Aberrant prefrontal beta oscillations predict episodic memory encoding deficits in schizophrenia

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**Abstract**

Verbal episodic memory is one of the core cognitive functions affected in patients with schizophrenia (SZ). Although this verbal memory impairment in SZ is a well-known finding, our understanding about its underlying neurophysiological mechanisms is rather scarce. Here we address this issue by recording brain oscillations during a memory task in a sample of healthy controls and patients with SZ. Brain oscillations represent spectral fingerprints of specific neurocognitive operations and are therefore a promising tool to identify neurocognitive mechanisms that are affected by SZ. Healthy controls showed a prominent suppression of left prefrontal beta oscillatory activity during successful memory formation, which replicates several previous oscillatory memory studies. In contrast, patients failed to exhibit such a left prefrontal beta power suppression. Utilizing a new topographical pattern similarity approach, we further demonstrate that the degree of similarity between a patient’s beta power decrease to that of the controls reliably predicted memory performance. This relationship between the observed encoding impairments in SZ in verbal memory.

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1. Introduction

In the healthy brain, the cognitive architecture is flexible. Such flexibility is regulated by synchronous and non-synchronous neural oscillatory activity within and between brain networks (Buzsáki and Draguhn, 2004; da Silva, 2013). Therefore, complex cognitive tasks rely on the fine-grained balance between synchronized and desynchronized brain networks. An increasing number of studies show that this balance between synchronized and desynchronized networks is affected in patients with schizophrenia (SZ), who show markedly aberrant patterns of oscillatory activity in the lower (i.e., delta, theta, alpha, beta) and higher (gamma) frequency ranges (Uhlhaas and Singer, 2015). Arguably, oscillations reflect basic neurocognitive operations with different frequency bands representing different univocal neural signatures, i.e., spectral “fingerprints”, of these operations (Siegel, Donner, and Engel, 2012). A promising approach for SZ therefore is to identify the affected neurocognitive operations by isolating spectral fingerprints in specific tasks in patients, ultimately identifying targets for brain stimulation or pharmaceutical interventions. This approach has proven to be quite successful in the domains of perception (i.e., feature binding in Gestalt perception, Grützner et al., 2013) and working memory (e.g., Griesmayr et al., 2014; Haenschel et al., 2009). Grützner et al. (2013) investigated the ability of healthy and schizophrenic participants to bind fragmented black and white parts into a coherent object representation, i.e., face perception. The authors observed a pronounced reduction in the gamma band power (higher than 60 Hz) in schizophrenic patients, when compared to healthy controls, in perceiving upright, but not inverted faces, unravelling aberrant high gamma power as the spectral fingerprint of lower-order processes deficits in schizophrenia. In a similar vein, Haenschel et al. (2009) showed aberrant alpha and beta cortical activity during encoding and aberrant theta and gamma bands during maintenance and retrieval of visual working memory content in a visual
delayed match to sample task; Griesmayr et al. (2014) extended these results with a similar task by showing aberrant increase of frontal theta band during encoding phase of to-be-manipulated working memory content. However, no study has so far investigated the oscillatory fingerprints of episodic memory formation deficits in SZ. This is quite remarkable given that episodic memory is particularly strongly affected in SZ (Schaefe, Giangrande, Weinberger, and Dickinson, 2013). We here aim to fill this gap and test whether specific oscillatory markers can explain cognitive deficits in verbal episodic memory encoding in SZ.

