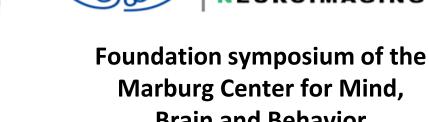
# Comparing multimodal data fusion approaches for simultaneous EEG-fMRI recordings





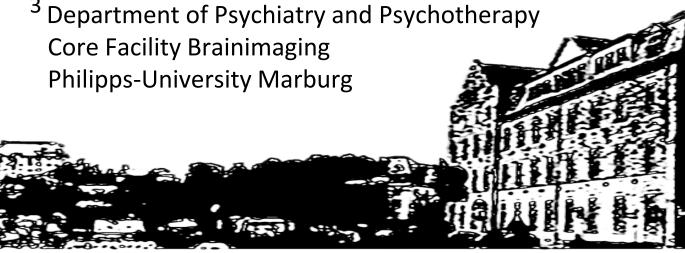


**Brain and Behavior** (MCMBB)

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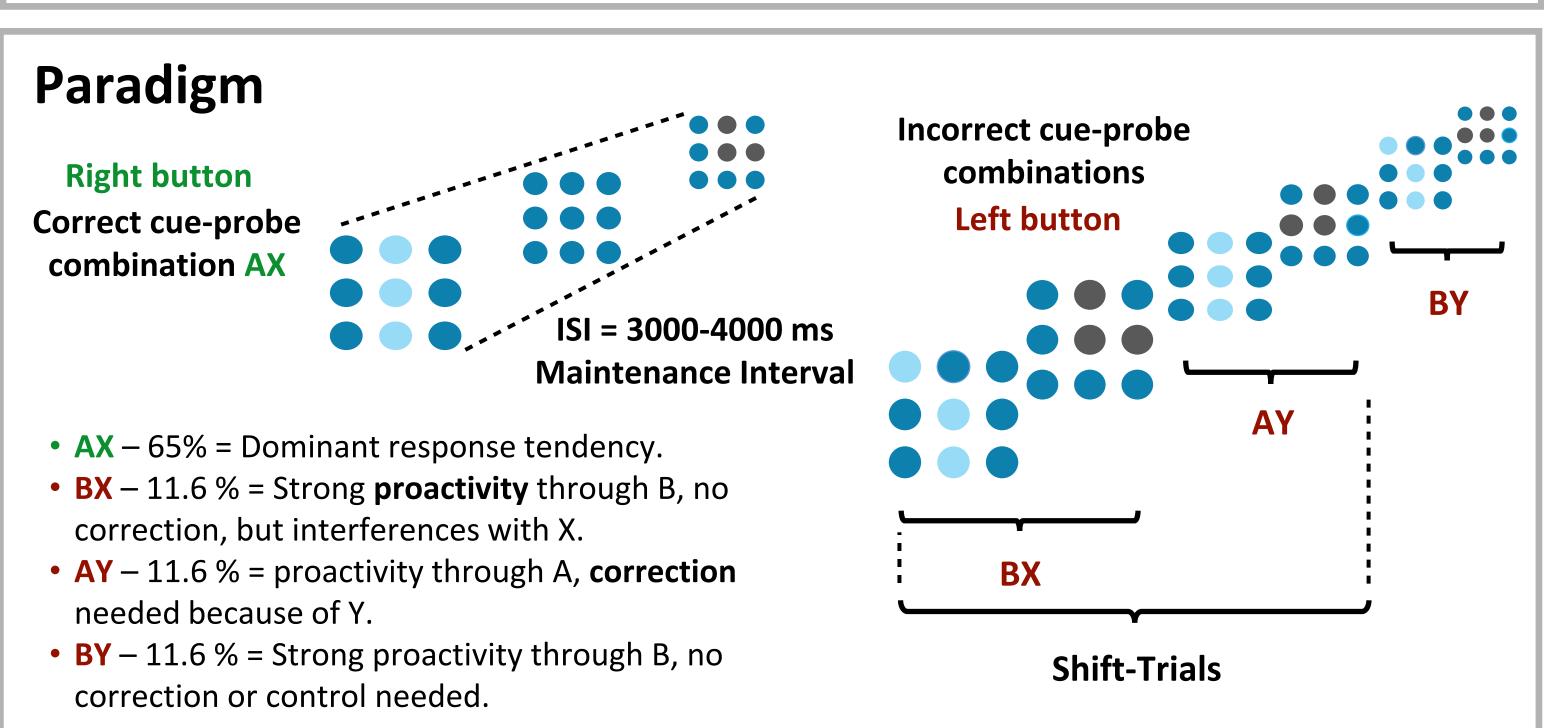
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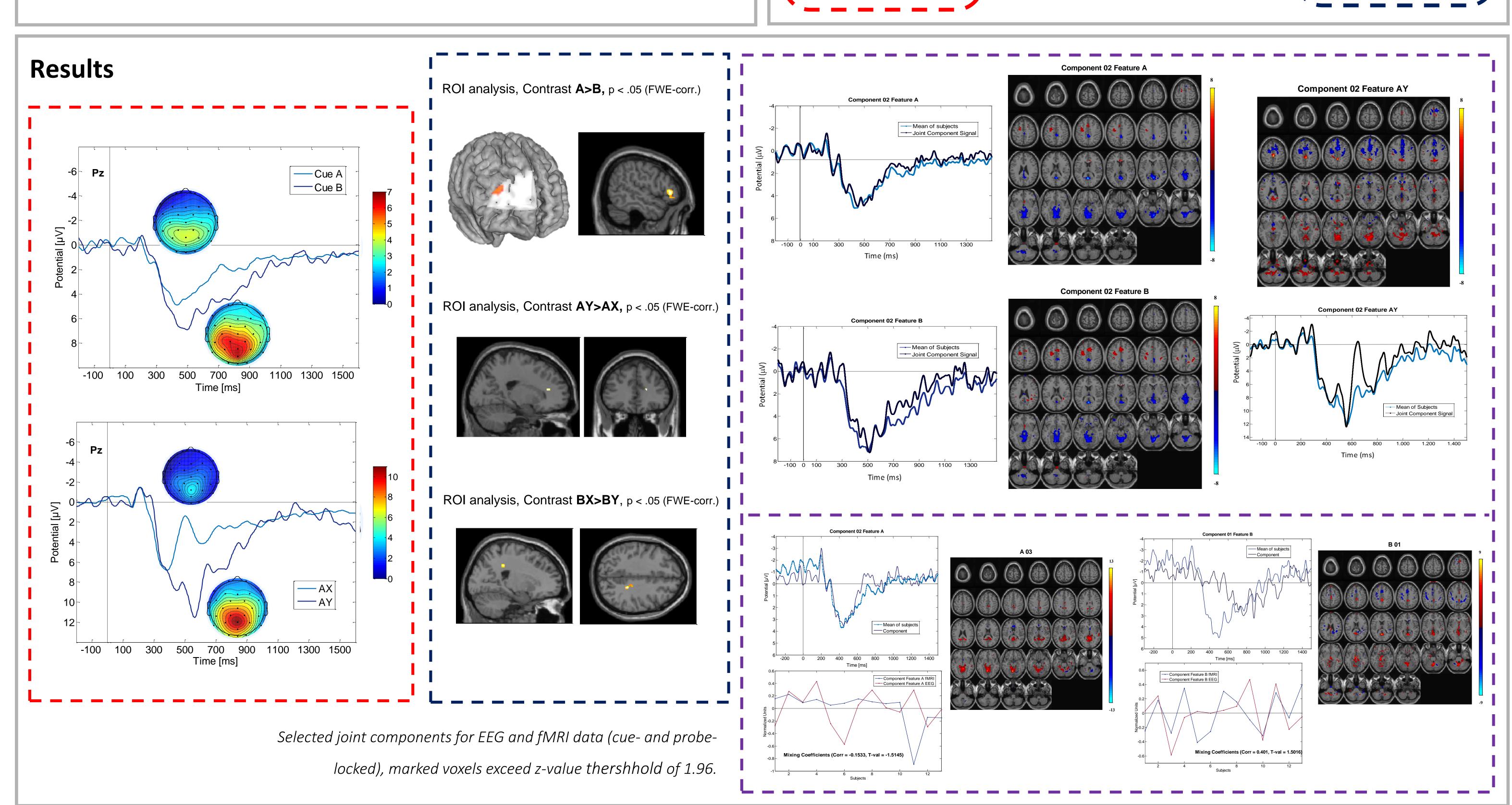
## **Theoretical Background**

