

Feature-Space Clustering for fMRI Meta-Analysis

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Human Brain Mapping : July 2001.

Scope

Exploratory data analysis

as a possible alternative to *inferential analyses*

Generalize the existing methodologies for clustering raw time series:
for *data reduction* &
meta-analysis of *single-voxel* neuroimaging analyses.

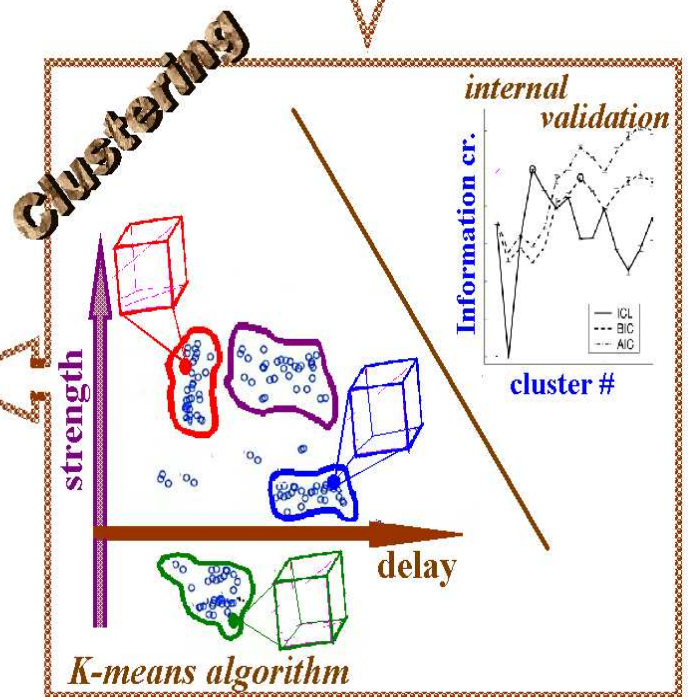
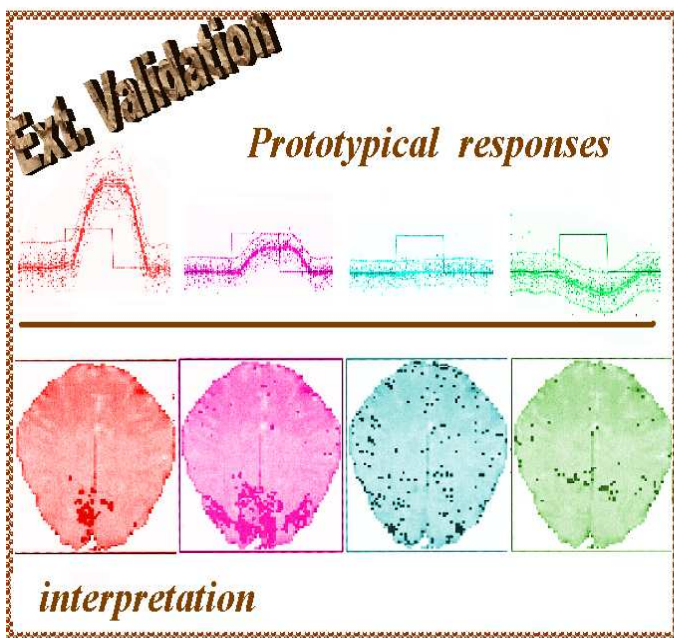
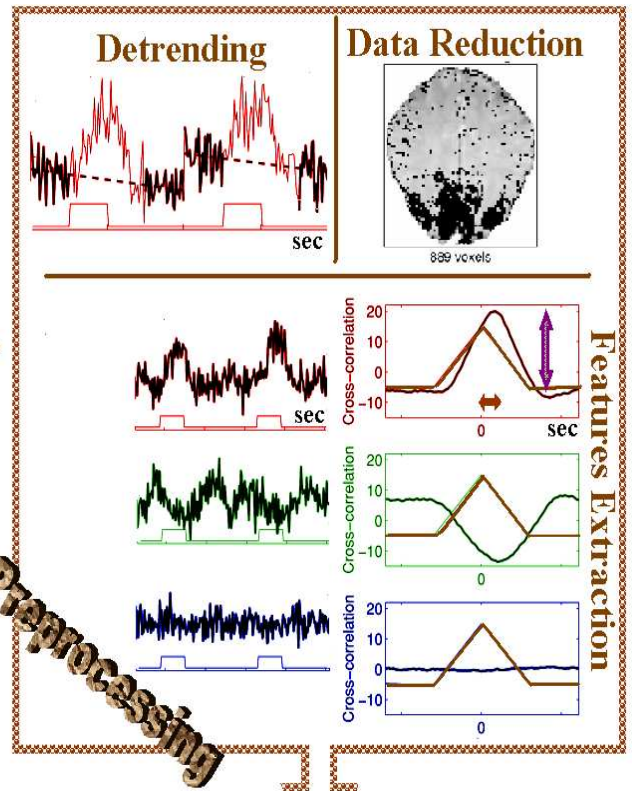
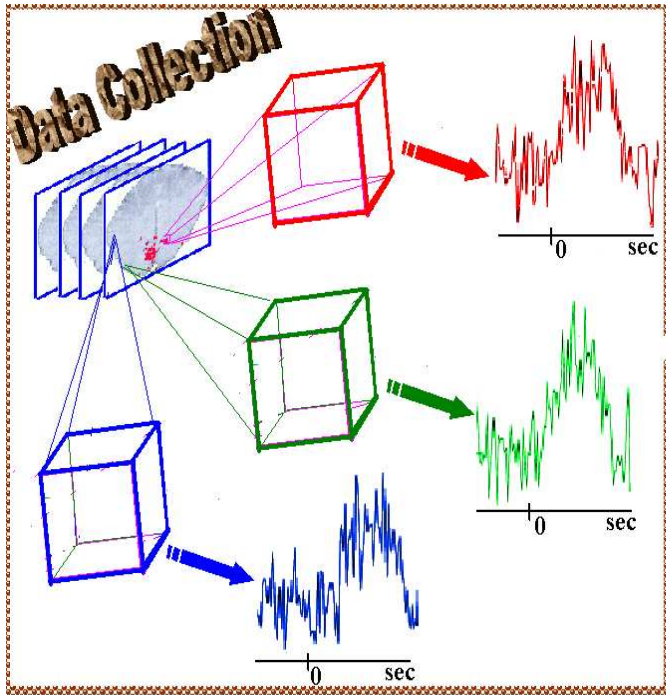
Baumgartner et al. "*Fuzzy clust. vs correlation analysis*" [Magn. Reson Imaging, 1998],

