

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

Arnaud Delorme

Lecture from Cyril Pernet, PhD, Edinburgh Imaging &
Centre for Clinical Brain Sciences

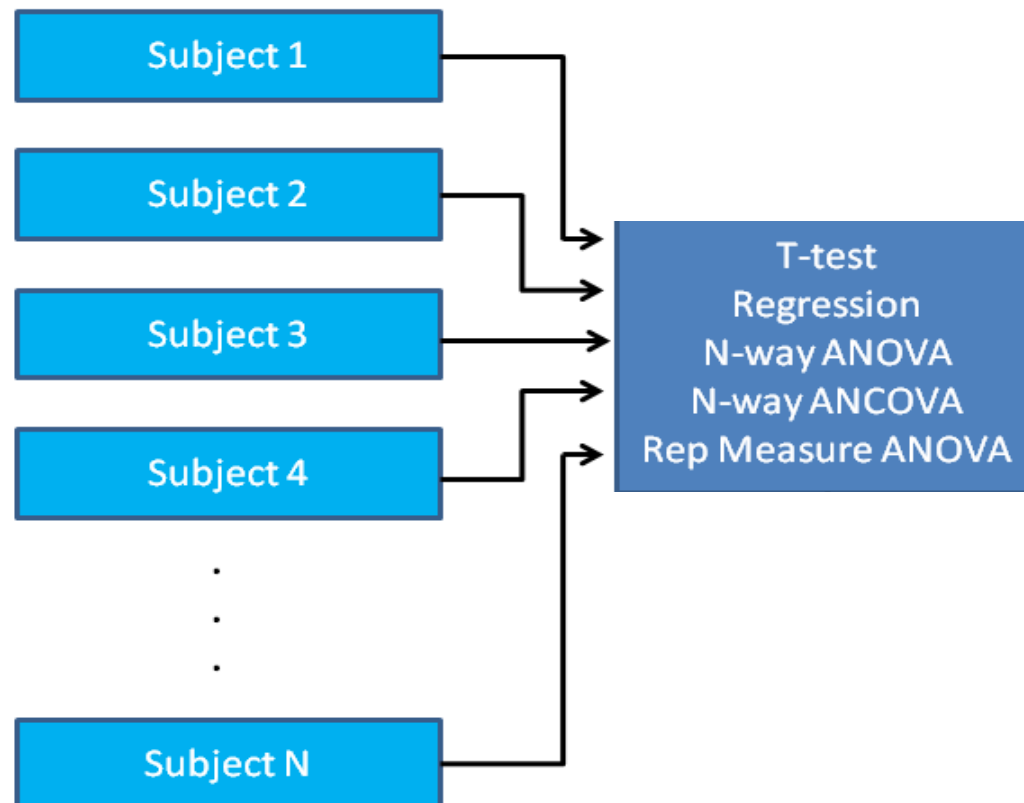
Hierarchical Linear Model

1st level analysis:

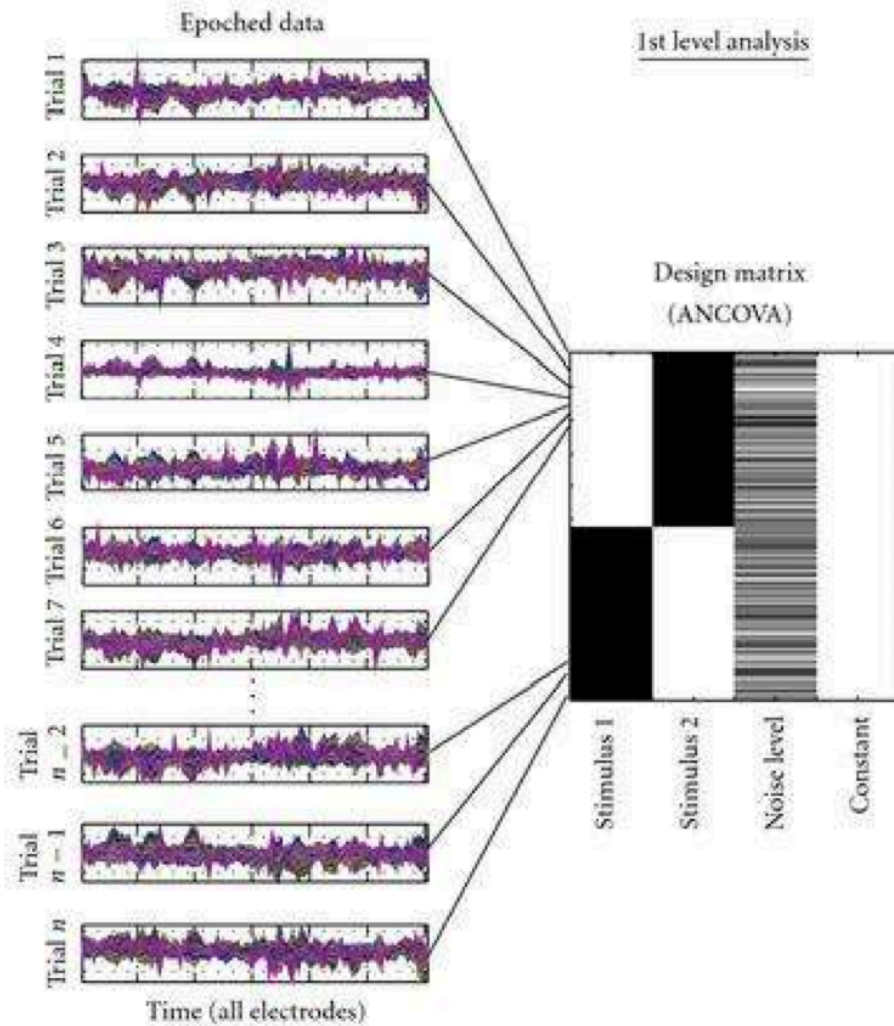
GLM: $Y = X\beta + \epsilon$
→ 1 β per column of X
(= within subject effects)

