

**Title:** Early neural signatures of visual short-term memory

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## **Abstract**

Visual Short-term memory (VSTM) relies on a distributed network including sensory-related, posterior regions of the brain and frontal areas associated with attention and cognitive control. To characterize the fine temporal details of processing within this network, we recorded event-related potentials (ERPs) while human subjects performed a recognition-memory task. The task's difficulty was graded by varying the perceptual similarity between the items held in memory and the probe used to access memory. The evaluation of VSTM's contents against a test stimulus produced clear similarity-dependent differences in ERPs as early as 156 ms after probe onset. Posterior recording sites were the first to reflect the difficulty of the analysis, preceding their frontal counterparts by about 50 ms. Our results suggest an initial feed-forward interaction underlying stimulus-memory comparisons, consistent with the idea that visual areas contribute to temporary storage of visual information for use in ongoing tasks. This study provides a first look into early neural activity underlying the processing of visual information in short-term memory.

## **Introduction**

Visual short-term memory (VSTM), a key aspect of human cognition, is thought to be supported by a distributed network of different brain areas, including parts of the prefrontal cortex (PFC) and posterior areas associated with sensory perception. A wealth of neuroimaging data (Courtney et al., 1997; Haxby et al., 2000; Mitchell and Cusack, 2007; Pessoa et al., 2002; Postle et al., 2003; Todd and Marois, 2004) show that activity in many brain regions correlates with memory demands, either when measured by contrasting a memory task with a purely perceptual task or by varying the amount

of information held in memory. Unfortunately, the limited temporal resolution of functional neuroimaging precludes the identification of the fine temporal details of this network. While VSTM has been studied using electroencephalography (EEG) and magnetoencephalography (MEG) (Agam and Sekuler, 2007; Croize et al., 2004; Morgan et al., 2008; Ruchkin et al., 1990; Ruchkin et al., 1997; Sreenivasan et al., 2007; Vogel and Machizawa, 2004; Vogel et al., 2005), techniques that afford higher temporal resolution, the very early neural events that underlie access to temporarily stored visual information have eluded full description. Knowing the timing and location of the earliest signatures of VSTM will provide clues about mechanisms fundamental to the operation of this important ability.

We had two objectives in carrying out the study reported here. First, we wanted to determine the latency associated with the brain's comparison of some visual input against stored visual memories. Second, we wanted to identify the direction of information flow within the brain; specifically, in which regions of the brain was evidence of the evaluation process first seen? These questions are important for understanding the nature of VSTM, but have been difficult to answer. One particular challenge is distinguishing between memory-related effects and those resulting from intrinsic properties of the stimuli used to probe memory. Verbal stimuli, for example, rely on long-term, language-based representations. Those extra-laboratory associations are difficult to control and may vary substantially in scope and topography between subjects and between memory exemplars. Some visual stimuli, such as pictures of familiar objects, may also lead to the use of verbal labels. Moreover, when objects are presented in the context of some larger scene, recognition is likely influenced by the early extraction of the scene's gist or other context-related information (Biederman et al., 1974). To overcome such complexities in studying VSTM, we used test materials that do not depend on extra-sensory representations and associations (Sekuler and Kahana, 2007). Our stimuli

were compound gratings that differed only in their vertical spatial frequency. These stimuli are advantageous in that they resist encoding with verbal labels, are unlikely to evoke long-term memory representations, and can easily be manipulated along a single dimension. With the burden of the task squarely on the processes of short-term memory, we could obtain a reliable neural index of access to VSTM: We manipulated the similarity between the probe stimulus and the stimuli held in memory, a variable that generates a broad range of recognition-memory performance (Kahana and Sekuler, 2002; Kahana et al., 2007; Sekuler and Kahana, 2008; Zhou et al., 2004). The similarity between memory items is known to be a core component of the neural computations for recognition, regardless of the nature of the items: words, objects, sounds etc. (Sekuler and Kahana, 2008). Therefore, results with our simplified, synthetic stimuli can be generalized to other types of sensory information.

