EEG oscillations during word processing predict MCI conversion to Alzheimer’s disease

Ali Mazaheri\textsuperscript{a,b,1}, Katrien Segaert\textsuperscript{a,b,1}, John Olichney\textsuperscript{c}, Jin-Chen Yang\textsuperscript{c}, Yu-Qiong Niu\textsuperscript{c}, Kim Shapiro\textsuperscript{a,b}, Howard Bowman\textsuperscript{a,b,d}

\textsuperscript{a} School of Psychology, University of Birmingham, United Kingdom
\textsuperscript{b} Centre for Human Brain Health, University of Birmingham, United Kingdom
\textsuperscript{c} Center for Mind and Brain and Neurology Department, University of California, Davis, CA, United States
\textsuperscript{d} School of Computing, University of Kent, United Kingdom

ABSTRACT

Only a subset of mild cognitive impairment (MCI) patients progress to develop a form of dementia. A prominent feature of Alzheimer’s disease (AD) is a progressive decline in language. We investigated if subtle anomalies in EEG activity of MCI patients during a word comprehension task could provide insight into the likelihood of conversion to AD. We studied 25 amnestic MCI patients, a subset of whom developed AD within 3-years, and 11 elderly controls. In the task, auditory category descriptions (e.g., ‘a type of wood’) were followed by a single visual target word either semantically congruent (i.e., oak) or incongruent with the preceding category. We found that the MCI converters group (i.e. patients that would go on to convert to AD in 3-years) had a diminished early posterior-parietal theta (3–5 Hz) activity induced by first presentation of the target word (i.e., access to lexico-syntactic properties of the word), compared to MCI non-converters and controls. Moreover, MCI converters exhibited oscillatory signatures for processing the semantically congruent words that were different from non-converters and controls. MCI converters thus showed basic anomalies for lexical and meaning processing. In addition, both MCI groups showed anomalous oscillatory signatures for the verbal learning/memory of repeated words: later alpha suppression (9–11 Hz), which followed first presentation of the target word, was attenuated for the second and third repetition in controls, but not in either MCI group. Our findings suggest that a subtle breakdown in the brain network subserving language comprehension can be foretelling of conversion to AD.

Mild cognitive impairment (MCI) is a syndrome characterized by cognitive decline that although not interfering with daily life, is greater than expected given an individual’s age. Roughly 60% of the individuals diagnosed with MCI progress to develop dementia within 5 years of MCI diagnosis (Gauthier et al., 2006; Portet et al., 2006). Identifying factors that predict conversion of MCI to dementia will lead the way to early pharmacological intervention, as well as secondary prevention by controlling risk factors such as blood pressure, inactivity, diet and cholesterol levels (Brookmeyer et al., 2016; Sjogren et al., 2006; Wiesmann et al., 2015).

The most prevalent underlying cause of dementia is Alzheimer’s disease (AD). A prominent feature of AD symptomatology is a progressive cognitive decline in faculties such as learning and memory, executive control, and language (Ferris and Farlow, 2013; Vestal et al., 2015). The deterioration of language abilities has been proposed to be of particular clinical relevance in terms of tracking the progression from moderate to severe stages of AD (Ferris and Farlow, 2013). The objective of the current study was to investigate if neuronal activity of MCI patients during a word comprehension task provides insight into the likelihood of conversion to AD dementia.

We examined the electroencephalogram (EEG) of MCI patients, a subset of whom developed dementia within 3 years, and healthy controls, while they performed a language comprehension task. In this task, patients and controls heard auditory phrases describing a category (e.g. ‘a type of wood’, ‘a breakfast food’), each of which was followed by the visual presentation of a single target word, which either fit (congruent, i.e. oak, pancake) or did not fit (incongruent nouns) with the preceding category statement.

The lexical processing of a target word requires the reader to access a range of different kinds of information about the word in the mental lexicon, including phonological, morphological and syntactic information, as well as the word’s semantic meaning. Early changes in EEG...
induced by target word presentation, irrespective of whether the word is congruent or not, have been suggested to be indicative of lexical processing (Huang et al., 2014; Ledoux et al., 2007). Lexical processing can be facilitated by repeated word presentation due to implicit memory for properties of the earlier presented word. Throughout the experiment we presented target words once, twice or three times. Changes in the EEG for congruent compared to incongruent target words are indicative of semantic or meaning processing. For congruent words, properties of the target word’s meaning have been pre-activated and can be integrated with the preceding phrase, facilitating meaning processing.

The data used in the current investigation was part of a previously published study (Olichney et al., 2008) using the N400 and P600 event related potentials (ERPs) to investigate memory encoding and retrieval processing deficits as predictors for conversion to AD dementia.

We focused our investigation on induced power changes in oscillatory activity generated by the onset of the target word. We examined oscillatory activity in theta (3–5 Hz), alpha (~10 Hz) and beta (15–20 Hz) frequency ranges given that prior studies have implicated these bands in various aspects of language processing, including lexical and semantic processing (Bastiaansen et al., 2005; Bastiaansen et al., 2008; Davidson and Indefrey, 2007; Hagoort et al., 2004; Hermes et al., 2014; Lam et al., 2016), as well as access to stored information and integration of information (Klimesch, 2012; Olesler and Weisz, 2012; Strauss et al., 2014; Weiss and Mueller, 2012). Finally, given that language comprehension involves the interaction of different brain regions across the cortex (Hagoort, 2013; Hermes et al., 2014), we also examined cross-frequency interactions of power across different brain areas during word comprehension (Mazaheri et al., 2009; Mazaheri et al., 2010; Mazaheri et al., 2014a).

