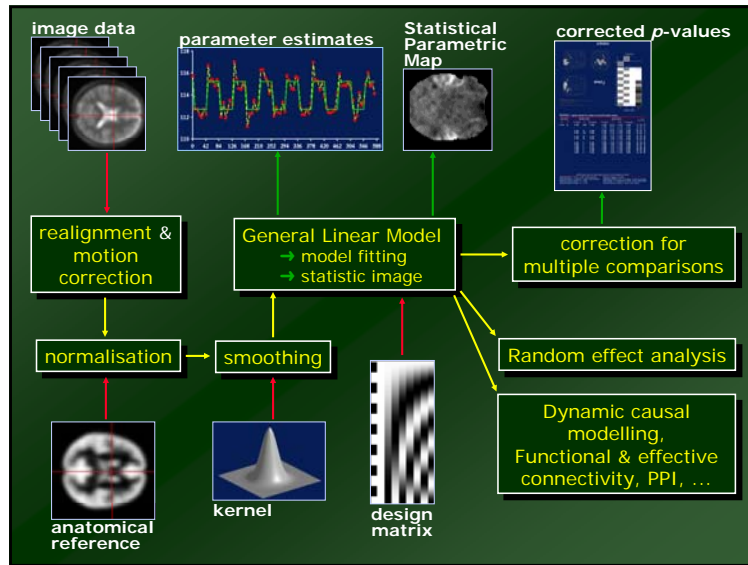


Experimental design

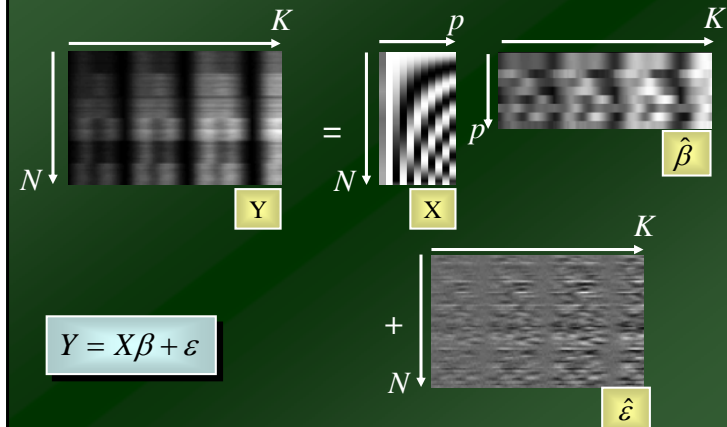


Contents

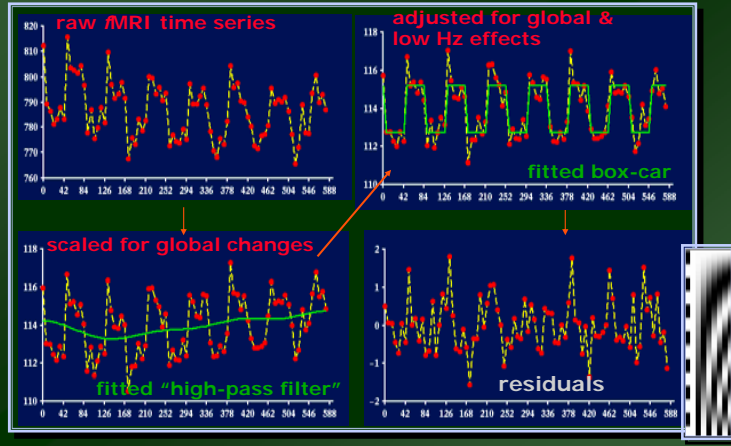
- Introduction & recap
- Experimental design
- « Take home » message