We recorded scalp EEG while subjects performed a Sternberg recognition memory task (Sternberg, 1966). To track the time course of recognition, we compared the ERPs time locked to the stimulus probe under different similarity conditions. Rather than focus on predefined ERP components, we made statistical comparisons among similarity conditions at consecutive time points relative to probe onset. This allowed us to identify the time of the earliest differences. This basic method has been used before to study the speed of visual processing (Rousselet et al., 2007; Thorpe et al., 1996), but not in the context of short-term memory. In other studies, the subject had to judge a picture against a fixed criterion (“animal?”). Here, in contrast, subjects had to update the content of memory on every trial. This allowed us to search for the earliest signature of access to working memory, here operationalized as the earliest effect of probe-item similarity.

## **Materials and Methods**

*Subjects:* Subjects were twelve volunteers (all right-handed, 8 female, ages 20-34, who had normal or corrected-to-normal Snellen acuity, and normal contrast sensitivity as measured with Pelli-Robson charts (Pelli and Robson, 1988). All subjects provided written informed consent to a protocol approved by Brandeis University's Committee for the Protection of Human Subjects.

*Stimuli:* Stimuli were compound gratings generated by adding vertical and horizontal sinusoidal luminance gratings of equal contrast (Kahana and Sekuler, 2002). Gratings were generated and displayed using Matlab (The Mathworks, Natick, MA) and extensions from the psychophysics and video toolboxes (Brainard, 1997; Pelli, 1997). Stimuli were centered on a 20-inch cathode ray tube display with 800x600 pixels screen resolution and 75 Hz refresh rate. Each stimulus' mean luminance was matched to the display's constant, background luminance, 60 cd/m<sup>2</sup>. As a result, the onset or offset of any stimulus produced no potentially distracting luminance transient. Stimuli subtended 5 degrees of visual angle, and were viewed from a distance of 57 cm from the monitor.

Prior to each subject's first experimental session, an adaptive psychophysical procedure (Wetherill and Levitt, 1965) determined the subject's discrimination threshold for spatial frequency. The threshold was defined as the value at which the subject could discriminate successively presented gratings 79% of the time. Expressed as a Weber fraction, the subjects' mean threshold value was 0.13, with a standard error of 0.06. To take into account differences in the accuracy with which subjects encoded grating's spatial frequency, the stimuli used to test each subject's recognition memory were scaled according to his or her own threshold.

During memory testing, subjects were presented sequentially with two study items ( $S_1$  and  $S_2$ ) on

each trial, followed by a probe *PB*. All three stimuli shared a common horizontal frequency, making any differences among  $S_1$ ,  $S_2$  and *PB* dependent solely on their vertical frequencies. Both horizontal and vertical frequencies varied randomly across trials. Horizontal frequencies were randomly drawn from a uniform distribution between 1.0 and 7.0 cycles per degree. The lowest possible vertical frequency was randomly drawn from a uniform distribution between 1.0 and 2.0 cycles per degree, but the actual range of possible vertical frequencies consisted of discrete increments from that minimum, using logarithmic frequency steps based on the individual subject's discrimination threshold, expressed in units of "just noticeable difference" (JND). In addition, a random phase shift was applied to the vertical frequency of each stimulus. This ensured that subjects could not successfully base memory judgments on residual retinotopic information.

The study stimuli,  $S_1$  and  $S_2$ , always differed from one another by four JNDs. On half of the trials  $S_1$ 's spatial frequency was higher than  $S_2$ 's; on half the trials, the reverse was true. Additionally, *Target* trials, on which *PB* matched one of the study items, and *Lure* trials, on which it matched neither study item, were equally probable. On half the *Target* trials, *PB* matched  $S_1$ ; on the other half, it matched  $S_2$ . On *Lure* trials, *PB* spatial frequency was drawn from a discrete uniform distribution whose spatial frequencies spanned a range centered on the geometric mean of the two study items. Specifically, this range for possible *PB* frequencies went from four JNDs below the spatial frequency of the lower frequency study item to four JNDs above the spatial frequency of the higher frequency study item. These equally-probable, candidate values for *PB* spatial frequency on *Lure* trials were  $\pm 6$ ,  $\pm 4$ , and 0 threshold units relative to the geometric mean of that trial's  $S_1$  and  $S_2$  frequencies. Expressed in these same coordinates, the study items' spatial frequencies, as well as *PB* frequency on *Target* trials, were  $\pm 2$  threshold units. For *Lure* trials, probe frequencies relative to the study items were equally probable and varied randomly.