1. Methods and materials

1.1. Participants

We examined the EEG recordings of 25 patients with amnestic MCI (mean age 73.2 years, range 55–84) and 11 normal elderly controls (mean age 74.1 years, range 57–79), who were all part of a previously published study (Olichney et al., 2008) in which the patients were tested annually with a semantic judgment task that has been found to reliably elicit the N400 and P600 components (Olichney et al., 2008).

In the present analyses, we focus on the initial baseline EEG session obtained after the MCI diagnosis in order to investigate patterns of responses that could be predictive of dementia conversion in the following 3 years. The MCI patients were recruited primarily through the Shiley-Marcos Alzheimer’s Disease Research Center (ADRC) at the University of California, San Diego. All participants were screened for treatable causes of cognitive impairment (e.g., vitamin B12 deficiency, thyroid dysfunction) and underwent a brain scan (generally MRI) prior to enrolment. The exclusion criteria included stroke, epilepsy, psychiatric conditions, as well as CNS-active medications. At the initial baseline EEG recording session, the patients all met the Petersen Criteria for MCI (Petersen, 2004), but not for dementia (American Psychiatric Association, 2000). Conversion to AD was defined according to the criteria set out by the National Institute of Neurological and Communicative Disorders and Stroke–Alzheimer’s Disease and Related Disorders Association (McKhann et al., 1984). Fifteen patients of the 25 with MCI subsequently converted to AD within 3 years of their initial baseline EEG recordings (mean number of years 1.62 ± 0.7 years). For more specific information of participant demographics, and neurocognitive testing of participants please refer to Olichney et al. (Olichney et al., 2008).

1.2. Semantic judgment task and study design

Our focus in this paper was purely on the encoding phase of the experiment. At encoding, participants were presented with an auditory phrase describing a category (e.g., ‘a type of wood’, ‘a breakfast food’) followed 1 s later by a single visual target word (300 ms, visual angle ~0.4°) that was either congruent (i.e., oak, pancake) or incongruent (i.e. nouns matched on word frequency and length) with the preceding category phrase. The probability of congruent versus incongruent target words was 0.5. The congruent target words were medium typicality exemplars of the selected category. The incongruent target words were incongruent with the category set by the preceding phrase but matched for frequency of usage and word length to the congruent targets. In addition, 2/3rds of the target words were repeated in a pseudorandom fashion between ~10 and 140 s later. These repeated target words always followed the same uniquely associated category statement as on the first presentation (congruent or incongruent). The participants were instructed to wait ~3-s after the onset of the word, then read the word aloud and state if it fit the preceding category with a “yes” or a “no” decision. Further details of this experimental paradigm have been published previously (Olichney et al., 2000; Olichney et al., 2008).

There were no differences between conditions and groups with respect to the behavioural accuracy of performing this task, as all participants performed near ceiling (Olichney et al., 2008). Healthy controls and patients were given as much time as required and reaction time was not recorded.

For our analyses of lexical processing associated with the presentation of the target word (Section 2.1 in the Results section), as well as for our analyses of semantic congruency (Section 2.3 in the Results section), we include only the first presentation of the target words, thus excluding target words that have been primed previously. For our analyses of the facilitatory effects of repeated word presentation on lexical processing (Section 2.2 in the Results section), we analysed the repeated target words that occurred on 1st, 2nd and 3rd presentations within the experiment.

1.3. EEG recordings

19 to 32 channel EEG was recorded at 250 Hz, band pass 0.016 to 100 Hz, and re-referenced off-line to linked mastoids. The pre-processing of the EEG data was performed using EEGLAB (Delorme and Makeig, 2004). Fieldtrip (Oostenveld et al., 2011) EEG epochs were locked to the onset of the visual target words and manually inspected for non-physiological artefacts. Independent component analysis was then applied to remove eye movement artefacts (Jung et al., 2001). Our classifications of frequency bands into alpha (9–11 Hz), theta (3–5 Hz) and beta (15–20 Hz) were based on prior literature (Bauer et al., 2014; Mazaheri et al., 2014b; Slagter et al., 2016; Van Diepen et al., 2015) and word-induced changes in power irrespective of the condition or participant group.

1.4. Oscillatory analyses

Time-frequency representations (TFRs) of power were calculated for each trial (1 s prior to word onset, and 1.5 s after) using sliding Hanning tapers having a varying time window of three cycles for each frequency (ΔT = 3/π). This approach has been used in a number of previous studies (Bengson et al., 2012; Van Diepen et al., 2015). Word presentation induced changes in the power of oscillatory activity, which was assessed in terms of change scores from baseline (ΔP) using the following formula:

\[ ΔP = (P_r - P_t) / P_t \]

where \( P_r \) was the mean power during the baseline period 500 ms to 100 ms before the onset of the word and \( P_t \) was the power at each specific time point.

