Early and Late Components of EEG Delay Activity Correlate Differently with Scene Working Memory Performance

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Abstract

Sustained and elevated activity during the working memory delay period has long been considered the primary neural correlate for maintaining information over short time intervals. This idea has recently been reinterpreted in light of findings generated from multiple neural recording modalities and levels of analysis. To further investigate the sustained or transient nature of activity, the temporal-spectral evolution (TSE) of delay period activity was examined in humans with high density EEG during performance of a Sternberg working memory paradigm with a relatively long six second delay and with novel scenes as stimuli. Sensor level analyses revealed transient rather than sustained activity during delay periods. Specifically, high amplitude activity encompassing the theta range was found early, during the first three seconds of the delay and correlated positively with subsequent ability to distinguish new from old probe scenes. Later, toward the end of the delay, lower amplitude activity encompassing the alpha/beta range was found that negatively correlated with subsequent ability to distinguish probes. Source level signal estimation implicated a right parietal region of transient early delay activity that correlated positively with working memory ability. Importantly, varying the trial temporal window duration and changing the baseline period for computation of TSE activity were found to influence the strength of the correlations obtained. This pattern of results adds to recent evidence that transient rather than sustained delay period activity supports visual working memory performance. The findings are discussed in relation to synchronous and desynchronous intra- and inter-regional neural transmission, and choosing an optimal baseline for expressing temporal-spectral delay activity change.

Keywords: encoding; maintenance; parietal cortex; theta; alpha; beta; synchrony; oscillations

Introduction

The ability to maintain information 'online' for a matter of seconds and then to use this

information to realize immediate goals is critical for what is known as working memory [1].

While the behavioral limits of working memory, including capacity and duration, have been

extensively documented [2, 3], it remains to be fully understood the neural dynamics that allow

stimuli to be held in mind for short periods of time [3, 4].

Early work in non-human primates suggested [5] that it was the sustained and elevated

firing of neurons in prefrontal and parietal cortices during the entire delay period since encoding

but before a memory test that was important. More recent work challenges the long-standing

view that elevated and sustained activity is necessary to support working memory [6].

Furthermore, work in monkeys suggests that working memory is supported by discrete and brief

gamma and beta bursting [7]. Recently, combined functional MRI (fMRI) and transcranial

magnetic stimulation (TMS) in humans shows evidence that the active representation of an item

in working memory can drop to baseline when attention shifts away, and a targeted pulse of

TMS can cause the item representation to re-emerge [8]. This latter finding also challenges the

sustained and elevated firing view, and suggests that the working memory neural representation

is dynamic, modifiable through cognitive control, and can be maintained by so-called 'activity

silent' mechanisms that do not necessarily involve sustained neuronal firing [9, 10].

Studies utilizing electroencephalography (EEG) in humans add to the complex picture

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regarding the sustained nature of oscillations during working memory delay periods [11-14].

Using scalp EEG, sustained synchronous alpha rhythms in parietal and occipital regions have

been observed during the delay period of trials where the subject subsequently makes a

successful probe choice [15, 16]. A more recent study points to beta activity, in addition to alpha activity, as playing a role in successful maintenance [4]. Oscillations in the theta range have been observed during the delay period as well [17]. The relationship of these electrical oscillations to widely used indirect measures of neural activity (e.g., BOLD-fMRI) remains a topic of ongoing investigation. Using intracranial EEG, which affords both high spatial and temporal resolution, delay period gamma band oscillation amplitude has been found to increase on electrodes located near fMRI activity, while theta band activity was elevated for electrodes located away from fMRI activity [18]. Synchronous frontal theta activity is reportedly also associated with active maintenance of information. Frontal theta activity is implicated in maintaining temporal information of encountered stimuli, whereas remembering item-specific details is associated with synchronous alpha and beta activity in parietal and occipital regions [19, 20].

A review of the literature on neural recording in non-human primates and on fMRI and EEG in humans shows that it remains to be completely characterized how neural activity evolves during the delay period of a working memory task. The long-standing elevated and sustained neural firing account is challenged by more recent studies that suggest transient dynamics. Transient in this context means that activity could rise or fall at different times during the delay period. Activity could also be heterogeneous across frequency bands, with different bands exhibiting different relationships to memory performance. In the present study, the main questions addressed are: How does scalp EEG activity measured during the delay period of a working memory task change as both a function of time and frequency? Furthermore, how do the changes in delay activity predict subsequent short-term memory performance?

In the present study, participants completed several trials of a Sternberg task, a classic paradigm to assay working memory that when coupled with neural recording allows for a disambiguation of activity related to encoding, maintenance (during the delay period), and a probe of recognition memory for stimuli encountered before the delay. Here, the Sternberg paradigm was modified to use complex naturalistic scenes as stimuli. The use of complex scenes as visual stimuli served the dual purpose of having a large number of stimuli that are novel, and also to minimize the chance that participants could rely exclusively on a strategy of recruiting their phonological loop through covert articulatory rehearsal. The first and main objective was to analyze scalp EEG collected during the performance of the task in order to quantify delay period activity as a function of both time and frequency. The second objective was to understand how patterns of delay activity predict subsequent working memory performance by performing a correlation of the time-frequency delay period activity patterns with subjects' ability to distinguish a probe stimulus presented immediately after the delay period.

To evaluate systematically the hypothesis of transient rather than sustained delay activity supporting working memory, a temporal-spectral evolution (TSE) analysis of EEG data was conducted with different baseline and temporal windows. TSE allows for quantification of event-related enhancement or suppression of EEG oscillation amplitude or power [21, 22]. In the present study, the event is the onset of the delay period and TSE is computed using different time windows relative to onset of the delay. Additionally, TSE is expressed as change relative to different baselines. For example, the entire delay period could serve as its own baseline and then each temporal unit of activity starting from the delay could be expressed relative to the mean activity taken across the delay. Alternatively, a period of time before or after the delay period could serve as the baseline, but then the resultant activity pattern would need to be interpreted

relative to the sensory and cognitive processes occurring during each of these different baselines.

The lack of systematic exploration in the literature motivates the final objective here of studying

how changing temporal windows and baselines affects temporal-spectral measures of EEG delay

activity, and the correlations between delay activity measures and memory performance.

Materials and Methods

Subjects

A total of 20 participants were recruited by flyers posted throughout the City College of New

York campus. Each participant provided written informed consent and completed the study

procedures according to a protocol approved by the Institutional Review Board of the City

College of New York. Participants were compensated \$15 per hour for participation. Five

participants were excluded due to excessive noise in their EEG signals, leaving a total of 15

participants (11 males, mean age 24.13 years, SD = 6.71) whose data are reported here.