The dynamic exertion of cognitive control in accordance to everyday demands is a key factor in executive functioning and goal directed behaviour. Both active maintenance and flexible updating of context information in working memory constitute efficient behavioural control. Therefore, two distinct modes of control can be identified: proactive (i.e. preparatory, context-driven) and reactive control (i.e. corrective, stimulus-driven) (1). Past research points to the dorsolateral prefrontal cortex (DLPFC) to be the core structure for both modes (4). Concerning Event-Related Potentials (ERPs), late fronto-parietal positivity associated with working memory updating and maintenance (i.e. P3b, Late Positive Potentials) should relate to the processing of predictive context cues and increased efforts to integrate new information into behavioural plans.

These findings have been observed in the past, but mostly isolated from one another. The aim of this study was the application of multimodal data fusion approaches to identify joint and distinct signal sources characterizing the different modes of control. These approaches vary in how much information they utilise, whether they are performed on the subject or trial level, which specific measures of fMRI or EEG are entered and in the physiological or statistical assumptions they make. In specific, the capabilities of **blind** and **semi-blind** source separation methods (i.e. joint and parallel Independent Component Analysis) for joint analyses are tested here.



#### **Experimental Setup and Design** fMRI Data (13 Subjects) **EEG Data (13 Subjects) T** timepoints **T** timepoints **X** voxels **C** electrodes M trials **M** trials T\*X by M matrix T\*C by M matrix Realignment **Realignment Parameters** Hemodynamic lag, Realignment temporally unmatched parameter-Hemodynamic informed **Response Function** average artifact **Alignment of temporal** subtraction (5) resolution with HRF **First Level Contrasts ICA** correction of cardioballistic Convolution of single subject EEG **Cues and Probes** artifacts (3) and single subject ERPs **Joint Data Space** [T\*C + T\*X] by M for sample N of both data modalities **Second Level Contrasts Grand Averages Joint and Parallel Independent** cues and probes cue- and probe-locked **Component Analysis** (jICA and pICA): Extraction of shared component weights matrix with complementary components (2)



## Conclusions

- Further evidence for fonto-parietal structures (i.e. DLPFC, rigth middle and inferior frontal gyrus) underlying cognitive control functioning in working memory: 1) More prefrontal and less posterior parietal activity for ambiguous as opposed to predictive cues 2) Higher, sustained positivity (P3b, LPPs) for reactive control correcting behaviour and predictive cues (i.e. goal maintenance) as opposed to early behavioural optimisation and ambigguous cues
- Multimodal Data Integration (i.e. jICA) as a way of examining shared information between features in a joint data space.
- Otherwise lost information in means, isolated contrasts or grand averages can be extracted.
- Means to observe simultaneous variation in data signals when investigating temporal and spatial dynamics of cognitive control functioning.
- Joint and Parallel ICA as tools for multimodal data fusion is limited by:
  - Assumption of exclusively linear assosciations
  - Decomposition is constrained to identical (or correlated) modulation of data sources across subjects
- Implemented as second level analysis, usage of higher order statistics

### Literature

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