013 14:50
limo_dataset_S2.fdt	21/01/2015 21:21
limo_dataset_S2.set	21/01/2015 21:21
limo_dataset_S2.set_single_trials_daterp.mat	21/01/2015 21:21

What have we done: results

- Image all (R2, condition, covariate)
- Course plots, topoplots



Robust Statistics

WHY & HOW?

Issues with standard stats

- Standard stats are all instantiations of a GLM using an Ordinary Least Square solution → implies looking at the mean
- the breakdown point of an estimator is the proportion of incorrect observations (e.g. arbitrarily large observations) an estimator can handle before giving an incorrect
- For data x_1 to x_n – the mean has a bkdp of 0 because we can make the mean large changing any x_i – the median has a bkdp of 50%

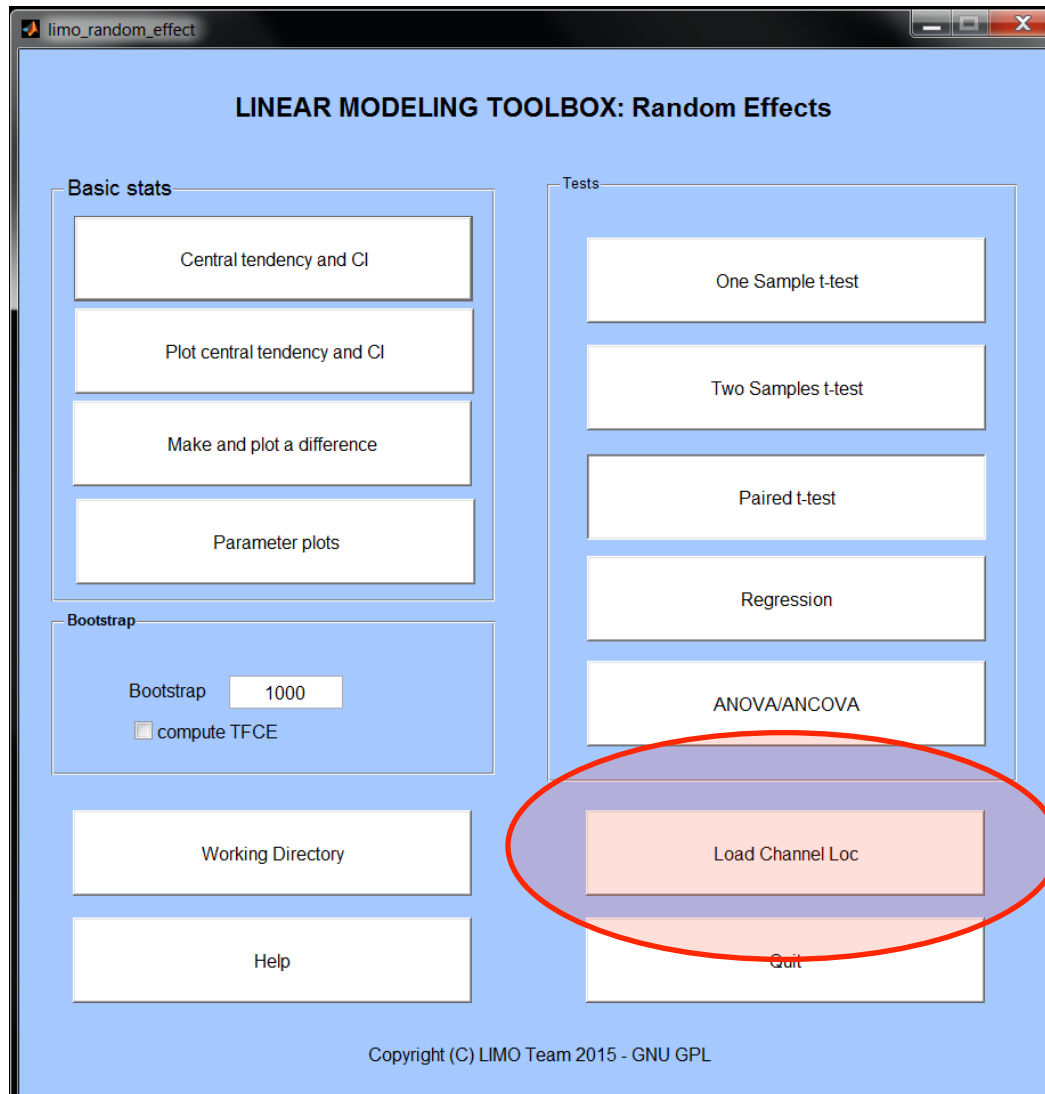
Yes but my data are Gaussian

- Are you sure?
- [Micceri \(1989\). The Unicorn, The Normal Curve, and Other Improbable Creatures. Psych Bul. 105, 156-166](#)
- If the data are Gaussian, the median, the trimmed mean is the same as the mean ! So no reason not to use alternative techniques.
- 1st level, uses weighted least square (weights down bad trials)
- 2nd level involves 20% trimmed mean (weights = 0 for bad subjects): t-tests, 1-way ANOVA, Repeated Measures ANOVA (soon)
- For regressions and N-way ANOVA/ANOVA we use an IRLS (all subjects have weights from 0 to 1)