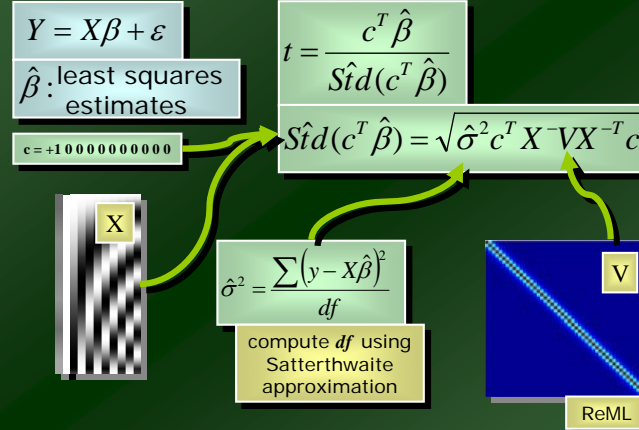
Mass univariate approach



GLM fitted

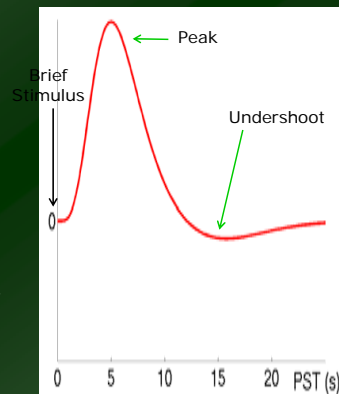


t-statistic - Computations



BOLD Impulse Response

- Function of blood oxygenation, flow, volume (Buxton et al, 1998)
- Peak (max. oxygenation) 4-6s poststimulus; baseline after 20-30s
- Initial undershoot can be observed (Malonek & Grinvald, 1996)
- Similar across V1, A1, S1...
- ... but differences across: other regions (Schacter et al 1997) individuals (Aguirre et al, 1998)



Advantages of Event-related fMRI

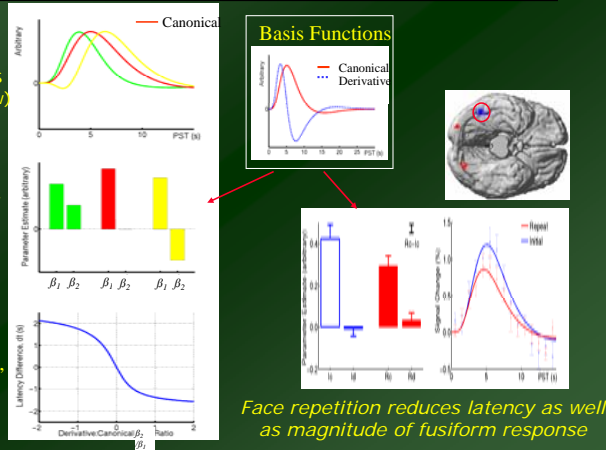
1. Randomised trial order
c.f. confounds of blocked designs
2. Post hoc / subjective classification of trials
e.g. according to subsequent memory
3. Some events can only be indicated by subject (in time)
e.g. spontaneous perceptual changes
4. Some trials cannot be blocked
e.g. "oddball" designs
5. More accurate models even for blocked designs?
e.g. "state-item" interactions

BOLD Response Latency (Linear)

Delayed Responses (green/ yellow)

Parameter Estimates

Actual latency, d_t , vs. β_2/β_1



Face repetition reduces latency as well as magnitude of fusiform response

Contents

- Introduction & recap
- **Experimental design**
- « Take home » message

Experimental design

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Subtraction Logic

Cognitive subtraction originated with reaction time experiments (F. C. Donders, a Dutch physiologist).

Measure the time for a process to occur by comparing two reaction times, one which has the same components as the other + the process of interest.

Example:

- T1: Hit a button when you see a light
- T2: Hit a button when the light is green but not red
- T3: Hit the left button when the light is green and the right button when the light is red

$T2 - T1$ = time to make discrimination between light color

$T3 - T2$ = time to make a decision

Assumption of pure insertion: You can insert a component process into a task without disrupting the other components.

Widely criticized (we'll come back to this when we talk about parametric studies)



Franciscus Cornelis Donders (1818-1889)

Activation and Baseline Conditions

Aim:

To reveal brain activation related to a cognitive or sensori-motor process of interest (PI)

Cognitive Subtraction:

Contrast Activation task (engages PI) to a Baseline task (no PI).
Difference = Brain regions associated with PI.

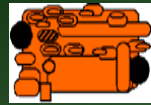
Example:

PI = Object recognition

Activation task: with PI



Baseline task: no PI



Difference = Brain regions associated with Object Recognition

Cognitive Subtractions:

Stimulus or task changes

Stimulus Change

Activation condition



Task: (constant) "View picture"

Baseline condition



"View picture"

= Object Recognition

Stimulus (constant)

Activation condition



Task: Change: "Name Object"

Baseline condition



"Say: "Yes"

= Name Retrieval

Cognitive Subtraction: Baseline-problems

• „Distant“ stimuli



→ Several components differ!