In the healthy brain, encoding of episodic memories is typically reflected by a well-orchestrated interplay of synchronized theta/gamma activity in the hippocampus and desynchronized low frequency activity in the neocortex (<30 Hz) with the first associated to information binding and the latter associated to the representation of the content of such information (Hanslmayr, Staresina, and Bowman, 2016). Within this framework, the active engagement of specific cortical modules is mediated by a pronounced desynchronization in the low frequencies which allow these modules to represent information in a rich manner, at least for verbal material (Hanslmayr, Staudigl, and Fellner, 2012). More specifically, while formation of non-verbal memories has been associated with cortical theta and gamma power increases as well as fronto-temporal phase-amplitude coupling (Mölle, Marshall, Fehm, and Born, 2002; Lega, Jacobs, and Kahana, 2012; Hanslmayr, Spitzer, and Bäuml, 2009; Osipova et al., 2006; Friese et al., 2013; Köster, Friese, Schöne, Trujillo-Barreto, and Gruber, 2014; Lega, Burke, Jacobs, and Kahana, 2016), a different picture arises concerning beta oscillations. Cumulating evidence revealed that when participants are required to intentionally encode a list of words (Hanslmayr and Staudigl, 2014; Hanslmayr et al., 2011; Hanslmayr et al., 2012) or to incidentally encode a list of words via accessing their semantic content (Hanslmayr et al., 2009, 2011; Fellner, Bäuml, and Hanslmayr, 2013) successful encoding of verbal material into episodic memory is reflected by a pronounced beta power decrease (13–20 Hz) in the left inferior prefrontal cortex (IFG), as revealed by MEG (Meeuwissen, Takashima, Fernández, and Jensen, 2011), simultaneous EEG-fMRI (Hanslmayr et al., 2011) and tMRS-EEG (Hanslmayr, Matuschek, and Fellner, 2014) studies. Left prefrontal beta power decreases have also been shown to reflect the level of processing of semantic material, hence stronger power decreases indicate deeper, i.e. semantic processing and therefore better memory performance (Hanslmayr et al., 2009). Taken together these results suggest that the oscillatory signature of memory formation varies with the neural processes involved during memory encoding, and that the beta power decreases in the left prefrontal cortex reflect a deeper (more semantic) level of processing, especially for verbal material. Interestingly, a recent TMS-EEG study suggested a link between aberrant (more semantic) level of processing, especially for verbal material. Results suggest that the oscillatory signature of memory formation varies

2. Methods

2.1. Patients

Twenty-four patients were initially recruited from the local psychiatric clinic (Regensburg, Germany). The data of four patients were excluded because too few trials remained after artifact rejection and after splitting trials into subsequently remembered (hits) and subsequently forgotten (misses), thus not allowing for a meaningful subsequent memory contrast (i.e. <13 trials; see Hanslmayr et al., 2009 for a justification for this trial number). The data of the first three patients were excluded because the paradigm was changed afterwards (i.e. lists were shortened because memory performance was too low); the data of two further patients were excluded because of incomplete datasets. The remaining sample consisted of 15 patients, all of whom suffered from schizophrenia of either the paranoid-hallucinatory (n = 10; ICD-10, F20.0) or disorganised subtype (n = 5; ICD-10, F20.1). The patients had a mean duration of illness of 66.6 months (SD, 71.02). 14 patients received atypical neuroleptics only; one patient received no medication at all. Mean chlorpromazine equivalent dosage (CPZ, Woods, 2003) was 559.63 mg/d (SD, 557.29). Two patients received antidepressant medication (selective serotonin reuptake inhibitors, or tetracyclic antidepressants). No patient had any neurological diseases, additional psychiatric diseases, or medication with tranquillizers. Current clinical symptoms were assessed with the positive and negative syndrome scale in all but one patient for whom this data is missing (PANSS, Kay, Fiszbein, and Opfer, 1987; N = 14; mean score, 73.6; SD, 14.3).

2.2. Controls

Twenty healthy volunteers served as control subjects, matched according to gender (4 females), educational level (M = 3 where 0: no graduation–4: university entrance diploma), handedness (all right-handed), amount of cigarettes smoked per day (M = 9.24, SD = 7.62), and premorbid IQ (M = 103.53, SD = 8.62) measured with the “Mehrfachwahl–Wortschatz–Intelligenztest” (Lehrl, 2005). Although healthy volunteers and patients were both groups of young adults, healthy volunteers were on average slightly younger than patients (M = 24.35, SD = 4.55; t(30) = −2.53 p = 0.02). None of the participants had any history of neurological or psychiatric diseases or medication of any kind. Three healthy volunteers were discarded from analysis because of incomplete data, leaving a sample of 17 full datasets of control participants.

Demographic and IQ data and the number of cigarettes smoked per day for the final sample of controls and patients are summarized in Table 1. All of the participants had normal or corrected-to-normal vision and were German native speakers. All gave written informed consent. The protocol was approved by the local ethical review board.