Task

Each participant completed a 100 trial modified Sternberg working memory task [23] with

naturalistic scenes as stimuli. Scenes were 24-bit color images randomly sampled from the SUN

database [24]. The task was programmed in SuperLab 5 (Cedrus Inc.). Each scene was displayed

on a 27-inch LED monitor with a refresh rate of 60 hertz (Hz) and a screen resolution of 1920-

by-1080. Participants sat 83.5 cm from the monitor and maintained stable viewing using a

combined forehead/chin rest. Each scene measured 800-by-600 pixels on the screen, and from

the subject's point of view occupied a horizontal viewing angle of 17.2 degrees and a vertical

viewing angle of 12.7 degrees. The experiment took place within a sound-attenuated booth (IAC

acoustics) to minimize auditory and visual distractions.

Each trial of the working memory task consisted of the following phases (Fig 1):

encoding of scenes, delay (maintenance), probe, and presentation of a phase-scrambled scene.

During the encoding phase, two scenes were presented sequentially each for 2 seconds for a total

of 4 seconds of encoding. During the delay (maintenance) period a white screen was visible to

the participant for a total of 6 seconds. After the delay, a probe scene was presented for 2

seconds. If the scene was one of the two scenes in the immediately preceding encoding phase

(50% chance) the participant was required to press the right (green) button on a RB-530 response

pad (Cedrus Inc); if the probe scene was not one of the previous encoding set, the subject was

required to press the left (red) button on the response pad. After the probe, a phase-scrambled

image was presented for 5 seconds.

[insert Fig 1 here]

Behavioral Analysis

Correct and incorrect trials were counted, and a signal detection analysis was performed to assess

each participant's performance. For the latter, a hit was counted when a scene from the

immediately preceding encoding set was signaled by the subject by pressing a button indicating,

correctly, that the stimulus had been previously seen (an old stimulus correctly classified as old).

A false alarm was counted when a new scene not presented in the immediately preceding

encoding set was indicated by the participant pressing a button indicating, incorrectly, that the

scene had been previously presented (a new stimulus incorrectly classified as an old stimulus).

For each participant, total hits and false alarms were expressed as proportions in each subject and

used to compute a measure of sensitivity as the difference in standardized normal deviates of hits minus false alarms: d-prime (d') = $Z(hit\ rate)$ – $Z(false\ alarm\ rate)$. The d-prime sensitivity measure represents the separation between the means of the signal and noise distributions, compared against the standard deviation of the signal or noise distributions [25]. It was used as the primary covariate for the EEG analyses described below to determine how brain electrical activity during the delay period is related to a participant's ability to distinguish stimuli presented during the last 10 seconds. The first 7 subjects were only presented 98 trials out of the planned 100 due to an error in task programming. Behavioral measures for these subjects were adjusted for this difference in trial number.

EEG Acquisition

Electroencephalography (EEG) data were sampled at 1 kHz using Pycorder software from 62 scalp locations (Fig 2) using an active electrode system with an actiCHamp amplifier (Brain Products). Electrodes were placed at standard locations specified by an extended 10-20 system (https://osf.io/ebvsr/). The recording ground (Fpz) was located at the frontal midline and the recording reference was located at the left mastoid (TP9) leaving 61 scalp recordings (sensor level labels in Fig 3 and Fig 4). Two additional channels were designated for left (LOC) and right (ROC) vertical electrooculography (VEOG) recordings for subsequent isolation of eye blink artifacts. Recordings to disk were initiated after electrode impedances fell below 25 K Ohms. Although the standard convention is to reduce impedance to 5 K Ohms or below [26], the acquisition system utilizes active electrodes with noise reducing techniques built into the amplifier thereby ensuring that impedances under 25 K ohms are sufficient for interpretable signals [27]. Channels with impedance values above 25 K ohms were interpolated using data

from neighboring electrodes with impedances below 25 K ohms. An auxiliary channel was used

to record from a photosensor placed directly on a corner of the LED monitor. A 10-by-10 pixel

square located under the photosensor was programmed to change from white to black during

onset of each scene stimulus; it changed from black to white during stimulus offset. Recording

the changing screen luminance from the photosensor signal at 1 kHz allowed for precise timing

of stimulus onset and offset with respect to the recorded EEG data.

[insert Fig 2 here]

EEG Analysis

EEG signals were processed using BESA Research (v6.1). First, notch (frequency 60 Hz, 2 Hz

width) and bandpass (low cutoff 1 Hz, type forward, slope 6dB/oct; high cutoff 40 Hz, type zero

phase, slope 12 dB/oct) filters were applied to all channels. Second, the signal on each channel

was visually inspected to find, mark, and exclude the duration of all muscle artifacts. Third, a

characteristic eye-blink was marked by finding an alternating deflection greater than 100

microvolts (µV) between the LOC and ROC signals. Then a template matching algorithm was

used to find all eye blink artifacts on all channels and remove the component of variance

accounted for by the eye blinks [28, 29]. Finally, additional artifacts were isolated and excluded

using amplitude (120 μ V), gradient (75 μ V), and low-signal (max. minus min) criteria (0.01). A

participant's data was used in further processing only if a minimum of 60% of trials survived this

final artifact scan.

[insert Fig 3 here]

After preprocessing, each participant's artifact-free EEG data were used to compute the temporal-spectral evolution (TSE) of the signal at each sensor location. TSE is a method for examining signal changes around an event (e.g., the onset of the delay) as a function of both time and frequency. At each sensor location, an average TSE amplitude matrix (2 Hz wide frequency bins, 4-50 Hz frequency range, 25 ms wide temporal bins) was computed as:

$$TSE = \frac{A(t,f) - A_{baseline}(f)}{A_{baseline}(f)} \square 100\%$$

where A(t,f) is the amplitude of activity during the specific timeframe of interest and frequency band and $A_{baseline}(f)$ is mean activity in a specific frequency band during a specified baseline period. The calculated value is expressed as a percent; the value is either a positive or negative change in spectral activity compared to the baseline period [30-32]. In other words, the amplitude for each time bin is normalized to the mean amplitude of the baseline epoch for that frequency. In the resulting matrix, the x-axis shows the time relative to the event, and the y-axis shows the frequencies. The TSE values in the matrix range from -100% to +infinity. A value of +100% means that activity is twice as high during the baseline epoch. Positive changes reflect event-related enhancement or synchronization as amplitude peaks around the event, while negative changes reflect event-related suppression or desynchronization as amplitude decreases around the event [31, 33].

In the present study, the timeframe (t) and baseline were systematically varied. In the first analysis (Fig 6), the timeframe was the entire 6 seconds delay period and the baseline was defined as the entire 6 seconds delay period. In the second analysis, the timeframe was the entire 17 second window starting at encoding through scrambled scene presentation with the baseline

defined as the entire 17 second window (Fig 7a). In the third analysis, the timeframe was the

entire 17 second window starting at encoding through scrambled image presentation with the

baseline defined as just the 4 seconds encoding phase of the task (Fig 7b). In the fourth analysis,

the timeframe was the entire 17 second window starting at encoding through scrambled scene

presentation with the baseline defined as the 5 seconds scrambled scene presentation that came

after the 2 seconds probe phase (Fig 7c) and before the next set of two scenes for the encoding

period of the next trial. TSE calculations included all correct and incorrect trials surviving the

eye-blink correction and artifact-scan, and for the sensor level analyses shown in Figs 6 and 7 the

average TSE matrix across sensors is displayed.