1.5. Cross-frequency coupling between theta and alpha/beta

Language comprehension involves the interaction between different
brain regions across the cortex. Our goal is to capture these interactions in the healthy and MCI groups by examining the cross-frequency connectivity between theta and alpha/beta power changes following word onset. Investigating connectivity between brain regions using EEG has been difficult due to problems associated with volume conduction, namely that nearby electrodes pick up activity from the same sources. Here, we attempt to circumvent this issue by focusing on trial-by-trial negative correlations (De Lange et al., 2008; Mazaheri et al., 2009; Mazaheri et al., 2010; Mazaheri et al., 2011a), given that it is unlikely that a common source generates a simultaneous increase and decrease in power of different frequencies at distant electrode sites. Across participants, the amplitude envelope of the theta and alpha/beta oscillations at distant electrodes of interest were correlated on a trial-by-trial basis for each condition. For each participant, the correlation coefficients were converted to z values using Fischer’s r-to-z transform to obtain a normally distributed variable (Bengson et al., 2012; Mazaheri et al., 2010; Mazaheri et al., 2012a). The analysis of these correlations was assessed within groups, between groups, and between conditions, using a one-sample t-test of the Fisher r-to-z transformed correlations.

1.6. Statistical analysis

We used a-priori defined latencies of interest and selection of electrodes for our statistical tests, which were guided by previous literature looking at oscillatory changes in EEG/MEG activity induced by word onset (Bastiaansen et al., 2005; Bastiaansen et al., 2008; Hagoort et al., 2004), as well as by the aggregated grand average data of conditions and participant groups collapsed (Brooks et al., 2017).

Repeated-measures ANOVAs were used to analyze the event-related changes in theta and alpha/beta between conditions. F-ratios were tested using degrees of freedom adjusted by the Greenhouse–Geisser procedure. Post hoc comparisons were Bonferroni corrected for multiple-comparisons.

The factors used were Word type (congruous, incongruous), Group (control, MCI non-convertors, MCI converters) and Time window (0–500 ms or 400–500 ms, and 500–1000 ms). For analyses of the effects of repeated word presentation we also used Word repetition, by contrasting the 1st, 2nd and 3rd presentations. Frequency ranges as well as time windows were guided by previous research (Bastiaansen et al., 2005; Hagoort et al., 2004; Kielar et al., 2014; Strauss et al., 2012).

The justification for time windows and frequency ranges of interest can be found in each individual section of the results. In addition to using pre-defined electrodes and time windows of interest, we also used a complementary approach where we assessed the difference in oscillatory theta power related to lexical and semantic processing between MCI converters and non-converters across all electrodes and time points between 0 and 1 s after the onset of words. The multiple comparison issue here (multiple time points and electrodes) was circumvented by obtaining probability values through whole-volume non-parametric permutation testing (Maris and Oostenveld, 2007). In this procedure, first a two-tailed independent t-test (MCI converters versus non-converters) was computed for each individual channel-time pairs and thresholded at a 5% significance level. Significant pairs were clustered by direction of effect and spatial proximity using the ‘distance’-method (which defines neighbouring sensors based on proximity). A probability value of this electrode-time cluster was obtained through the Monte Carlo estimate of the permutation p-value of the cluster by randomly swapping the group label 1000 times and calculating the maximum cluster-level test statistic across the entire volume. A similar procedure has been used in a number of previous studies (Van Diepen et al., 2015; van Diepen et al., 2016; Van Diepen and Mazaheri, 2017).

It should be noted that these two analysis approaches are complementary, with the first being guided by previous research (i.e. prior precedent), while the second is data driven, but controls false positive rates by inferring maximum clusters under the null across the entire volume. To foreshadow our results, findings from the two approaches converge.

2. Results

2.1. Lexical processing

Firstly, we will focus on the oscillatory changes in the EEG associated with the lexical processing of the first presentation of single target words. During lexical processing, a reader accesses a range of different kinds of information about each word in the mental lexicon, including phonological information (e.g., segmental and metrical structure), morphological information (e.g. stems and affixes), syntactic information (e.g. grammatical class), and the word’s semantic meaning.

The onset of a word previously has been found to induce an early increase in theta activity over posterior regions, related to the processing of word form (Bastiaansen et al., 2005; Bastiaansen et al., 2008; Hagoort et al., 2004). Later alpha suppression at posterior sites has been associated with further post-perceptual processing of sensory information (Pluetscheller, 2001) and allocation of resources according to processing demands (Foxe et al., 1998). We were guided by the literature in our analysis approach, and analysed oscillatory changes at Pz, in the theta (3–5 Hz) and alpha (9–11 Hz) range, respectively 0–500 ms and 500–1000 ms following the onset of the target word. Of particular interest are any effects of Group.

2.1.1. Theta power increase related to lexical processing was attenuated in MCI converters

The onset of the word induced an increase in theta activity across all three groups, peaking at around 0.25 s after word onset and maximal over the midline-parietal electrode Pz (Fig. 1A and B). For this early theta increase, we found a main effect of Group (F(2,33) = 6.8, p = 0.003). MCI converters had a significantly diminished theta increase compared to the MCI non-converters (converters: 22% increase vs non-converters 50% increase, p = 0.046) and compared to the healthy elderly control group (MCI converters: 22% increase vs controls: 59% increase, p = 0.004) (Fig. 1A).

The theta increase was followed by alpha activity suppression 0.5 to 1 s after word onset (blue in Fig. 1A). The alpha suppression effect did not differ between groups (F(2,33) = 0.24 p = 0.78).