• „Related“ stimuli



→ P implicit in control task ?

„Queen!“

„Aunt Jenny?“

• Same stimuli, different task



→ Interaction of process and task ?

Name Person!

Name Gender!

Cognitive Subtractions: Serial subtraction

Baseline condition for one contrast acts as activation condition for another contrast

Example:

Condition A.



Stimulus:

Task: Name Object

Condition B.



Say: "Yes"

Condition C.



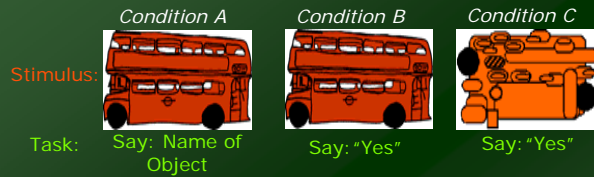
Say: "Yes"

A-B = Name Retrieval

B-C = Object Recognition

Very limited...

Problems with Serial Subtractions



Assumptions:
 A - B = only changes processing associated with Name Retrieval
 B - C = only changes processing associated with Object Recognition

- BUT**
1. There may be *implicit* naming in condition B. In which case: naming component is removed from A-B and introduced into B-C.
 2. Name Retrieval may increase the demands on object recognition (*Interactions*).

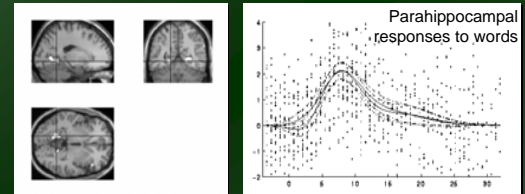
i.e. A - B : May reveal Object recognition NOT Name retrieval.
 B - C : May reveal Object Recognition AND Name Retrieval

Implicit processing and interactions between processing components make it difficult to find baseline tasks that control for all but the process of interest.

Evoked responses

Differential event-related fMRI

SPM{F} testing for evoked responses



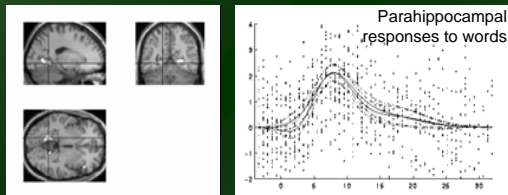
BOLD EPI fMRI at 2T, TR 3.2sec.
 Words presented every 16 secs; (i) studied words or (ii) new words

- "Baseline" here corresponds to session mean (and thus processing during "rest")
- Null events or long SOAs essential for estimation
- "Cognitive" interpretation hardly possible, but useful to define regions generally involved in the task

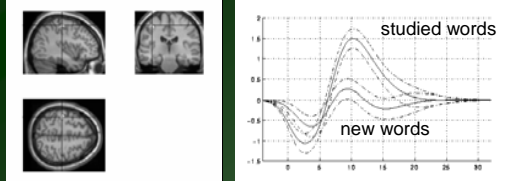
Differential responses

Differential event-related fMRI

SPM{F} testing for evoked responses



SPM{F} testing for differences



BOLD EPI fMRI at 2T, TR 3.2sec.
 Words presented every 16 secs; (i) studied words or (ii) new words

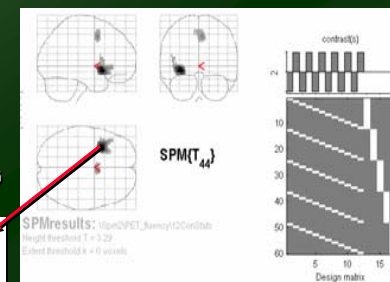
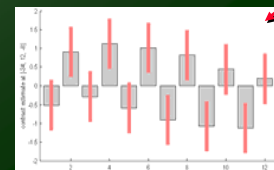
Peri-stimulus time (secs)

A categorical analysis

Experimental design

Word generation G
 Word repetition R

R G R G R G R G R G



G - R = Intrinsic word generation

...under assumption of pure insertion

Overview




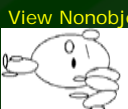
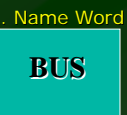

- **Categorical designs**
 - Subtraction - Pure insertion, evoked / differential responses
 - Conjunction - Testing multiple hypotheses
- **Parametric designs**
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Conjunctions

- One way to minimise the baseline/pure insertion problem is to isolate the same process by two or more separate comparisons, and inspect the resulting simple effects for commonalities
- A test for such activation common to several independent contrasts is called "Conjunction"
- Conjunctions can be conducted across a whole variety of different contexts:
 - tasks
 - stimuli
 - senses (vision, audition)
 - etc.
- But the contrasts entering a conjunction have to be truly independent!