The sensor level TSE analyses as described above were also carried out in source space

[34, 35] using the BR Brain Regions montage in BESA Research (Fig 5) and an age appropriate

(20-24 year old) average adult head model [36]. The montage represents source space as 15 brain

areas divided among left and right frontal, midline, parietal, temporal, and occipital regions.

[insert Fig 5 here]

Statistical analyses of sensor and source space TSE matrices with participant d-prime

performance measures was done in BESA Statistics v2.0 with corrections for multiple

comparisons. Significant TSE-performance correlations are depicted in Figs 6 through 8 as dark

red or blue shaded clusters overlaid on the average TSE activity matrices. Nonparametric cluster

permutation tests (N=1,000) were computed in BESA Statistics v2.0 [37, 38]. For correlations,

cluster permutation tests include two steps. First, spatially contiguous clusters of coherent r

values exceeding a certain threshold along selected dimensions (time-by-frequency averaged

over electrodes) are detected in the data. Here significant correlations were indicated by an a

priori corrected p-value of less than or equal 0.05 (marginally significant by less than 0.10).

Second, summed r values of the clusters are compared to a null distribution of r sums of random

clusters obtained by permuting data order across subjects. This controls for type I errors due to

multiple comparisons. Thus, the null hypothesis behind the permutation test for correlations

assumes that the assignment of the covariate per subject (d-prime in this case) is random and

exchangeable. Clusters of r values subjected to permutation are built across different dimensions.

In the case of the sensor level analyses, clustering was performed on the time-frequency window

after averaging over electrodes. In the case of the source estimation analyses, clustering was

performed on 15 time-frequency windows each representing different source brain regions. The

time and frequency centers of each identified cluster were extracted to report locations with

respect to well-known frequency bands, including theta (4-8 Hz), alpha (8-13 Hz), beta (13-40

Hz). The delta (<4 Hz) and gamma (40-90 Hz) ranges were not considered in this study.

Results

Behavioral

Participants performed well above chance on the task (percent correct M = 93.13, SD = 7.86;

choice reaction time M = 1009 ms, SD = 91.66 ms). The mean number of hits (out of 50) was

45.87 (SD = 3.48), and the mean number of false alarms was 0.87 (SD = 1.77). The resulting

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average sensitivity, d-prime, to distinguish old from new scenes was 3.70 (SD = 0.70).

Sensor Level TSE

Delay Window Baseline

The average TSE matrix including artifact-free correct and incorrect trial data averaged over

sensors is shown for the delay period window (Fig 6a). An area of negative change, relative to

the whole delay window baseline, can be seen spanning 6 to 26 Hz, including from the start of

the delay period, delay onset time=0, to 600 ms. A weak positive change from baseline is present

from around 500 ms to 2550 ms spanning frequencies 8 to 18 Hz. A stronger positive change is

present in the 4 to 7 Hz range from around 200 ms to 1300 ms. Finally, widespread, weak

negative change is present from approximately 4000 ms to the end of the maintenance period

spanning the 6 to 28 Hz range and also from 2700 ms to 5350 ms spanning the 4-7 Hz range.

Statistical analysis of the correlation between TSE and participant d-prime measures for this

window and baseline revealed three significant clusters (Fig 6b). The most significant cluster,

Cluster 1, occurred from 5000-6000 ms, spanned a frequency range of 6 to 38 Hz, and contained

activity that was negatively correlated (Fig 6e, r = -.71, p = .02) with d-prime. The next most

significant cluster, Cluster 2, occurred from 1400 to 2425 ms, spanned a frequency range of 4 to

24 Hz, and was positively correlated (r = .75, p = .03) with d-prime. The third most significant

cluster, Cluster 3, occurred from 3850 to 4800 ms, spanned a frequency range of 8 to 24 Hz, and

was negatively correlated (r = -.73, p = .04) with d-prime.

[insert Fig 6 here]

Whole Window Baseline

The average TSE matrix including artifact-free correct and incorrect trial data and averaged over

sensors is shown for the entire trial window, from the start of encoding to the end of the

scrambled image presentation (Fig 7a). The baseline for this analysis is the entire trial window.

The delay period is indicated by 0 to 6000 ms on the x-axis. A period of positive change is

noticeable from approximately 425 ms to about 4400 ms ranging from 8 to 26 Hz. There is a

short period of weaker positive change between 350 ms to 1800 ms ranging from 4 to 7 Hz.

Weak negative change occurs between 0 and 400 ms ranging from 4 to 22 Hz and also from

4600 ms to the end of the delay period at 6000 ms in the 4 to 8 Hz range. Statistical analysis of

the correlation between TSE and participant d-prime measures revealed one significant cluster

ranging from 1400 to 3150 after delay onset, spanning frequencies 4 to 30 Hz, and that was

positively correlated with d-prime (r = .68, p = .04).

[insert Fig 7 here]

Encoding Period Baseline

The average TSE plot including artifact-free correct and incorrect trial data averaged over

sensors is shown for the entire trial window, but with only the encoding period serving as

baseline (Fig 7b). A positive change is visible after the onset of the delay period from 450 ms to

about 6000 ms spanning the 8 to 24 Hz range, with a short period of weaker positive change

between 400 and 1800 ms spanning the 4 to 8 Hz range. Statistical analysis of the correlation

between TSE and participant d-prime measures revealed one significant cluster (Fig 7b) ranging

from 925 to 3150 ms after delay onset, spanning frequencies 4 to 32 Hz, and positively

correlated with d-prime (r = .65, p = .02).

Scrambled Scene Baseline

The average TSE matrix including artifact-free correct and incorrect trial data averaged over

sensors is shown for the entire trial window, but with only the post-probe scrambled scene period

serving as baseline (Fig 7c). There is a visible positive change from 475 ms to about 4500 ms in

the 8 to 24 Hz range after the onset of the delay period. There is weak negative change from 0 to

400 ms in the 4 to 22 Hz range, and also from 4700 ms to the end of the delay period in the 4 to 7

Hz range. Statistical analysis of the correlation between TSE and participant d-prime measures

revealed no significant clusters.