*Procedure:* Fig. 1A summarizes the behavioral paradigm. A trial began with a fixation cross at the center of the computer display. The fixation cross was succeeded in turn by  $S_1$ ,  $S_2$ , and  $PB$ , each lasting 750 ms. An inter-stimulus interval (ISI) of  $275 \pm 75$  ms separated fixation,  $S_1$  and  $S_2$ ; a  $500 \pm 75$  ms retention interval separated  $S_2$  and probe. By pressing designated “Yes” and “No” keys subjects indicated their judgment of whether  $PB$  matched one of the study items, or failed to match either of them. An inter-trial interval (ITI) of 3 seconds separated successive trials. Subjects were instructed to refrain from blinking except during the ITI. In addition, every 20 trials subjects received feedback on their accuracy and response times. Each subject participated in four sessions of 320 trials each, during which both behavioral performance and EEG signals were recorded.

*EEG recording:* We recorded from 129 electrode sites at 250 Hz using an Electrical Geodesics (Eugene, OR) system. All channels were referenced to the vertex during data acquisition, and were adjusted so that scalp impedances were below 100 kOhm. Data were cleaned of bad channels (chosen by visual inspection), re-referenced to the grand average and reduced to a montage comprising 27 standard electrode locations, two vertical, bipolar channels above and below each eye, and one horizontal, bipolar eye channel, using BESA (MEGIS Software GmbH, Munich). Data were then baseline corrected (100 ms prestimulus), averaged and analyzed using MATLAB (The Mathworks, Natick, MA). Blink artifacts were eliminated by rejecting epochs in which the difference between the maximum and minimum voltage at any EOG channel exceeded  $150 \mu\text{V}$ . This resulted in the removal, on average, of 5% (SD: 4%) of each subject’s trials. Data were high-pass filtered at 0.1 Hz, low-pass filtered at 30 Hz, and notch-filtered at 60 Hz.

*Statistical Analysis:* To examine the divergence between ERPs for different probe-study item

relationships, we performed a repeated-measures ANOVA (with probe frequency, relative to the study items, as the main factor) at each time sample. Our goal was to identify consecutive periods of time during which the ERPs consistently differed from each other. The appropriate significance thresholds ( $P$ -value and number of consecutive samples) were determined by a random permutation procedure: We created 1,000 surrogate sets of data by randomly assigning trials to one of the three probe frequency groups, while preserving the temporal and spatial correlations between electrodes (Supplementary Fig. 1A). This provided an empirical estimate of the probability of a Type I statistical error.

Supplementary Fig. 1B shows a flowchart of the random permutation procedure. For this statistical method, the two variables that need to be determined are  $PV$ , the  $P$ -value threshold for each individual comparison, and  $N$ , the number of consecutive samples below  $PV$  required to produce  $F$ , the false discovery rate (FDR). Either  $PV$  or  $N$  can be chosen by the experimenter, and the other variable is then determined empirically. For this study, we decided to control  $PV$  and searched for a matching  $N$ . For each of the 1,000 surrogate datasets, we performed the same statistical test (ANOVA) for each electrode and time point, as if we were dealing with the real data. For the data presented here, we assigned  $PV$  to be 0.05 and set our target  $F$  to 1%. By choosing a sufficiently low FDR, we ensured the reliability of the differences seen in the real (unshuffled) data: At 1%, we can expect no more than 0.27 out of 27 electrodes, on average, to produce false positives. The number of consecutive samples,  $N$ , required to produce 1% FDR turned out to be 26, which corresponded to 104 ms of data for which  $P < 0.05$ .