2.3. Stimuli and procedure

Stimuli were comprised of 80 words drawn from the MRC Psycholinguistic Database (Coltheart, 1981), translated into German. The words were split into 5 lists of 16 words each. The lists were matched according to word frequency (M: 52.95, SD: 51.12), number of letters (M: 5.36, SD: 1.15), syllables (M: 1.69, SD: 0.54), concreteness (mean, 542.9; SD, 42.5), and imageability (M: 563.24, SD: 32.3). Stimuli were presented in white font on a black background of a 17” computer screen with a refresh rate of 70 Hz. Participants performed a verbal long-term memory task, which was comprised of three phases: (i) encoding; (ii) distracter
task; and (iii) free recall. The task was programmed using Presentation (© Neurobehavioral Systems). In the encoding phase, 16 words were sequentially presented in white font on a black background. Each word was presented for 2.5 s, preceded by a fixation cross with a variable duration of 1.5–2.5 s. Participants were instructed to try to keep the presented words in memory as they will be required to recall them later. Half-way through the list a reminder was presented prompting the participant to try to keep the presented words in memory. After the encoding phase a distractor task was carried out for 30 s, requiring the participants to rate the attractiveness of faces. The purpose of the distractor task was to minimize the contribution of working memory. Thereafter, an instruction was presented asking the participant to try to recall as many words as possible. Participants were given 1 min to recall the 16 words. Then the next study-test cycle started with a new list of words. Overall, 5 such study-test runs were conducted. The structure of one study-test run is depicted in Fig. 1. Most subjects also participated in a separate attention paradigm which is reported elsewhere (Hanslmayr et al., 2013).

2.4. EEG recording and analysis

The EEG was recorded from 62 channels, distributed equidistantly over the scalp and placed on an elastic cap (EasyCap) referenced to Cz. The EEG was re-referenced offline to the average reference. One additional external electrode was placed below the left eye to record vertical EOG. Impedance of the electrodes was kept below 20 kΩ. EEG and vertical EOG signals were amplified between 0.1 and 100 Hz with a notch filter at 50 Hz and digitized at a sampling rate of 500 Hz.

2.4.1. EEG—analysis

EEG data was analyzed with MATLAB (©The Mathworks, Munich, Germany) using the open-source FieldTrip toolbox (Oostenveld et al., 2011; http://fieldtrip.fcdonders.nl/) and in-house Matlab routines. EEG data were first segmented into epochs of 6 s, starting 2 s before word onset. The epoched data was visually inspected to discard large artifacts from further analysis. Remaining preprocessing steps included Independent Component Analysis for ocular artefacts correction and re-referencing to average reference. After removing trials which were contaminated by eye and muscle artefacts, an average of 30.5 trials (range: 13–52) remained for later remembered condition and 41.5 trials (range: 22–59) for later forgotten condition for the control subjects. For patients, an average of 19.4 trials (range: 13–39) and 52.8 (range: 38–61) for later remembered and later forgotten conditions remained for data analysis, respectively. The lower amount of trials that remained for the patients will be taken into account to ensure that any observed effect cannot be exclusively related to a worse signal to noise ratio driving results.

The segmented epochs (6 s) were then subjected to a wavelet transformation using Morlet wavelets (with a width of seven cycles) as implemented in fieldtrip to extract time-frequency characteristics optimally for each frequency. The frequency range of interest was 1–40 Hz with a time resolution of 50 ms. Power values were calculated for each single trial, and averaged across trials within the two conditions (later remembered, later forgotten). These power values were then transformed to relative power changes from baseline using an interval of −0.75 to −0.25 s before word onset (according to the formula: Power = Powerbaseline − Powertrial/Powertimebase as implemented in fieldtrip). We finally subtracted power values for later forgotten words from power values for later remembered words to compute the subsequent memory effect (i.e., SME) for each participant.

We confined statistical analysis to the left hemisphere in the frequency range including alpha and beta bands (8–20 Hz), according to our hypothesis. To control for multiple comparisons nonparametric randomization tests were conducted (using 1000 permutations) averaging the power data across all channels of the left hemisphere, along the time-window of word presentation (2.5 s). We performed this analysis separately for each group first (i.e. controls and patients) and then directly contrasted the SMEs between the two groups.