Source Level TSE

The average TSE matrix including artifact-free correct and incorrect trial data is shown

for the entire trial window, from the start of encoding to the end of the scrambled scene

presentation, for source signals representing 15 brain regions (Fig 8d). The baseline for this

analysis is the entire trial window. Strong positive change was evident after the onset of the

delay period, while negative change was evident during the encoding period (before the delay) as

well as during the probe period after the delay. This pattern of changes was particularly

noticeable in the medial frontal pole, left and right posterior temporal lobe, the parietal lobe, and

the medial occipital pole. Statistical analysis of the correlation between the 15 source TSE

signals and participant d-prime measures revealed one cluster (Fig 8b, blue) ranging from -3750

to +250 ms, spanning 4 to 38 Hz, that was significantly negatively correlated with d-prime (r = -

.76, p = .01). A second cluster (Fig 8b, red) was also revealed, ranging from 1250 to 3150 ms

and spanning 4 to 24 Hz, that was positively correlated (r = .65, p = .07) with d-prime and

reached marginal significance. These clusters were located in the right parietal source region

(PR_BR, region of source signal depicted orthogonally in Fig 8a). The cluster of negative

correlation occurred during the encoding period of the task, while the cluster of positive

correlation occurred during the delay period of the task. An additional source analysis was done

with a restricted window spanning just the delay period, with the same window of time serving

as baseline. Statistical analysis of the correlation between the 15 source TSE signals and

participant d-prime measures revealed one cluster (Fig 8d) ranging from 1400 to 2450 ms,

spanning from 4 to 24 Hz, and that was significantly positively correlated with d-prime (r = .79,

p = .03) after a multiple comparison correction including the number of source regions. This

cluster was also located in the right parietal source region.

[insert Fig 8 here]

Discussion

There are three main findings of the present study. The first is that evidence from the temporal-

spectral evolution method suggests transient rather than sustained activity during a relatively

long 6 seconds delay period. It has been argued that findings of sustained activity may be a

byproduct of averaging over trials and recording locations [7]. In the present study, averaging

over trials and locations resulted in transient patterns of TSE activity early and late in the delay

period. When examined at both the sensor and source levels, and when using different baselines

for comparison, an increase in TSE amplitude during the beginning of the delay period was

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followed by a decrease in TSE amplitude toward the end of the delay period.

The second main finding is that memory performance is predicted differently by the early and late patterns of delay period TSE activity. Early in the delay period, TSE amplitude in the 4-24 Hz range is *positively* correlated with d-prime, or the ability of subjects to distinguish a probe from the scenes presented at encoding. This means that the higher the TSE amplitude earlier in the delay, the more likely subjects will be able to distinguish whether an upcoming probe scene was one of the two previously presented scenes. Later in the delay period, however, TSE amplitude in the 4-30 Hz range is *negatively* correlated with d-prime. This means that the lower the TSE amplitude later in the delay, the more likely subjects will be able to distinguish whether a probe was one of the two previously presented scenes. Thus, the pattern of correlations of TSE activity with memory performance is also by definition transient, since TSE amplitude differently predicts subsequent short-term memory performance as a function of time from the beginning of the delay.

The third main finding of the present study is that the patterns of correlation between TSE amplitude and memory performance change when both the temporal window and baseline used for computation of TSE are varied. When the temporal window is restricted to just the 6 seconds delay period and the baseline consists of the mean activity across that 6 seconds window, an early positive cluster of correlation is found in addition to two later clusters of negative correlation (Fig 6). When both the window of analysis and baseline are expanded to include the entire trial (Fig 7a), only the single early cluster of positive correlation remains. When using a whole trial window, and restricting the baseline for TSE computation to the 4 seconds encoding period, a similar pattern also emerges (Fig 7b): a single cluster of positive correlation between TSE and performance remains early in the delay period. Finally, when using

the whole trial window and restricting the baseline for TSE computation to the 5 seconds scrambled scene period following the probe, there are no significant clusters of either positive or negative correlation anywhere in delay period or anywhere in the trial window.

The increase in TSE amplitude early in the delay period is consistent with event-related synchronization, with the event or time-locking being the start of the delay period [32, 39]. One interpretation of increased synchronization early in the delay period is that it reflects a turning of attention inward after the visual input stops. This pattern is consistent with cognitive processing associated with the attempt to maintain the two previously presented scenes in an online buffer. Since applying a verbal label to the complex scenes used here is effortful, and this strategy was not encouraged during this study, maintenance of the scenes in a buffer may last only a few seconds [40, 41] after which decay occurs. Maintenance for a short period of time followed by decay is consistent with the observed elevated activity early in the delay, only lasting for the first 2 or 3 seconds followed by the tapering of activity [4, 15, 16, 19, 20, 42]. This interpretation is supported by the finding of a positive correlation of only the early TSE amplitude with subsequent memory performance: the higher the TSE early in the delay period, the better subjects are at distinguishing old from new scenes during the subsequent probe phase.

The decrease in TSE amplitude toward the end of the delay period is consistent with event-related desynchronization, with time-locking being the beginning of the delay period [32, 39]. One interpretation of this late pattern of activity is that it reflects the motor planning that subjects make in anticipation of the end of the delay [33, 43]. The greater the decrease in TSE amplitude during the end of the delay period, the better subjects are at making a button press after distinguishing old from new scenes during the probe phase. An alternative to the motor planning interpretation is that this desynchronization instead reflects a reactivation of the

encoded scenes late in the delay period as subjects anticipate the approaching probe comparison [44, 45]. This alternative reactivation interpretation is supported by a similar pattern of negative correlation between TSE amplitude and memory performance during the actual encoding period, when the stimuli are visibly present, that was obtained in the source analysis in the right parietal region (Fig 8b).

An additional novel contribution of this report is the demonstration that changing the temporal window of analysis and baseline can alter the TSE-performance correlations obtained. The most extreme example in the sensor analyses performed here occurred when the baseline for the TSE analysis was switched to the 5 seconds period of scrambled scene presentation occurring after the probe scene was presented (Fig 7c). All clusters of significant correlation between TSE and memory performance disappeared, even though the pattern of higher TSE early in the delay and lower TSE during encoding and around the probe remained. One explanation for this is that, as in the early portion of the delay period, during the presentation of scrambled scenes subjects' attention is turned inward and focused on the previously presented scenes. Thus, when the scrambled period is used as a baseline, it effectively cancels out the similar activity during the delay period. During the viewing of scrambled scenes immediately after the probe, it is possible that the EEG signal reflects patterns of activity similar to the activity present early in the delay period. This is plausible especially for the 50% of the trials where the immediately preceding probe was a positive probe, and could have reminded the subject of one of the two previously encoded scenes. This scenario is one explanation for why using the scrambled period of activity as a baseline reduced the TSE delay activity correlation with memory performance.

This transient pattern of activity that was observed during the delay period is inconsistent with some previous research [4, 15, 16, 19, 20, 42]. Studies using MEG [46-48] and EEG [49,

50] have found sustained, not transient, increases in alpha power (8-12 Hz) during the delay period for stimuli that were successfully remembered compared to forgotten stimuli. These studies have also reported sustained activity in the theta range (4-8 Hz) [19, 20, 42]. Increases in beta oscillations (12-30 Hz) have also been observed throughout the delay period of working memory tasks [4, 48]. In addition, sustained decreases in alpha and beta power are typically associated with semantic encoding of stimuli [51] and similar decreases in theta power have also been observed throughout task encoding [52] but not retention periods.