2.2. Facilitatory effects of lexical processing due to repeated word presentation

Repeated word presentation facilitates lexical processing, since there is implicit memory for properties of the word that are activated upon first presentation (Schacter et al., 1993). Thus we compared 1st to 2nd and 3rd presentations of the target words (collapsing congruent and incongruent words).

As discussed in Section 1, lexical processing is associated with an early (0–500 ms) theta increase (3–5 Hz) and a late (500–1000 ms) alpha suppression effect (9–11 Hz) at Pz. This guided our decision to examine both oscillatory signatures for the effects of word repetition.

It should be noted that we are protected here from criticisms of double dipping (Kriegeskorte et al., 2009), since under the null hypothesis, effects of repetition would be orthogonal to effects on first presentation, and the basic lexical processing effects identified in Section 1 were indeed on first presentations (Friston et al., 2006).

2.2.1. Alpha suppression attenuates with each word repetition, but only for healthy controls and not for either MCI patient group

We did not observe any significant changes in the early theta increase as a result of word repetition (not shown). However for the alpha suppression, there was a Word Repetition by Group interaction (F(4,66) = 4.2, p = 0.005) (Fig. 2). In the healthy elderly group, the
alpha suppression diminished with each repetition of the word (F(2,20) = 24.1, p < 0.001; simple effects: 1st vs 2nd word presentation: p = 0.008; 2nd vs 3rd word presentation: p = 0.072; 1st vs 3rd word presentation: p < 0.001). There was no word repetition effect in either the MCI non-convertors (F(2,18) = 0.45, p = 0.61) or the MCI convertors (F(2,28) = 0.24, p = 0.76).

2.3. Semantic or meaning processing

We now focus on differential oscillatory changes in the EEG for congruous versus incongruous target words. For congruous words, the preceding category description has pre-activated the category exemplar information, facilitating access to meaning information for the congruous target words. Moreover, it is likely that participants semantically integrate the target’s word meaning with the preceding phrase. Previous research has found modulation of central theta associated with semantic processing in the classic N400 time window (note that this is distinct from the earlier and more posterior theta peaking at around 250 associated with lexical processing, as discussed above) (Hagoort et al., 2004; Lam et al., 2016). The N400 is an ERP indicative of semantic processing of single words, as well as the meaning integration of these words into context. This context can be with preceding words as well as with semantic knowledge about the world (Kutas and Federmeier, 2011). The N400 is attenuated for words congruous with the context. For the changes in theta power (3–5 Hz) related to semantic processing, we focused our analyses at Cz in the N400 time window (400–500 ms). We also investigated modulations of late alpha (9–11 Hz) at Cz 500–1000 ms following word onset, since later occurring alpha activity over central regions has been suggested to facilitate access to stored information and be related to semantic processing demands (Klimesch, 2012). Additionally, changes in beta over left inferior frontal gyrus has been associated with integration of information (Wang et al., 2012) guiding us to analyze beta activity (15–20 Hz) at Cz 500–1000 ms following word onset.
2.3.1. MCI convertors show an attenuated central theta increase at 400 ms and a late beta suppression related to processing of semantic congruency

For the differential effect between congruent and incongruent words at Cz, we found a transient increase in theta activity 400–500 ms following word onset (F(1,33) = 17.36, p = 0.001) (Fig. 3). This congruency effect differed between the groups (Word Type x Group interaction: F(2,33) = 4.8, p = 0.014). The theta congruency effect for the MCI non-convertors significantly differed from the MCI convertors (MCI non-convertors: 37% vs MCI convertors: 1%, p = 0.014). For the healthy elderly control and MCI convertor groups, a similar theta congruency difference was revealed but was not statistically significant (healthy controls: 23% vs MCI convertors: 1%, p = 0.205). There was no difference in the theta congruency effect between the MCI non-convertors and healthy elderly controls (p = 0.82). For alpha (9–11 Hz), the effect of semantic congruency did not differ between groups (Word Type x Group interaction: F(2,33) = 0.855, p = 0.44).

There was a suppression of beta activity at Cz for congruent compared to incongruent words, occurring 0.5 s after word onset and lasting until 1 s post onset (Fig. 3). While the degree of beta suppression (15–20 Hz) was not significantly different for congruent vs. incongruent words over all groups (F(1,33) = 1.37, p = 0.250), there was a clear Word Type by Group interaction (F(2,33) = 5.58, p = 0.008). Post-hoc analysis revealed that MCI convertors had significantly more beta suppression to congruent than incongruent words, unlike the MCI non-convertors (MCI convertors: −10% vs MCI non-convertors: 3%, p = 0.017) or the healthy controls (MCI convertors: −10% vs controls: 1%, p = 0.038).

2.3.2. Cross-frequency interactions for congruent words are only present for the healthy elderly controls, not for either MCI patient group

As mentioned above, one interpretation of the semantic congruency effect is that the meaning of the target word is integrated with the to-be-integrated information retrieved from temporal regions (associated with modulations of theta). The dynamic interaction between unification of information (frontal regions) and lexical retrieval processes (temporal regions) is thought to be critical during language comprehension (Hagoort, 2013; Wang et al., 2012).