Tonini & Edelman. "*Functional Clustering in Neuroimaging Data*" [Neuroimage, 1998]

Baune et al. "*Dynamical Cluster Analysis of cortical fMRI activation*" [Neuroimage, 2001]

Baumgartner et al. "*Assessing coactivation in fMR via MST*" [Neuroimage, 2001]

OUTLINE

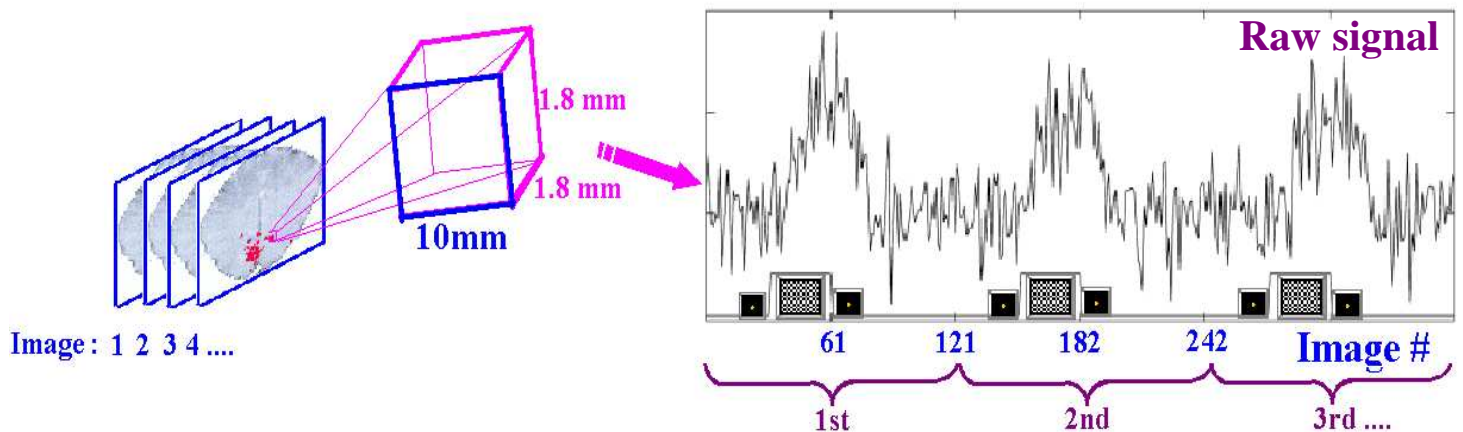
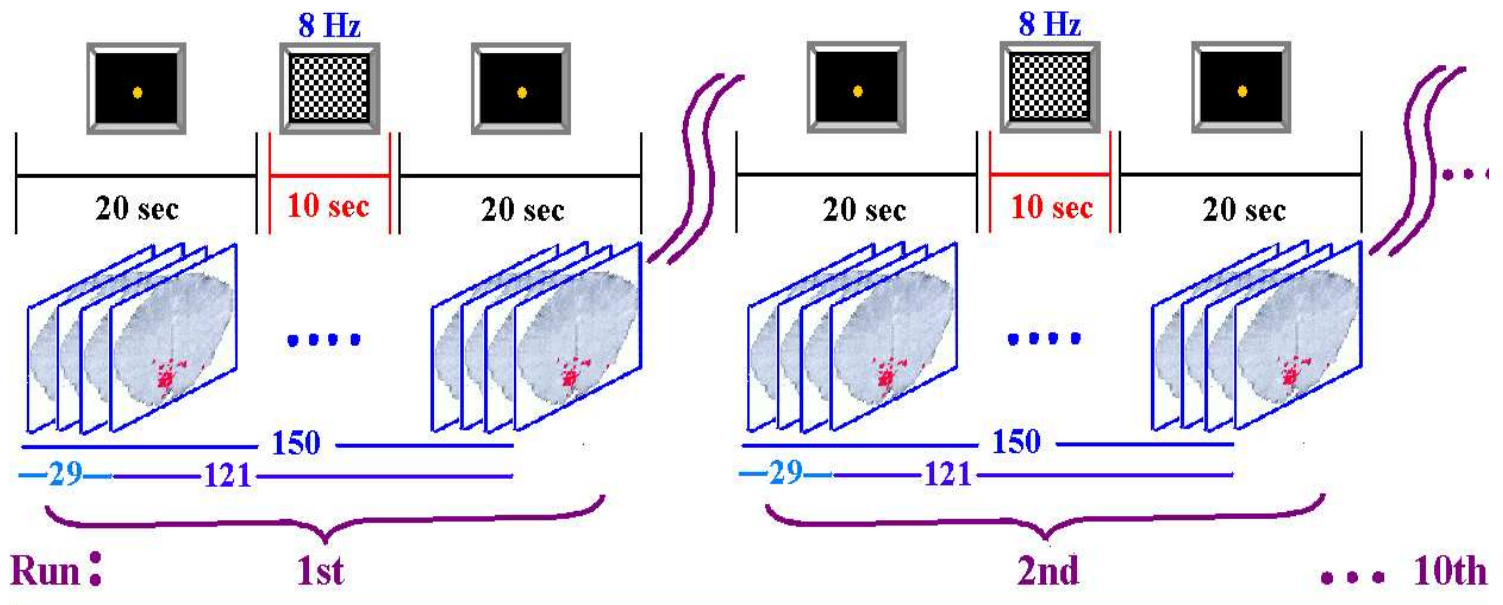


DATA

Images (128 x 128 pels) : FOV of 230 mm, 10mm slice thickness, were acquired in a para-axial orientation parallel to the *calcarine sulcus*.