More recently, delay activity has also been associated with enhanced activity in the medial temporal lobe (MTL) [53]. According to one framework [44], synchronization and desynchronization reflect a division of labor between hippocampus and neocortex respectively. In the neocortex, information is stored in the cortex in material-specific regions. The hippocampus, a structure in MTL, that binds information together among widespread cortical regions. In this framework, it is postulated that synchronization and desynchronization mechanisms interact in the formation of episodic memories, but the same processes may be at work during the initial processing of novel stimuli in working memory. Information transmission between neocortex and hippocampus is believed to be related to activity in the theta range, and more specifically the cross coupling of theta and gamma oscillations [44, 54, 55]. Within one theta cycle, different patterns of neurons activate during each gamma cycle, and represent an object that is held in memory. Multiple objects can be held together, each represented by a different gamma cycle [56, 57]. Moreover, phase synchronization of theta activity between prefrontal and posterior brain regions reflects engagement of attention or central executive systems [53]. Thus, the observed theta activity may also reflect communication between the medial temporal structures and neocortex to facilitate the binding of encoded information for

later retrieval. Alpha and beta activity observed during working memory processes are also implicated in the processing and storage of information. Decreased activity in both the alpha and beta range is related to the indexing of information in cortex during information processing. Reactivation is associated with activity in the same material-specific regions where activity was observed during initial processing [58]. Because desynchronous alpha and beta activity appear to be one correlate of a mechanism for storing information during memory encoding, observation of this activity during the later portion of the delay period raises the question of whether it may represent reactivation of encoded stimuli in preparation for comparison with a subsequent probe stimulus.

In the present study, the correlation of delay activity with memory performance was affected by changing the baseline period used in the TSE computation. Patterns of significant correlations were observed most clearly when the entire delay period itself served as its own baseline. When the scrambled scene period was used as the baseline, the significant activity correlations with performance disappeared. The scramble scene period after the probe is a time when the participant may be directing attention inward and mostly ignoring visual input [39]; hence, it could be argued that this baseline resembles a pre-stimulus baseline in anticipation of another upcoming set of stimuli to be encoded. Alternatively, the scrambled scene occurred immediately after the probe, a time during which on at least half of the trials the subject would have been just reminded of one of the encoded stimuli. Therefore, reactivation of the encoded stimuli could be reflected in the activity during this scrambled scene period, and this activity could be similar to the patterns early in the delay period. Such a scenario would explain why delay activity performance correlations disappeared when the scrambled scene is used as a baseline. Many previous studies employ a baseline period that occurs before the presentation of

stimuli [31, 32] and participants are instructed to focus on a fixation cross [4, 46, 59]. Recently it has been suggested that a baseline period in which the participant is not engaged in any cognitive activity could bias the interpretation of the spectral changes [60]. Therefore, future studies should examine how cognitive processes interact with different visual inputs in baseline periods before encoding, during the delay period, and after probe presentation.

Previous studies have often utilized relatively short retention periods of approximately 3 seconds or less [4, 15, 16]. The present study used a delay period twice as long. Yet, the pattern of synchronous activity that was found early in this long delay period had a duration (~3 seconds) that was similar to the aforementioned studies. Extending the delay period in the present study resulted in activity that is temporally restricted to only a few seconds, but about the same length as reported in other studies. One interpretation is that the delay period durations used in previous EEG studies may have been too short to reveal transient, early activity. In addition, many studies that examine delay activity have excluded activity from trials where the subject subsequently made incorrect responses. In this study, both correct and incorrect trials were included in the TSE matrix computations to be able to compute accurately correlations with the d-prime performance sensitivity measure, which is computed based on both correct and incorrect trials. It would not be optimal to correlate only correct trial activity patterns with a measure of performance computed from both correct and incorrect behavioral choices. It is likely that correct and incorrect responses produce different patterns of delay activity [61], but this is reflected in the average TSE matrices and in the subsequent correlations with d-prime.

Finally, the present study utilized complex naturalistic stimuli, as compared to many studies that have employed simple shapes, letters, or numbers. Some working memory studies have used faces [16, 46] or real-world scenes [62]. Although the early pattern of activity

exhibited during the delay is consistent with studies that have used simple stimuli, it was not sustained throughout the entire delay period. The early positive activity that faded after 3 seconds may reflect visual memory decay associated with the difficulty of maintaining complex scenes without effortful covert rehearsal. If subjects had difficulty attaching a verbal label to the complex stimuli, they may not have been able to employ rehearsal mechanisms. While some stimuli were easier to assign a verbal label (e.g., scene with a building), other stimuli may have been more difficult (e.g., scene with trees, a valley, and mountains). When memory for stimuli that were difficult to label began to fade, underlying synchronous activity may have become more desynchronous around 3 seconds. If on the other hand, participants were able to verbally label the scenes and rely on rehearsal, they may have been able to maintain the stimuli throughout the retention period thus resulting in a sustained pattern of synchronous activity through the whole delay period. It is worth noting that in the source analyses, there were no significant correlations of TSE delay activity with performance in any left frontal region, the brain area implicated in covert articulatory rehearsal. Rather, the only significant TSE activity performance correlation was obtained in the right parietal region. This region has is consistent with the neural substrate of the working memory visuospatial sketchpad's inner scribe and visual cache. The phonological loop's region of articulatory rehearsal [1] is typically localized to left frontal regions. Future source estimation studies should examine the transient versus sustained nature of the delay activity when the phonological loop is recruited during covert rehearsal with explicit verbal labels and also during control conditions with and without articulatory suppression (e.g., repeating 'the' instead of a scene-based verbal label).

Conclusions

The temporal spectral evolution of the delay period EEG suggests that short-term visual memory maintenance involves a period of complex, transient activity. Changes toward the beginning and toward the end of a 6 second fixed delay period contribute differently to subsequent memory performance. Early in the delay period, synchronous neural oscillations are positively correlated with subsequent successful performance on a working memory task; later in the delay period, desynchronous neuronal oscillations are correlated with subsequent performance but with an opposite, negative relationship. The significance of the delay activity-performance correlations depends on the time window and baseline periods used in the analysis. These results add to the growing literature supporting transient rather than sustained activity changes during working memory delay periods. Additional research is needed to understand how different cognitive processes involved in maintaining complex visual information modulates underlying oscillatory activity, and how synchronous and desynchronous patterns reflect intra- and inter-regional information transmission.

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Author Contributions

Conceived and designed the experiments: T.M.E.

Collected data: C.R. and K.N. Analyzed data: C.R. and K.N. Wrote the paper: T.M.E. and C.R.

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Figures and Legends

Figure 1. Example Scene Working Memory Trials. In a load 2 working memory trial, two encoding images are each presented for 2 seconds followed by a 6 seconds delay (maintenance) period. After the delay, a probe image is shown for 2 seconds followed by a 5 seconds scrambled scene baseline. Some trials contain a negative probe (a) where the probe stimulus is not one of the previously presented encoding images. Other trials contain a positive probe (b) where the probe stimulus is one of the previously presented encoding images. An example average EEG trace from one subject's 94 artifact free trials is shown (c) with trial phases labeled. Onset of the delay period is marked by a longer vertical red line.

