

# Methods & Models for fMRI Analysis 2019

# TUTORIAL

---

Sandra Iglesias

[iglesias@biomed.ee.ethz.ch](mailto:iglesias@biomed.ee.ethz.ch)

Translational Neuromodeling Unit (TNU)  
Institute for Biomedical Engineering (IBT)  
University and ETH Zürich



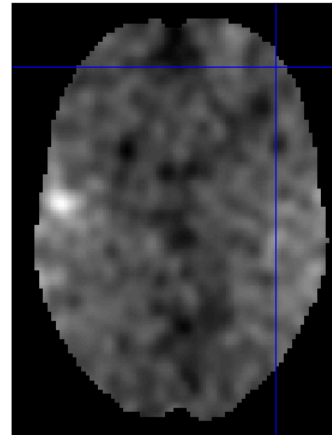
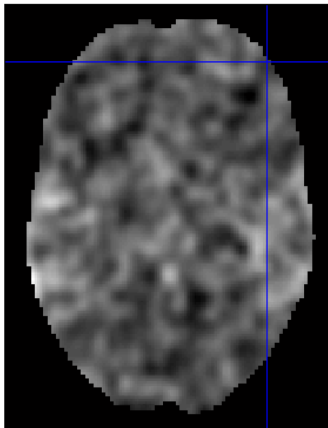
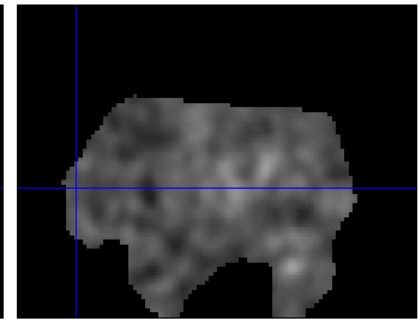
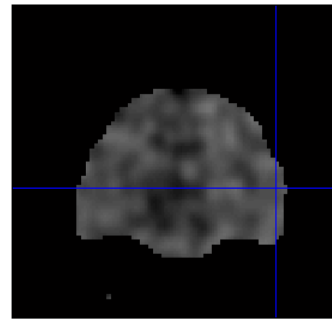
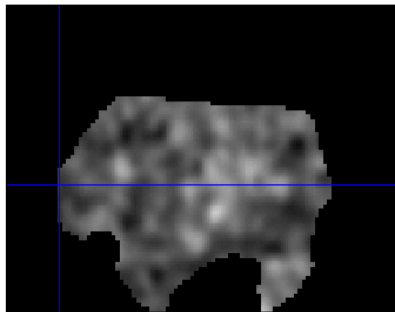
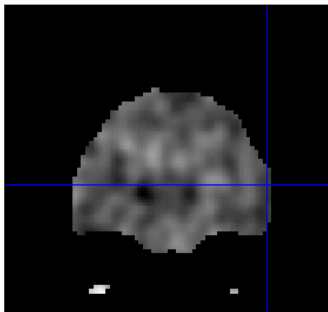
University of  
Zurich UZH

**ETH**

Eidgenössische Technische Hochschule Zürich  
Swiss Federal Institute of Technology Zurich



Translational Neuromodeling Unit



**Beta value** = % change above global mean for a particular condition

Crosshair Position	
mm:	-58.0 -12.0 6.0
vc:	69.0 51.0 29.0
Intensity:	4.54226

right [mm]	0
forward [mm]	0
up [mm]	0
pitch [rad]	0
roll [rad]	0
yaw [rad]	0
resize [x]	1
resize [y]	1
resize [z]	1

Reorient images... Reset...

File: **..lysism\_ove2|beta\_0005.img**

Dimensions: 79 x 95 x 69

Datatype: float32

Intensity: Y = 1 X

**spm\_spm:beta (0005) - Sn(1) StV\*bf(1)**

Vox size: 2 x 2 x 2

Origin: 40 57 26

Dir Cos: 1.000 0.000 0.000  
0.000 1.000 0.000  
0.000 0.000 1.000

Full Volume	Hide Crosshairs
World Space	bilin interp
Auto Window	Add Blobs

Crosshair Position	
mm:	-58.0 -12.0 6.0
vc:	69.0 51.0 29.0
Intensity:	23.356

**Con. value** = summation of all relevant betas.

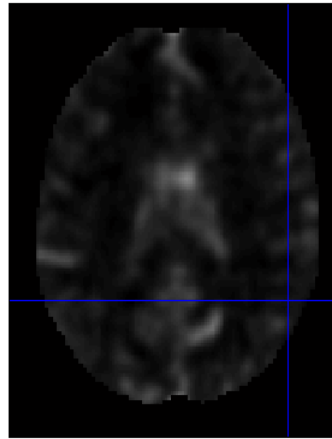
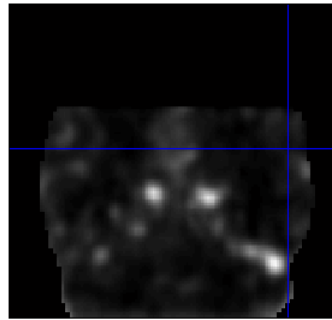
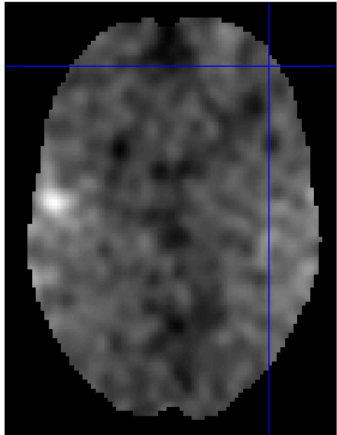
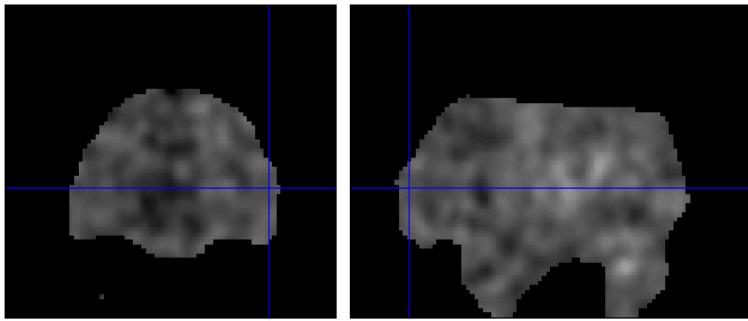
File: **..alysism\_ove2|con\_0005.img**

Dimensions: 79 x 95 x 69

Datatype: float32

Intensity: Y = 1 X

SPM contrast - 5: stv



1g Unit

**ResMS.img** =  $\hat{\sigma}_i^2$   
 residual sum of squares  
 or variance image and is  
 a measure of within-  
 subject error at the 1<sup>st</sup>  
 level (or between-subject  
 error at the 2<sup>nd</sup>).

Crosshair Position	
mm:	-58.0 -12.0 6.0
vc:	69.0 51.0 29.0
Intensity:	23.356

File:..alysismove2lcon\_0005.img  
 Dimensions: 79 x 95 x 69  
 Datatype: float32  
 Intensity: Y = 1 X  
 SPM contrast - 5: stv

Crosshair Position	
mm:	-58.0 -12.0 6.0
vc:	69.0 51.0 29.0
Intensity:	2.04154
right [mm]	0
foward [mm]	0
up [mm]	0
pitch [rad]	0
roll [rad]	0
yaw [rad]	0
resize [x]	1
resize [y]	1
resize [z]	1

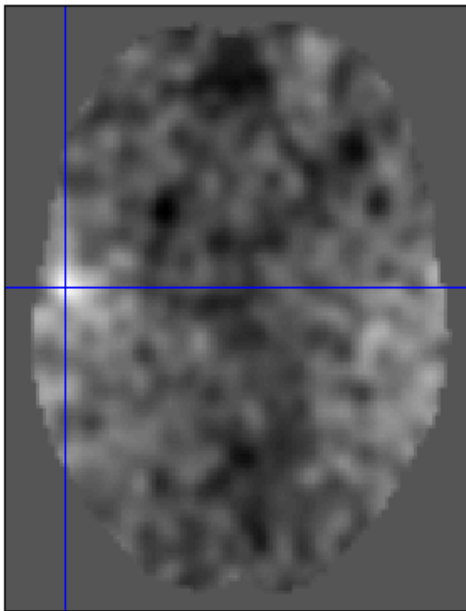
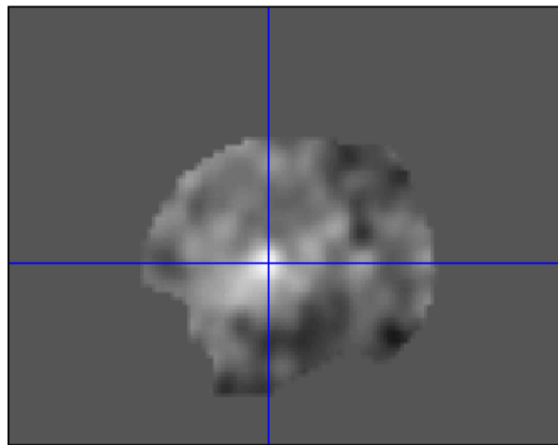
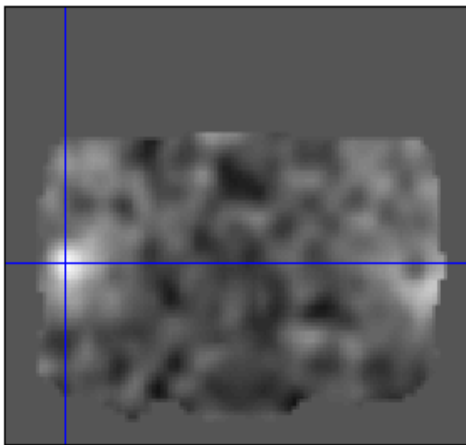
Reorient images... Reset...