2nd level analysis:

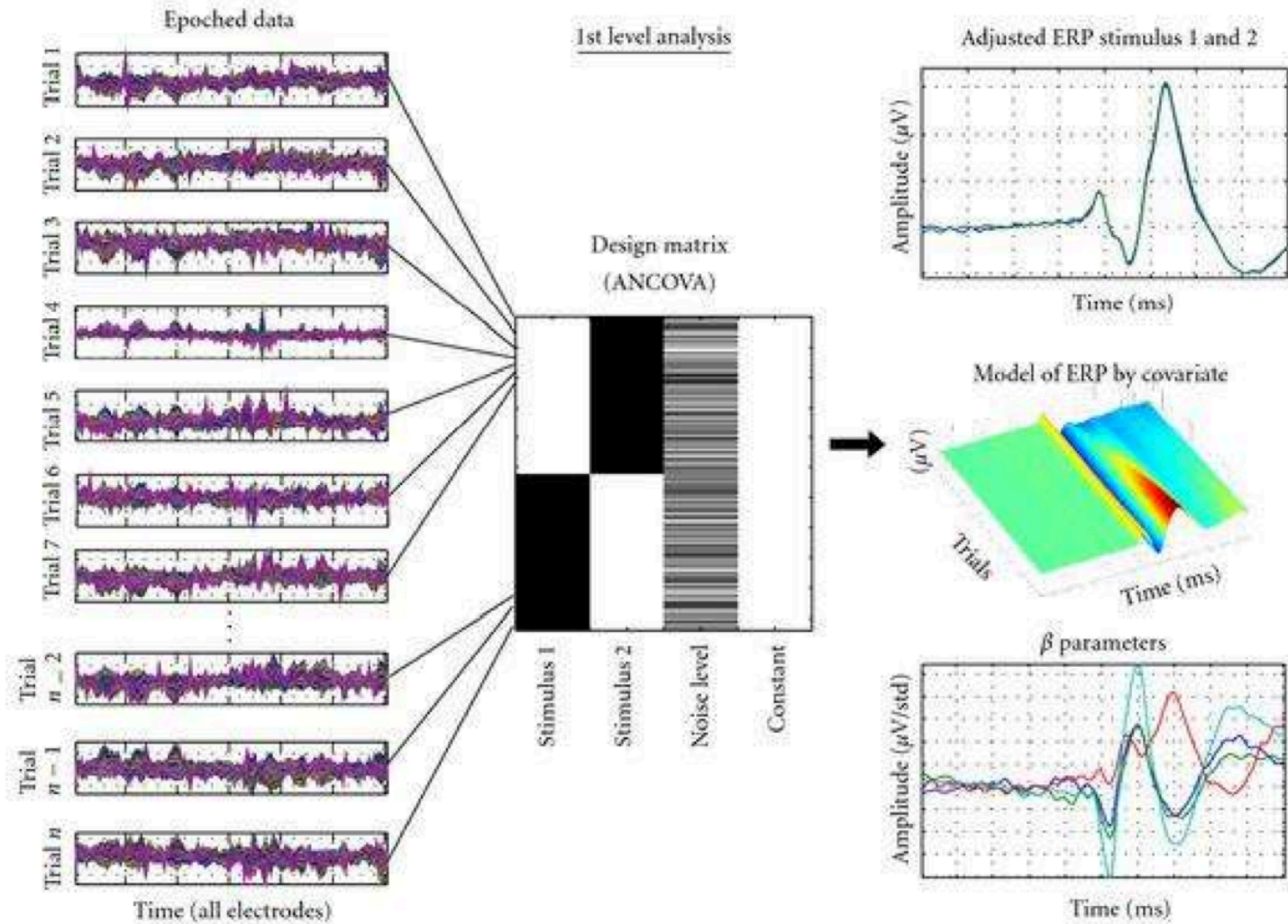
Robust stats (Yuen t-tests, robust GLM, robust Hotelling T²)