Practical

- One sample t-test on 'noise' regressor
 - You can select files by hand, but it's easier to build lists – right click/run sheffield_mklist.m
 - This makes a list_of_Betas.txt we can use
 - From the GUI, choose 'Random Effect'

Practical

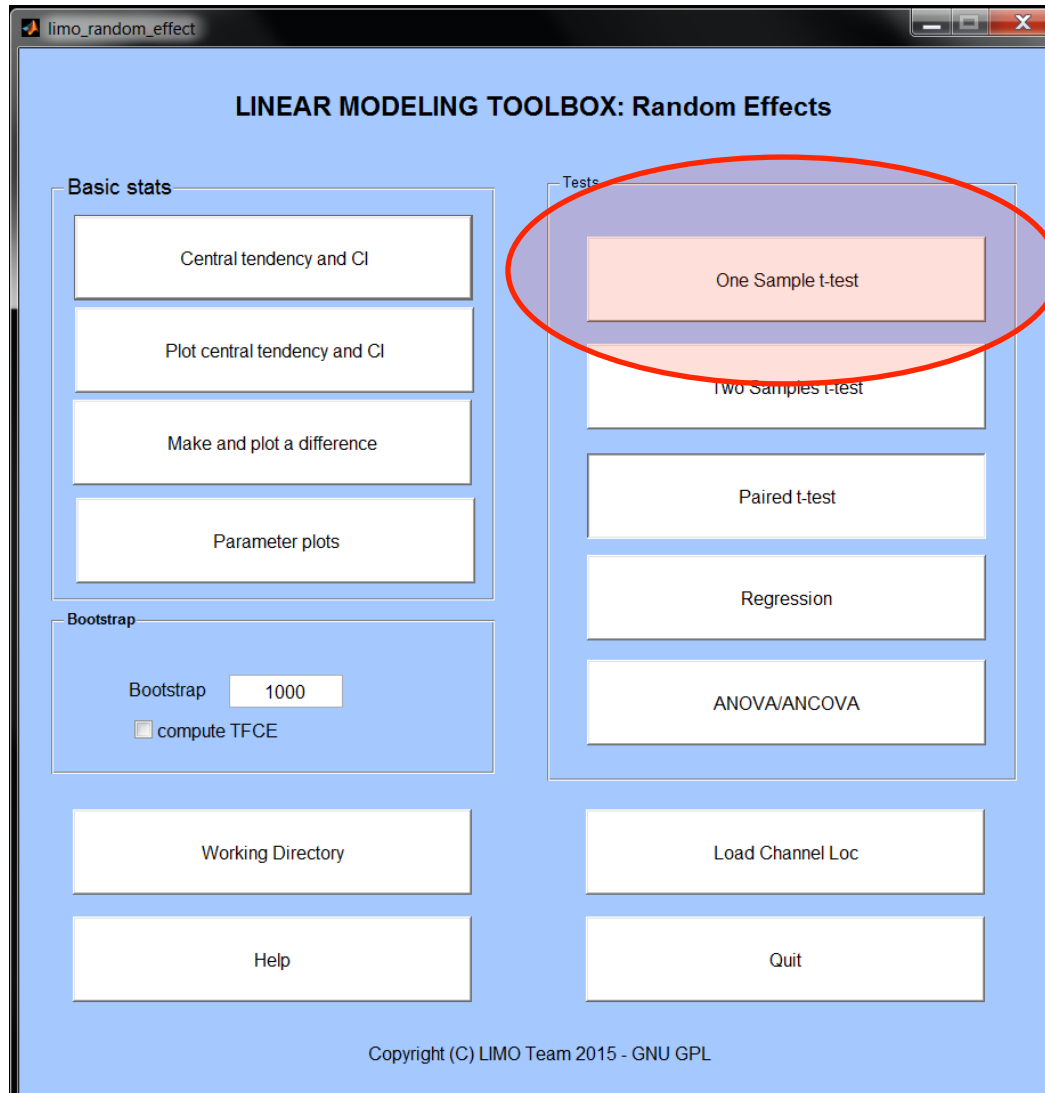


LIMO EEG expect you to build a template cap to use across subjects (LIMO TOOLS) because only valid electrodes are analysed per subject