File:..lanalysismove2lResMS.img  
 Dimensions: 79 x 95 x 69  
 Datatype: float64  
 Intensity: Y = 0.00118203 X  
 spm\_spm: Residual sum-of-squares

Vox size: -2 x 2 x 2  
 Origin: 40 57 26  
 Dir Cos: 1.000 0.000 0.000  
 0.000 1.000 0.000  
 0.000 0.000 1.000

Full Volume	Hide Crosshairs
World Space	bilin interp
Auto Window	Add Blobs

Con. value is combined with  
 ResMS value at that voxel to  
 produce a T statistic or spm.T.img.



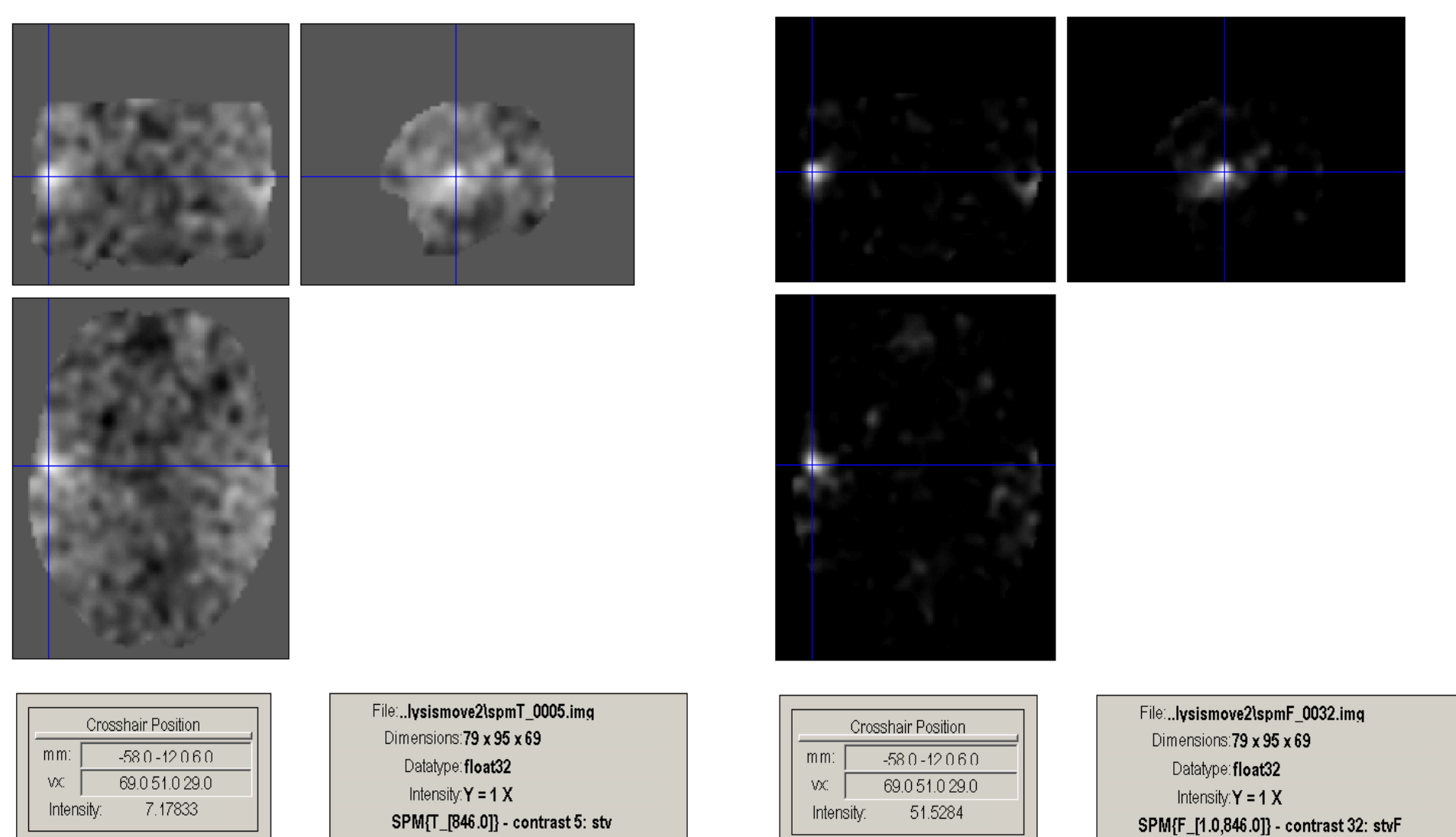
spmT.img  
Thresholded using the  
results button.

Test statistic:

$$T = \frac{\text{contrast of estimated parameters}}{\sqrt{\text{variance estimate}}}$$

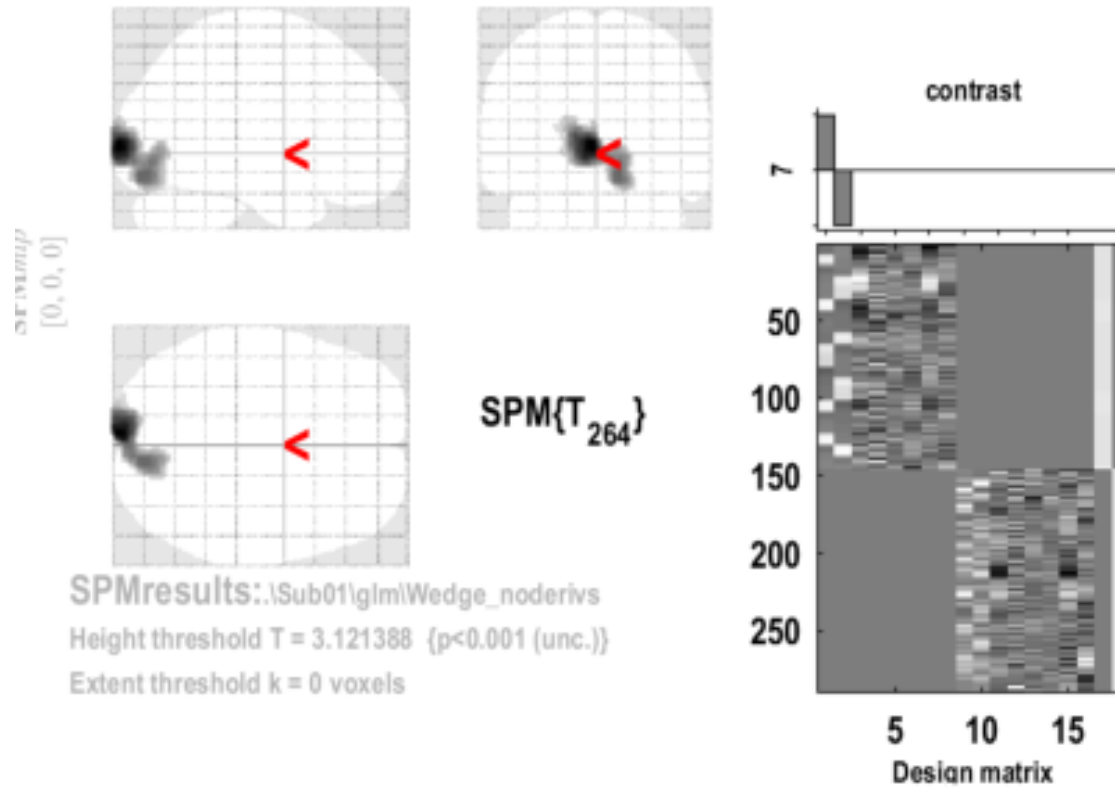
Crosshair Position	
mm:	-58.0 -12.0 6.0
vx:	69.0 51.0 29.0
Intensity:	7.17833