Linear Modeling of EEG data



Linear Modeling of EEG data

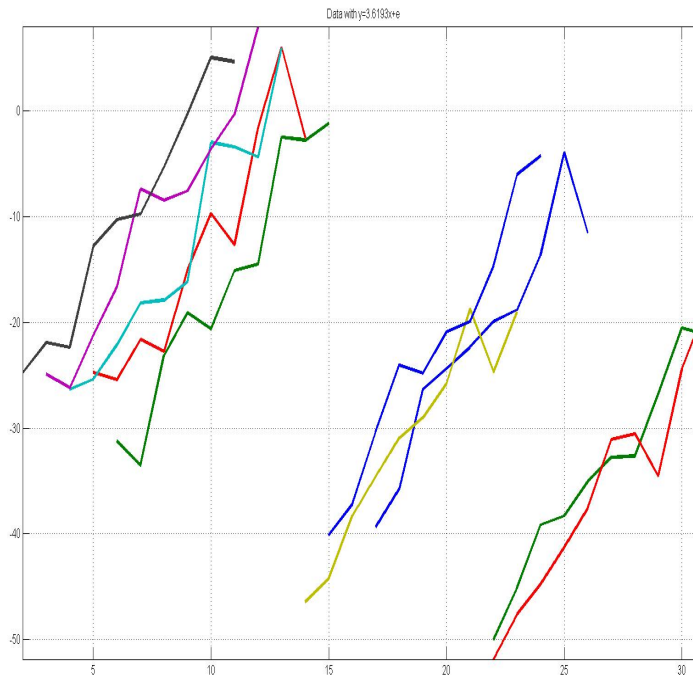


Random Effect Model

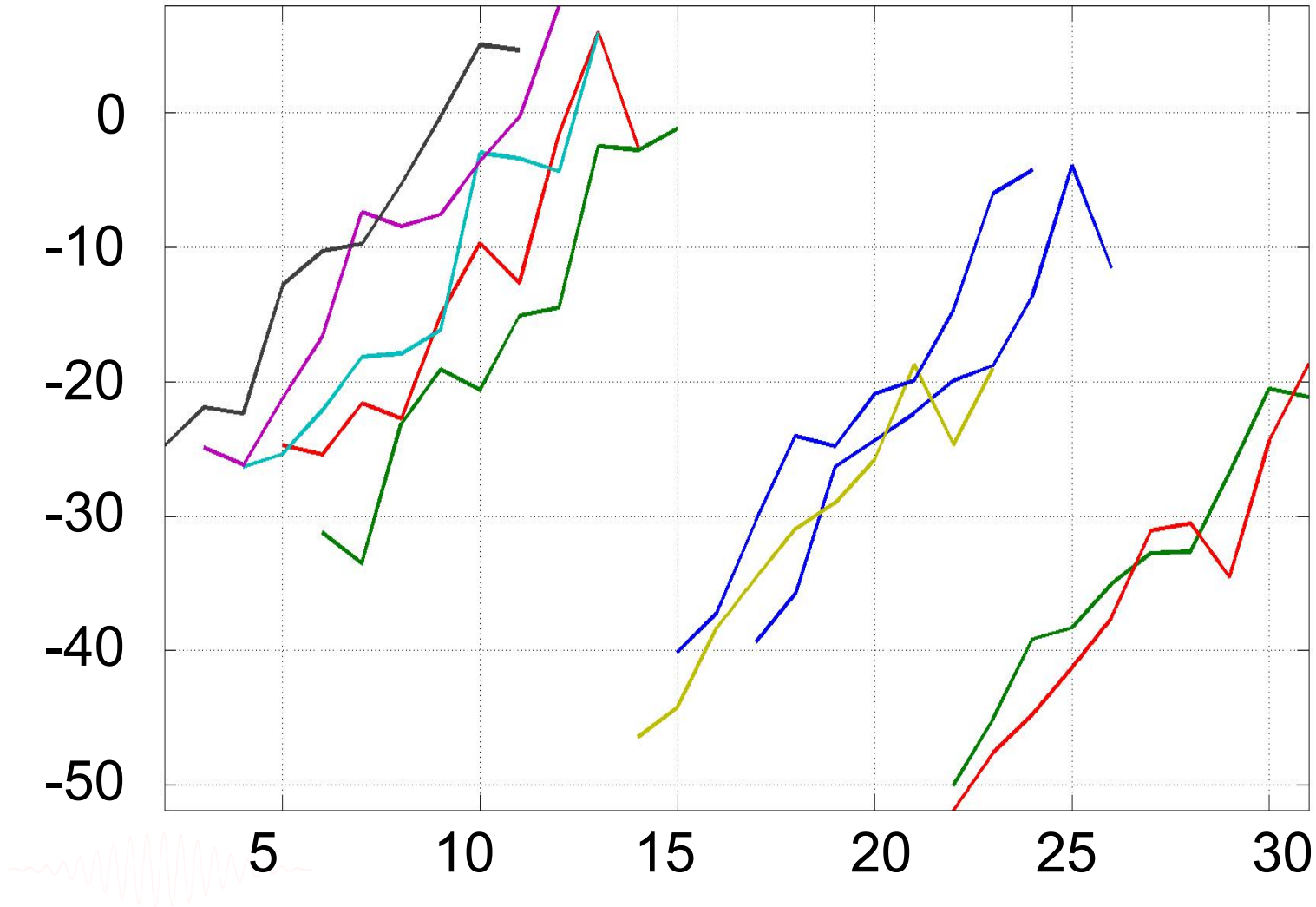
Model the data with fixed effects (the experimental conditions) and a random effect (subjects are allowed to have different overall values – considering subjects as a random variable)