= no interpolated values at the subject level
= LIMO EEG deals with missing data

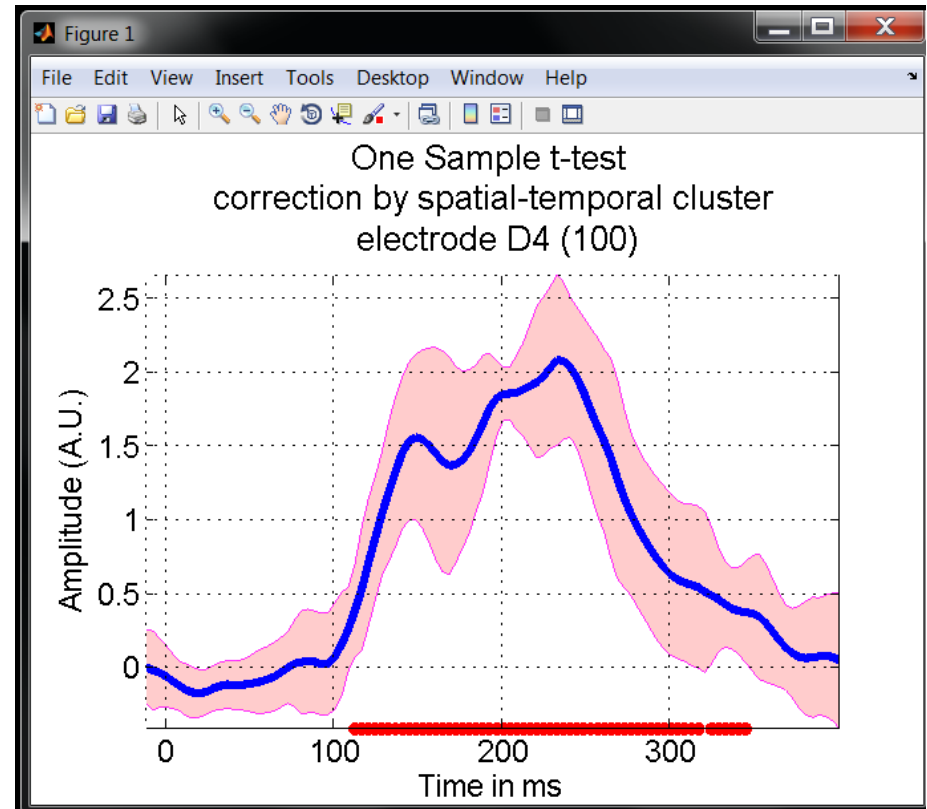
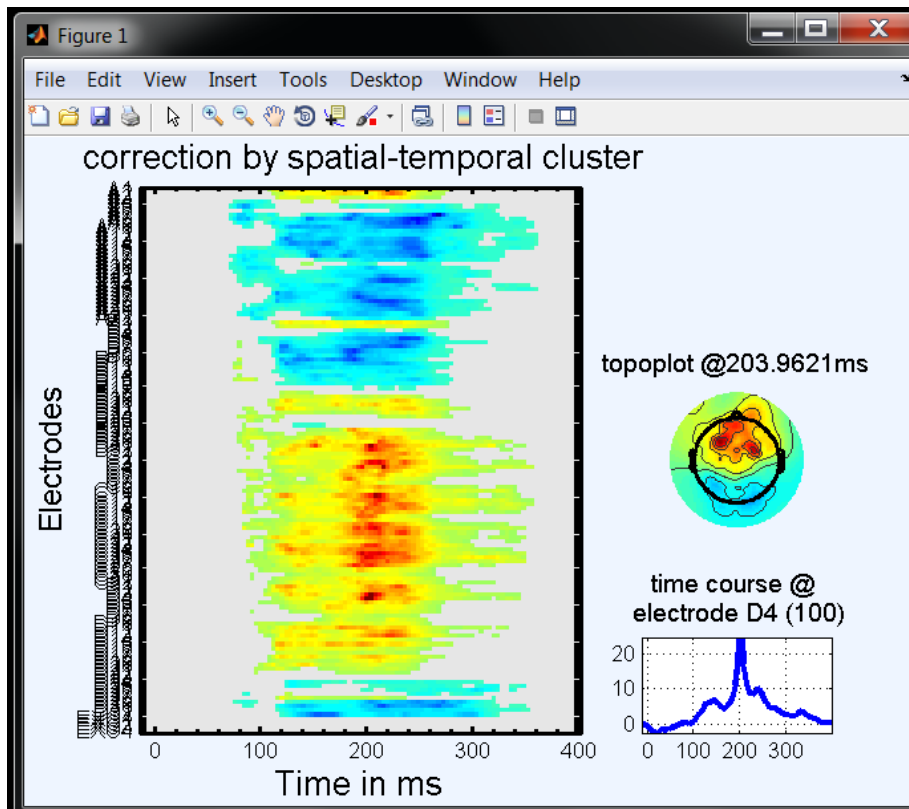
The expected channlocs is in the gp_effects directory

Practical



Compute a one-sample t-test on betas parameter 3 i.e. the Effect of stimulus phase information on ERP

Review gp level results



Design questions!

- Let's think how to analyse your data!
- Nb of conditions / covariates
- contrasts
- 1st level covariates
- 2nd level covariates

Design questions!

- Typical 2*2 design
→ 1st level vs 2nd level, where to model the interaction
- Testing the effect of covariate within or between conditions?