Figure 2. Scalp Electrode Montage Used for EEG Recordings. Scalp electrode positions used in the recordings are shown displayed on a head model. Contour lines represent 3D voltage amplitude mapping in a single subject 2000 ms after delay period onset. Note right posterior focus.

Figure 3. Averaged Whole Trial Window EEG Traces. An average EEG trace is shown at each sensor (a) for a single subject with 94 trials surviving artifact detection. Electrode T7 is shown in zoomed view (b) revealing the structure of the average EEG signal as a function of encoding, delay (maintenance), probe and scramble stimuli presentation. The T7 channel overplot view is also shown (c) showing superimposition of all 94 trial EEG traces from the beginning to end of the time window. Note that a negative potential is a upward deflection in these plots.

Figure 4. Temporal Spectral Evolution at the Sensor Level. At each sensor, the average TSE for a single subject with 94 trials surviving artifact detection is shown. Positive TSE change is visible after the onset of the delay period (vertical red line), while negative TSE change is evident before and after the onset and offset of the delay period. At each sensor, the x-axis of the TSE matrix shows time relative to the event, and the y-axis displays linearly scales with frequency. TSE amplitude intensities are color coded where blue is negative and red is positive. Note positive red change dominates during the delay period, whereas blue is most prominent during the encoding and probe phases.

Figure 5. Temporal Spectral Evolution at the Source Level. A brain region source montage shows the average TSE for a single subject with 94 trials surviving artifact detection. Positive TSE change (red) is evident after the onset of the delay period (vertical red line), while negative TSE change (blue) is evident before and after the onset and offset of the delay period. These changes are particularly noticeable in medial frontal pole (FpM_BR), left posterior temporal lobe (TPL_BR), left parietal lobe (PL_BR), medial parietal lobe (PM_BR), right parietal lobe (PR_BR), right posterior temporal lobe (TPR_BR) and medial occipital pole (OpM_BR).

Figure 6. Average Delay Window TSE and Significant Clusters of Correlation with Performance Across All Subjects. Average delay period TSE plot (computed using whole delay window as baseline) with no statistical significance overlay (a) shows initial negative change from 6 to 26 Hz at the start of the delay period 0 to 600 ms. Weak positive change (red)

is then present from around 500ms to 2550 ms at 8-18 Hz. Stronger positive change is present in the 4-7 Hz from around 200 ms to 1300 ms. A switch to negative change (blue) is present from approximately 4000 ms to the end of the delay period in the 6-28 Hz range and from 2700 ms to 5350 ms in the 4-7 Hz range. Correlation of TSE with subject d-prime revealed three significant clusters (b) in descending order from most significant. Significant clusters are represented by darker shades of red or blue overlaid on TSE activity. Cluster 1 (c, latency 5000-6000 ms and frequency range 6-38 Hz) and Cluster 3 (d, latency 3850-4800 ms and frequency range 8-24 Hz) were negatively correlated with d-prime (Cluster 1: r = -.71, p = .02; Cluster 3: r = -.73, p = .04) and Cluster 2 (e, latency 1400-2425 ms and frequency range 4-24 Hz) was positively correlated with d-prime (r = .75, p = .03). Dark black lines on the graphs (c-e) are linear regression fits and dotted lines are 95% confidence intervals.

Figure 7. Effect of Baseline Window on TSE-Performance Correlations. When the whole 17 seconds window from encoding to scrambled scene presentation was used as baseline in computation of TSE (a) a single cluster (latency 1400-3150 ms and frequency range 4-30 Hz) was found that positively correlated with d-prime (r = .68, p = .04). When the 4 seconds period of encoding was used as baseline in computation of TSE (b) a single cluster (latency 925-3150 ms, frequency range 4-32 Hz) was found that correlated positively with d-prime (r = .65, p = .02). When the 5 seconds period of scrambled scene presentation after the probe stimulus was used as the baseline in computation of TSE (c), there were no significant clusters found that correlated with d-prime.

Figure 8: Source Space Analysis Reveals TSE-Performance Correlation in a Right Parietal Region. Source TSE analysis finds that PR_BR (a) includes a cluster of negative correlation with d-prime (blue, r = -.76, p = .01) during encoding and a positive correlation with d-prime during the delay (red, r = .65, p = .07) in a whole window analysis (b) and a cluster of positive correlation with d-prime (red, r = .79, p = .03) in a smaller temporal window that included just the delay period (c). Location of the PR_BR relative to the other sources used to compute TSE is shown in a top down view (d).

References

- 1. Baddeley A. Working memory: looking back and looking forward. Nat Rev Neurosci. 2003;4(10):829-39. doi: 10.1038/nrn1201. PubMed PMID: 14523382.
- 2. Cowan N. What are the differences between long-term, short-term, and working memory? Prog Brain Res. 2008;169:323-38. doi: 10.1016/S0079-6123(07)00020-9. PubMed PMID: 18394484; PubMed Central PMCID: PMCPMC2657600.
- 3. Fukuda K, Awh E, Vogel EK. Discrete capacity limits in visual working memory. Curr Opin Neurobiol. 2010;20(2):177-82. doi: 10.1016/j.conb.2010.03.005. PubMed PMID: 20362427; PubMed Central PMCID: PMCPMC3019116.
- 4. Bashivan P, Bidelman GM, Yeasin M. Spectrotemporal dynamics of the EEG during working memory encoding and maintenance predicts individual behavioral capacity. Eur J Neurosci. 2014;40(12):3774-84. doi: 10.1111/ejn.12749. PubMed PMID: 25288492.
- 5. Goldman-Rakic PS. Cellular basis of working memory. Neuron. 1995;14(3):477-85. PubMed PMID: 7695894.
- 6. Shafi M, Zhou Y, Quintana J, Chow C, Fuster J, Bodner M. Variability in neuronal activity in primate cortex during working memory tasks. Neuroscience. 2007;146(3):1082-108. doi: 10.1016/j.neuroscience.2006.12.072. PubMed PMID: 17418956.
- 7. Lundqvist M, Rose J, Herman P, Brincat SL, Buschman TJ, Miller EK. Gamma and Beta Bursts Underlie Working Memory. Neuron. 2016;90(1):152-64. doi: 10.1016/j.neuron.2016.02.028. PubMed PMID: 26996084; PubMed Central PMCID: PMCPMC5220584.
- 8. Rose NS, LaRocque JJ, Riggall AC, Gosseries O, Starrett MJ, Meyering EE, et al. Reactivation of latent working memories with transcranial magnetic stimulation. Science. 2016;354(6316):1136-9. doi: 10.1126/science.aah7011. PubMed PMID: 27934762; PubMed Central PMCID: PMCPMC5221753.
- 9. Stokes MG. 'Activity-silent' working memory in prefrontal cortex: a dynamic coding framework. Trends Cogn Sci. 2015;19(7):394-405. doi: 10.1016/j.tics.2015.05.004. PubMed PMID: 26051384; PubMed Central PMCID: PMCPMC4509720.
- 10. Wolff MJ, Ding J, Myers NE, Stokes MG. Revealing hidden states in visual working memory using electroencephalography. Front Syst Neurosci. 2015;9:123. doi: 10.3389/fnsys.2015.00123. PubMed PMID: 26388748; PubMed Central PMCID: PMCPMC4558475.
- 11. Gevins A, Smith ME, Le J, Leong H, Bennett J, Martin N, et al. High resolution evoked potential imaging of the cortical dynamics of human working memory. Electroencephalogr Clin Neurophysiol. 1996;98(4):327-48. PubMed PMID: 8641154.