## Results

As expected, response time (Fig. 1B) and the probability of a “Yes” response (Fig. 1C) both increased as a function of the similarity between the probe and the study items (Kahana et al., 2007; Zhou et al., 2004). In the case of *Lure* trials, this increase reflected the difficulty in detecting a memory-stimulus mismatch. Because probe frequency strongly influenced task difficulty, differences between ERPs for the three *Lure* conditions provide an early indication of a comparison between the probe and memory items. As both the behavioral and the ERP data (see Supplementary Figs. 2 and 3) were highly symmetrical around the spatial frequency midway between the two study items (the geometric mean of their respective frequencies), we used three aggregated sets of *Lure* trials, sorted according to their similarity to the memory items. This aggregation produced three sets of relative probe frequencies: *Lure* trials with the probe falling either  $\pm 6$  JNDs,  $\pm 4$  JNDs, or 0 JNDs from the mid-point.

For each time sample (4 ms), the corresponding data were subjected to a repeated-measures ANOVA; we searched for time periods over which consecutive samples showed significant differences between ERPs (Giard and Peronnet, 1999; Rieger et al., 2005; Rousselet et al., 2007; Rugg et al., 1995; Shams et al., 2001; Thorpe et al., 1996). Unlike previous implementations of this statistical method, we used random permutations of the data to find the appropriate number of consecutive samples required to declare significance (see Methods). We found significant differences ( $P < 0.05$  for at least 104 ms) in 19 out of 27 regions of interest, the earliest of which appeared at 156 ms. Fig. 2A shows the ERP traces and the corresponding  $P$ -values for each sample. As any choice of  $PV$  is an arbitrary one, we wanted to make sure that our results hold true with choices of  $PV$  other than the one we used, namely 0.05. We therefore repeated this analysis with  $PV = 0.01$  and  $PV = 0.005$ , for which the values of  $N$  were 12 (48 ms) and 9 (36 ms), respectively. For  $PV = 0.01$ , 24 electrodes met the significance criterion, and the earliest difference onset was at 160 ms. For  $PV =$

0.005, 22 electrodes reached significance, with the first difference showing at 160 ms. Note that the global probability of a Type I error is the same for any choice of  $PV$ , because that probability depends on  $F$ , the target FDR (see Methods).

We were interested not only in how early stored memories can be evaluated against visual inputs, but also in the spatial distribution of the effects. Specifically, we sought to the flow of activation — from posterior visual areas to frontal areas, or vice versa — during the comparison process. As we shall discuss below, this has important theoretical implication. Fig. 2B shows topographical distributions of the  $P$ -values obtained by the aforementioned ANOVA tests (see also Supplementary Fig. 4 for distributions mapped with higher temporal resolution). Note the initial, largely posterior modulation at 175 ms, which precedes frontal modulation by about 50 ms. Fig. 2C shows distributions of onset times for ERP differences at posterior and anterior locations. Indeed, the distributions suggest an earlier posterior effect. To further quantify the difference between anterior and posterior activity, we combined all the anterior and all the posterior locations into two representative ERPs. As shown in Fig. 2D, the posterior ERPs show significant differences earlier than their anterior counterparts; The combined posterior ERPs took 8 ms less than anterior ERPs to reach  $P < 0.05$ , 28 ms less to reach  $P < 0.005$ , and 56 ms less to reach  $P < 0.0005$ .

We evaluated some alternative, non-memory related explanations for our results. First, the differences between the ERPs might have reflected differences in the physical properties of the probe (i.e., lower or higher absolute vertical frequency), rather than its relation to the memory items. This proposal is unlikely because two of the three *Lure* conditions were a combination of two probe frequencies, one lower and one higher than the mid-point of the frequency range. In fact, ERPs for the probes at the lower spatial frequencies and those at the higher frequencies show similar trends