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

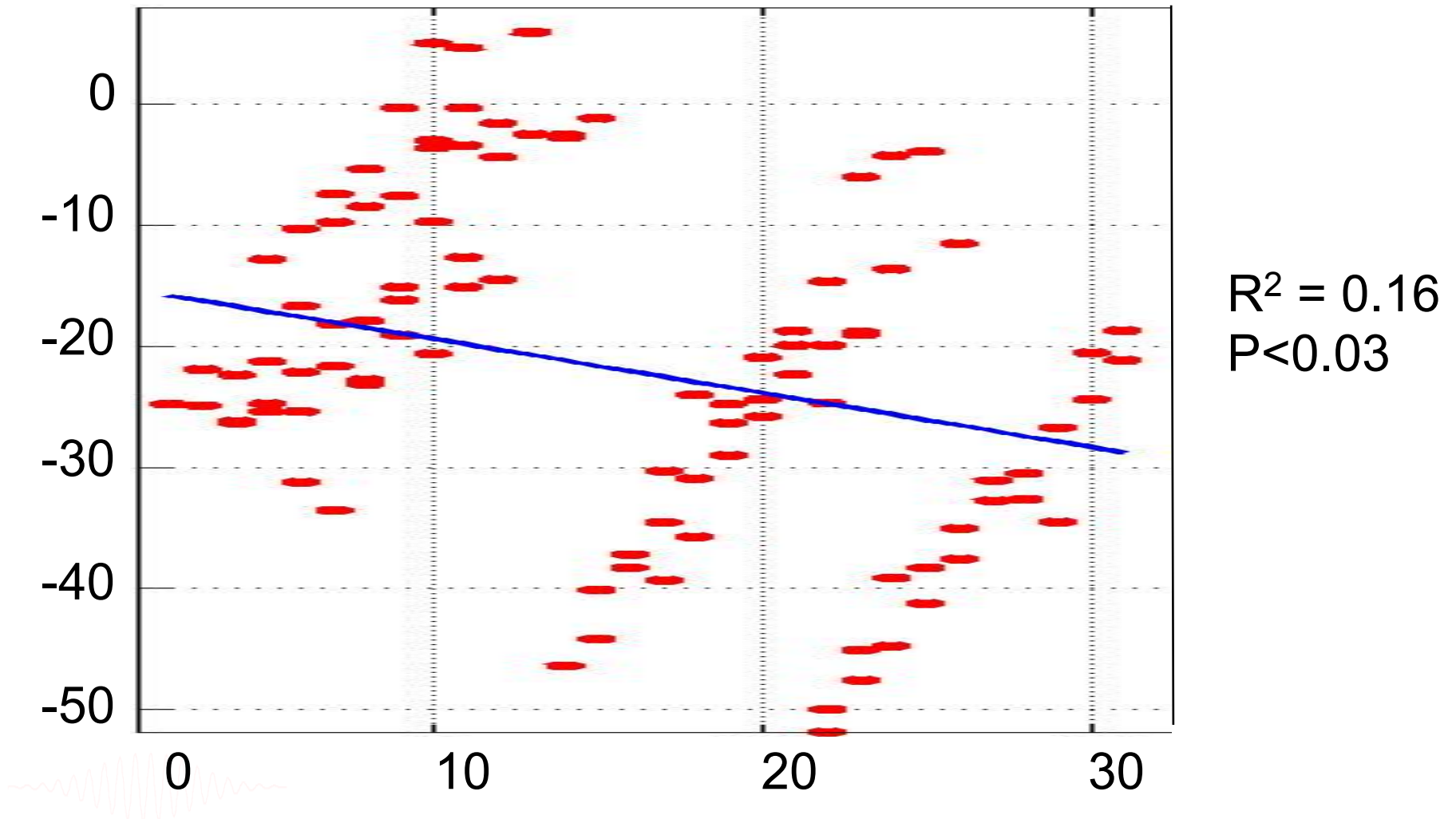
→ Plot the data per intensity



Fixed Effect

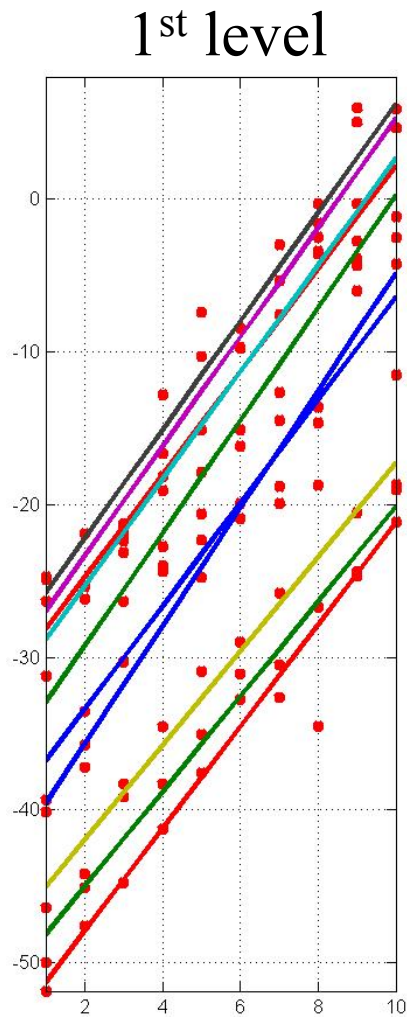


Fixed effect



Fixed effect = average across subjects → negative correlation?

Random Effect Model



2nd level

Slope: [3.43
3.51 3.39 3.41
3.62 3.55 ...]