Using repeated measures ANOVA with the factors of word-type and group, we determined if there was a difference between congruent and incongruent words with regards to the theta and alpha/beta coupling,
and if this difference was observed in each group. In addition, rather than just testing for differences in cross-frequency correlations between condition and groups, we directly tested if the correlations were significant within each groups. This approach for looking at differences in cross-frequency interactions between patient populations and control groups has previously been successfully used (Mazaheri et al., 2014a).

The power of the theta activity 0.4–0.5 s after word onset, at electrode Pz was correlated (Spearman) with the late alpha/beta (10–20 Hz) power suppression at frontal electrode Fz on a trial-by-trial basis and converted to z-values with Fischer’s r-to-z transform. Here we move the spectral components of the averaged ERP, from the epoch captures both non-phase locked (i.e., induced) activity as well as the spectral representation of the word locked ERP. It is therefore possible that the increase in oscillatory activity observed after word onset could simply be due to ERPs locked to word onset. One approach used in previous work (Cacace and McFarland, 2003) has been to remove the spectral components of the averaged ERP, from the ‘total’ spectra measured in each single trial. This approach is not without caveats (Mazaheri and Picton, 2005) with one being that the latency variability in the ERP can underestimate the spectrum of the ERP on a single trial. Nonetheless, we performed this subtraction (Fig. 5) and found that the majority of the theta increase in the 0–0.5 s window after word onset is non-phase-locked. Thus, it is unlikely that ERP differences for the early word responses can account for the oscillatory differences we observed between the healthy and patient groups.

2.4. Word onset induces mostly non-phase-locked increases in theta activity

Obtaining time-frequency representations of power in each EEG epoch captures both non-phase locked (i.e., induced) activity as well as the spectral representation of the word locked ERP. It is therefore possible that the increase in oscillatory activity observed after word onset could simply be due to ERPs locked to word onset. One approach used in previous work (Cacace and McFarland, 2003) has been to remove the spectral components of the averaged ERP, from the ‘total’ spectra measured in each single trial. This approach is not without caveats (Mazaheri and Picton, 2005) with one being that the latency variability in the ERP can underestimate the spectrum of the ERP on a single trial. Nonetheless, we performed this subtraction (Fig. 5) and found that the majority of the theta increase in the 0–0.5 s window after word onset is non-phase-locked. Thus, it is unlikely that ERP differences for the early word responses can account for the oscillatory differences we observed between the healthy and patient groups.

Fig. 4. The trial-by-trial cross-frequency coupling between the late theta increase (at the central parietal electrode Pz) and the late alpha/beta suppression (at Fz following congruent versus incongruent words) distinguishes healthy controls from both MCI converters to Alzheimer’s disease and non-convertors. We correlated (Spearman) the theta power (3–5 Hz, 0.4 to 0.5 s) at Pz with the alpha/beta (10–20 Hz, 0.5–1.0 s) power at frontal electrode Fz on a trial-by-trial basis, and then converted to z-values with Fischer’s r-to-z transform. The correlation of theta and alpha/beta in each individual is illustrated, separately for congruous and incongruous words. For the healthy participants, there was an anti-correlation (i.e. below zero) between the late theta increase and late alpha/beta suppression following the congruous words which was not present to incongruous words. There was no anti-correlation for congruous words in either MCI group, nor was there a significant difference in correlation values between congruous and incongruous words in either MCI group. The thick coloured lines represent the difference in mean cross-frequency correlation between the congruent and incongruent words, whereas the dots and thin lines represent individual participants. (For interpretation of the references to color in this figure legend, the reader is referred to the online version of this chapter.)

Fig. 5. Word onset induces mostly non-phase-locked changes in theta activity. The amplitude envelope of the ‘total’ theta activity locked to onset of the word (obtained at a single trial level and averaged) is represented in the dark-line. The amplitude envelope of the phase-locked theta activity measured through the time-frequency transformation of the ERP locked to the onset of word is represented as the grey-line. The subtraction of the phase-locked theta activity from the total theta activity shown in the red-line should consist of primarily non-phase-locked change (although see Mazaheri and Picton, 2005, for caveats of this assumption). We thus found the majority of the increased theta activity in the 0–0.5 s window after word onset to be primarily non-phase locked (compare the red vs grey lines).
Our results by selecting individual electrodes or choosing time intervals, these results taken together suggest that not only are we not biasing relative to MCI non-convertors in a cluster of electrodes consisting of: 0.548 s after word onset was attenuated in the MCI convertors group relative to MCI non-convertors (Monto Carlo estimated $p < 0.009, \text{Fig. 6A}$). For semantic processing (congruent minus incongruent words), theta activity between 0.35 s and 0.548 s in a cluster (high-lighted with $\ast$) consisting of left temporal, left occipital as well as midline electrodes (T5, O1, Cz, and Pz) was attenuated in the MCI converters group relative to MCI non-convertors (Monto Carlo estimated $p < 0.0129$).