File: **..lysismove2\spmT\_0005.img**  
 Dimensions: **79 x 95 x 69**  
 Datatype: float32  
 Intensity: Y = 1 X  
**SPM{T\_[846.0]} - contrast 5: stv**



spmT.img and corresponding spmF.img

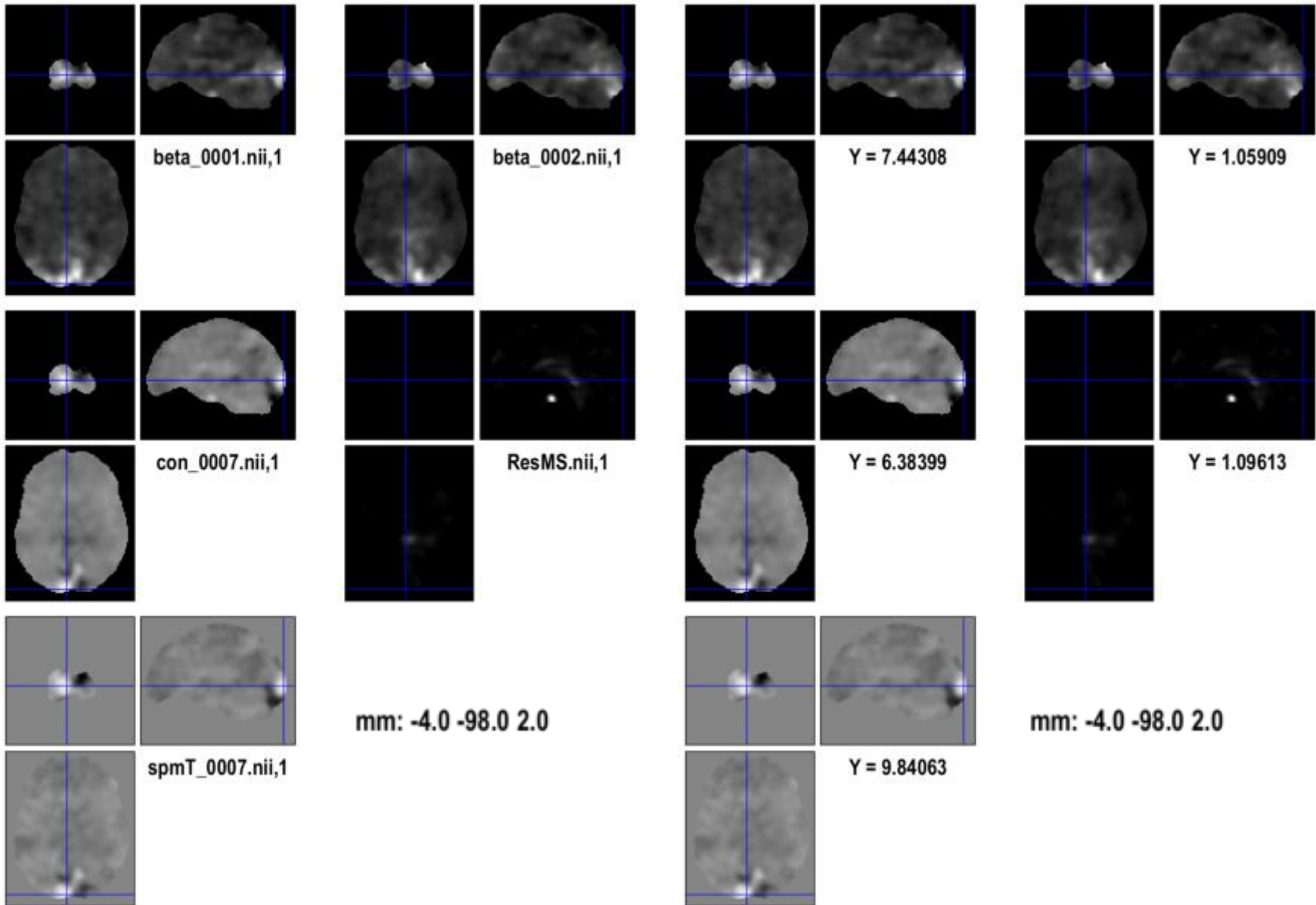
### Left > Right run1



Statistics: *p-values adjusted for search volume*

cluster-level				peak-level					mm mm mm		
$p_{FWE-corr}$	$q_{FDR-corr}$	$k_E$	$p_{uncorr}$	$p_{FWE-corr}$	$q_{FDR-corr}$	$T$	$(Z_{...})$	$p_{uncorr}$			
0.000	0.000	1282	0.000	0.000	0.000	9.84	Inf	0.000	-4	-98	2
				0.000	0.000	6.63	6.37	0.000	10	-86	-14
				0.000	0.000	6.27	6.05	0.000	12	-80	-6

spm\_contrasts.m → Compute SPM{t} image line 213

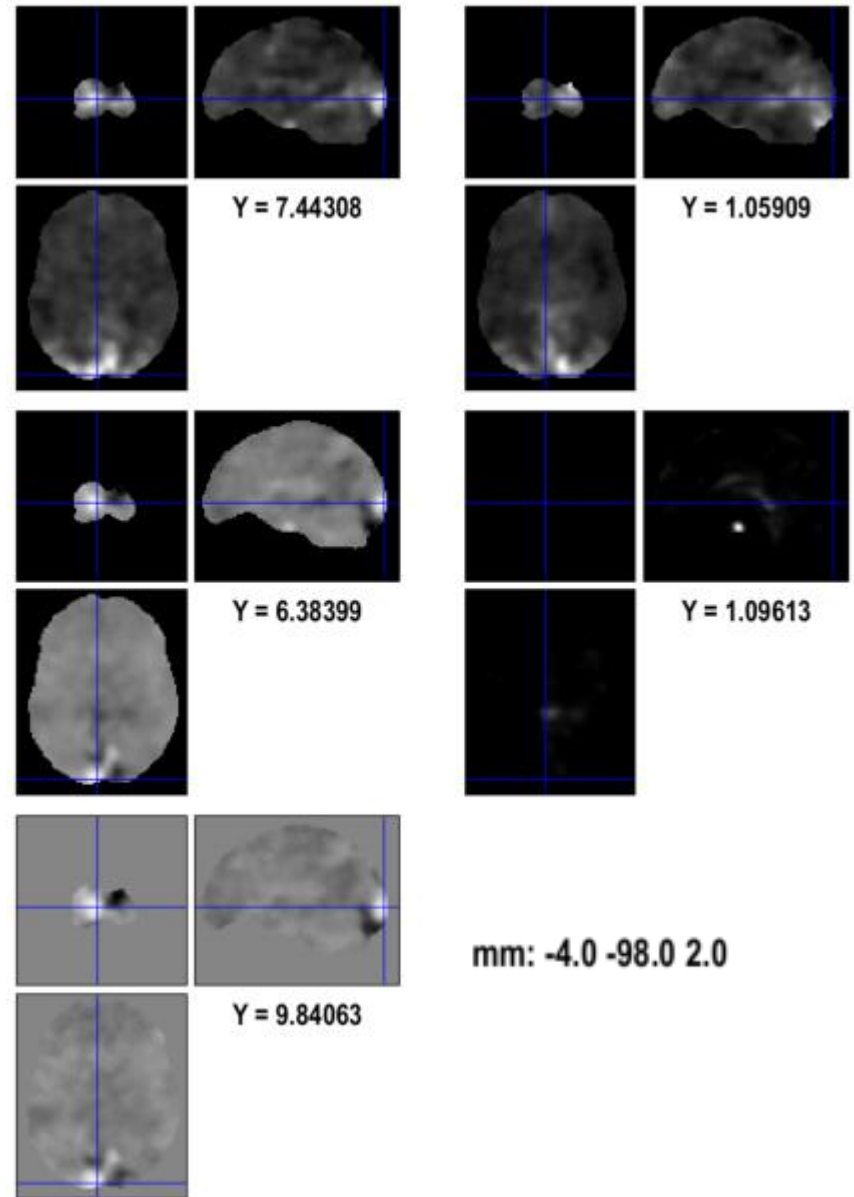


spm\_contrasts.m → Compute SPM{t} image line 213

c = contrast weights  
c=[1 -1 zeros(1,16)]

ResMS\_intensity = 1.09613  
con\_image\_intensity = 6.38399

$Vc=c*SPM.xX.Bcov*c'$   
 $SE = \sqrt{Vc*ResMS\_intensity}$   
con\_image\_intensity/SE





# So, which images?

- **beta images** contain information about the size of the effect of interest.
- Information about the error variance is held in the **ResMS.img**.
- beta images are linearly combined to produce relevant **con.img**.
- The design matrix, contrast, constant and ResMS.img are subjected to matrix multiplication to produce an estimate of the st.dev. associated with each voxel in the con.img.
- The **spmT.img** are derived from this and are thresholded in the results step.

# Worth to check:

## Review Design

- Matrix
- Orthogonality

## Explore Design:

- Sessions

## mask.nii

## Correlation among regressors:

%% SPM.xX.xKXs.X: matrix of trials and betas (columns) in each trial

%% R=corrcoef(X) calculates a matrix R of correlation coefficients

```
corr_pre=corrcoef(SPM.xX.X);
```

```
figure; imagesc(corr_pre, [-1 1]);colorbar
```