CONJUNCTION DESIGNS

Conjoint effects from multiple contrasts.

- | | | | |
|------------|---|--|--|
| | A. Name Object
 | B. View Nonobject
 | A - B =
Name retrieval
<i>confounded with</i>
object recognition |
| Contrast 1 | — | | |
| | C. Name Colour
 | D. View Nonobject
 | C - D =
Name retrieval
<i>confounded with</i>
colour processing |
| Contrast 2 | — | | |
| | E. Name Word
 | F. View XXXs
 | E - F =
Name retrieval
<i>confounded with</i>
word recognition. |
| Contrast 3 | — | | |

Name Retrieval = Areas activated by Conjunction of A-B and C-D and E-F

Conjunctions

Example:

Which neural structures support object recognition, independent of task (naming vs viewing)?

Visual Processing V
Object Recognition R
Phonological Retrieval P

Task (1/2)

Viewing Naming

Stimuli (A/B)		
Objects	Colours	
	A1	A2
	B1	B2

(Object - Colour viewing) & (Object - Colour naming)

→ [1 -1 0 0] & [0 0 1 -1]

→ [R,V - V] & [P,R,V - P,V]

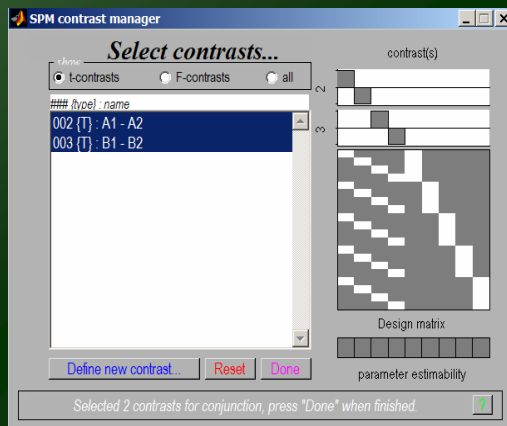
= R & R = R

Price et al, 1997



(assuming no interaction $R \times P$; see later)

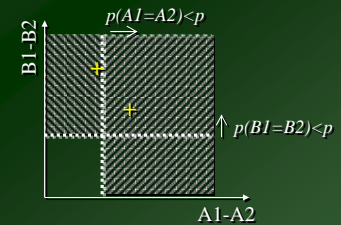
Conjunctions



Two flavours of inference about conjunctions

SPM5/8 offers two general ways to test the significance of conjunctions:

- Test of global null hypothesis: Significant set of consistent effects
 - ➔ "which voxels show effects of similar direction (but not necessarily individual significance) across contrasts?"
- Test of conjunction null hypothesis: Set of consistently significant effects
 - ➔ "which voxels show, for each specified contrast, effects > threshold?"
- Choice of test depends on hypothesis and congruence of contrasts; the global null test is more sensitive (i.e., when direction of effects hypothesised)



Friston et al. (2005). *Neuroimage*, 25:661-7.
Nichols et al. (2005). *Neuroimage*, 25:653-661.

Overview

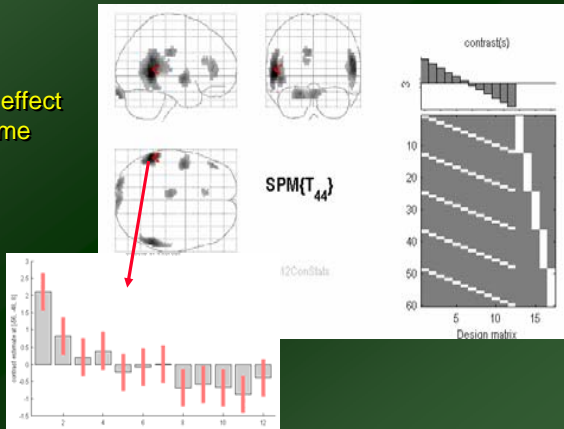
- **Categorical designs**
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Parametric Designs: General Approach