### Table 1
Demographic data of control participants and patients.

<table>
<thead>
<tr>
<th>Gender (m/f)</th>
<th>Handedness RH/LH</th>
<th>Education</th>
<th>Age (mean±SD)</th>
<th>Cigarettes per day (mean±SD)</th>
<th>Verbal IQ (mean±SD)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Controls (N = 17)</td>
<td>13/4</td>
<td>17/0</td>
<td>3</td>
<td>24.35 (4.55)</td>
<td>9.24 (7.62)</td>
</tr>
<tr>
<td>Patients (N = 15)</td>
<td>10/5</td>
<td>14/1</td>
<td>2.5</td>
<td>29.2 (6.26)</td>
<td>13.33 (12.28)</td>
</tr>
<tr>
<td>Stat. Difference</td>
<td>x² = 0.379, p = 0.7</td>
<td>x² = 1.17, p = 0.45</td>
<td>Z = 1.05, p = 0.29</td>
<td>t(30) = 1.062 p = 0.3</td>
<td>t(30) = 0.711, p = 0.48</td>
</tr>
</tbody>
</table>

The table reports mean values and standard deviations (in brackets) of age, cigarettes per day and verbal IQ. The number of male and female participants and their handedness are also reported. The values of the statistical tests in the last row of the table show that the two groups are matched for age, cigarettes smoked per day, verbal IQ, gender and handedness. Median values are reported for educational values.
2.5. Topographical similarity analysis

We computed a topographical similarity measure between patients and control participants in order to quantify to which degree the distribution on the scalp of the EEG pattern reflecting SME in a given patient diverged from control subjects. In order to obtain individual patient topographies, power values for each electrode were averaged over a 1 s long time-window centered at the point of maximal power decreases (later remembered – later forgotten) in each patient in a frequency window between 8 and 20 Hz (i.e., individual patients’ SMEs). Individual topographical distribution of power in this time-frequency ranges were used to calculate correlation indexes (i.e., Pearson’s $r$) between these individual topographies and the topography of power of the grand average of controls. Power values of the topography of controls were averaged over a one second long time window centered at the point of maximal power decreases as shown by $t$-values obtained with statistical analysis depicted in Fig. 2A (middle panel).

2.6. Source analysis

For source reconstruction, a frequency-domain adaptive spatial filtering dynamic imaging of coherent sources (DICS) algorithm (Groß et al., 2001), as implemented in fieldtrip was applied. As source model a standardized boundary element model was used, which was derived from an averaged T1-weighted MRI dataset (MNI, www.mni.mcgill.ca). Source analysis was carried out for the frequency range which revealed significant results on the scalp level.

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**Fig. 2.** Subsequent memory effects. A. Left panel shows controls’ subsequent memory effect (SME; 15–16 Hz, 0.5–1.5 s), dots represent significant cluster of electrodes, blue color indexes power decreases for later remembered words compared to later forgotten ones. The time-frequency plot shows significant SMEs ($t$-values) averaged over all electrodes of the left hemisphere (central panel); the right panel shows the topography ($t$-values) of the beta SME. B. Time-course of the power decrease at 16 Hz for subsequently remembered (blue) and forgotten (red) words of healthy controls (left panel). Source localization of the SME observed at the scalp level in healthy controls (15 Hz, 0.5–1.5 s; MNI: $-60\ 21\ 6$; right panel). C. Time-course of the power decrease at 16 Hz for subsequently remembered (blue) and forgotten (red) words in SZ participants (left panel). Time course of the contrast between controls’ (blue) and patients’ (light blue) SMEs (right panel); light grey rectangle represents the time-window (0.4–1 s) where the difference between SMEs has been observed. Shaded areas represent mean standard error. Circles in the pictures of the electrode positions on the scalp represent the averaged electrodes of the left hemisphere where SMEs and the contrast between groups’ SMEs have been observed. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

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2.7. Control analysis

As previous findings also reported power increases associated to SMEs during verbal and non-verbal memory tasks in the gamma and theta frequency bands, we performed analysis in order to look at possible increases of gamma frequency bands (i.e. in 40–150 Hz frequency window) and in theta frequency band (i.e. 1–7 Hz).