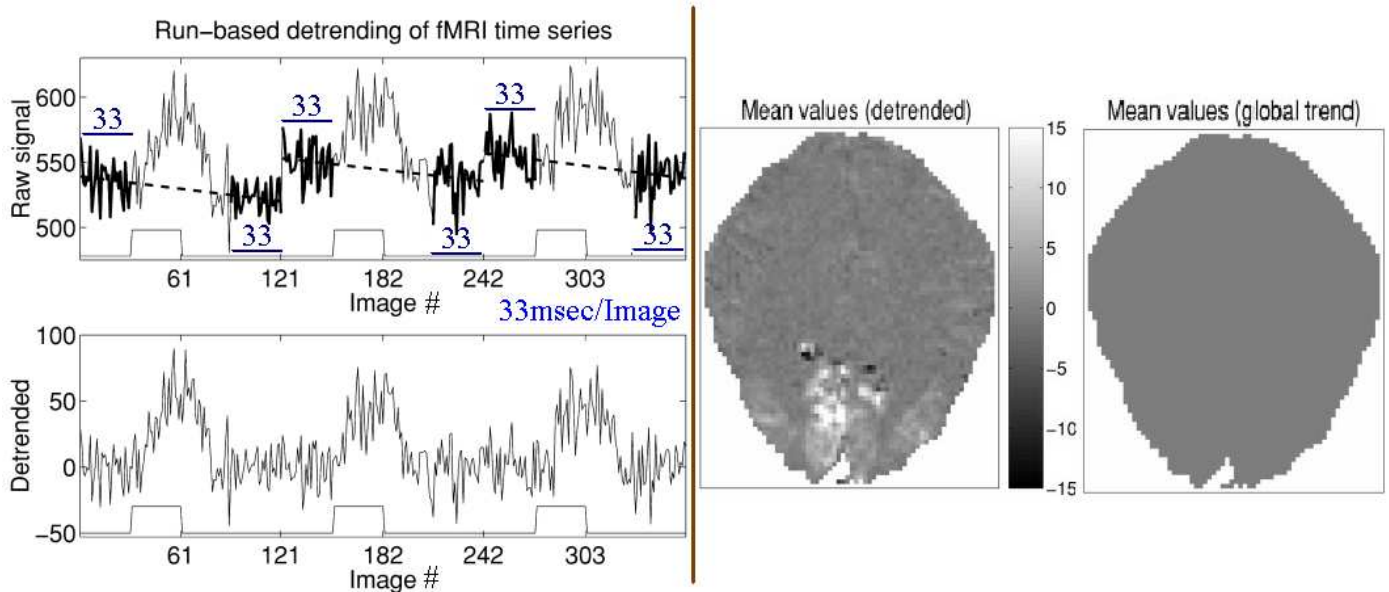
ROI was limited to a 68 x 82 2D voxel map (each voxel: 1.8 x 1.8 x 10 mm)

Visual paradigm: 20 sec of rest period (darkness / a light fixation dot),
10 sec of full-field checkerboard reversing at 8 Hz
20 sec of rest period



PREPROCESSING

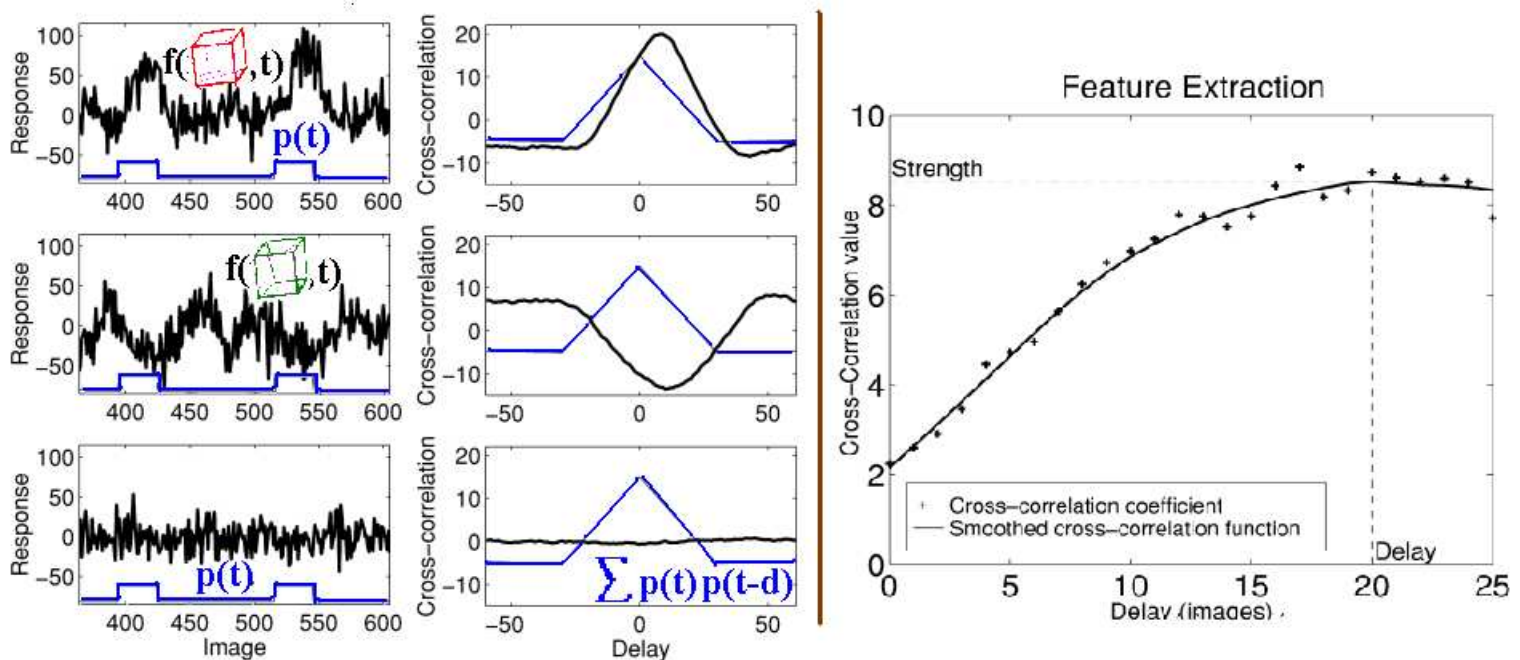
➡ Detrending



➡ Data reduction (*omnibus F test*)