(Supplementary Figs. 2 and 3). Second, it is conceivable that the ERP differences were due to adaptation of neural populations following the study items' presentation (Kimura et al., 2006): More similar probes presumably recruit overlapping neural representations (responsible for encoding spatial frequency) and therefore are subject to repetition suppression, the reduced response to repeated stimuli (Grill-Spector et al., 2006; Henson et al., 2004). To study the degree of adaptation, we analyzed separately the probes with lower spatial frequency (-4 and -6 JNDs) and those with higher frequencies (+4 and +6 JNDs). Significant adaptation would predict weaker ERPs when the probe is closer in frequency to the second study item than when it is closer to the first, as less time has been available to recover from suppression. We compared ERPs for each group of probes under two conditions: the first study item having a lower or higher frequency than the second item. If adaptation occurs, it should lead to attenuated ERPs for the lower frequency probes when the second item is lower in frequency, and vice versa for the higher frequency probes. However, as can be seen in Supplementary Figs. 5 and 6, this is not the case. In fact, not a single location showed a reliable frequency-order effect. It therefore seems that the early differences in the ERPs indeed depend on retrieval from VSTM, rather than on lower-level features or on adaptation.

## **Discussion**

Our results suggest that information in short-term memory exerts a very early influence on the evaluation of a subsequent visual test stimulus. In fact, such evaluation adds no substantial delay to the speed of processing when compared to other ERP studies whose paradigms impose no demand on short-term memory. Those studies (Rousselet et al., 2007; Thorpe et al., 1996) used a categorization task, in which each trial's test stimulus was compared to the same stable criterion that could be held in memory over many trials (e.g., an "animal" template in long-term memory). As subjects saw and

categorized just a single stimulus on each trial, some preparatory memory processing was likely done prior to stimulus presentation in order to establish the categories that would be judged. In contrast, our task required that the contents of memory and the decision criterion had to be updated on every trial and held in VSTM until the probe presentation. In addition, it has been suggested that the differences seen in those studies at around the same latency arise from differences in the physical features of the stimuli (Johnson and Olshausen, 2003); Our design overcomes this confound.

It is by now well accepted that prefrontal brain regions and posterior, “sensory” cortex both contribute to short-term memory. Visual selectivity has been observed not only in posterior regions of the visual system, but also in PFC (Everling et al., 2006; Funahashi et al., 1990). Similarly, both regions show delay-period activity during working memory tasks (Freedman et al., 2003; Funahashi et al., 1990; Haxby et al., 2000; Pessoa et al., 2002; Postle et al., 2003; Rainer et al., 1998; Zaksas and Pasternak, 2006), suggesting they are both important for temporary maintenance of information. The role of each component in this distributed network, however, is unsettled. The results of the present study imply that the comparison process in VSTM involves a significant role for early flow of activation in a posterior-anterior direction. A few caveats, however, should be noted: First, 156 ms is long enough to allow feedback connections from prefrontal cortex to more posterior areas (Foxe and Simpson, 2002); For example, ERP experiments have shown that attention modulated visual processing as early as 100 ms poststimulus (Hillyard et al., 1998). Second, ERP measurements provide only coarse information about about the localization of activity in specific brain areas, which makes it difficult to know the sources of our ERPs with precision. Moreover, a single dipole may have generated both posterior and anterior differences, in which case the earlier posterior difference could be explained by higher sensitivity in those electrodes. More precise spatial identification, then, would likely require invasive recordings. With these caveats in mind, we suggest that the earlier

posterior differences, as well as the early onset of the ERP differences, support the hypothesis that the actual storage of mnemonic visual information relies on brain areas that are also used for visual perception (Postle, 2006). In this view, the role of PFC may be to distribute attention between the representations of the remembered items and the representations activated by the presence of the visual stimulus.