2.8. Statistical analysis

Mean proportions of accurate recall words was averaged across blocks and lists and compared between patients and controls via an independent samples t-test. For correlations, Pearson’s correlations were calculated unless stated otherwise.

3. Results

3.1. Behavioral results

As expected, controls recalled significantly more words in the verbal memory task compared to patients diagnosed with SZ (see Fig. 1B; t(30) = 3.85, p = 0.001). On average controls recalled 44% (SD ± 13.7), whereas patients recalled only 28% (SD ± 8.87), even though there is one patient who performed >2 SD above the average (53.75%; see Fig. 1); no outliers have been registered in healthy participants. No significant correlations between memory performance and CPZ (r = 0.21; p = 0.24), duration of illness (r = 0.20; p = 0.25) or age of the patients (r = −0.17; p = 0.18) were obtained either with or without exclusion of the outlier, suggesting that these variables were not the main contributors to the observed memory deficits in SZ.

3.2. EEG results

In a first step, subsequent memory effects (SMEs) were analyzed for control participants. A cluster analysis over the whole scalp for control participants revealed a significant SME in a cluster of left fronto-temporal electrodes (pcorr = 0.023) in the beta frequency range (15–16 Hz) from 0.5 to 1.5 s after stimulus onset (Fig. 2A). This beta oscillatory SME was due to a power decrease for subsequently remembered compared to subsequently forgotten words (Fig. 2B), replicating previous findings (Hanslmayr et al., 2009, 2011, 2014; see also Kim, 2011 for a meta-analysis reporting consistent SMEs for verbal information). Based on this result, further analysis were restricted to electrodes in the left hemisphere for both groups. Brain oscillatory SMEs are plotted in Fig. 2 for both controls and patients. A significant SME was observed in control participants by contrasting oscillatory activity for subsequently remembered and subsequently forgotten trials in the whole window of word presentation in a frequency range between 8 and 20 Hz and considering all electrodes of the left hemisphere (Fig. 2A middle and right panels). Source level analysis revealed that this effect was mainly driven by sources in the left inferior frontal gyrus (MNI: −60 21 6; BA 45, see Fig. 2B). Overall, these results show a normal subsequent memory effect in the control participants, i.e. stronger beta power decreases in the left prefrontal cortex during successful verbal memory formation (Meeuwissen et al., 2011; Hanslmayr et al., 2009, 2011, 2014). In contrast, patients failed to show a significant SME (pcorr > 0.7; Fig. 2C). Moreover, a direct comparison in the left hemisphere of the beta SMEs between patients and controls in a 600 ms time-window (0.4–1 s) in a refined frequency-range between 10 and 18 Hz revealed a significant difference between patients and controls (pcorr < 0.05) showing that the beta SME in the left hemisphere for controls was stronger compared to patients (Fig. 2D).

3.3. Topographical similarity results

Although the above results suggest that, on average, patients lack a left frontal beta SME compared to controls, inspection of individual datasets revealed that there was quite some variance with some patients actually showing left frontal beta power decreases quite similar to those obtained in controls (see Supplementary Fig. 1). If these left frontal beta power decreases are functionally linked to the deficits in the patients’ memory performance, we should be able to predict the patients’ memory performance from their beta SME patterns. In order to formally test this hypothesis we computed a topographical similarity analysis between the SME patterns in patients and controls. To this end, we calculated correlation indexes (i.e., Pearson’s rs) between the individual patient SME topographies and the average SME topography obtained from the control subjects (Fig. 3A; see Materials and Methods) and correlated this similarity index with memory performance. For this correlation analysis the outlier in memory performance (see Fig. 1) was excluded (see Supplementary Fig. 1 for the individual patient topographies). A remarkably strong positive correlation was observed between the patients’ topographical similarity values and memory performance (r = 0.683, p = 0.004). In other words, the more similar the beta SME pattern of a patient was to the pattern observed in the control group, the better the patient performed in the memory task. The correlation results are depicted in Fig. 3. Critically, this correlation still holds when including the outlier and calculating Spearman’s correlation (p = 0.635 p = 0.005), and when controlling for several critical aspects such as medication (r = 0.667 p = 0.006) duration of illness (r = 0.669 p = 0.006), the alpha/beta peak frequency where the strongest power decrease was observed (r = 0.530 p = 0.031), age of patients (r = 0.639 p = 0.009) and the individual number of trials (r = 0.684 p = 0.005), which is lower for patients than controls, due also to the lower accuracy rate. It is important to report that individual patients’ SMEs did not correlate with patients’ memory performance (r = −0.06 p = 0.85), this result was in further support of the functional role of the topographical distribution of beta power decreases (i.e., in the frontal regions of the left hemisphere) in memory performance. Additionally, we also correlated topographical similarity with the negative scale of the PANSS, however, no significant correlation was obtained (r < 0.1; p > 0.5).