➡ Features extraction from cross-correlation

between the Raw signal $f(x,y,z,t)$ and the paradigm $p(t)$



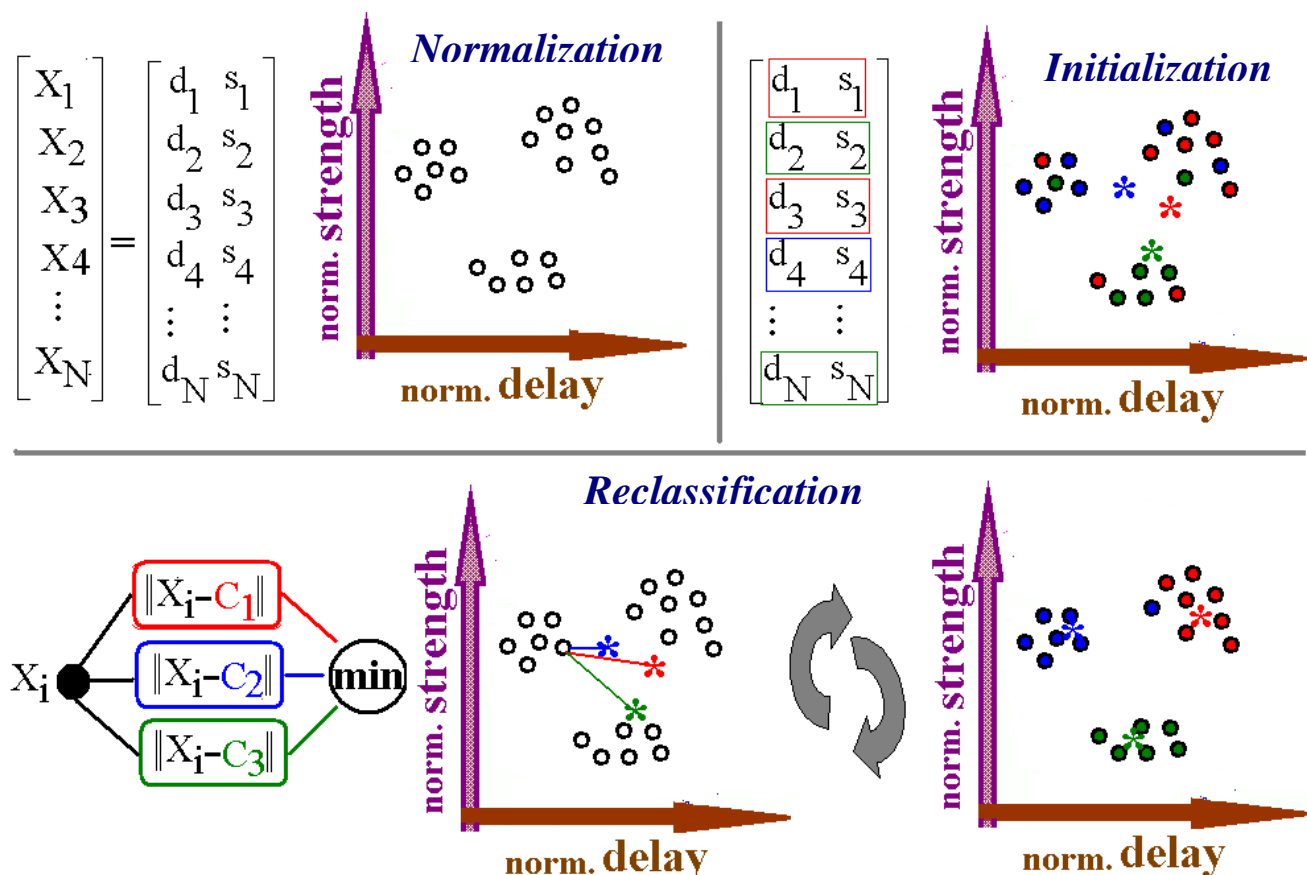
➡ Features extraction for *Metaclustering*

fusion of attributes from standard voxelwise analyses:

- ◆ standard Student t test (rest   VS activation) : - *t statistic*
- ◆ Kolmogorov – Smirnov test : - *d statistic*
- ◆ Correlation with the paradigm (delayed 7 sec) : - *r coefficient*
- ◆ FIR filter model, fitted on the fMRI signal :
 - *std of the fitted signal*
 - *delay estimated from the FIR filter*
- ◆ Gamma filter model, fitted on the fMRI signal :
 - *strength parameter from the fitted signal*
 - *delay estimated from filter*