- Parametric designs approach the baseline problem by:
 - Varying the stimulus-parameter of interest on a continuum, in multiple ($n > 2$) steps...
 - ... and relating blood-flow to this parameter
- Possible tests for such relations are manifold:
 - Linear
 - Nonlinear: Quadratic/cubic/etc.
 - „Data-driven“ (e.g., neurometric functions)

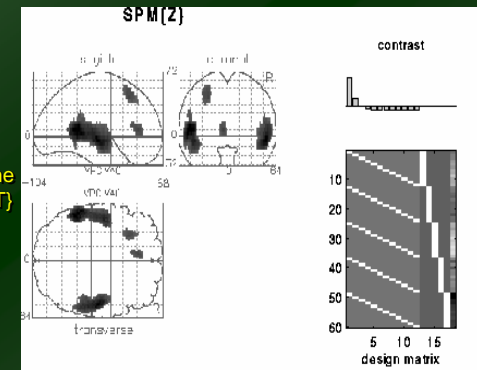
A linear parametric contrast

Linear effect of time

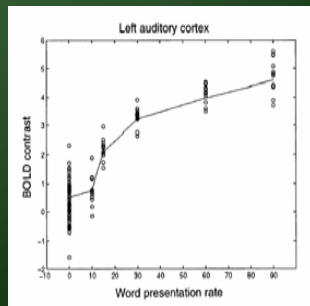


A nonlinear parametric contrast

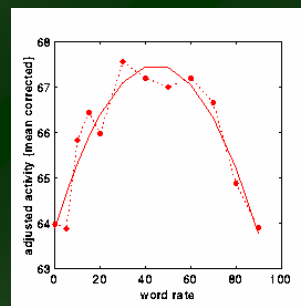
The nonlinear effect of time assessed with the SPM(T)



Nonlinear parametric design matrix



versus



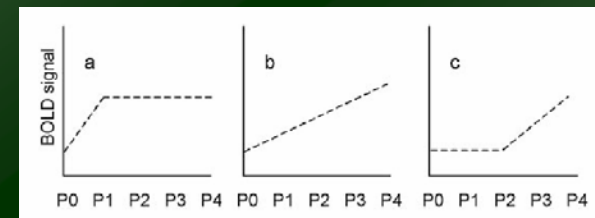
Inverted 'U' response to increasing word presentation rate in the DLPFC

Rees, G., et al. (1997). *Neuroimage*, 6: 27-78

Parametric Designs: Neurometric functions

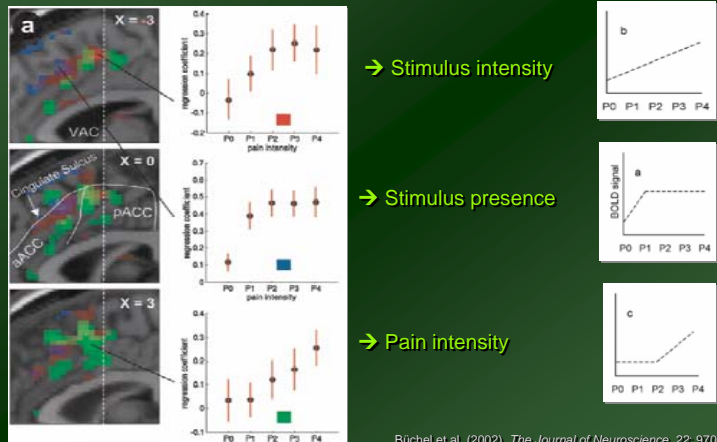
Coding of tactile stimuli in Anterior Cingulate Cortex:
Stimulus (a) presence, (b) intensity, and (c) pain intensity

- Variation of intensity of a heat stimulus applied to the right hand (300, 400, 500, and 600 mJ)
- **Assumptions:**



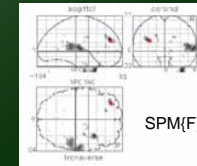
Büchel et al. (2002). *The Journal of Neuroscience*, 22: 970-6

Parametric Designs: Neurometric functions



Nonlinear parametric design matrix

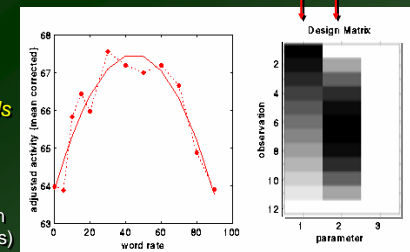
E.g. F-contrast [0 1 0] on Quadratic Parameter =>
Inverted 'U' response to increasing word presentation rate in the DLPFC