3.4. Theta and gamma frequency bands

We also tested for SMEs in the gamma or theta frequency bands. In line with previous findings showing gamma synchronization associated with incidental encoding of non-verbal rather than intentional encoding of verbal material we did not observe any gamma activity (min p = 0.67). We did not either observe any theta synchronization after Montecarlo permutation (min p = 0.3). These null-results with regards to theta and gamma are largely consistent with our and other previous SME studies reporting that alpha/beta power decreases are the most robust signatures of memory encoding in EEG (Hanslmayr et al., 2012).

4. Discussion

In the present study we show that deficits in episodic memory encoding of verbal material in patients with SZ are associated with aberrant left prefrontal beta oscillations. This conclusion is based on two findings (i) SZ patients did not show the typical left prefrontal subsequent memory effect compared to controls; (ii) the degree to which an individual patient’s memory was impaired was strongly related to the dissimilarity between their beta oscillatory memory pattern and that of controls. This latter finding is especially important because it suggests that aberrant left prefrontal beta oscillations reflect memory encoding deficits in SZ.

On a behavioral level, healthy participants could recall significantly more words than patients, which is in good agreement with a recent meta-analysis showing that episodic memory is especially impaired in
SZ (Schaefer et al., 2013). In the EEG, we found that successful encoding of verbal material in controls was indicated by a left inferior prefrontal beta power decrease, which replicates a number of previous studies (see Hanslmayr et al., 2012; Hanslmayr and Staudigl, 2014, for reviews). However, patients with SZ failed to show such a typical SME pattern, which confirms our initial hypothesis. This result is in good agreement with previous fMRI studies showing decreased BOLD signal in the left inferior prefrontal cortex for SZ compared to controls during memory encoding (Achim and Lepage, 2005), as well as with a previous TMS study showing a link between aberrant prefrontal beta oscillations in SZ and memory (Ferrarelli et al., 2012). Importantly, our study goes beyond these previous findings in measuring beta oscillations during a memory encoding task and linking these oscillations directly to memory impairments in SZ.