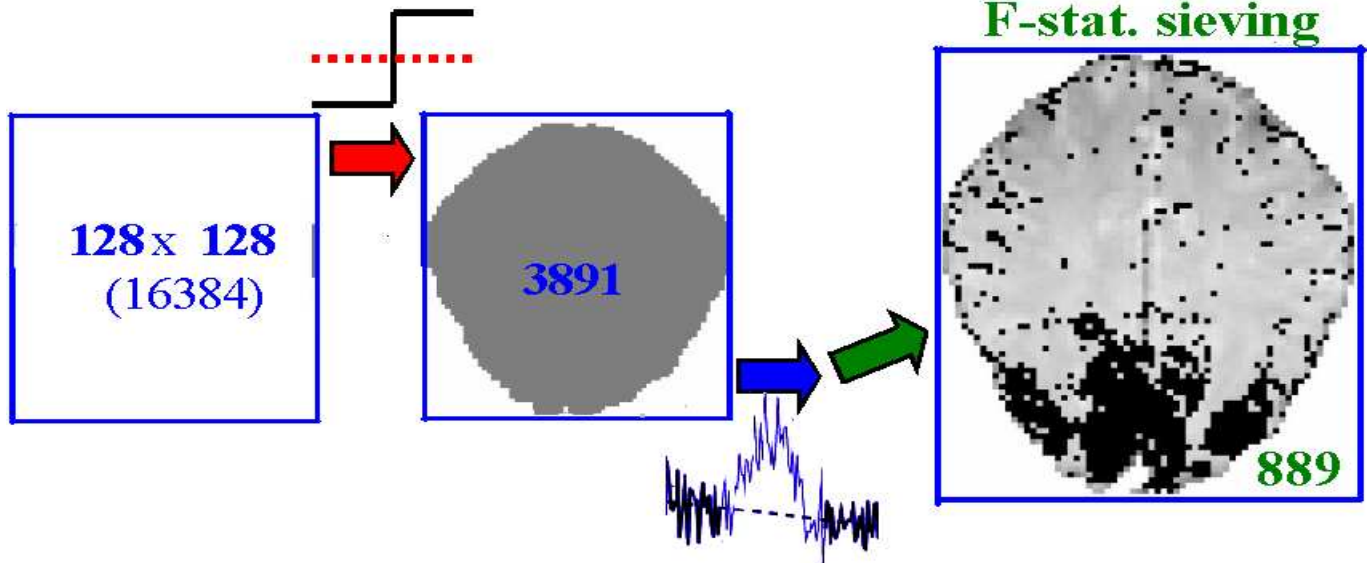
CLUSTERING

➡ *K-means Algorithm*

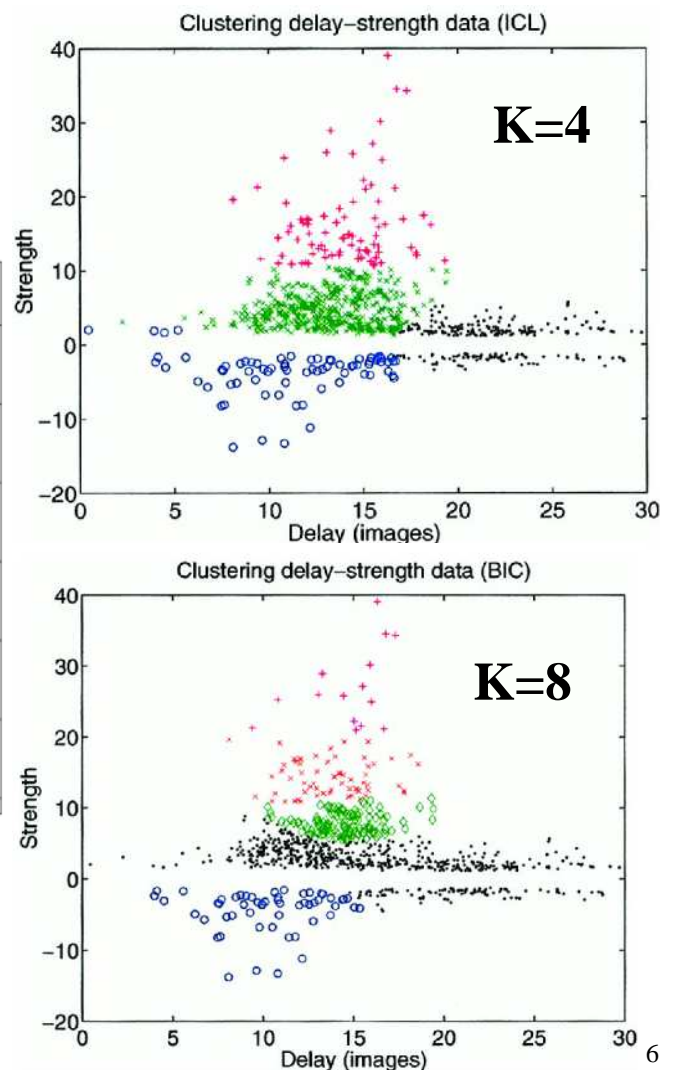
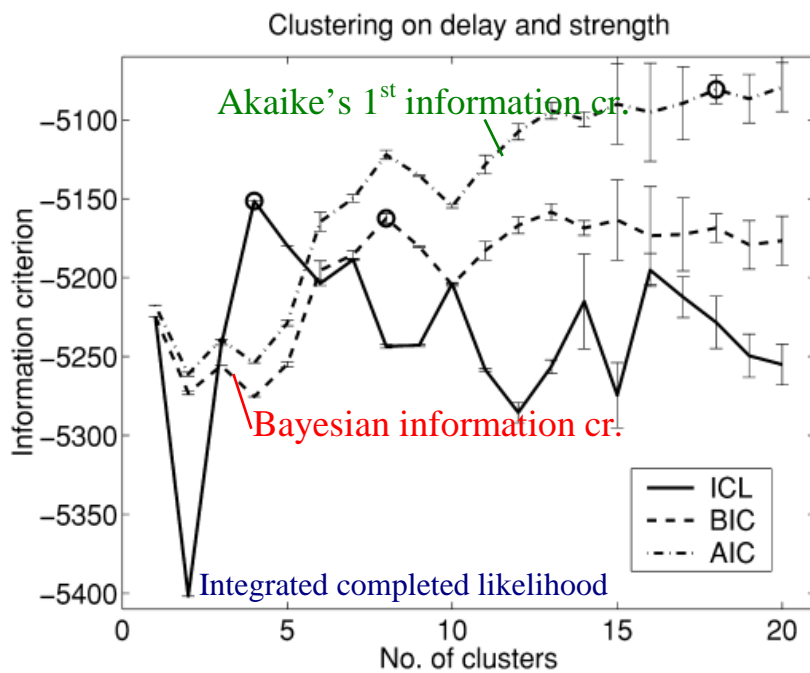


