

# Effect of cross-modality on the precision of working memory representations and fMRI activity patterns

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## Psychophysics

### INTRODUCTION

- There is an ongoing debate on working memory resources:
  - Does WM contain separate and independent memory storages for visual and auditory information (Baddeley, 2010) or are resources domain-general (Cowan, 2011)
  - Is memory based on discrete slots (Zhang & Luck, 2008) or continuous resources (Bays & Husain, 2008)
- Precision of representations sets the limit for both visual (Bays & Husain, 2008; Salmela & Saarinen, 2013; Wilken & Ma, 2004) and auditory (Kumar et al., 2013) memory performance, suggesting resource sharing
- We tested whether memory resources are shared across representations in different sensory modalities

### METHODS

- Sinewave gratings presented simultaneously with sinewave tones (Fig. 1)
- Visual features: spatial frequency and orientation
- Auditory features: pitch and tone duration
- Memory precision => delayed discrimination threshold
- Two-interval 2-1 staircase => 70.7%
- Each feature separately and all possible feature combinations
- Only the memory load was varied while keeping the stimuli similar across conditions

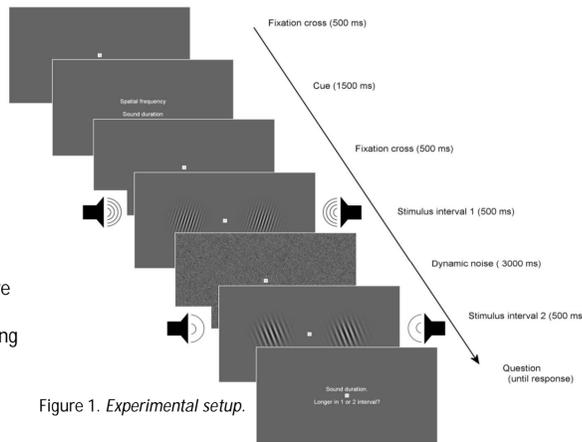
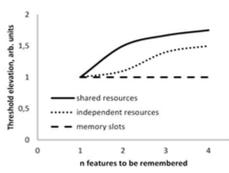


Figure 1. Experimental setup.

### RESULTS - Precision

- Possible outcomes



- Memory precision for a single feature was close to perceptual threshold (Fig. 2)
- Precision collapsed as the number of features to-be-remembered was increased above one (Fig. 2;  $p < .01$ )

- Resource sharing function fits the data well (Table 1):

$$Th = \left(\frac{1}{n}\right)Th_{n=1} + \left(1 - \frac{1}{n}\right)As$$

( $n$  = number of stored features;  $Th_{n=1}$  = threshold with one item;  $As$  = constant asymptote)

Table 1. Constant parameter and  $R^2$  of the fitted functions.

|       | Average | SF         | pitch   | OR       | dur     |
|-------|---------|------------|---------|----------|---------|
| $As$  | 3.02    | 0.62 c/deg | 50.2 Hz | 21.8 deg | 98.7 ms |
| $R^2$ | .98     | .96        | .95     | .98      | .83     |

### RESULTS - Cross-modality

- Possible outcomes

- No cross-modal benefit in any condition.
- Cross-modal thresholds were different than intra-modal thresholds only in 1/8 conditions (orientation & pitch/duration)
- Simultaneously storing one visual and one auditory feature had an equal effect on memory precision as simultaneously storing two visual or two auditory features.

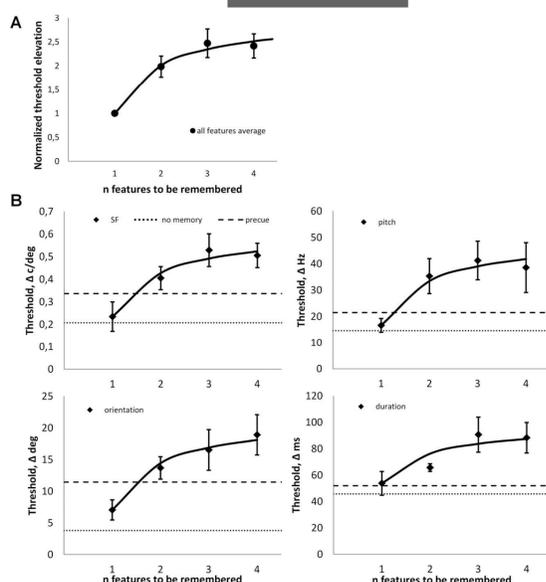


Figure 2. Discrimination threshold as a function of memory load. (precision = 1/threshold)