$p < 10^{-12}$



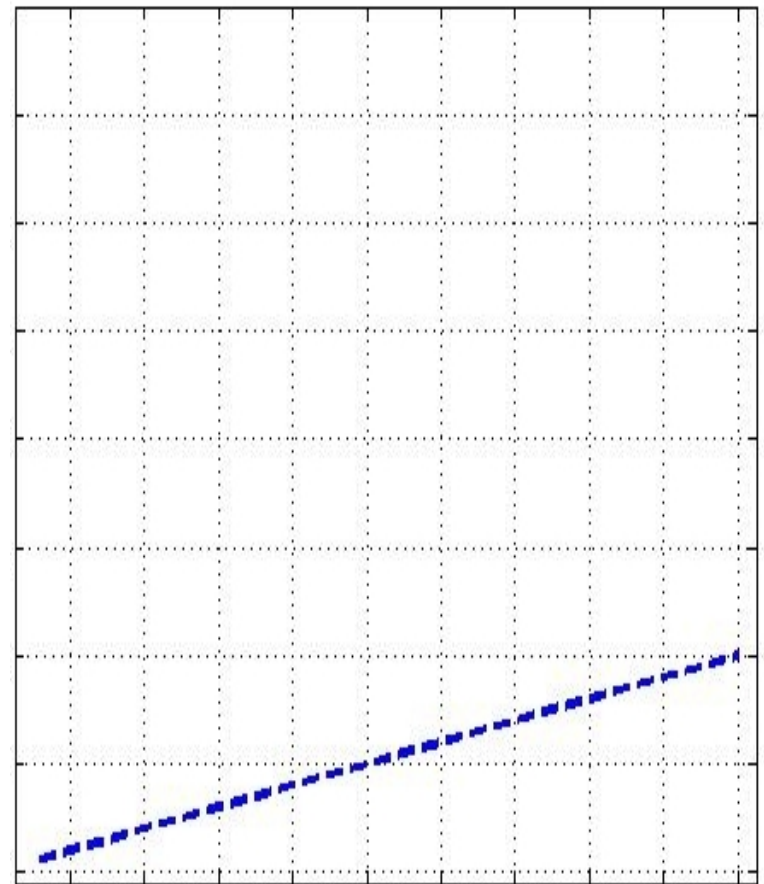
Linearity

- 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>

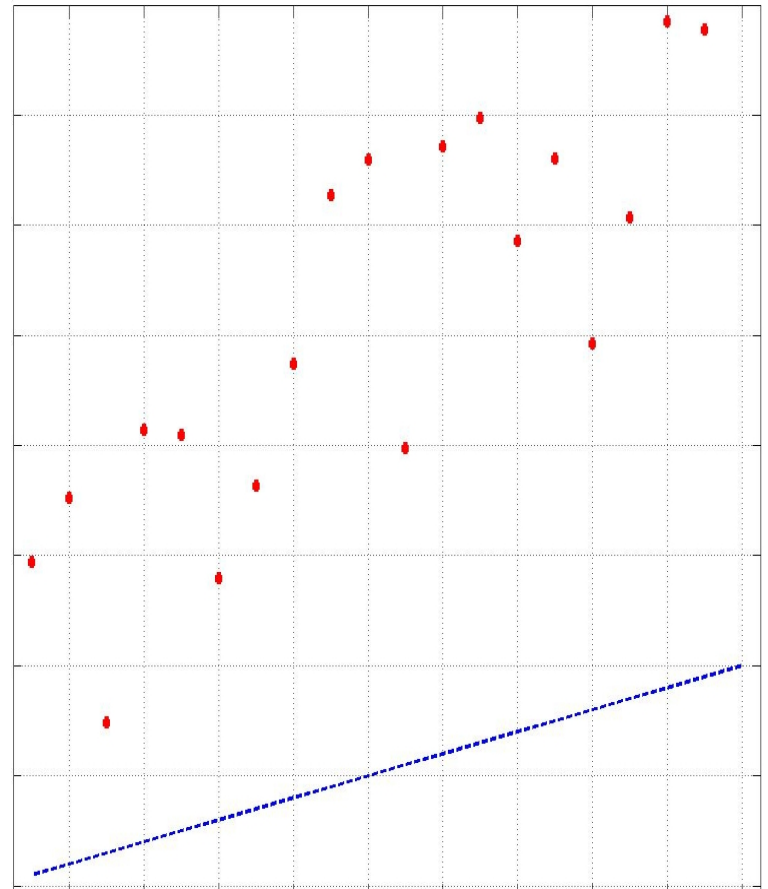
A regression is a linear model

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



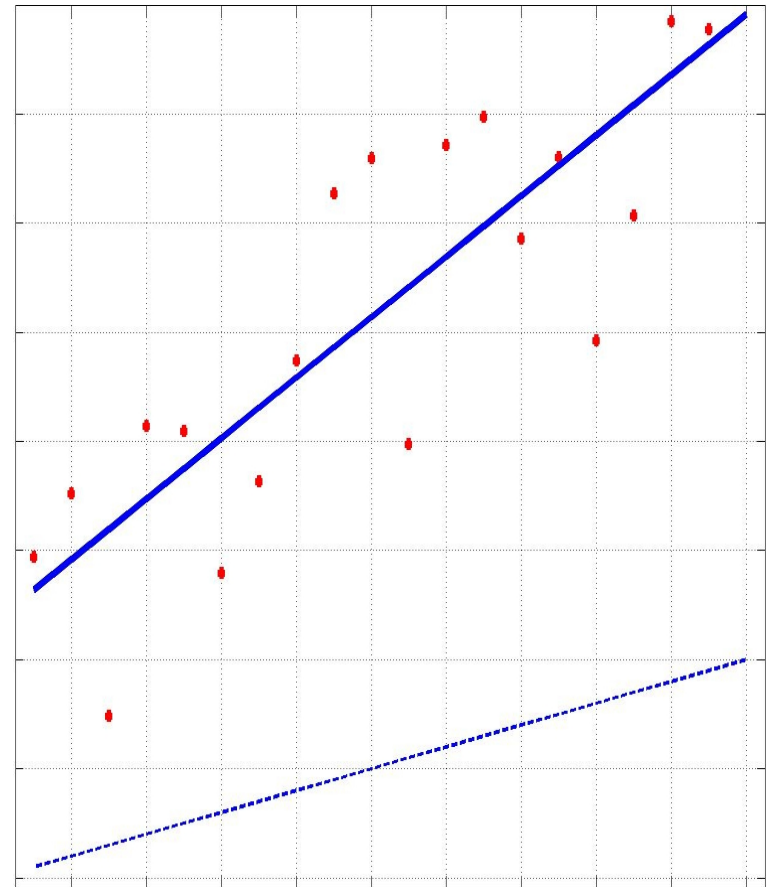
A regression is a linear model

- 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)



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
- $y^{\wedge} = 2.7x + 23.6$



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

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: $y = \beta_1x + \beta_2 + \varepsilon$
- Multiple regression: $y = \beta_1x_1 + \beta_2x_2 + \beta_3 + \varepsilon$
- One way ANOVA: $y = u + \alpha_i + \varepsilon$
- Repeated measure ANOVA: $y = u + \alpha_i + \varepsilon$
- ...

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
4	4
9	4

Design matrix