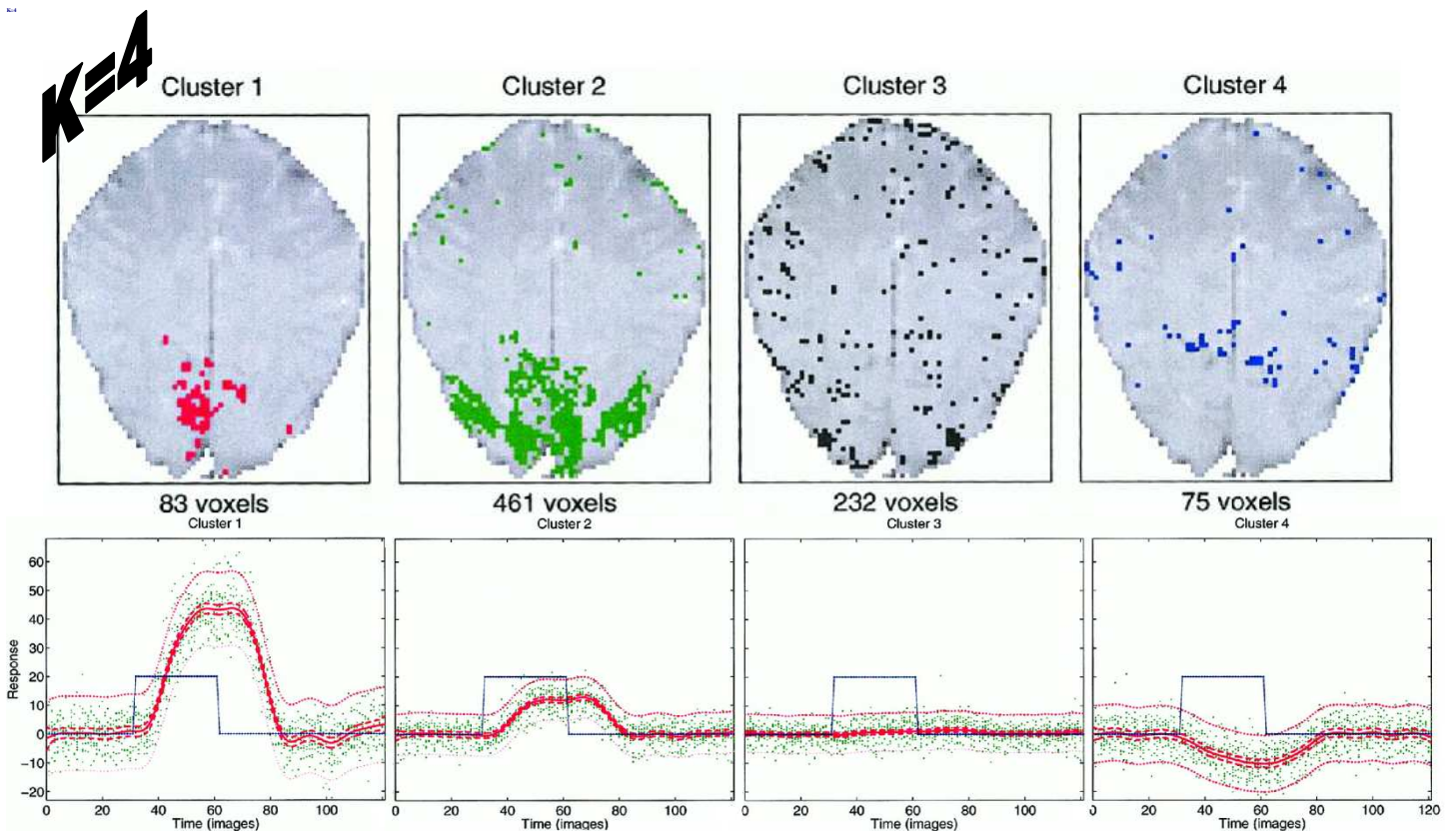
RESULTS

➡ Preprocessing

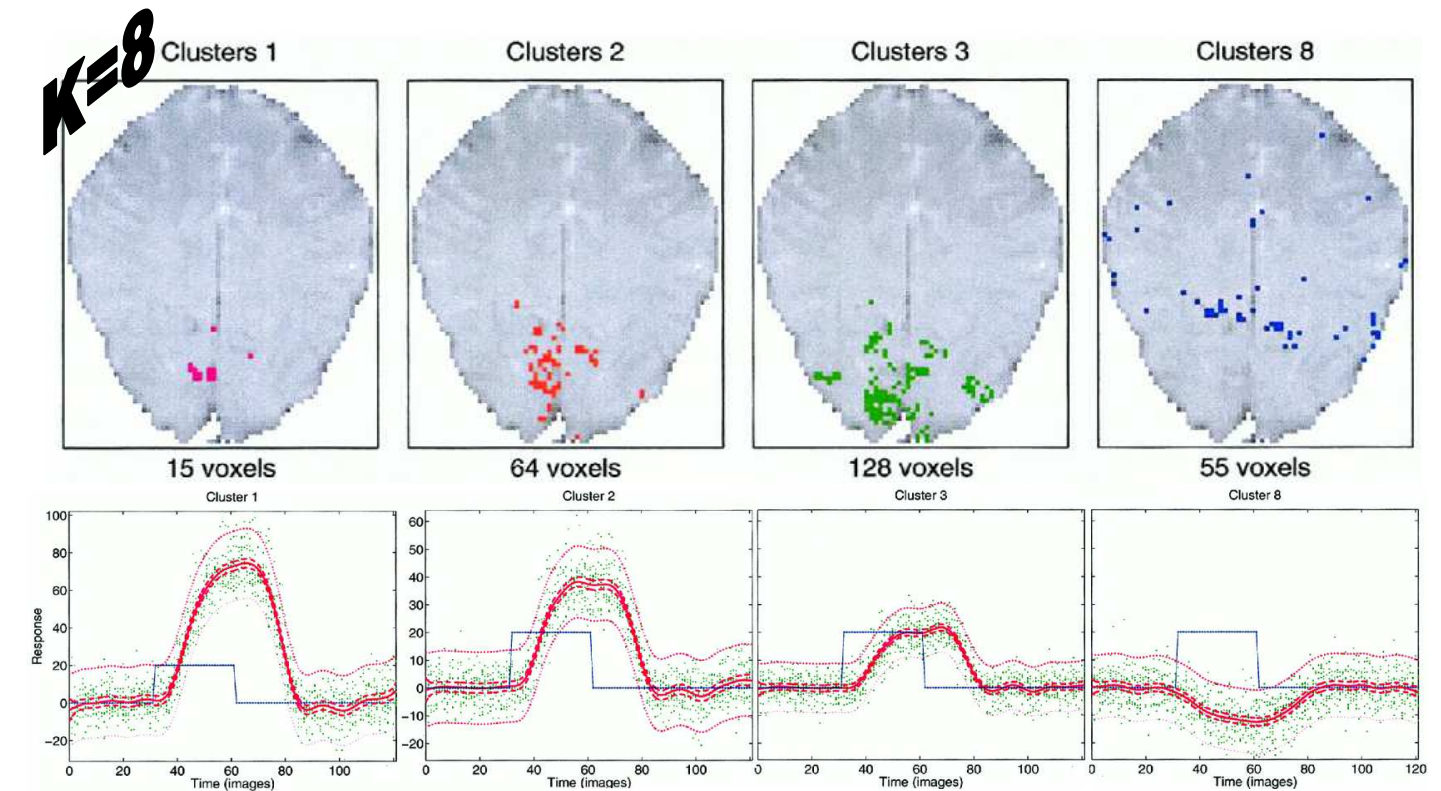


➡ Clustering on delay and strength





- ❶ The data in clusters 1, 2 & 4 display very significant activations of various strength.
- ❷ The delay to 90% of maximum activation is 6.5, 6.2, 6.8 sec
- ❸ There is a significant undershoot in cluster 1 around images 85-105.