- 12. Gevins A, Smith ME, McEvoy L, Yu D. High-resolution EEG mapping of cortical activation related to working memory: effects of task difficulty, type of processing, and practice. Cereb Cortex. 1997;7(4):374-85. PubMed PMID: 9177767.
- 13. Offen S, Schluppeck D, Heeger DJ. The role of early visual cortex in visual short-term memory and visual attention. Vision Res. 2009;49(10):1352-62. doi: 10.1016/j.visres.2007.12.022. PubMed PMID: 18329065; PubMed Central PMCID: PMCPMC2696572.
- 14. Yantis S, Schwarzbach J, Serences JT, Carlson RL, Steinmetz MA, Pekar JJ, et al. Transient neural activity in human parietal cortex during spatial attention shifts. Nat Neurosci. 2002;5(10):995-1002. doi: 10.1038/nn921. PubMed PMID: 12219097.
- 15. Jensen O, Gelfand J, Kounios J, Lisman JE. Oscillations in the alpha band (9-12 Hz) increase with memory load during retention in a short-term memory task. Cereb Cortex. 2002;12(8):877-82. PubMed PMID: 12122036.
- 16. Tuladhar AM, ter Huurne N, Schoffelen JM, Maris E, Oostenveld R, Jensen O. Parieto-occipital sources account for the increase in alpha activity with working memory load. Hum Brain Mapp. 2007;28(8):785-92. doi: 10.1002/hbm.20306. PubMed PMID: 17266103.
- 17. Raghavachari S, Kahana MJ, Rizzuto DS, Caplan JB, Kirschen MP, Bourgeois B, et al. Gating of human theta oscillations by a working memory task. J Neurosci. 2001;21(9):3175-83. PubMed PMID: 11312302.
- 18. Khursheed F, Tandon N, Tertel K, Pieters TA, Disano MA, Ellmore TM. Frequency-specific electrocorticographic correlates of working memory delay period fMRI activity. Neuroimage. 2011;56(3):1773-82. doi: 10.1016/j.neuroimage.2011.02.062. PubMed PMID: 21356314; PubMed Central PMCID: PMCPMC3085578.
- 19. Hsieh LT, Ekstrom AD, Ranganath C. Neural oscillations associated with item and temporal order maintenance in working memory. J Neurosci. 2011;31(30):10803-10. doi: 10.1523/JNEUROSCI.0828-11.2011. PubMed PMID: 21795532; PubMed Central PMCID: PMCPMC3164584.
- 20. Roberts BM, Hsieh LT, Ranganath C. Oscillatory activity during maintenance of spatial and temporal information in working memory. Neuropsychologia. 2013;51(2):349-57. doi: 10.1016/j.neuropsychologia.2012.10.009. PubMed PMID: 23084981; PubMed Central PMCID: PMCPMC3546228.
- 21. Pfurtscheller G, Aranibar A. Evaluation of event-related desynchronization (ERD) preceding and following voluntary self-paced movement. Electroencephalogr Clin Neurophysiol. 1979;46(2):138-46. PubMed PMID: 86421.
- 22. Salmelin R, Hari R. Spatiotemporal characteristics of sensorimotor neuromagnetic rhythms related to thumb movement. Neuroscience. 1994;60(2):537-50. PubMed PMID: 8072694.

- 23. Sternberg S. High-speed scanning in human memory. Science. 1966;153(3736):652-4. PubMed PMID: 5939936.
- 24. Xiao J, Hays J, Ehinger KA, Oliva A, Torralba A, editors. Sun database: Large-scale scene recognition from abbey to zoo. Computer vision and pattern recognition (CVPR), 2010 IEEE conference on; 2010: IEEE.
- 25. Stanislaw H, Todorov N. Calculation of signal detection theory measures. Behav Res Methods Instrum Comput. 1999;31(1):137-49. PubMed PMID: 10495845.
- 26. Teplan M. Fundamentals of EEG measurement. Measurement science review. 2002;2(2):1-11.
- 27. Brain Products. Products by Application. 2016.
- 28. Picton TW, van Roon P, Armilio ML, Berg P, Ille N, Scherg M. The correction of ocular artifacts: a topographic perspective. Clin Neurophysiol. 2000;111(1):53-65. PubMed PMID: 10656511.
- 29. Ille N, Berg P, Scherg M. Artifact correction of the ongoing EEG using spatial filters based on artifact and brain signal topographies. Journal of clinical neurophysiology. 2002;19(2):113-24.
- 30. Lopes da Silva F, Pfurtscheller G. Basic concepts on EEG synchronization and desynchronization. Event-Related Desynchronization Handbook of Electroencephalography and Clinical Neurophysiology Revised Series, vol 6. Swammerdam Institute for Life Sciences (SILS): Elsevier Science; 1999.
- 31. Pfurtscheller G. Functional brain imaging based on ERD/ERS. Vision Res. 2001;41(10-11):1257-60. PubMed PMID: 11322970.
- 32. Pfurtscheller G, Lopes da Silva FH. Event-related EEG/MEG synchronization and desynchronization: basic principles. Clin Neurophysiol. 1999;110(11):1842-57. PubMed PMID: 10576479.
- 33. Neuper C, Pfurtscheller G. Event-related dynamics of cortical rhythms: frequency-specific features and functional correlates. Int J Psychophysiol. 2001;43(1):41-58. PubMed PMID: 11742684.
- 34. Hoechstetter K, Bornfleth H, Weckesser D, Ille N, Berg P, Scherg M. BESA source coherence: a new method to study cortical oscillatory coupling. Brain topography. 2004;16(4):233-8.
- 35. Scherg M. Functional imaging and localization of electromagnetic brain activity. Brain topography. 1992;5(2):103-11.