**2.5. Investigating word onset differences in theta activity between MCI groups at all time points and electrodes**

In addition to using a-priori defined electrode locations and time intervals of interest, we also utilized a mass-univariate data driven approach looking at all electrodes and time points from 0 to 1 s after word onset (while correcting for multiple comparisons, i.e. controlling familywise error rates). This was to ensure that we were not overlooking any other differences between MCI converters and MCI non-convertors, or biasing our results through our a-priori choices. For lexical processing (congruent and incongruent words collapsed), theta activity between 0.252 s and 0.752 s after word onset was significantly attenuated in the MCI convertor group relative to the non-convertor group in a cluster of electrodes (highlighted with $\ast$) consisting of: T5, O1, O2, Pz, and Cz (Monto Carlo estimated $p < 0.009$). (B) For semantic processing (congruent minus incongruent words), theta activity between 0.35 s and 0.548 s in a cluster (highlighted with $\ast$) consisting of left temporal, left occipital as well as midline electrodes (T5, O1, Cz, and Pz) was attenuated in the MCI converters group relative to MCI non-convertors (Monto Carlo estimated $p < 0.0129$).

**2.6. Associations between EEG measures and language and verbal memory abilities (assessed by neuropsychological testing)**

In exploratory analyses across all subjects, we found that the late theta increase to either congruent or incongruent words correlated significantly with performance on the animal fluency test, in which participants named as many different animals as possible within one minute (Table 1A). Boston Naming Test (BNT) scores tended to correlate with the early lexical theta activity ($\rho = 0.33$, uncorrected $p = 0.048$). Two EEG oscillatory measures were strongly correlated with learning and memory on the California Verbal Learning Test (CVLT (Delis et al., 1987)). Both the change in the alpha suppression from 1st to 3rd presentation, as well as the strength of the anti-correlation between late parietal theta and frontal alpha-beta power, were correlated with learning, Delayed Recall and Recognition (Table 1B).

**2.7. Classification analysis**

In practice, biomarkers need to be applied at the individual level. Accordingly, we performed a leave-one-out classification to assess discrimination accuracy. Importantly, criticisms of over-fitting can be levelled even if cross-validation is applied, since one can effectively fish in the choice of classifier or classifier hyper-parameters (Skocik et al., 2016). To guard against such problems, we just applied two very standard classifiers: Linear Discriminant Analysis (LDA) (Balakrishnama and Ganapathiraju, 1998) and Support Vector Machine with Gaussian kernels, i.e. radial-basis functions (SVMrbf) (Suykens and Vandewalle, 1999). Linear Discriminant Analysis is probably the most basic classifier one could use, which places a single hyperplane in the data space. We also wanted to have the capacity to classify in the presence of linear inseparability, and the most standard classifier to do this is a Support Vector Machine with Gaussian kernels.

We trained both LDA and SVMrbf classifiers to distinguish MCI to Alzheimer’s converters from the other two groups – MCI stable and

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**Table 1A**

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**Table 1B**

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<td>CVLT list A 1–5</td>
<td>0.51</td>
<td>0.001</td>
<td>–</td>
</tr>
<tr>
<td>CVLT long delay cued recall</td>
<td>0.58</td>
<td>0.004</td>
<td>–</td>
</tr>
<tr>
<td>CVLT discriminability</td>
<td>0.48</td>
<td>0.003</td>
<td>–</td>
</tr>
</tbody>
</table>

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$^d_p < 0.01.$

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In exploratory analyses across all subjects, we found that the late theta increase to either congruent or incongruent words correlated significantly with performance on the animal fluency test, in which participants named as many different animals as possible within one minute (Table 1A). Boston Naming Test (BNT) scores tended to correlate with the early lexical theta activity ($\rho = 0.33$, uncorrected $p = 0.048$). Two EEG oscillatory measures were strongly correlated with learning and memory on the California Verbal Learning Test (CVLT (Delis et al., 1987)). Both the change in the alpha suppression from 1st to 3rd presentation, as well as the strength of the anti-correlation between late parietal theta and frontal alpha-beta power, were correlated with learning, Delayed Recall and Recognition (Table 1B).
controls. We did this on each relevant variable in our data set, and on combinations of these variables. The relevant variables were theta power during lexical retrieval (window from 0 to 0.5 s after word onset) at electrode Pz for the Congruent and Incongruent words separately, as well as theta power during semantic processing at Cz (0.4–0.5 s after word onset) for Congruent and Incongruent words.

We will present results for all four condition-electrode combinations in order to assess how classification performance changes across variables. First, we did also explore classifying with combinations of variables, i.e. multiple predictors. However, there was no apparent improvement in classification performance. We report sensitivity (i.e. hit rate), specificity (i.e. correct reject rate) and d-prime (i.e. overall discriminability).

The results of our classification analyses at the individual level are outlined in Table 2. Discriminability varies considerably across condition. However, the electrode-condition combination that has the most prior precedent in the literature is Cz incongruent. For example, Hagoort et al. (2004) identified high theta at Cz for a semantically incongruent condition. Consistent with this, it is the incongruent condition at Cz at which we obtain the greatest discriminability, with the highest linear discriminant d-prime of 1.752 and the highest support vector machine d-prime of 2.51. Interestingly, both the LDA and SVM have very high specificity for this condition (both 0.952) and the increase in d-prime gained with the SVM arises from a large increase in sensitivity (from 0.533 to 0.8). The practical implication of this is that the MCI-converter group may contain two subgroups with different patterns. This will be the subject of future investigations.