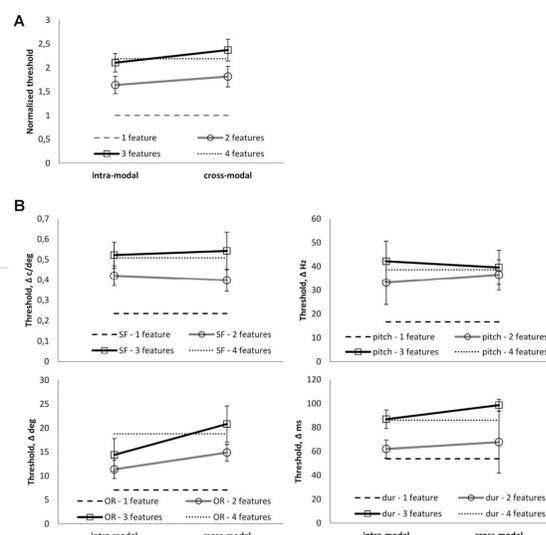


Figure 3. Thresholds in intra-modal and cross-modal conditions.

### CONCLUSIONS

- Working memory performance is limited by the precision of the stored representations, not by discrete memory slots
- Working memory can be described as a common resource pool shared between sensory modalities

## fMRI

### INTRODUCTION

- No sustained activity in the visual cortex during memory maintenance, but activity patterns predict memory performance (Harrison & Tong, 2009)
- Pattern classification accuracy decreases as a function of memory load and the decrease correlates inversely with memory precision (Emrich et al. 2013)
- We tested whether BOLD-responses and activity patterns in sensory cortices differ in intra-modal and cross-modal conditions.
- We measured the effect of attention, dual-tasking & cross-modality on
  - time-course of BOLD signal
  - event related responses to stimulus encoding and memory maintenance
  - multi-voxel activity patterns during encoding and memory maintenance

### METHODS

- Two-interval setup (Fig. 4)
- Similar to psychophysics, but:
  - retention time 7000 ms, no noise
  - auditory features: average pitch and frequency range of tone stream ("musical rain"): 14 tones, 40 ms, breaks 10 ms
  - discrimination task, constant difference
  - 4 diff stimuli in each category
  - five conditions, memorize:
    - pitch (one auditory feature)
    - orientation (one visual feature)
    - pitch and orientation (cross-modal)
    - orientation and spatial freq. (visual intra-modal)
    - pitch and freq. range (auditory intra-modal)

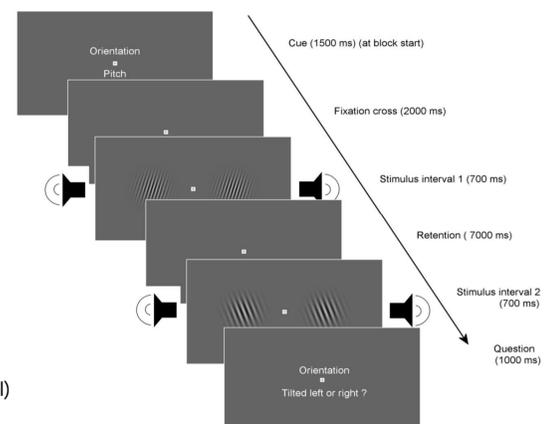


Figure 4. Experimental setup.

### RESULTS

- Behavioral performance, ~75% correct (Fig. 5)

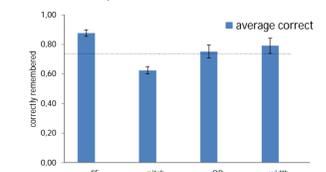


Figure 5. Average of correct answers.