Polynomial expansion:
 $f(x) \sim b_1 x + b_2 x^2 + \dots$

...up to (N-1)th order for N levels

(SPM8 GUI offers polynomial expansion as option during creation of parametric modulation regressors)



Correlation design

- No need to find baseline that controls for all but the process of interest
- Segregates areas showing differential effects (linear and nonlinear effects)

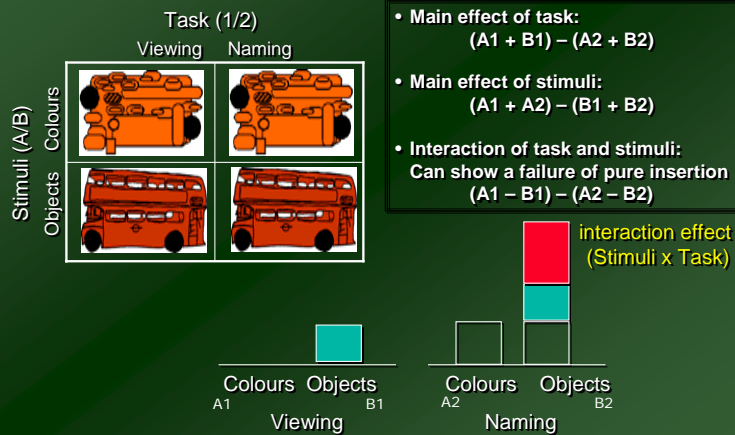
But:

- Common effects can not be revealed without a baseline.
- Limited to continuous variables (e.g. duration, frequency, word length, R.T.s etc)

Overview

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Factorial designs: Main effects and Interactions

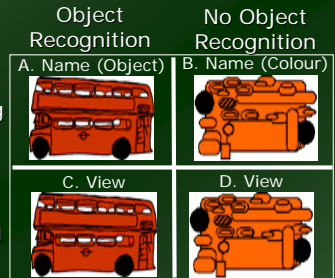


Factorial Designs

dissociate interactions

$B2 - B1 =$
Object Recognition
during naming

Naming



$A2 - A1 =$
Object Recognition
during viewing

No Naming

The Interaction effect
 $(B2-B1) - (A2-A1)$

i.e. The effect of Naming on Object recognition

$(B2-A2) - (B1-A1)$

i.e. The effect of object recognition on Naming.

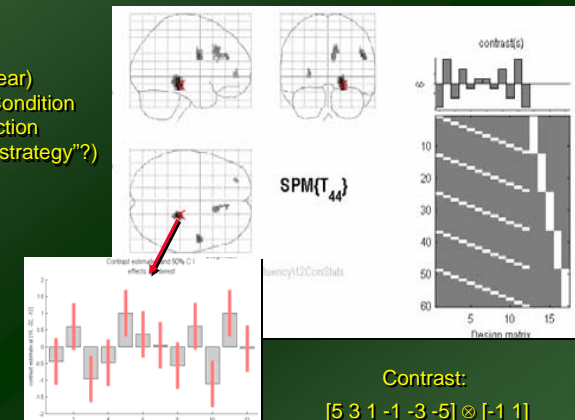
Conjunction of B2-B1 and A2-A1 reveals areas involved in object recognition irrespective of naming.
Conjunction of B2-A2 and B1-A1 reveals areas involved in naming irrespective of object recognition.

Overview

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Linear Parametric Interaction

A (Linear)
Time-by-Condition
Interaction
("Generation strategy"?)



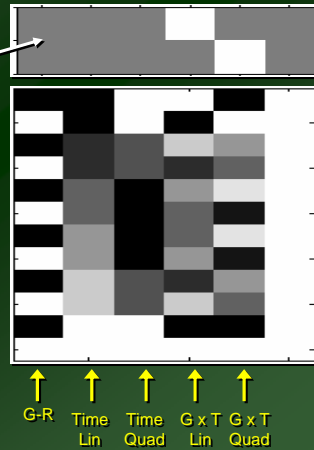
Nonlinear Parametric Interaction

F-contrast tests for nonlinear Generation-by-Time interaction (including both linear and Quadratic components)

Factorial Design with 2 factors:

1. Gen/Rep (Categorical, 2 levels)
2. Time (Parametric, 6 levels)

Time effects modelled with both linear and quadratic components...