3. Discussion

In the current study, we investigated the neurophysiological differences during a language comprehension task among MCI patients who later go on to convert to Alzheimer’s disease (AD) within 3 years (MCI converters), MCI patients who remained stable (MCI non-converters) and healthy elderly controls. We focused on the oscillatory changes in the EEG induced by the visual presentation of words, for which the semantic context was established by preceding category descriptions. Two critical features of language comprehension are the rapid access to the form and meaning of words (i.e., lexical processing), and the rapid integration of this information with meaning set up by the preceding linguistic context and world knowledge (i.e. semantic, or meaning, integration) (Hagoort et al., 2004). In addition, lexical processing is facilitated when words are repeatedly presented via learning and implicit memory.

We found that lexical retrieval (i.e., lexical and semantic access) during a word processing task in individuals with MCI who progressed to Alzheimer’s disease elicited significantly diminished theta activity (3–5 Hz) relative to healthy controls and MCI non-converters. Theta attenuation thus provides an oscillatory signature for distinguishing those who will convert from MCI to AD. Furthermore, individuals with more severe attenuation in this lexical theta activity tended to have poorer lexical retrieval abilities as measured by the BNT score.

Additional analyses reveal that later alpha suppression (9–11 Hz), which also followed first presentation of the target words, was attenuated for the second and even more so for the third repetition in healthy elderly controls, but not for either of the MCI groups. Thus alpha suppression attenuation provides a useful oscillatory signature for both the MCI converters and MCI non-converters alike, distinguishing both MCI patient groups from controls. This loss of the normal attenuation of alpha suppression as words repeat, may reflect an impairment in both implicit and explicit memory processes. The degree of alpha suppression attenuation was strongly correlated with CVLT measures of learning and declarative memory in this study, arguing that it reflects or is important for declarative memory.

MCI converters to Alzheimer’s disease also revealed a different oscillatory signature for processing semantically congruent words. We found that, whereas a central theta increase associated with word congruency was attenuated in MCI converters, this group exhibited a more pronounced late beta suppression over the frontal cortex. However, we did not observe any significant cross-frequency connectivity between theta and alpha/beta power changes in the MCI converters, or in the MCI non-converters. This coupling, indicative of the complex interplay between retrieving linguistic information and binding linguistic information, was only present for congruous words in the healthy controls, and also demonstrated robust correlations with verbal learning and memory on the CVLT. We discuss each of these findings in more detail below.

During lexical processing, a reader accesses a range of different kinds of information about the word in the mental lexicon, including phonological, morphological and syntactic information, and the word’s semantic meaning. The onset of a word has previously been found to induce an early increase in theta activity over posterior and temporal regions, related to the process of lexical retrieval (Bastiaansen et al., 2005; Bastiaansen et al., 2008; Hagoort et al., 2004). In our study, changes in the theta band associated with the lexical processing of single words served as a conversion marker for MCI patients, suggesting MCI converters have difficulties with this basic aspect of language comprehension.

Repeated presentation of words typically results in facilitated lexical processing, given that implicit memory (and possibly also declarative memory) are activated upon first presentation of that word (Schacter et al., 1993). We observed in the healthy elderly group that a late alpha suppression diminished with each repetition of the word. This repetition effect was absent in both MCI groups. Latent alpha suppression at posterior sites has been associated with active processing of sensory information (Pfurtscheller, 2001) and allocation of resources according to processing demands (Foxe et al., 1998). The reduction in the amount of alpha suppression after each word repetition in the healthy controls suggests the emergence of implicit and/or explicit memory processes, which translates to less resource allocation required after each subsequent presentation. The absence of such memory processing signatures for repeated words in both MCI groups could represent a mediating factor of their impairments in verbal learning and memory.

For semantic processing, previous research has found modulations of central theta activity in the classic N400 time window (Hagoort et al., 2004; Lam et al., 2016). This increase in theta activity has been interpreted as indexing the retrieval of semantic information (Bastiaansen et al., 2005; Bastiaansen et al., 2008). Attenuation of this central theta in the MCI converters may reflect a breakdown in the brain networks involved in semantic processing. Our correlational analyses, which found that induced central theta activity to either congruent or incongruent words were both predictive of higher semantic fluency, provides even stronger evidence that central theta activity is a critical neurophysiological substrate for semantic processes (cf., Brickman et al., 2005). MCI converters also showed an increased beta suppression (15-

### Table 2

Results of leave-one-out classification analysis, to assess discriminability at the single participant level. Each row is a classification of individual’s theta power during the lexical (Pz, 0–0.5 s after word onset) and semantic stage (Cz, 0.4–0.5 s) of word processing (congruent or incongruent) words for a particular classifier (Linear Discriminant Analysis (LDA) or Support Vector Machine (SVM)).

<table>
<thead>
<tr>
<th></th>
<th>Sensitivity</th>
<th>Specificity</th>
<th>d prime</th>
</tr>
</thead>
<tbody>
<tr>
<td>Pz cong LDA</td>
<td>0.67</td>
<td>0.81</td>
<td>1.30</td>
</tr>
<tr>
<td>Pz incong LDA</td>
<td>0.6</td>
<td>0.81</td>
<td>1.13</td>
</tr>
<tr>
<td>Cz cong LDA</td>
<td>0.73</td>
<td>0.86</td>
<td>1.69</td>
</tr>
<tr>
<td>Cz incong LDA</td>
<td>0.53</td>
<td>0.95</td>
<td>1.75</td>
</tr>
<tr>
<td>Pz cong SVM</td>
<td>0.6</td>
<td>0.81</td>
<td>1.13</td>
</tr>
<tr>
<td>Pz incong SVM</td>
<td>0.69</td>
<td>0.76</td>
<td>0.96</td>
</tr>
<tr>
<td>Cz cong SVM</td>
<td>0.6</td>
<td>0.90</td>
<td>1.56</td>
</tr>
<tr>
<td>Cz incong SVM</td>
<td>0.8</td>
<td>0.95</td>
<td>2.51</td>
</tr>
</tbody>
</table>