%% after whitening:

```
corr=corrcoef(SPM.xX.xKXs.X);
```

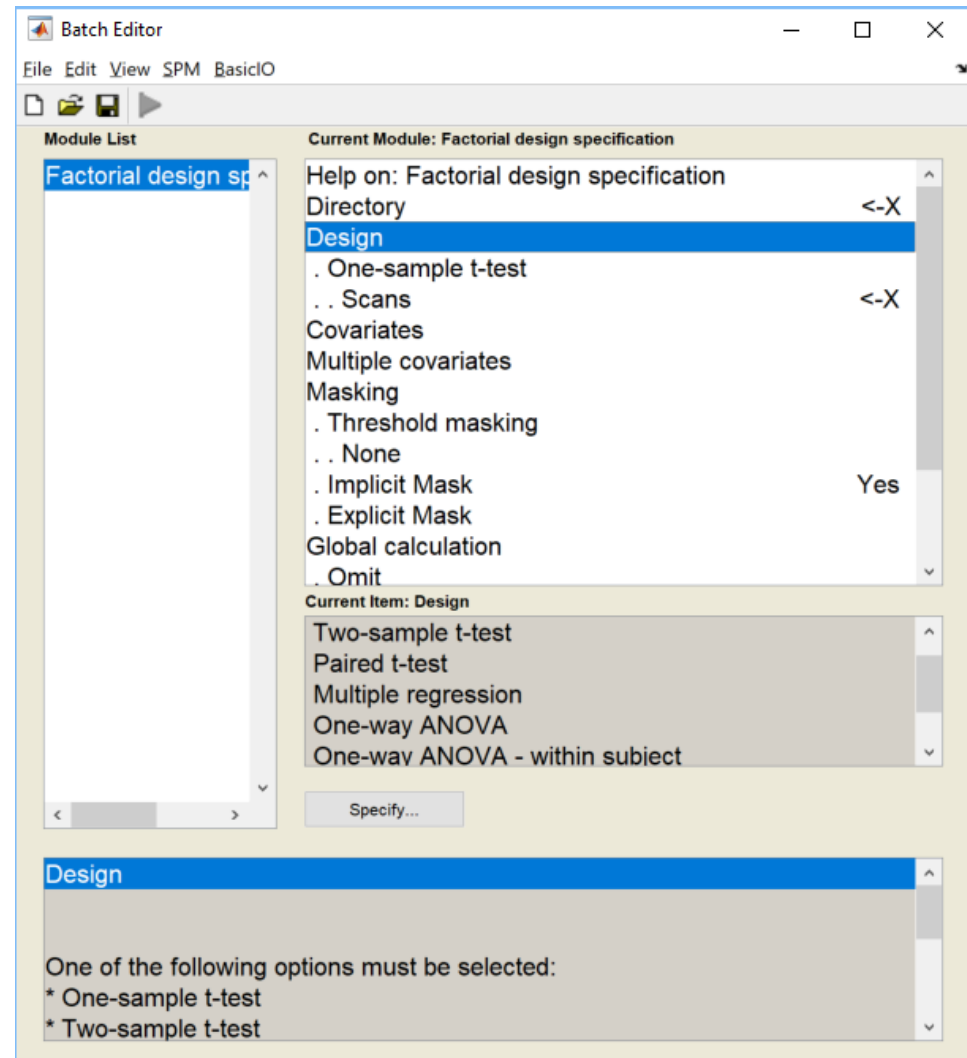
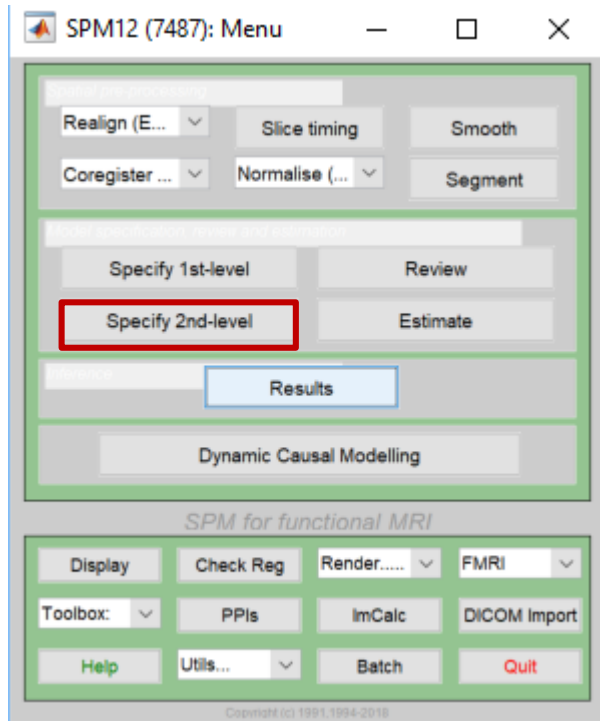
```
figure; imagesc(corr, [-1 1]);colorbar
```

# SPM interface: factorial design specification

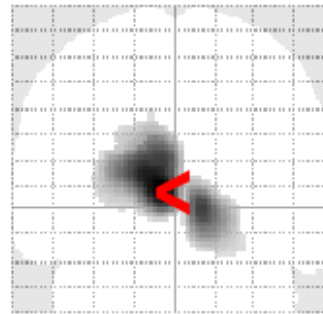
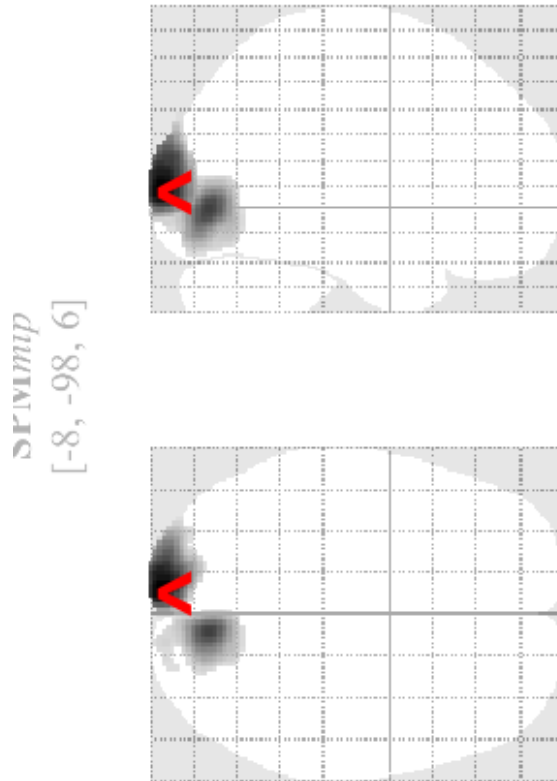
## Options:

- FFX
- One-sample t-test
- Paired t-test
- One-way ANOVA – within subject
- Flexible factorial

# SPM interface: factorial design specification

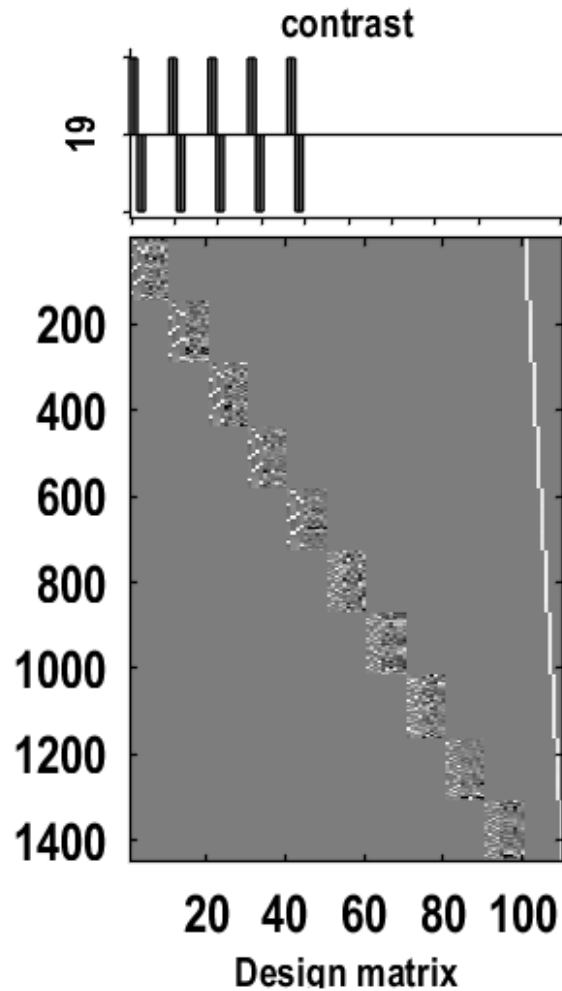


## TLBR > TRBL: Block



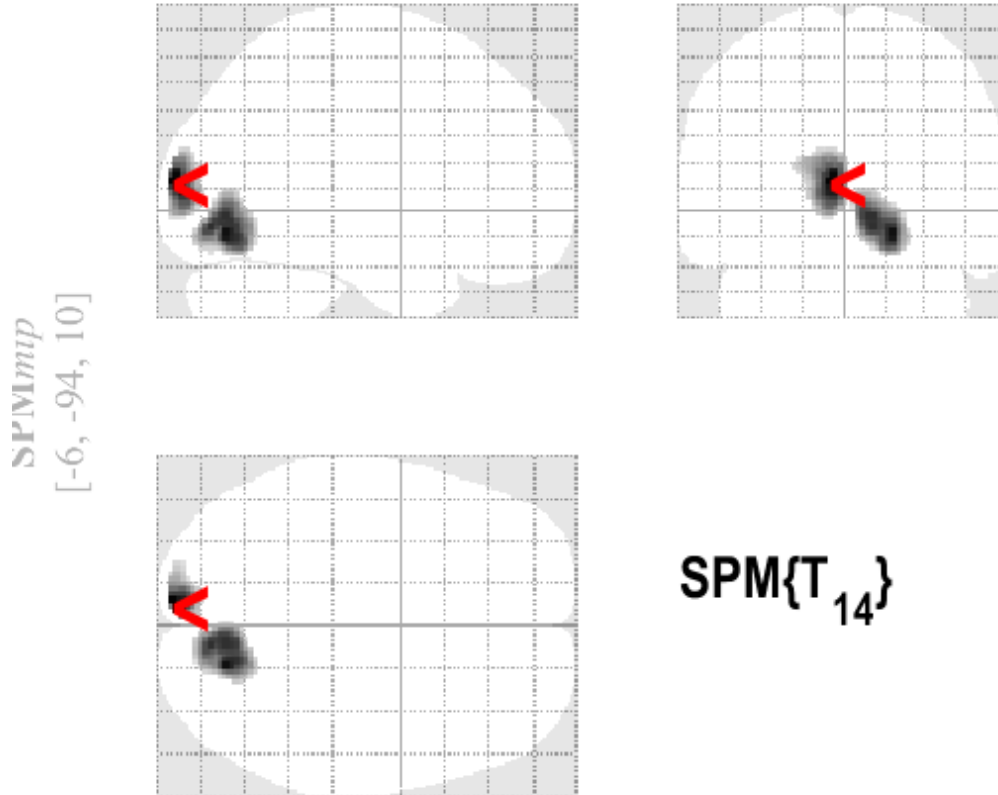
SPM{T<sub>1300</sub>}

SPMresults:\Group\ffx\Individual  
Height threshold T = 3.096514 {p<0.001 (unc.)}  
Extent threshold k = 1248 voxels



