# Statistical Analysis of Sensor Data

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#### 1 Introduction

2 Within-subject analysis

3 Between-subject analysis

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Litvak et al. (2011), Comp Intell and Neurosc



#### Multiple comparisons





Kilner et al. (2005), Neurosc Letters

## Multimodal face study



Henson et al. (2009); SPM8 Manual: Chapter 37

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### Within subject: Sensor-Time (3D)



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#### Within-subject: Time-Frequency







Difference Faces > Scrambled 50 100 150 200

250

300

0.5

1

1.5

Design matrix

2

2.5

Height threshold T = 3.736016 {p<0.05 (FWE)} Extent threshold k = 0 voxels

#### Statistics: p-values adjusted for search volume

set-level			cluster-level					∐z me				
p	С	₽ <sub>FWE-cor</sub>	r q <sub>FDR-corr</sub>	κ <sub>Ε</sub>	p <sub>uncorr</sub>		₽ <sub>FWE-corr</sub>	$q_{\rm FDR-corr}$	T	$(Z_{\equiv})$	₽ <sub>uncorr</sub>	112 1115
0.001	2	0.000 0.005	0.006 0.092	79 32	0.003 0.092		0.000 0.013	0.002 0.262	5.40 4.12	5.28 4.06	0.000 0.000	5 185 12 100

table shows 3 local maxima more than 8.0mm apart

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Height threshold: T = 3.74, p = 0.000 (0.050)	Degrees of freedom = [1.0, 334.0]
Extent threshold: k = 0 voxels, p = 1.000 (0.050)	FWHM = 7.5 58.5 Hz ms ; 7.5 11.7 {voxels}
Expected voxels per cluster, <k> = 13.420</k>	Volume: 28980 = 5796 voxels = 63.7 resels
Expected number of clusters, <c> = 0.05</c>	Voxel size: 1.0 5.0 Hz ms ; (resel = 87.91 voxels)
FWEp: 3.736, FDRp: 5.396, FWEc: 32, FDRc: 79	
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nenson et al. (2008), Neuronnage

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#### Between subjects: Time-Frequency









set-level		cluster-level				peak-level						Ha me	
D	С	D FWE-con	FDR-con	. <sup>К</sup> е	Duncorr	D FWE-cor	D FWE-corr Q FDR-corr T (Z ) D		Duncorr	112 1115			
0.161	2	0.059 0.441	0.169 0.814	12 1	0.085 0.814	0.082 0.507	0.240 0.991	4.88 3.65	3.81 3.09	0.000 0.001	14 6	80 80	



#### Parametric design at group level (MEG)





Talmi et al. (2012), NeuroImage

## MEG: Alignment of sensor data



Taulu et al. (2005), IEEE Trans on Signal Processing; Software MaxFilter (Elekta NeuroMag)

with respect to sensors

### MEG between-subjects

#### Analysis over subjects (2<sup>nd</sup> Level)

NOTE for MEG: Complicated by variability in head-position SOLUTION: Virtual transformation to same position in sessions, subjects

Without transformation to Device Space



#### With transformation to Device Space



Stats over 18 subjects, planar gradiometers

#### **Improved with transform:** more blobs, larger T values

Taylor & Henson (2008) Biomag; Taulu et al. (2005), IEEE Trans on Sig Process

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#### 

- Matched filter theorem: Different data needs different kernels
- Smoothing helps aligning the data
- Random field theory assumptions
- Evoked responses:
  - Some studies have used Gaussian kernel with FWHM of 8 - 10 mm in space (after transforming in voxel-space), and 8 - 10 ms in time
  - Note: Some temporal smoothness already because of lowpass filter

Litvak et al. (2011), Comp Intell and Neurosc

## Significant results?

Potentially useful in practice:

- Use strong prior hypothesis
  - reduces search volume (lower corrected p-values)
  - can be more compelling
  - use small volume correction
  - equivalent to fMRI: use uncorrected p-values in case of hypothesis
- Functional localizer
  - For example: Localize ROIs for response to faces in each subject
- EEG/MEG: More subjects necessary than for equivalent fMRI studies?

#### References

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## Thank you MAX PLANCK INSTITUTE

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