Model selection

• Model must fit i.e. model assumptions met

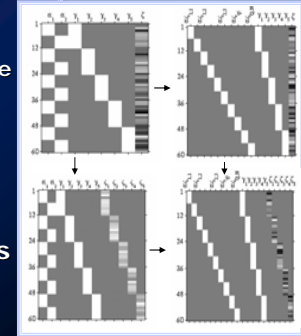
- at **every** voxel

• Omitting relevant effects

- effects contribute to variance
 - ⇒ residuals not *iid.* Normal
 - ⊗ model not valid
- outcomes?
 - variance ↑ (usually, but can ↓)
 - increased residual *d.f.*
 - invalid inference

• Including irrelevant effects

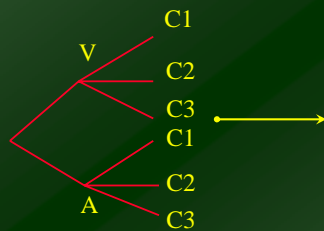
- ⊖ "waste" degrees of freedom
- ⊖ conservative tests
- ⊕ but safest!



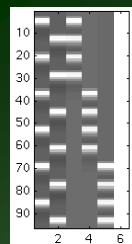
A real example (almost !)

Experimental Design ↔ Design Matrix

Factorial design with 2 factors : modality and category
 2 levels for modality (eg. Visual/Auditory)
 3 levels for category (eg. 3 categories of words)

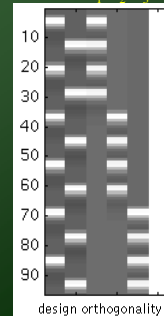


V A C₁ C₂ C₃



Asking ourselves some questions ...

V A C₁ C₂ C₃



Test C1 > C2 : c = [0 0 1 -1 0 0]

Test V > A : c = [1 -1 0 0 0 0]

Test C1, C2, C3 ? (F) c = [0 0 1 0 0 0]

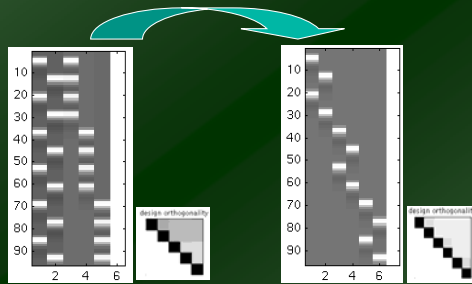
c = [0 0 0 1 0 0]

c = [0 0 0 0 1 0]

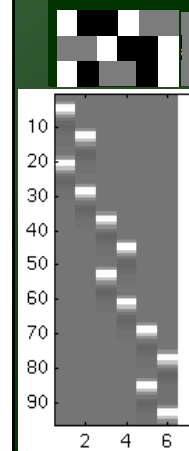
Test the interaction MxC ?

- Design Matrix not orthogonal
- Many contrasts are non estimable
- Interactions MxC are not modelled

Modelling the interactions



Asking ourselves some questions ...



Test $C1 > C2$: $c = [1 \ 1 \ -1 \ -1 \ 0 \ 0 \ 0]$

Test $V > A$: $c = [1 \ -1 \ 1 \ -1 \ 1 \ -1 \ 0]$

Test the category effect :

$$c = \begin{bmatrix} 1 & 1 & -1 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & -1 & -1 & 0 \\ 1 & 1 & 0 & 0 & -1 & -1 & 0 \end{bmatrix}$$

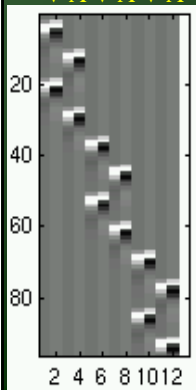
Test the interaction $M \times C$:

$$c = \begin{bmatrix} 1 & -1 & -1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & -1 & 1 & 0 \\ 1 & -1 & 0 & 0 & -1 & 1 & 0 \end{bmatrix}$$

- Design Matrix orthogonal
- All contrasts are estimable
- Interactions $M \times C$ modelled
- If no interaction ... ? Model is too "big" !

Asking ourselves some questions ... With a more flexible model

$C_1 \ C_1 \ C_2 \ C_2 \ C_3 \ C_3$
 $V \ A \ V \ A \ V \ A$



Test $C1 > C2$?