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## Figure Legends

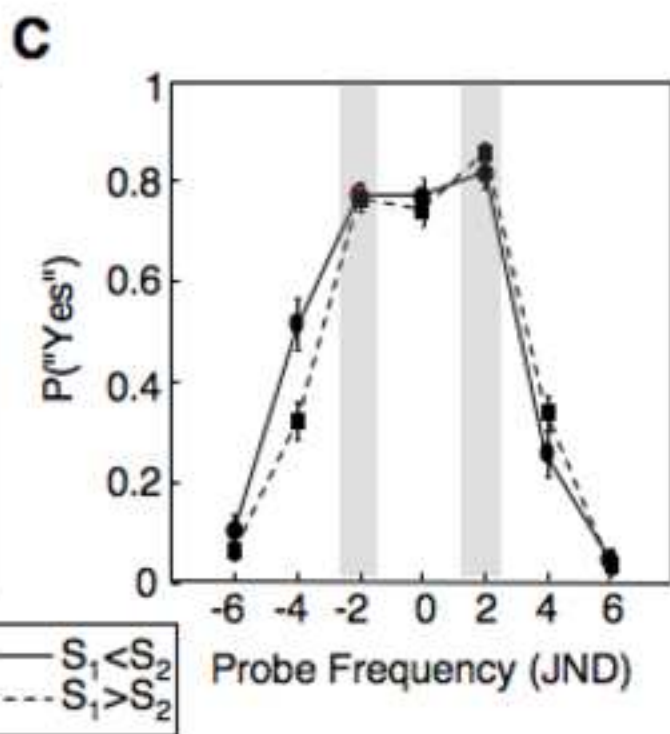
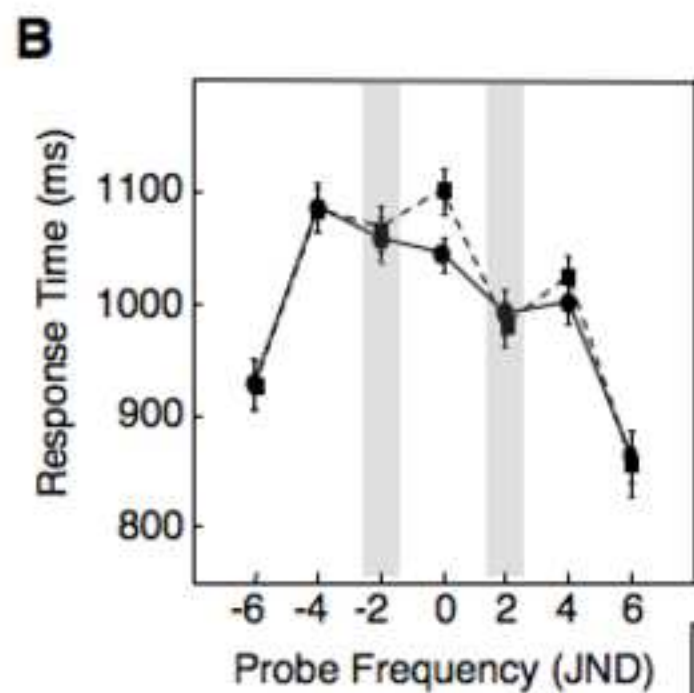
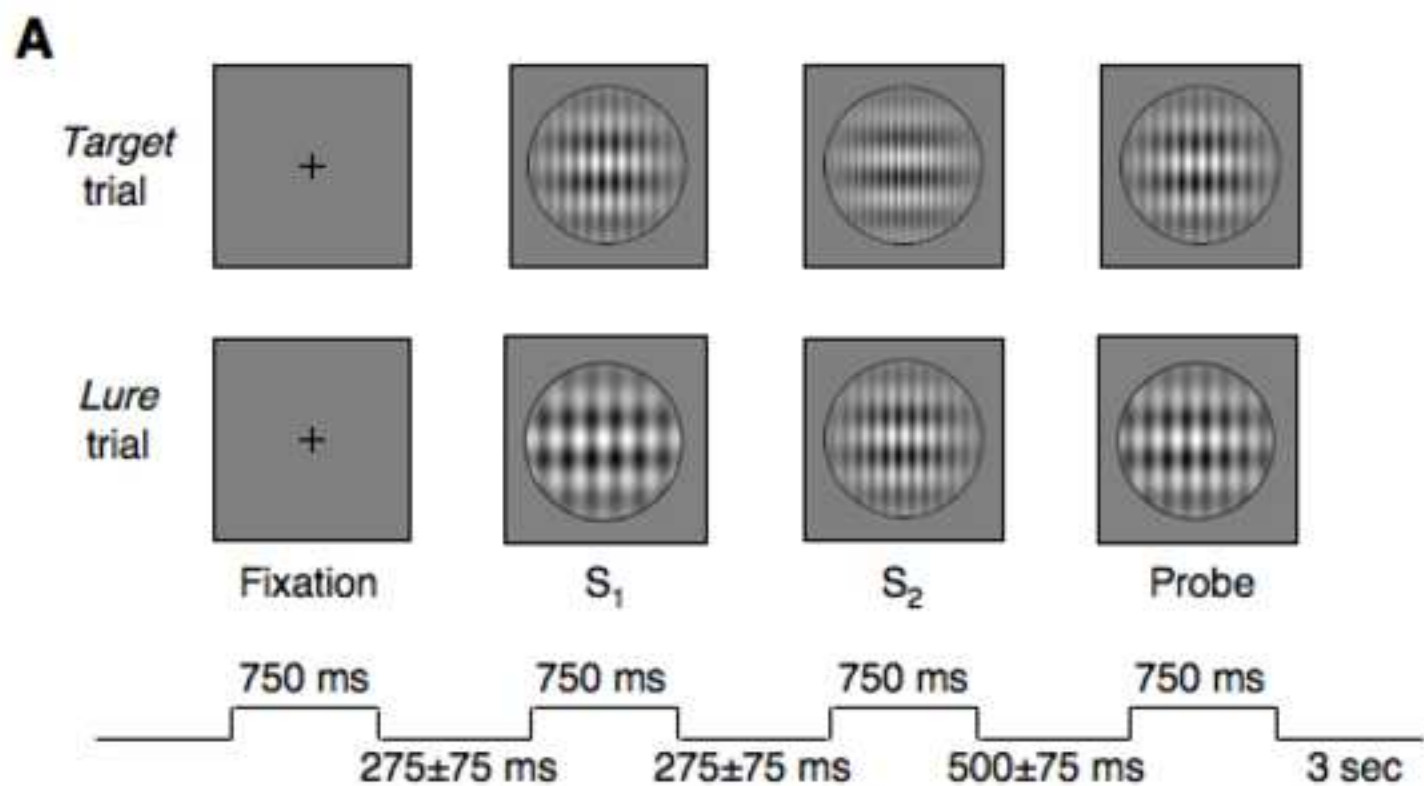
**Figure 1:** Experimental paradigm and behavioral results. A: Schematic diagram of a trial. Top example is for a *Target* trial, in which the probe matches one of the study items, and the bottom one is for a *Lure* trial, in which the probe is different from both study items. B: Response time as a function of the probe spatial frequency with respect to the study items. Solid and dashed lines indicate trials of the opposite frequency order. C: The probability of a “Yes” response plotted against probe frequency. Note the high symmetry of the results around the central frequency (the geometric mean of the study items' frequencies). Vertical gray bars indicate the study items' frequencies. Error bars are within-subject SEM (Loftus and Masson, 1994).