G₁ G₂ G₃ G₄ C

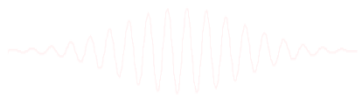
	G ₁	G ₂	G ₃	G ₄	C
		■	■	■	
		■	■	■	
		■	■	■	
	■		■	■	
	■		■	■	
	■		■	■	
	■	■		■	
	■	■		■	
	■	■		■	
	■	■	■		
	■	■	■		
	■	■	■		
	■	■	■		
	■	■	■		

$$\begin{aligned}
 y(1..3)1 &= 1x_1 + 0x_2 + 0x_3 + 0x_4 + c + e_{11} \\
 y(1..3)2 &= 0x_1 + 1x_2 + 0x_3 + 0x_4 + c + e_{12} \\
 y(1..3)3 &= 0x_1 + 0x_2 + 1x_3 + 0x_4 + c + e_{13} \\
 y(1..3)4 &= 0x_1 + 0x_2 + 0x_3 + 1x_4 + c + e_{13}
 \end{aligned}$$

$$\begin{pmatrix} 8 \\ 9 \\ 7 \\ 5 \\ 7 \\ 3 \\ 3 \\ 3 \\ 4 \\ 1 \\ 6 \\ 4 \\ 9 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \end{pmatrix} * \begin{pmatrix} \beta_1 \\ \beta_2 \\ \beta_3 \\ \beta_4 \\ c \end{pmatrix} + \begin{pmatrix} e_{11} \\ \\ \\ \\ \\ \\ \\ \\ \\ e_{13} \end{pmatrix}$$

Design considerations

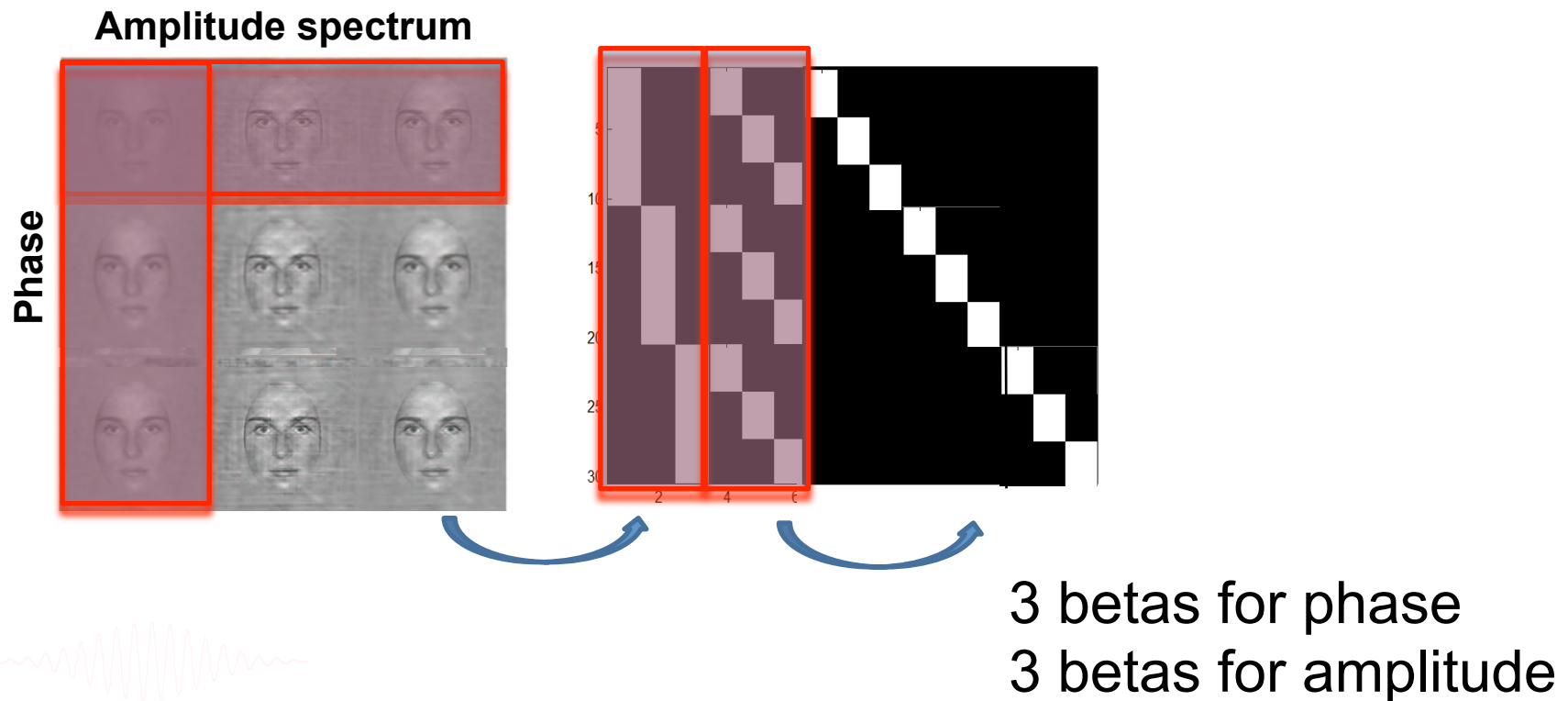
Illustration with a set of studies looking at the effect of stimulus phase information