# One-sample t-test:

TLBR > TRBL: all

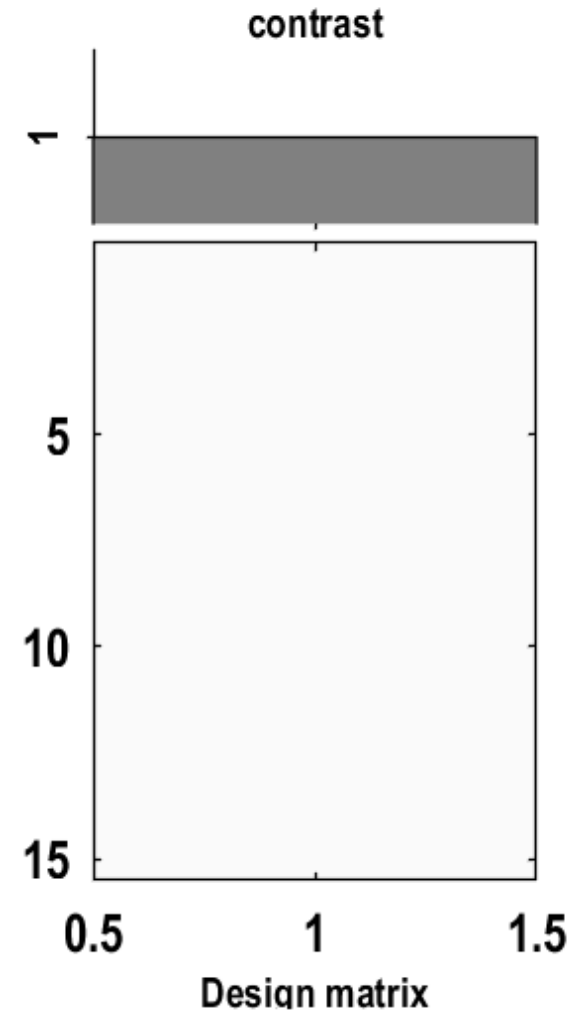


SPMresults:\Group\OneSample\3

Height threshold  $T = 3.787390$  { $p < 0.001$  (unc.)}

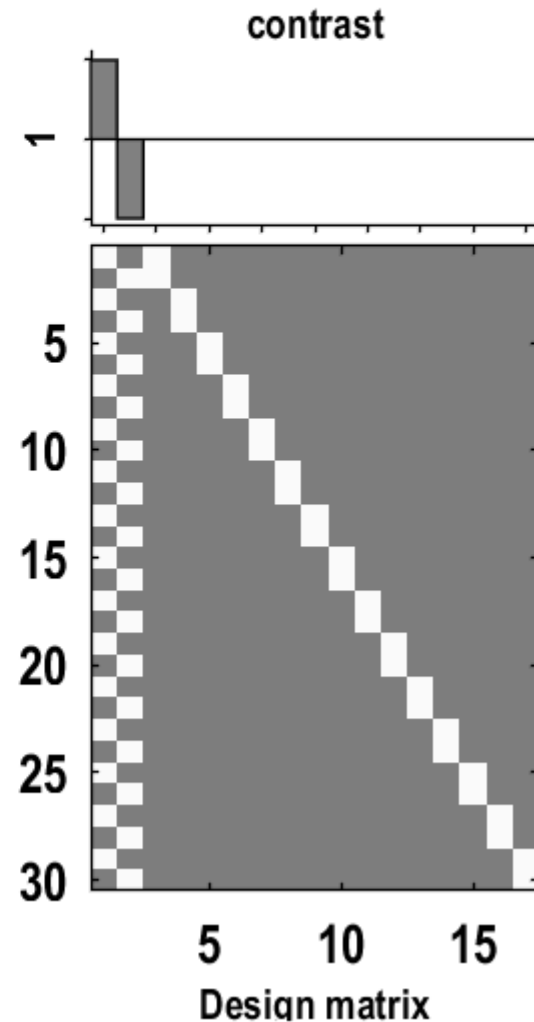
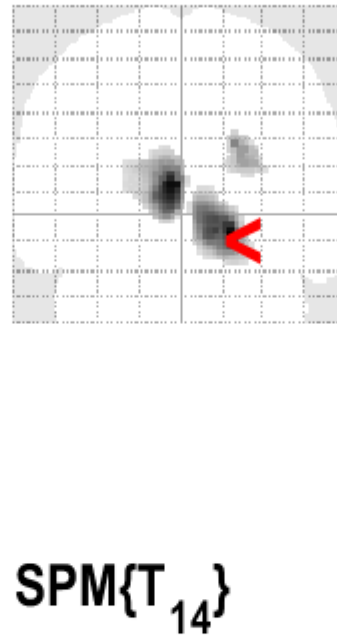
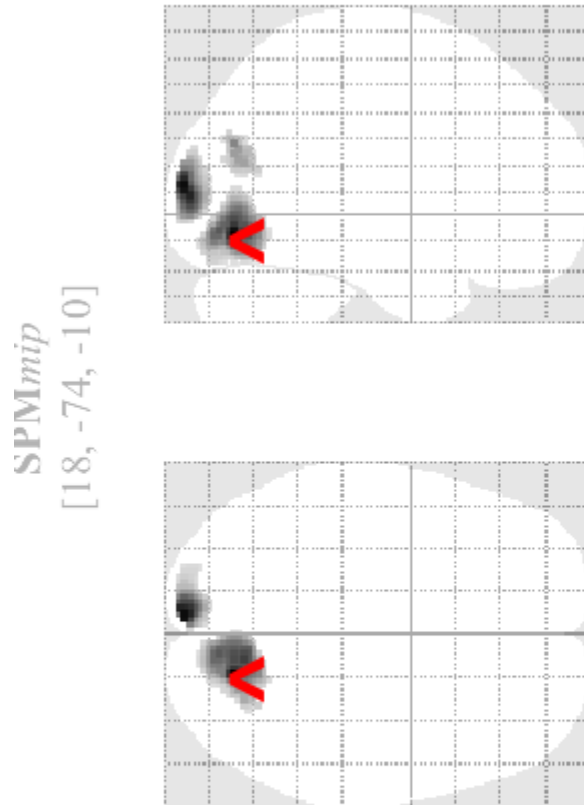
Extent threshold  $k = 387$  voxels

$p < 0.05$  FWE at cluster-level, with CDT: $p < 0.001$



# Paired t-test:

## TLBR vs TRBL: block

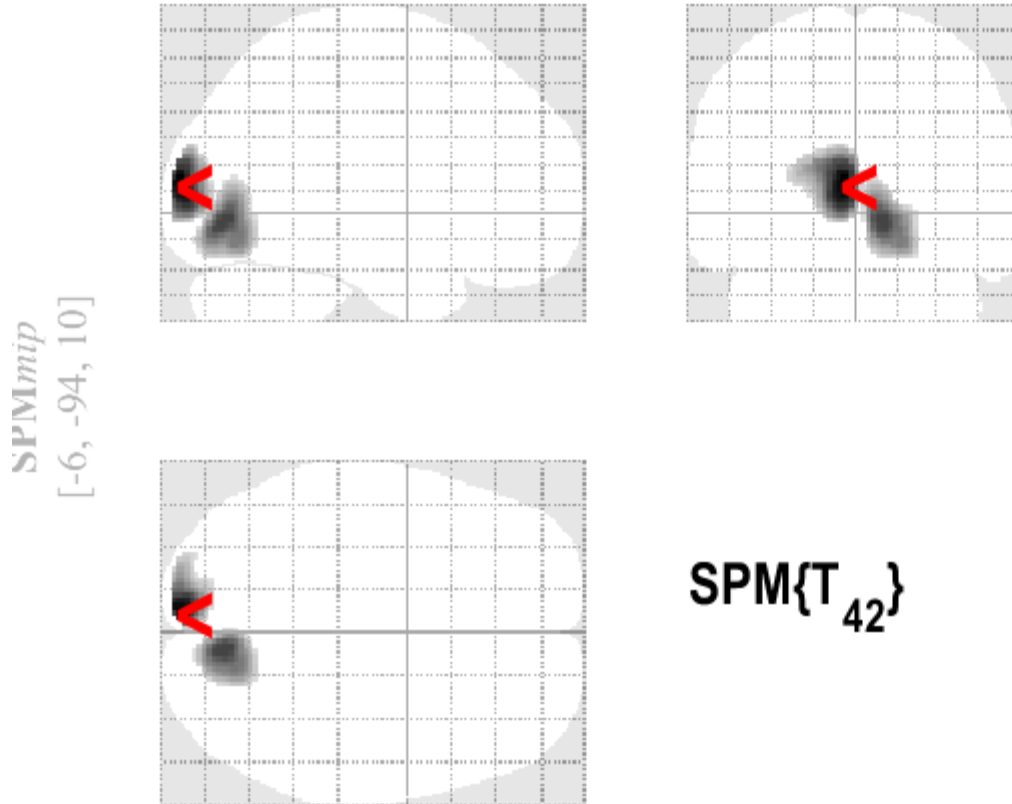


SPMresults:\Paired\Retinotopy\block  
Height threshold  $T = 3.787390$  { $p < 0.001$  (unc.)}  
Extent threshold  $k = 146$  voxels