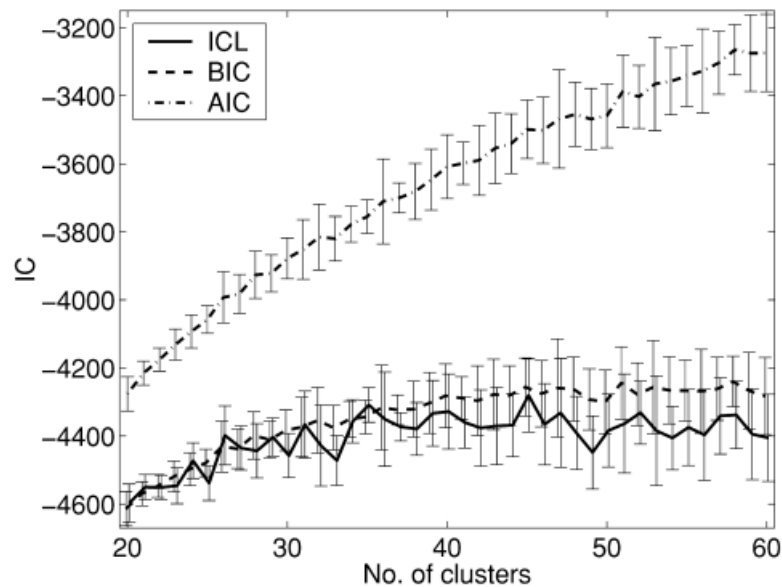


- ❶ 4 out of 8 resulting clusters display some kind of activation.
- ❷ The 3rd cluster shows lateral activation in two smaller areas that correspond to V5
- ❸ The delays for the negative activation are 1-1.5 sec sorter

than the delays for the positive activation.

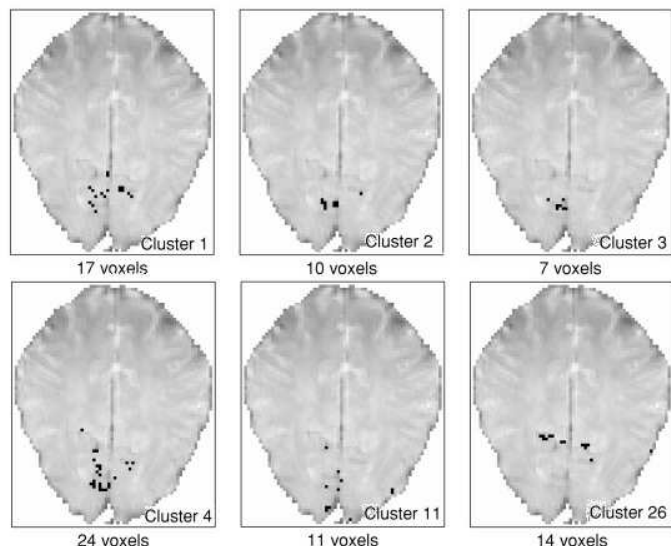
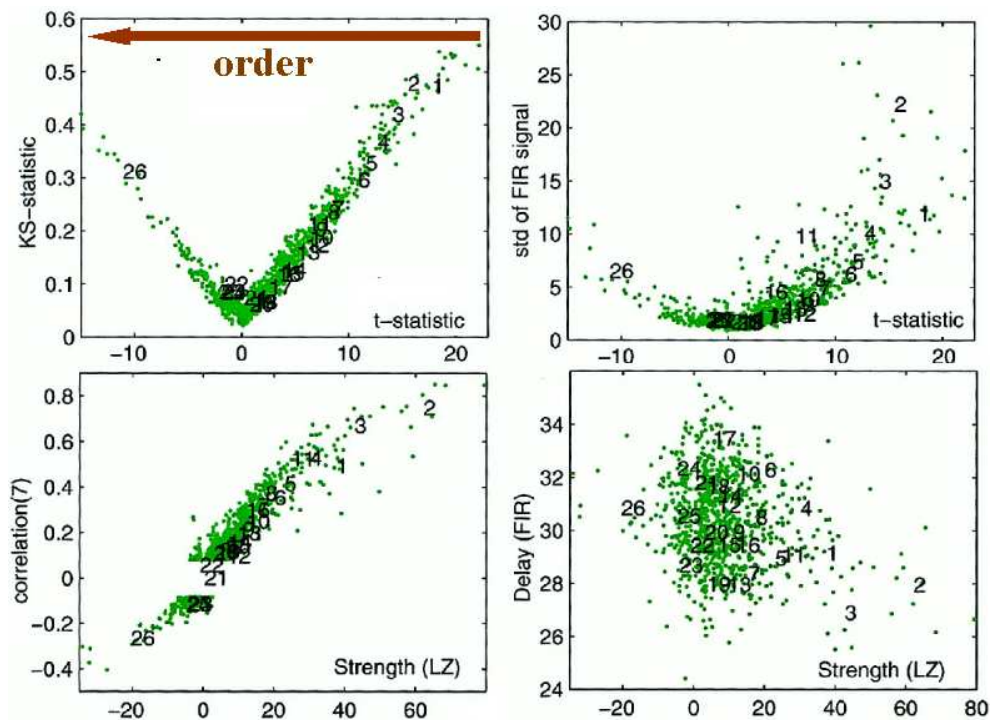
➡ **Meta clustering** in the 7D feature-space