### Voxel-wise analysis

- Functional localizer activated clear clusters in auditory and visual cortex (Fig. 6)
  - visual localizer = contrast-reversing checkerboard
  - auditory localizer = 2 s stream of random tones

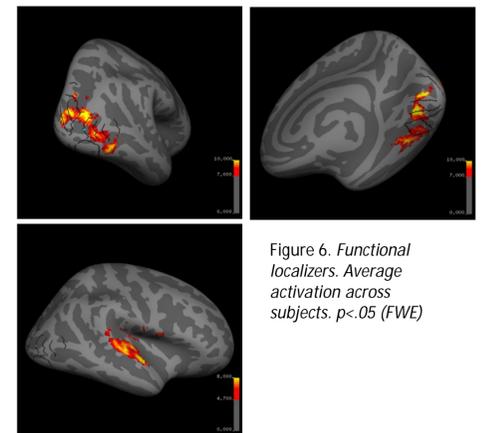


Figure 6. Functional localizers. Average activation across subjects.  $p < .05$  (FWE)

- Time course of BOLD-signal: overall increase due to dual-tasking and attention (Fig. 7)

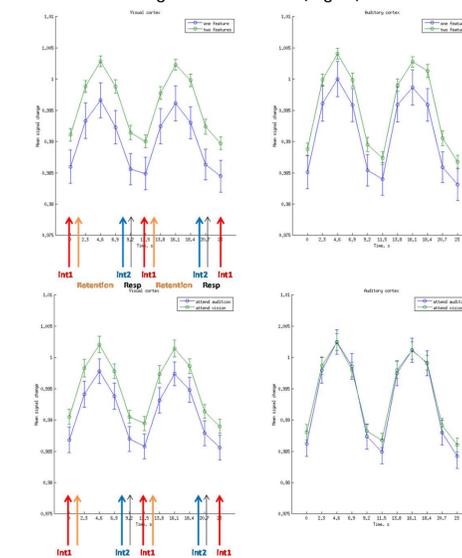


Figure 7. Time-course of BOLD-signal

- No activity during memory maintenance (Fig. 8)
- Responses to stimuli increased due to attention, dual-tasking and cross-modality (Fig. 8)

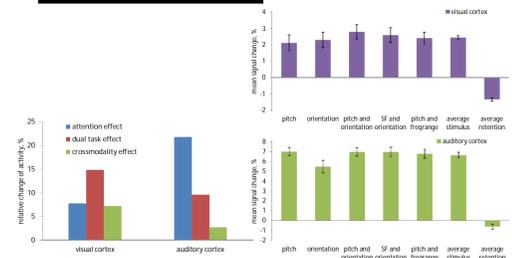


Figure 8. Event-related responses and the effect of attention, dual-tasking and cross-modality.

### Multivariate pattern analysis

- Not yet completed
- Searchlight analysis (circular, radius 3 voxels, volume ~100 voxels) will be conducted to find voxel clusters that contain most relevant information (Kriegeskorte et al., 2006)
- Representational similarity Analysis (Kriegeskorte et al., 2008) will be used to classify different conditions (Fig. 9)
- Within the searchlight, we will calculate correlation of the activity patterns to representational matrices corresponding to: feature type, feature value, modality, task type, and all combinations of these

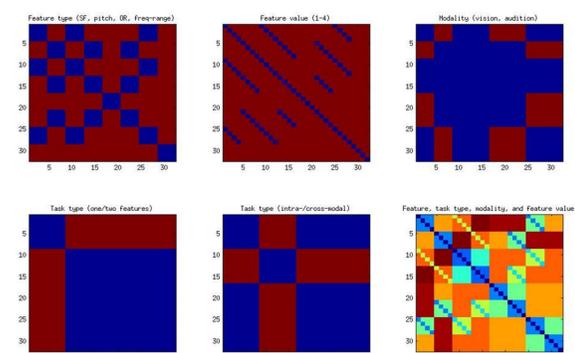


Figure 9. Classification of experimental conditions with representational matrices.

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