Left prefrontal beta power decreases have not only been linked to memory encoding, but also to semantic/conceptual processing of verbal material (Singh, 2012; Wang et al., 2012). Since processing words in a semantic/conceptual fashion has been shown to be a particularly efficient (i.e. deep) memory encoding strategy (Craik and Lockhart, 1972), left prefrontal beta power decreases likely indicate the depth of such semantic processing and therefore predict whether a given word is being later remembered or not (Hanslmayr and Staudigl, 2014). Indeed, instructing subjects to focus on semantic aspects of items during memory encoding leads to stronger left prefrontal beta power decreases, compared to less semantic encoding instructions (Hanslmayr et al., 2009). Given this background, our findings suggest that patients might not be able to process items in such a deep, semantic fashion as controls, thereby limiting their encoding abilities. This explanation fits with the fact that conspicuous semantic processing (i.e. disorganised speech, neologisms, etc.) is a core diagnostic criteria of SZ. For instance, previous studies attempted to explain how schizophrenic patients have difficulties in verbal communication. Patients with SZ show reduced engagement of brain areas involved in the semantic organization of speech and grammar articulation (Kircher et al., 2001; Kircher, Oh, Brammer, and McGuire, 2005) and abnormal hemodynamic activity in response to either semantic associations (Kuperberg, Deckersbach, Holt, Goff, and West, 2007) and in semantic decision tasks (Sommer, Ramsey, and Kahn, 2001). Psycholinguistic studies of semantic processing in schizophrenia have further disclosed difficulties of patients with SZ in using context or combinatorial linguistic information, such as syntactic rules, to interpret appropriately the meaning of single words or of the whole sentences (see Kuperberg, 2010a, 2010b for reviews). Impaired semantic processing in SZ is also indicated by decreased amplitudes of the N400 component of event-related potentials (Jackson et al., 2014), which is known to be strongly linked to semantic processing (Kutas and Hillyard, 1980; Kutas and Federmeier, 2011) and has been shown to strongly correlate with left prefrontal beta power decreases (Wang et al., 2012). Together, verbal memory encoding deficits, possibly due to impaired semantic processing, observed in SZ patients seem to be consistently associated to weaker left prefrontal beta power suppression. Therefore, although the sample size of the SZ group was small, the here presented results suggest a new target for therapeutic interventions tailored to alleviate fundamental problems in memory encoding in SZ. A previous combined rTMS-EEG study in healthy subjects has shown that left prefrontal beta oscillations can be modified via rhythmic transcranial magnetic stimulation (rTMS) and that this modulation affects memory encoding (Hanslmayr et al., 2014). An interesting future application of rTMS therefore is to develop protocols in order to suppress left prefrontal beta power and to increase memory encoding as a result thereof. Alternatively, transcranial alternating current stimulation could be used to target left prefrontal beta oscillations in a suppressive manner. Although these protocols are yet to be developed and tested in healthy subjects, our study might inspire the development of such neuro-stimulation protocols, which are being increasingly recognized as treatment options for psychiatric conditions (Downar, Blumberger, and Daskalakis, 2016). Moreover, oscillatory activity has been shown to depend on anatomical integrity and neurotransmitter systems, which are both abnormal in schizophrenia (Uhlhaas and Singer, 2010). Beta frequency band is supposed to be generated by neurotransmitter systems of glutamate, NMDA receptors and GABAa receptor activity (Traub, Bibbig, LeBeau, Buhl, and Whittington, 2004; Yamawaki, Stanford, Hall, and Woodhall, 2008). Thus, our results might stimulate the development of pharmacological intervention that target these neurotransmitters as aberrances in oscillatory activity can be quantified through non-invasive procedures such as MEG and EEG (Haenschel, Baldeuw, Croft, Whittington, and Gruzelier, 2000).

It is important to point out some potential caveats of our study, which were inevitably introduced by the clinical setting (i.e. patient availability, memory performance). For instance, the sample of patients was comprised of two different subtypes of SZ (i.e. 10 suffered from paranoid-hallucinatory and 5 from disorganised subtype). Furthermore the duration of illness and medication they were receiving at the time of data collection did vary considerably between patients; and patients were on average older than controls. However, arguably none of these aspects was the driving force in the observed correlation between beta similarity and memory performance as shown by the partial correlation analyses, further strengthening the argument that a lack of prefrontal beta suppression represents a core neurophysiological...
correlate of memory encoding impairments in schizophrenia. Finally, we cannot rule out a possible involvement of other oscillatory mechanisms which are not so readily accessible to non-invasive EEG (i.e. theta/gamma oscillations in the hippocampus), which is an important question for future studies to address.

5. Conclusions

We investigated the oscillatory signature of episodic verbal memory formation in groups of healthy and schizophrenic participants. In line with previous studies, we observed a beta desynchronization SME in the left inferior frontal gyrus in healthy controls. This ubiquitous signature of episodic memory formation was absent in SZ patients. However, patients who showed a more similar topographical pattern of beta desynchronization compared to that of healthy controls, also performed better in the memory task. Together, these findings demonstrate that aberrant left prefrontal beta oscillations are closely linked to the memory impairments in SZ in a verbal memory task. These findings might ultimately inspire the development of neuro stimulation and pharmaceutical therapies which target left prefrontal beta oscillations in order to improve memory in SZ.

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