$p < 0.05$  FWE at cluster-level, with CDT:  $p < 0.001$

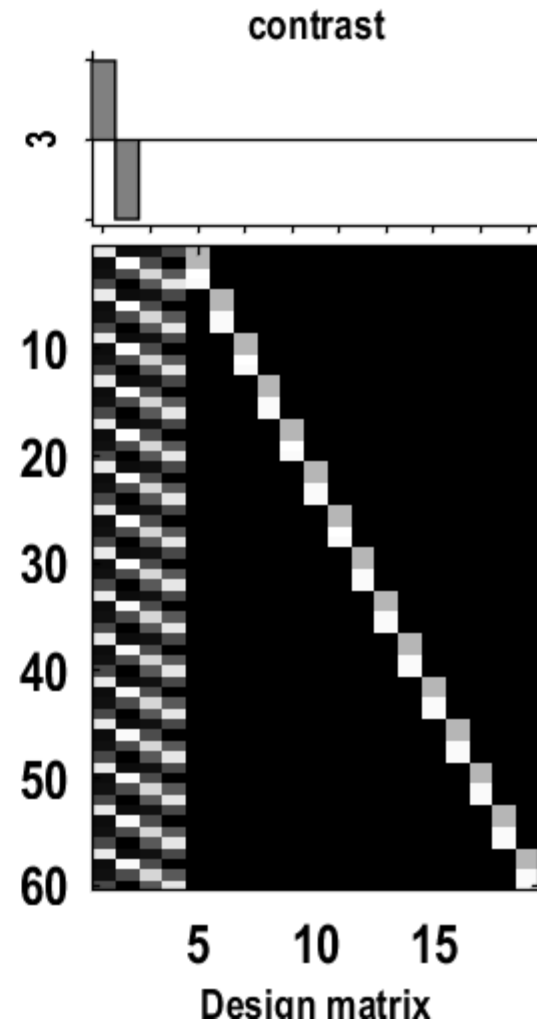
# wANOVA:

## TLBR vs TRBL: block



SPMresults:\Group\wAnova\Retinotopy  
Height threshold  $T = 3.295951$  { $p < 0.001$  (unc.)}  
Extent threshold  $k = 0$  voxels

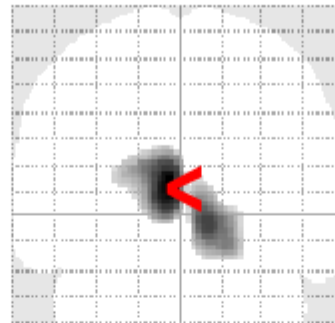
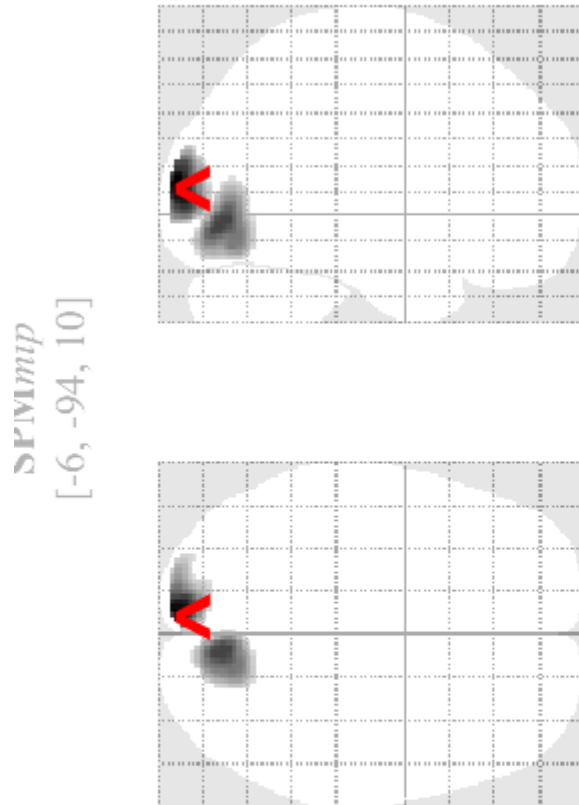
$p < 0.05$  FWE at cluster-level, with CDT:  $p < 0.001$





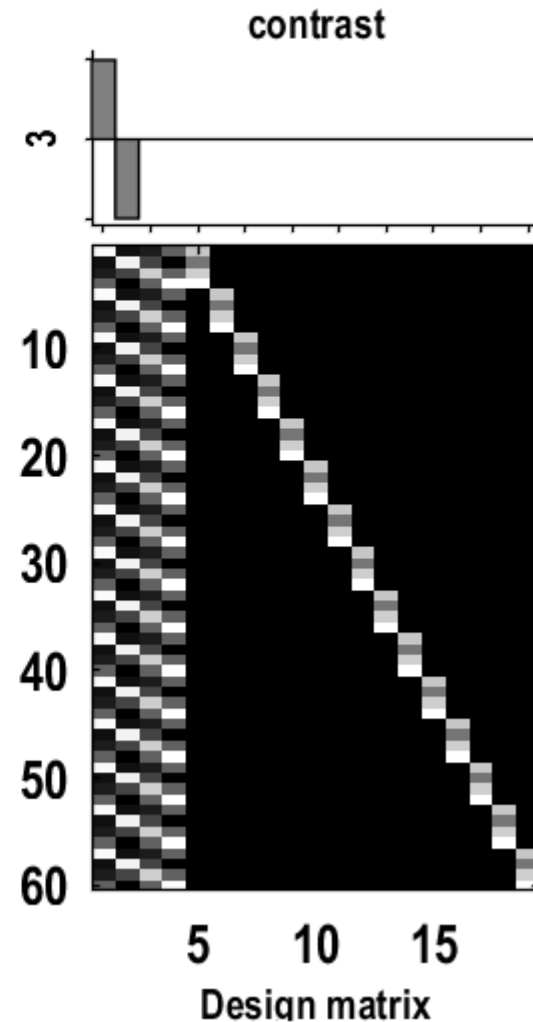
# flexANOVA:

## TLBR vs TRBL: block



$SPM\{T_{42}\}$

SPMresults:.\Group\flexAnova\Retinotopy  
Height threshold  $T = 3.295951$  { $p < 0.001$  (unc.)}  
Extent threshold  $k = 0$  voxels



$p < 0.05$  FWE at cluster-level, with CDT:  $p < 0.001$

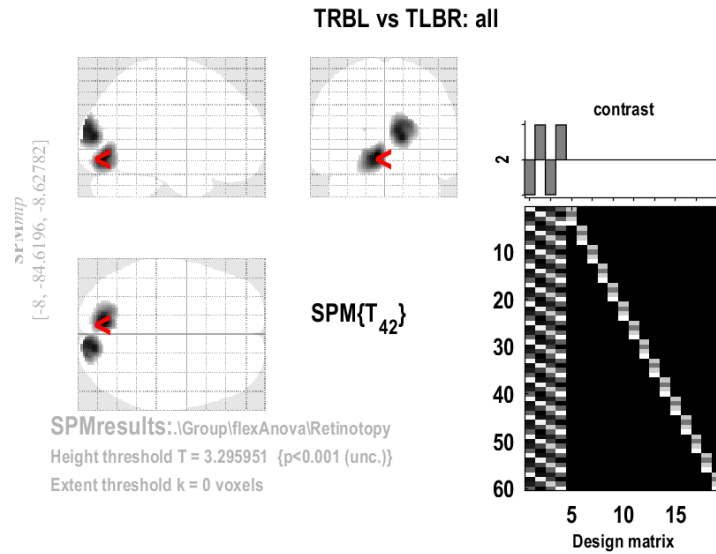
# Importance of sphericity specification:

Compare one-way within-subject ANOVA using:

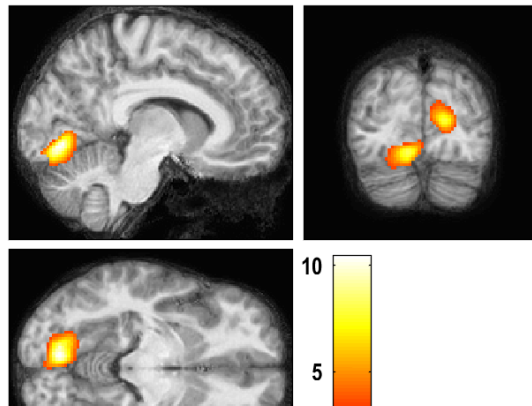
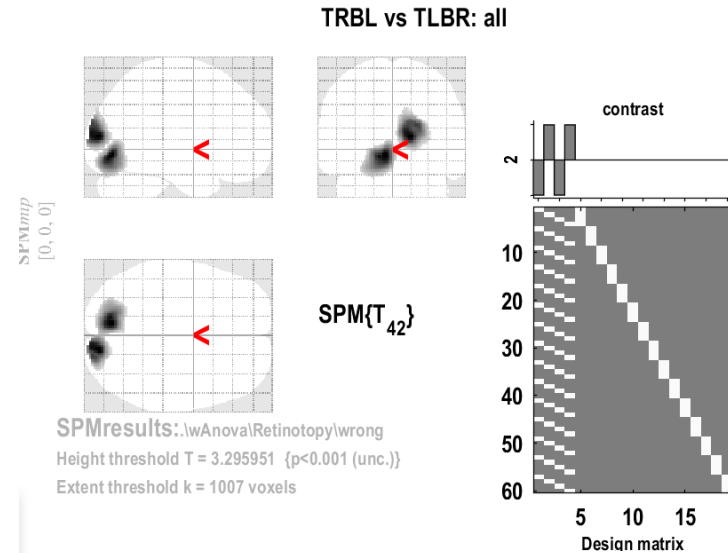
- correct sphericity specification vs.
- wrong sphericity specification