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

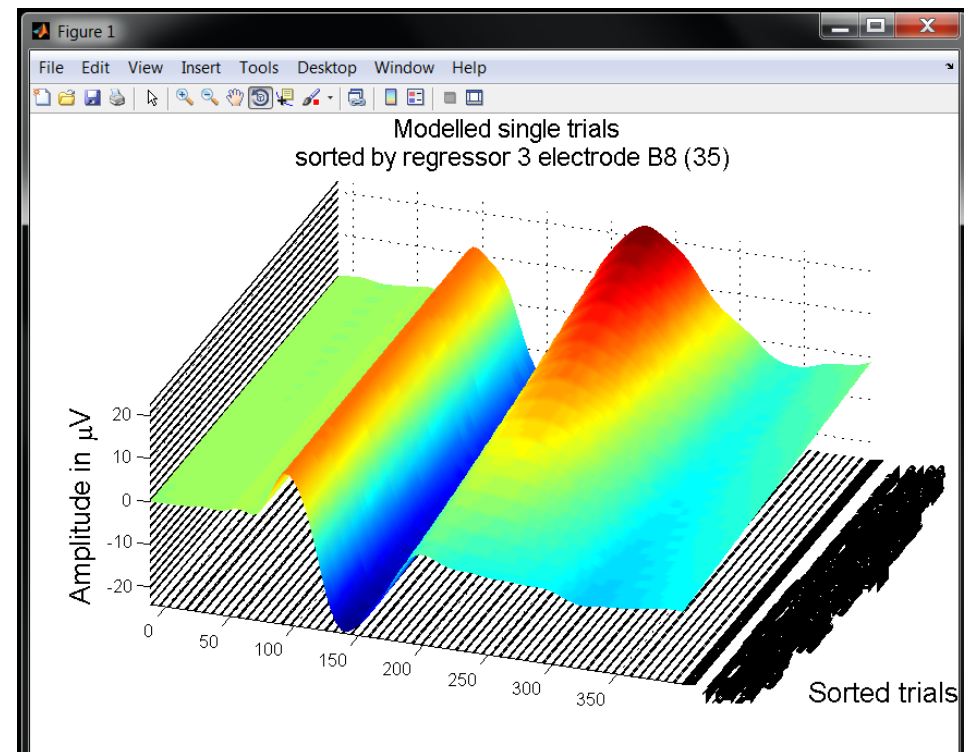
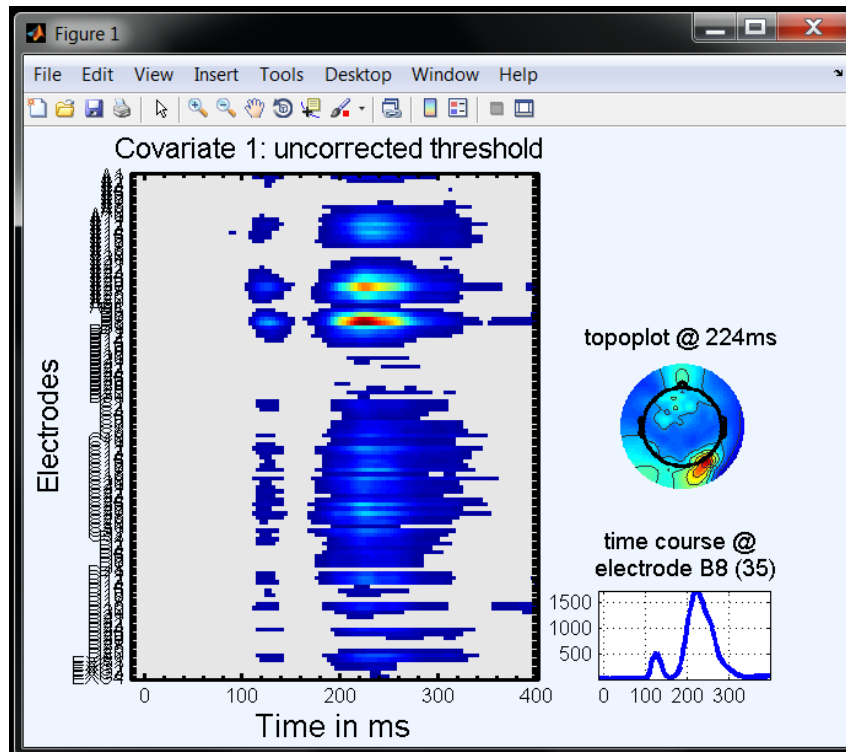
Factorial Designs: $N*N*N*...$