**Figure 2:** ERP results. A: ERP traces for each of 27 interpolated recording sites (anterior locations at the top). Each color corresponds to a different probe frequency relative to the study items'

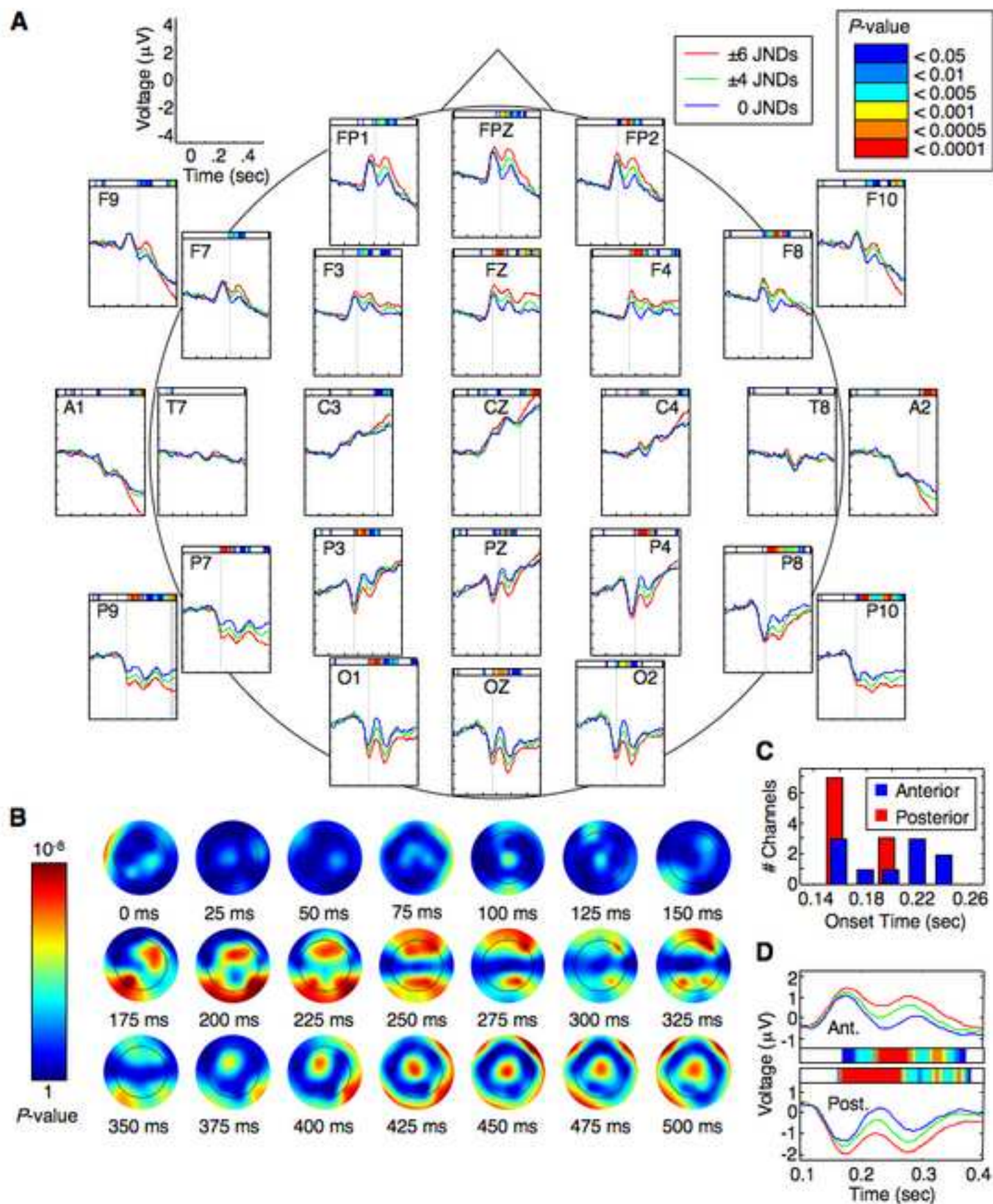
frequencies (aggregated across positive and negative distances from the mean). Time 0 is the onset of the probe stimulus. Vertical dashed lines denote the time of the first significant difference between the ERPs (at  $P < 0.05$ ), for locations that met the significance criterion. Colored stripes above each panel indicate the result of the ANOVA at every time point.  $P$ -values above 0.05 are not included. Number of trials for each condition (red, green, blue; mean $\pm$ SD): 242 $\pm$ 11.38, 244.2 $\pm$ 9.9, 121.8 $\pm$ 6.2.

B: Topographical maps of the  $P$ -values at time points separated by 25 ms. Red regions indicate strong modulation of the ERPs by probe frequency, and blue regions indicate little or no modulation.

C: Onset times of ERP differences. The histogram shows the time at which ERPs started to diverge ( $P < 0.05$ ) at ten posterior (red) and ten anterior (blue) locations. D: ERP traces for aggregated sets of anterior and posterior electrodes. Colored stripes represent significance values (top right color scale); note the earlier significance for the posterior ERPs.



5. Figure  
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**6. Supplementary Material**

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