# Sphericity specification:

correct sphericity specification



wrong sphericity specification



Statistics: *p*-values adjusted for search volume

set-level		cluster-level				peak-level				mm mm mm			
<i>p</i>	<i>c</i>	<i>p</i> <sub>FWE-corr</sub>	<i>q</i> <sub>FDR-corr</sub>	<i>k</i> <sub>E</sub>	<i>p</i> <sub>uncorr</sub>	<i>p</i> <sub>FWE-corr</sub>	<i>q</i> <sub>FDR-corr</sub>	<i>T</i>	( <i>Z</i> <sub>≡</sub> )	<i>p</i> <sub>uncorr</sub>			
1.000	3	0.000	0.000	984	0.000	0.000	0.000	10.41	7.28	0.000	-8	-78	-8
		0.000	0.000	807	0.000	0.000	0.000	9.73	7.00	0.000	14	-88	10

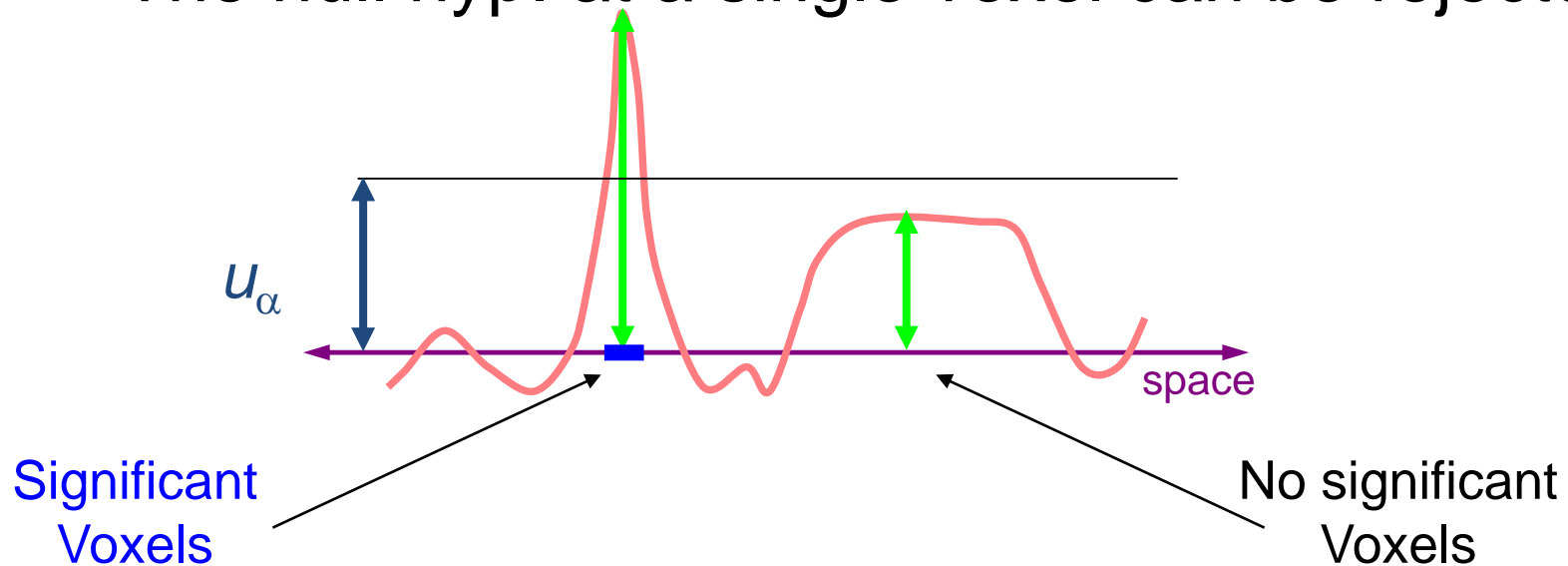
Statistics: *p*-values adjusted for search volume

set-level		cluster-level				peak-level				mm mm mm			
<i>p</i>	<i>c</i>	<i>p</i> <sub>FWE-corr</sub>	<i>q</i> <sub>FDR-corr</sub>	<i>k</i> <sub>E</sub>	<i>p</i> <sub>uncorr</sub>	<i>p</i> <sub>FWE-corr</sub>	<i>q</i> <sub>FDR-corr</sub>	<i>T</i>	( <i>Z</i> <sub>≡</sub> )	<i>p</i> <sub>uncorr</sub>			
0.000	2	0.000	0.000	1007	0.000	0.000	0.000	16.00	In∞	0.000	14	-88	10
		0.000	0.000	12.07	In∞	0.000	0.000	12.07	In∞	0.000	20	-94	20
		0.000	0.000	1408	0.000	0.000	0.000	15.92	In∞	0.000	-8	-78	-8

$p < 0.05$  FWE at cluster-level, with CDT:  $p < 0.001$

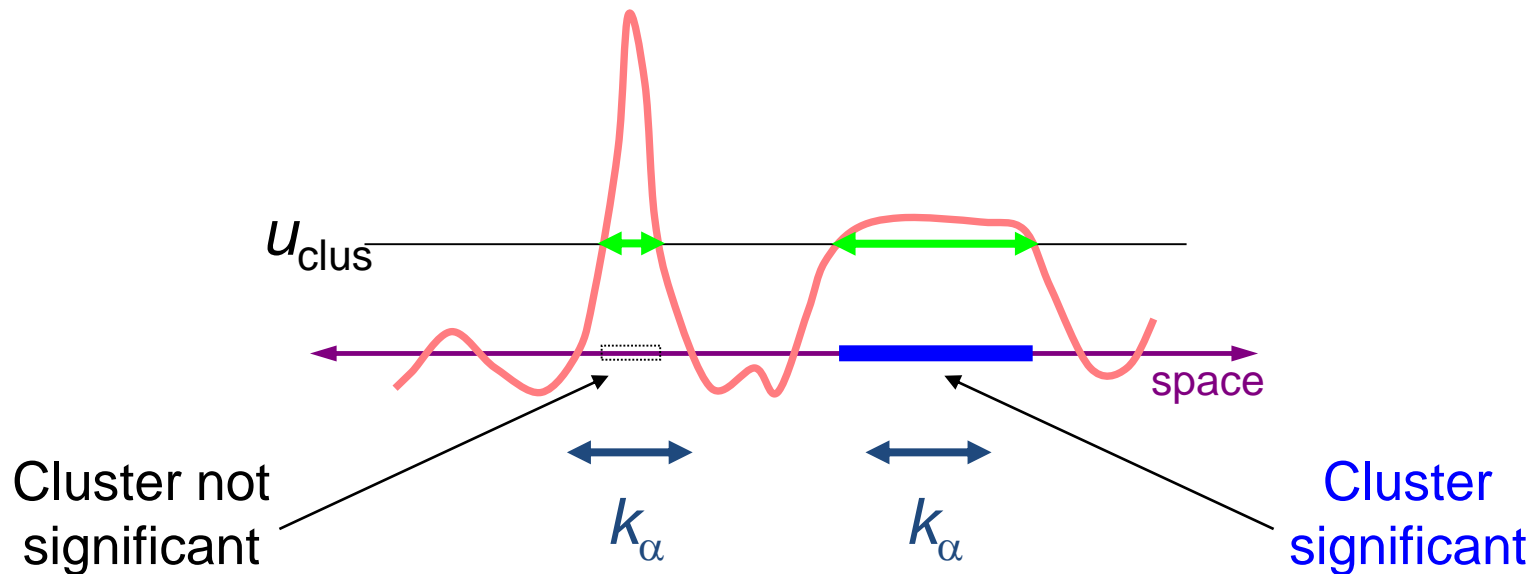
# Voxel-level Inference

- Retain voxels above  $\alpha$ -level threshold  $u_\alpha$ 
  - Signal magnitude
- Gives best spatial specificity
  - The null hyp. at a single voxel can be rejected