Selected K=26

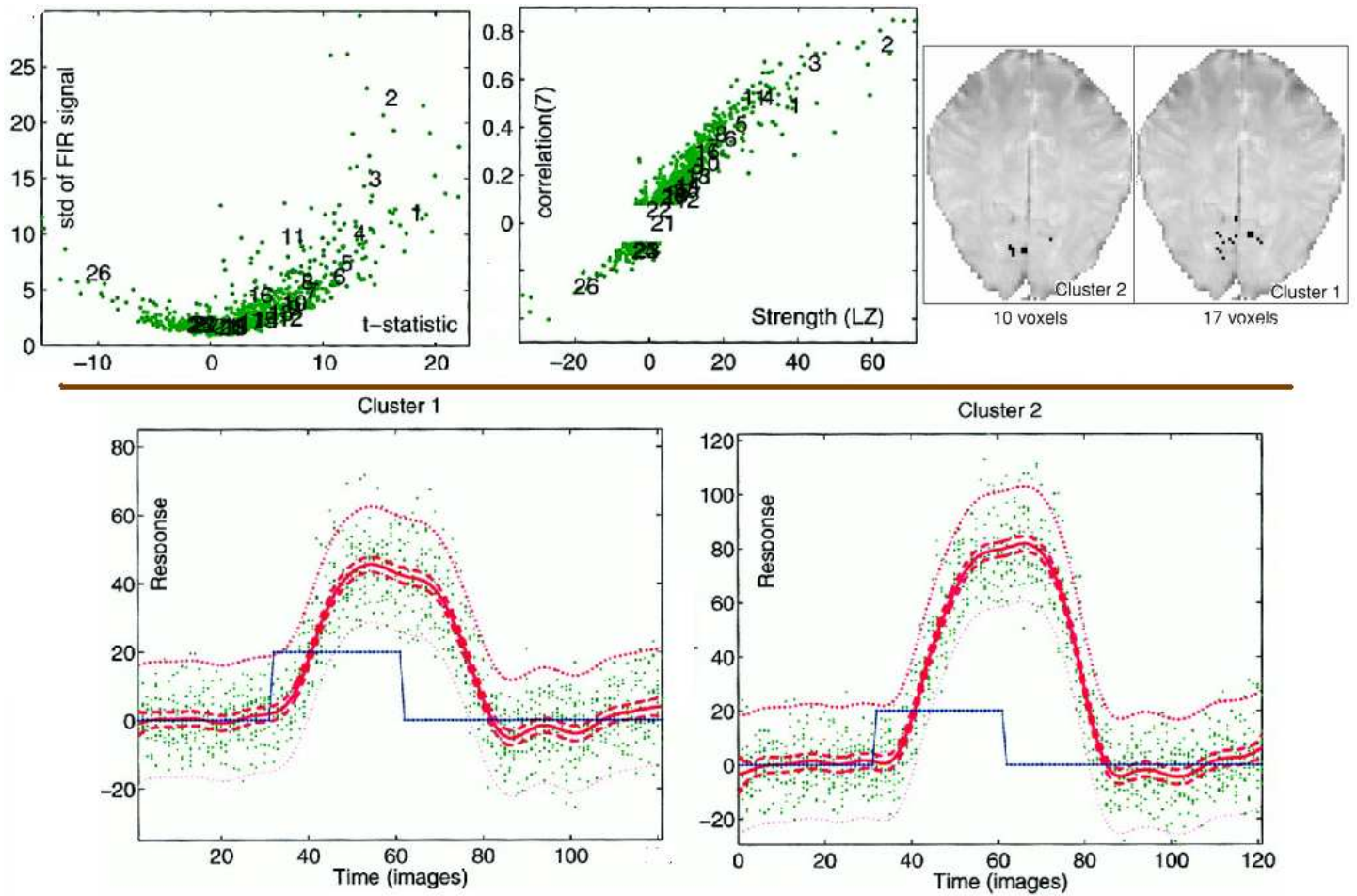


The projection of CA on 2D plots, as a means to compare the different features

The ordering of clusters according to each feature to access the consensus in the features
e.g. FIR-std assigns similar rank to cluster with positive-negative activation (t-stat. based [1] vs [26])

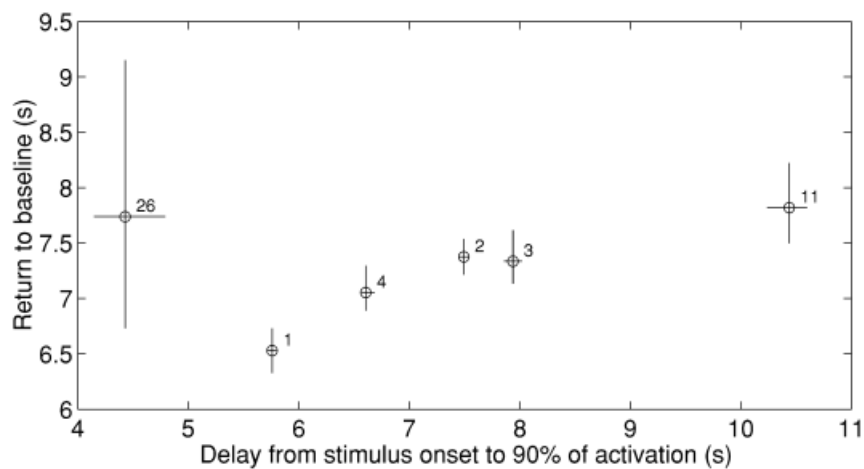


🚗 conventional SPMs criticism



t-stat. & KS-stat. rely heavily on a square wave design.

🕒 Timing of the haemodynamic response.



CONCLUSIONS

- ◆ Existence of voxel-groups characterized by negative pattern of activation.
- ◆ No evidence for “*Initial dip*”
- ◆ Parsimonious description of the Data
- ◆ Complementary information revealed by the plethora of different approaches for voxel-based analysis

Discussion

- ① Theoretical considerations
(metric / convergence / K-means modeling limitations etc.)
- ② Fusion of neuroimaging modalities
Anatomical vs Functional clustering (anat.MRI & fMRI)
- ③ Time-series Clustering
for exploring/testing putative Temporal-Coding schemes

Friedrich & Laurent “*Dynamic Optimization of Odor Representation*”,
Science, vol 291, Feb 2001