Test $C1$ different from $C2$?

from

$$c = [1 \ 1 \ -1 \ -1 \ 0 \ 0 \ 0]$$

to

$$c = \begin{bmatrix} 1 & 0 & 1 & 0 & -1 & 0 & -1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & -1 & 0 & -1 & 0 & 0 & 0 & 0 \end{bmatrix}$$

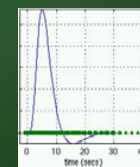
becomes an F test!

Test $V > A$?

$$c = [1 \ 0 \ -1 \ 0 \ 1 \ 0 \ -1 \ 0 \ 1 \ 0 \ -1 \ 0]$$

is possible, but is OK only if the regressors coding for the delay are all equal

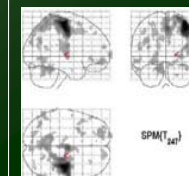
Convolution model



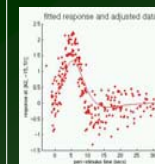
Design and contrast



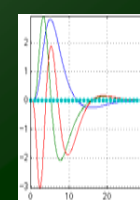
SPM(t) or SPM(F)



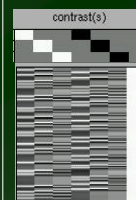
Fitted and adjusted data



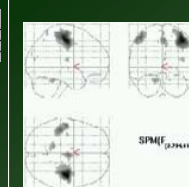
Convolution model



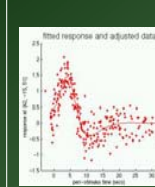
Design and contrast



SPM(t) or SPM(F)



Fitted and adjusted data



Toy example: take home ...

- *F tests have to be used when*
 - Testing for >0 and <0 effects
 - Testing for more than 2 levels
 - Conditions are modelled with more than one regressor
- *F tests can be viewed as testing for*
 - the additional variance explained by a larger model wrt a simpler model, or
 - the sum of the squares of one or several combinations of the betas (here the F test $b_1 - b_2$ is the same as $b_2 - b_1$, but two tailed compared to a t-test).

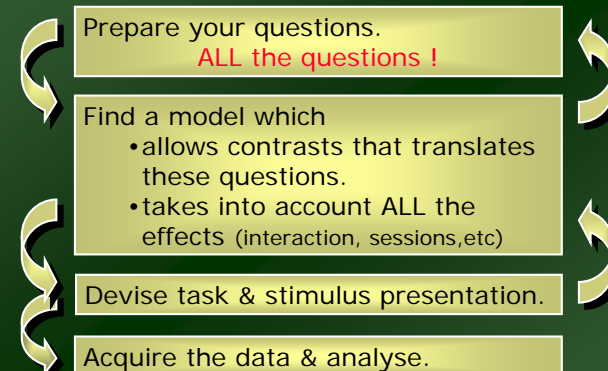
Contents

- Introduction & recap
- Experimental design
- « Take home » message

Conclusions...

- **General Linear Model**
 - (simple) standard statistical technique
 - temporal autocorrelation – a *Generalised Linear Model*
 - single general framework for many statistical analyses
 - flexible modelling \Leftarrow basis functions
 - design matrix visually characterizes model
 - fit data with combinations of columns of design matrix
 - statistical inference: *contrasts*...
 - *t*-tests: planned comparisons of the parameters
 - *F*-tests: general linear hypotheses, model comparison

Way to proceed



Not the other way round!!!

Three Stages of an Experiment

1. Sledgehammer Approach

- brute force experiment : powerful stimulus & don't try to control for everything
- look at what was done before or by others
- run a couple of subjects -- see if it looks promising
- if it doesn't look great, tweak the stimulus or task
- try to be a subject yourself so you can notice any problems with stimuli or subject strategies

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2. Real Experiment

- at some point, you have to stop changing things and collect enough subjects run with the same conditions to publish it
- how many subjects do you need
 - some psychophysical studies test two or three subjects, many studies test 6-10 subjects, random effects analysis requires at least 15 subjects,...
 - some subjects WILL be rejected, so acquire more than the minimum !
- can run all subjects in one or two days
 - pro: minimize setup and variability
 - con: "bad magnet day" means a lot of wasted time
 - make sure all the data are treated the "same way". (script)

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1. Sledgehammer Approach

2. Real Experiment

3. "Whipped Cream" experiment

- after the real experiment works, then think about a "whipped cream" version
- going straight to whipped cream is a huge endeavor, especially if you're new to imaging
- and it gives you a second paper !