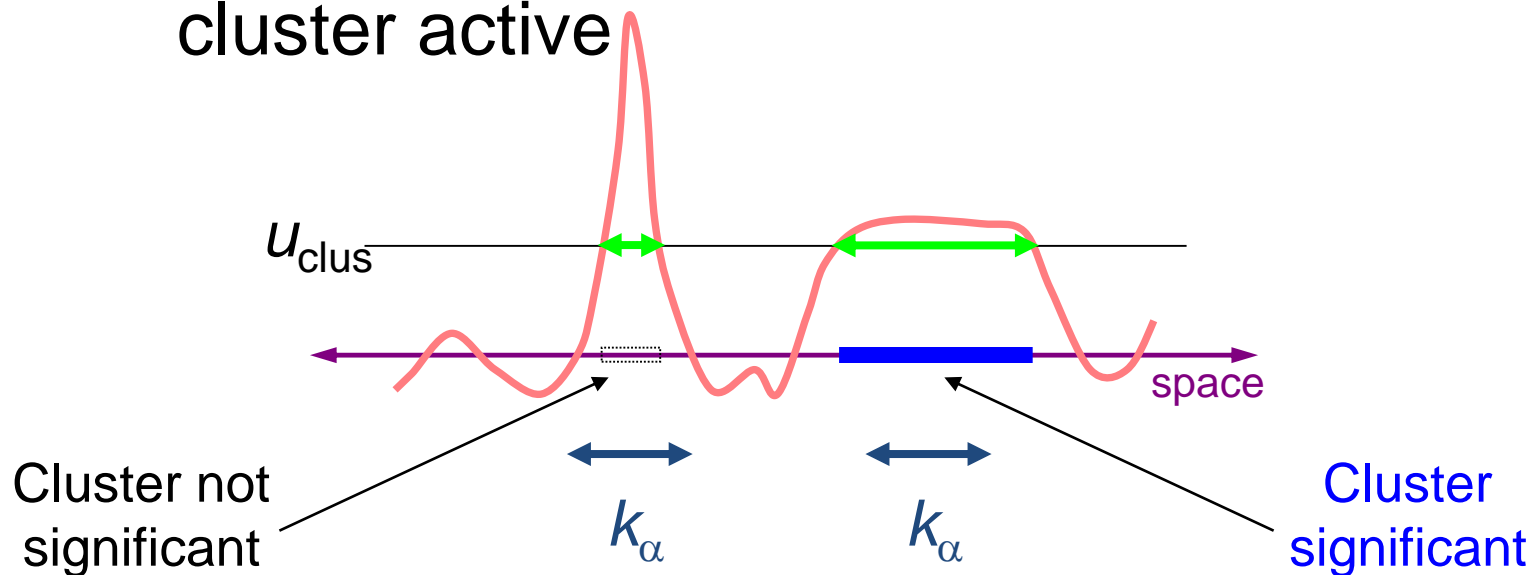
# Cluster-level Inference

- Two step-process
  - Define clusters by threshold  $u_{\text{clus}}$
  - Retain clusters larger than  $\alpha$ -level threshold  $k_{\alpha}$ : Spatial extent



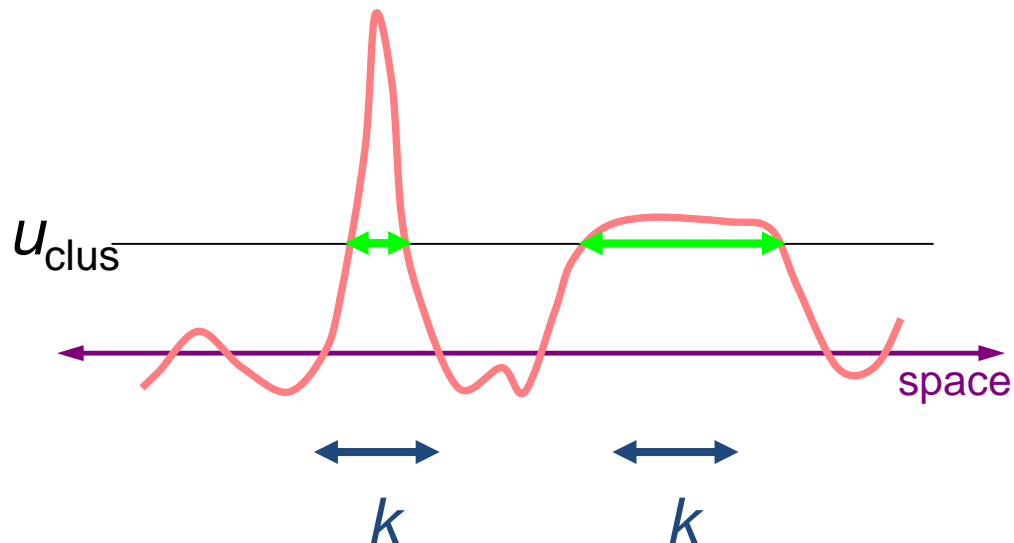
# Cluster-level Inference

- Typically better sensitivity
- Worse spatial specificity
  - The null hyp. of entire cluster is rejected
  - Only means that *one or more* of voxels in cluster active



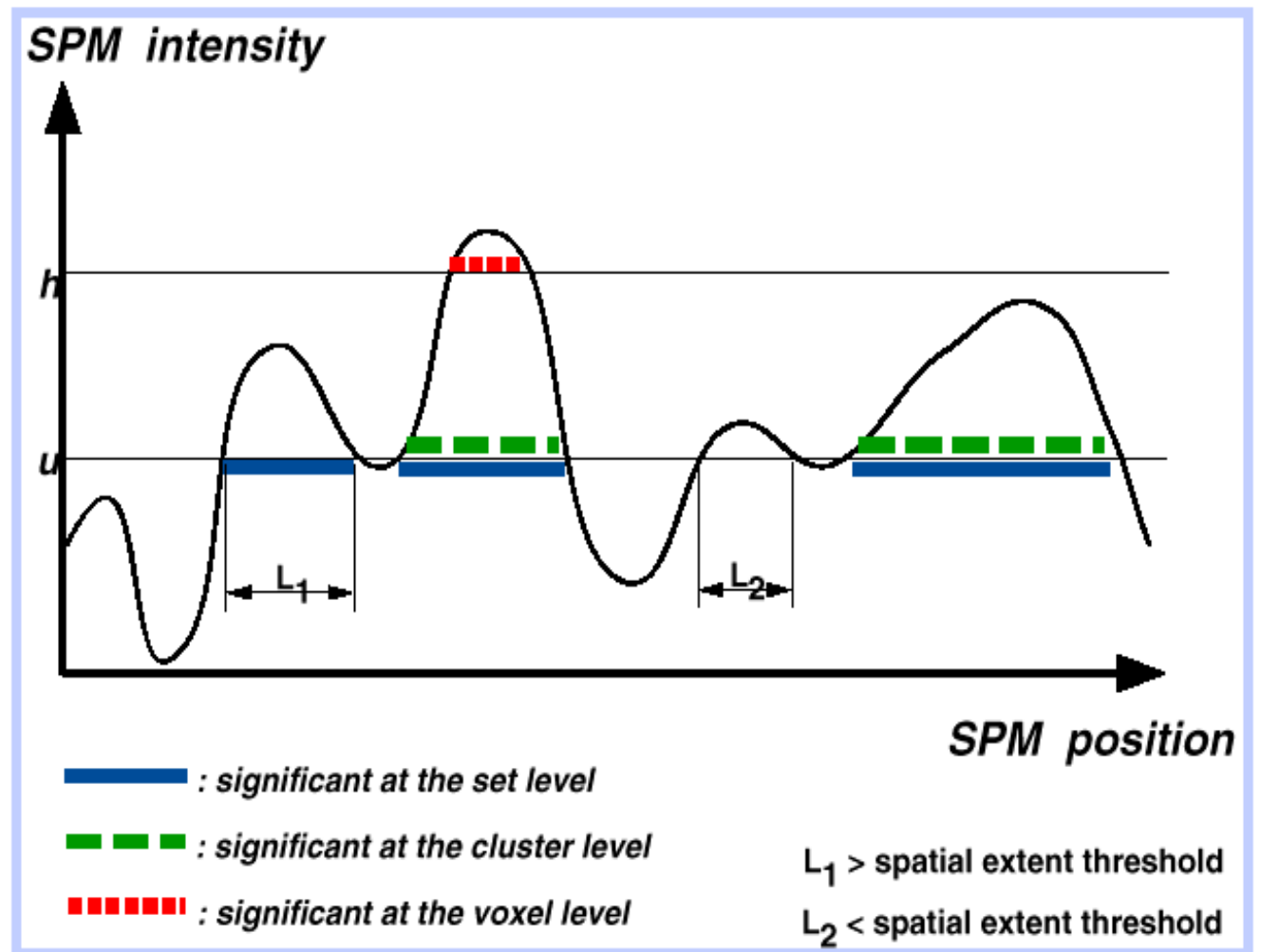
# Set-level Inference

- Count number of blobs  $c$ 
  - Minimum blob size  $k$
- Worst spatial specificity
  - Only can reject global null hypothesis



Here  $c = 1$ ; only 1 cluster larger than  $k$

# Review: Levels of inference & power



Set level...

Cluster level...

Voxel level...



# Options for displaying contrasts:

## Apply masking

- none
- contrast
- image
- atlas

Maximum probability tissue labels derived from the "MICCAI 2012 Grand Challenge and Workshop on Multi-Atlas Labeling"<sup>7</sup> are available in files `tpm/labels_Neuromorphometrics.{nii,xml}`. These data were released under the Creative Commons Attribution-NonCommercial (CC BY-NC) with no end date. Users should credit the MRI scans as originating from the OASIS project<sup>8</sup> and the labeled data as "provided by Neuromorphometrics, Inc."<sup>9</sup> under academic subscription". These references should be included in all workshop and final publications. See `spm_templates.man` for more details about the generation of this file.

# Options for displaying contrasts:

## Apply masking (image) vs. small volume correction

- $p$ -values will change
  - e.g. Positive effect of condition\_1 / left middle frontal gyrus

## Apply explicit mask vs. small volume correction

- $p$ -values should be the same
- can differ as your ReML-estimation will be based on the ROI

## Save thresholded image

- thresholded SPM
- all clusters
- current cluster