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

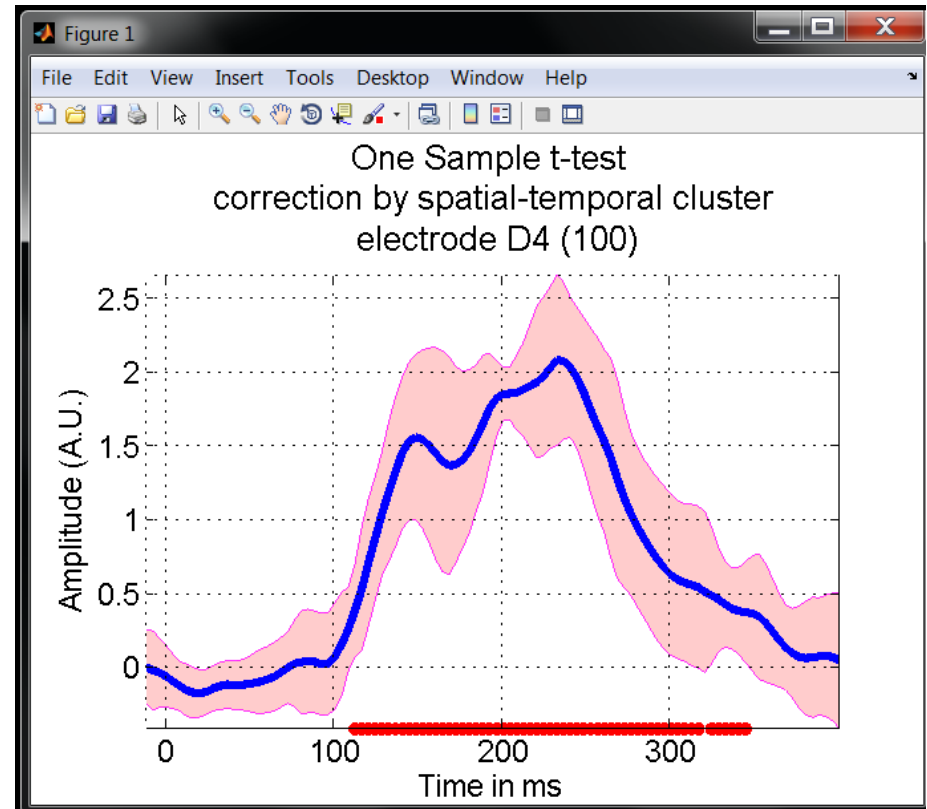
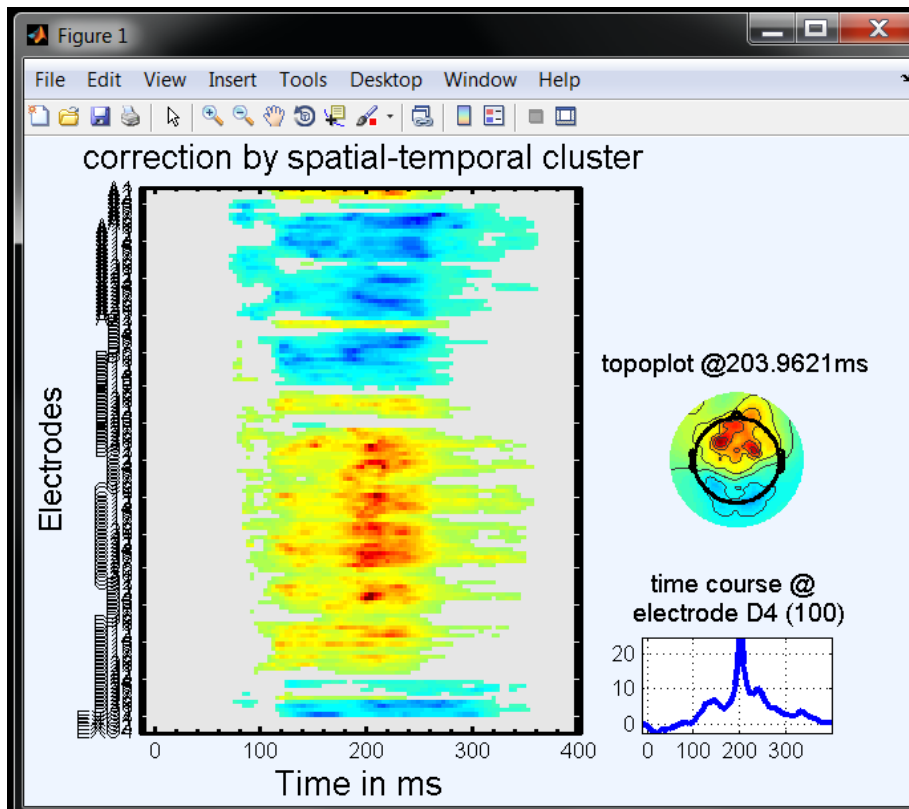


What have we done: results

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



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