- 36. Richards JE, Sanchez C, Phillips-Meek M, Xie W. A database of age-appropriate average MRI templates. Neuroimage. 2016;124(Pt B):1254-9. doi: 10.1016/j.neuroimage.2015.04.055. PubMed PMID: 25941089; PubMed Central PMCID: PMCPMC4630162.
- 37. Maris E, Oostenveld R. Nonparametric statistical testing of EEG- and MEG-data. J Neurosci Methods. 2007;164(1):177-90. doi: 10.1016/j.jneumeth.2007.03.024. PubMed PMID: 17517438.
- 38. Maris E. Statistical testing in electrophysiological studies. Psychophysiology. 2012;49(4):549-65. doi: 10.1111/j.1469-8986.2011.01320.x. PubMed PMID: 22176204.
- 39. Klimesch W. Memory processes, brain oscillations and EEG synchronization. Int J Psychophysiol. 1996;24(1-2):61-100. PubMed PMID: 8978436.
- 40. Jensen O, Mazaheri A. Shaping functional architecture by oscillatory alpha activity: gating by inhibition. Front Hum Neurosci. 2010;4:186. doi: 10.3389/fnhum.2010.00186. PubMed PMID: 21119777; PubMed Central PMCID: PMCPMC2990626.
- 41. Klimesch W, Sauseng P, Hanslmayr S. EEG alpha oscillations: the inhibition-timing hypothesis. Brain Res Rev. 2007;53(1):63-88. doi: 10.1016/j.brainresrev.2006.06.003. PubMed PMID: 16887192.
- 42. Jensen O, Tesche CD. Frontal theta activity in humans increases with memory load in a working memory task. Eur J Neurosci. 2002;15(8):1395-9. PubMed PMID: 11994134.
- 43. Kaiser J, Birbaumer N, Lutzenberger W. Event-related beta desynchronization indicates timing of response selection in a delayed-response paradigm in humans. Neurosci Lett. 2001;312(3):149-52. PubMed PMID: 11602332.
- 44. Hanslmayr S, Staresina BP, Bowman H. Oscillations and Episodic Memory: Addressing the Synchronization/Desynchronization Conundrum. Trends Neurosci. 2016;39(1):16-25. doi: 10.1016/j.tins.2015.11.004. PubMed PMID: 26763659; PubMed Central PMCID: PMCPMC4819444.
- 45. Polyn SM, Natu VS, Cohen JD, Norman KA. Category-specific cortical activity precedes retrieval during memory search. Science. 2005;310(5756):1963-6. doi: 10.1126/science.1117645. PubMed PMID: 16373577.
- 46. Jokisch D, Jensen O. Modulation of gamma and alpha activity during a working memory task engaging the dorsal or ventral stream. J Neurosci. 2007;27(12):3244-51. doi: 10.1523/JNEUROSCI.5399-06.2007. PubMed PMID: 17376984.
- 47. Meeuwissen EB, Takashima A, Fernandez G, Jensen O. Increase in posterior alpha activity during rehearsal predicts successful long-term memory formation of word sequences. Hum Brain Mapp. 2011;32(12):2045-53. doi: 10.1002/hbm.21167. PubMed PMID: 21162031.
- 48. Palva JM, Monto S, Kulashekhar S, Palva S. Neuronal synchrony reveals working memory networks and predicts individual memory capacity. Proc Natl Acad Sci U S A.

2010;107(16):7580-5. doi: 10.1073/pnas.0913113107. PubMed PMID: 20368447; PubMed Central PMCID: PMCPMC2867688.

- 49. Berger B, Omer S, Minarik T, Sterr A, Sauseng P. Interacting memory systems-does EEG alpha activity respond to semantic long-term memory access in a working memory task? Biology (Basel). 2014;4(1):1-16. doi: 10.3390/biology4010001. PubMed PMID: 25545793; PubMed Central PMCID: PMCPMC4381213.
- 50. Khader PH, Jost K, Ranganath C, Rosler F. Theta and alpha oscillations during working-memory maintenance predict successful long-term memory encoding. Neurosci Lett. 2010;468(3):339-43. doi: 10.1016/j.neulet.2009.11.028. PubMed PMID: 19922772; PubMed Central PMCID: PMCPMC3951969.
- 51. Fellner MC, Bauml KH, Hanslmayr S. Brain oscillatory subsequent memory effects differ in power and long-range synchronization between semantic and survival processing. Neuroimage. 2013;79:361-70. doi: 10.1016/j.neuroimage.2013.04.121. PubMed PMID: 23664950.
- 52. Burke JF, Zaghloul KA, Jacobs J, Williams RB, Sperling MR, Sharan AD, et al. Synchronous and asynchronous theta and gamma activity during episodic memory formation. J Neurosci. 2013;33(1):292-304. doi: 10.1523/JNEUROSCI.2057-12.2013. PubMed PMID: 23283342; PubMed Central PMCID: PMCPMC3711714.
- 53. Fell J, Axmacher N. The role of phase synchronization in memory processes. Nat Rev Neurosci. 2011;12(2):105-18. doi: 10.1038/nrn2979. PubMed PMID: 21248789.
- 54. Buzsaki G. Neural syntax: cell assemblies, synapsembles, and readers. Neuron. 2010;68(3):362-85. doi: 10.1016/j.neuron.2010.09.023. PubMed PMID: 21040841; PubMed Central PMCID: PMCPMC3005627.
- 55. Lisman J, Buzsaki G. A neural coding scheme formed by the combined function of gamma and theta oscillations. Schizophr Bull. 2008;34(5):974-80. doi: 10.1093/schbul/sbn060. PubMed PMID: 18559405; PubMed Central PMCID: PMCPMC2518638.
- 56. Lisman J. Working memory: the importance of theta and gamma oscillations. Curr Biol. 2010;20(11):R490-2. doi: 10.1016/j.cub.2010.04.011. PubMed PMID: 20541499.
- 57. Lisman JE, Jensen O. The theta-gamma neural code. Neuron. 2013;77(6):1002-16. doi: 10.1016/j.neuron.2013.03.007. PubMed PMID: 23522038; PubMed Central PMCID: PMCPMC3648857.
- 58. Hanslmayr S, Staresina BP, Bowman H. Oscillations and Episodic Memory: Addressing the Synchronization/Desynchronization Conundrum. Trends in neurosciences. 2016;39(1):16-25.
- 59. Hanslmayr S, Spitzer B, Bauml KH. Brain oscillations dissociate between semantic and nonsemantic encoding of episodic memories. Cereb Cortex. 2009;19(7):1631-40. doi: 10.1093/cercor/bhn197. PubMed PMID: 19001457.

- 60. Jafarpour A, Horner AJ, Fuentemilla L, Penny WD, Duzel E. Decoding oscillatory representations and mechanisms in memory. Neuropsychologia. 2013;51(4):772-80. doi: 10.1016/j.neuropsychologia.2012.04.002. PubMed PMID: 22561180.
- 61. Haegens S, Osipova D, Oostenveld R, Jensen O. Somatosensory working memory performance in humans depends on both engagement and disengagement of regions in a distributed network. Hum Brain Mapp. 2010;31(1):26-35. doi: 10.1002/hbm.20842. PubMed PMID: 19569072.
- 62. Osipova D, Takashima A, Oostenveld R, Fernandez G, Maris E, Jensen O. Theta and gamma oscillations predict encoding and retrieval of declarative memory. J Neurosci. 2006;26(28):7523-31. doi: 10.1523/JNEUROSCI.1948-06.2006. PubMed PMID: 16837600.