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20 Hz) for semantic congruency. Beta oscillations have been linked to binding of linguistic information, or unification (Wang et al., 2012), see review (Weiss and Mueller, 2012). In the current language comprehension task, we assume that the meaning of the target word is integrated with the meaning set-up by the preceding phrase. Following this assumption, one would expect that the suppression of beta activity (often found to be localized over the frontal cortices) is involved in semantic unification operations, with the to-be-integrated information retrieved from temporal regions (associated with modulations of theta). Whereas we did not observe any differences in beta suppression between congruent vs. incongruent words in the healthy elderly group and MCI non-convertors, we did find that MCI converters had significantly more beta suppression to congruent words. We do not believe however that we should interpret this finding for the MCI converters as an index of successful binding of semantic information, given the clear deficits in the initial retrieval of lexical and semantic information, and our cross-frequency connectivity findings across the groups, which we will discuss next.

The dynamic interaction between unification of information (frontal regions) and lexical retrieval processes (temporal regions) which is so important to language comprehension can be captured by examining the cross-frequency connectivity between theta (i.e., retrieval of information) and alpha/beta power changes (i.e., binding of information). In the healthy control group, semantic processing of congruent words induced significant coupling between the posterior theta increase and frontal alpha/beta suppression, which we interpret as reflecting communication between brain regions to obtain an integrated meaning representation of the target with the preceding semantically congruent linguistic context in memory. Although the MCI non-convertors showed intact oscillatory signatures of retrieving lexical and semantic information, the coupling effect was absent in this MCI group, suggesting intact retrieval processing but a breakdown in the binding process. Moreover, the MCI converters did not show this coupling, which, taken together with the anomalous oscillatory signatures of retrieval of information, suggest that the MCI converters have difficulties with the retrieval of single word meaning as well as binding multiple words. Our finding that the strength of coupling between parietal theta and frontal alpha-beta is strongly correlated with learning and delayed memory on the CVLT suggests there may be a fundamental neurophysiological mechanism for memory ‘binding’ and integration of a complex event into a single representation and/or memory trace. We also found significant correlations with verbal learning and memory for both the changes in the alpha suppression from 1st to 3rd presentation. Both of these EEG abnormalities may be responsible for the memory deficits that characterize amnestic MCI. Further replication in other cohorts is recommended and future studies would benefit from a focus on the degree to which these oscillatory changes predict trial-by-trial learning.

We should note that posterior theta oscillations do not have a role tied exclusively to language processes, but also to memory recollection (Addante et al., 2011). A future investigation is needed to examine how the aberrant pattern of theta activity in the MCI patients corresponds to performance domains other than word processing. We hypothesise that the attenuated theta response we observe during the various stages of word processing is not solely tied to the language domain, and likely reflects the subtle breakdown of the neural architecture involved in several domains of cognitive processing, including recollective memory. Interestingly, work by Addante et al. (2011) has found that theta activity prior to the onset of a stimulus played a critical role in predisposing memory recollection. While in the current study, we focussed on the post-word change in theta activity, theta activity is also present in the EEG signal at rest. One future endeavour could be investigating if the pre-stimulus theta, particularly its variability in time, could be another factor accounting for variance in MCI conversion to dementia.

A number of previous studies have investigated word processing in amnesic patients using event-related potentials (Addante et al., 2012; Addante, 2015). The (time-domain) event related averaging approach (i.e. ERPs) is only sensitive to brain activity that is both time-locked and consistently phase-locked to the onset of an experimental event, such as a word. In contrast to this time-domain event-related averaging approach, looking at the spectral changes in the EEG with respect to experimental events affords the possibility to look at brain activity that is time-locked, but not necessarily phase locked to an experimental event. In addition, the neurophysiological origins of oscillatory activity might be better defined than the event-related potentials, particularly for the late occurring sustained potentials often linked to higher cognitive processes such as semantic processing and working memory (please refer to (Mazaheri and Jensen, 2008) for discussion on this).

In summary, our findings suggest that MCI patients who progress to Alzheimer’s disease, as a group, have subtle deficits in language comprehension that are revealed by oscillatory changes in the EEG during word processing. Our results suggest that the breakdown of the brain network subserving language comprehension could be foretelling of the emergence of AD as well as the underlying memory failures observed in these patients in the prodromal stage of the disease. Additionally, our leave-one-out classification analysis, with a highest d-prime of 2.51, suggests that our findings may contain a workable early-stage biomarker for individuals facing impending conversion from MCI to AD. Future work with larger sample sizes will need to be undertaken to fully assess the effectiveness of our findings in providing individual-level markers. Another potential direction for future research would be to determine if such anomalies are also present in MCI individuals who later convert to other forms of dementia (e.g., Vascular, Frontotemporal or Lewy Body), testing if these EEG findings enable stratification for other forms of dementia.

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Conflicts of interest

All the authors declare no conflict of interest.

References


