Attention to language: Novel MEG paradigm for registering involuntary language processing in the brain

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**Abstract**

Previous research indicates that, under explicit instructions to listen to spoken stimuli or in speech-oriented behavioural tasks, the brain's responses to senseless pseudowords are larger than those to meaningful words; the reverse is true in non-attended conditions. These differential responses could be used as a tool to trace linguistic processes in the brain and their interaction with attention. However, as previous studies relied on explicit instructions to attend or ignore the stimuli, a technique for automatic attention modulation (i.e., not dependent on explicit instruction) would be more advantageous, especially when cooperation with instructions may not be guaranteed (e.g., neurological patients, children etc). Here we present a novel paradigm in which the stimulus context automatically draws attention to speech. In a non-attend passive auditory oddball sequence, rare words and pseudowords were presented among frequent non-speech tones of variable frequency and length. The low percentage of spoken stimuli guarantees an involuntary attention switch to them. The speech stimuli, in turn, could be disambiguated as words or pseudowords only in their end, at the last phoneme, after the attention switch would have already occurred. Our results confirmed that this paradigm can indeed be used to induce automatic shifts of attention to spoken input. At ~250 ms after the stimulus onset, a P3a-like neuromagnetic deflection was registered to spoken (but not tone) stimuli indicating an involuntary attention switch to them. The speech stimuli, in turn, could be disambiguated as words or pseudowords only in their end, at the last phoneme, after the attention switch would have already occurred. Our results confirmed that this paradigm can indeed be used to induce automatic shifts of attention to spoken input. At ~250 ms after the stimulus onset, a P3a-like neuromagnetic deflection was registered to spoken (but not tone) stimuli indicating an involuntary attention switch. Later, after the word-pseudoword divergence point, we found a larger oddball response to pseudowords than words, best explained by neural processes of lexical search facilitated through increased attention. Furthermore, we demonstrate a breakdown of this orderly pattern of neurocognitive processes as a result of sleep deprivation. The new paradigm may thus be an efficient way to assess language comprehension processes and their dynamic interaction with those of attention allocation. It does it in an automatic and task-free fashion, indicating its potential benefit for assessing uncooperative clinical populations.

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1. Introduction

Accurate assessment of patients with cognitive and neurological impairments is often a challenge even to the most experienced clinicians. More specifically, in evaluating language function – whether for the purpose of pre-surgical mapping, to assess a child’s development, effects of brain injury or to monitor therapy progress – one has always had to rely on behavioural observation or on a combination of behavioural/verbal responses with some neuropsychological tools (e.g., Baron-Cohen, Hoekstra, Knickmeyer, & Wheelwright, 2006; Benton, 1994; Folstein, Folstein, & McHugh, 1975; Shewan & Kertesz, 1980). This raises the problem of non-cooperative subjects. A brain-damaged individual may not be able to properly respond verbally because of collateral lesion-related deficits. A young child, particularly with a speech deficit, may not be willing or able to cooperate with those assessing his or her condition. A “locked-in” person may be conscious but does not have any means to give signs of this behaviourally due to a complete absence of motor control (Laureys, Owen, & Schiff, 2004; Ragazzoni, Grippo, Tozzi,
The spatial resolution is generally inferior to metabolic imaging (for reviews, see Pettigrew, Murdoch, Chenery, & Kei, 2004; Pulvermüller, Shtyrov, & Hauk, 2009). For this reason, fast neurophysiological imaging techniques like electro- or magnetoencephalography (EEG, MEG) have been used extensively in the cognitive neuroscience of language (see e.g., Friederici, 2002). These tools track mass neural activity (rather than blood metabolism) with millisecond precision, even though their spatial resolution is generally inferior to metabolic imaging (Dale & Halgren, 2001).

Still, how could one register the activity of neural networks underlying language function without having to apply linguistic tasks such as lexical decision or semantic judgements that require at least the subjects’ focussed attention and frequently involve their active participation? To address language function in cases when no active participation is possible, such tasks are clearly ruled out. To address this, a number of experiments used task-free methodologies successfully used in such studies is the so-called passive oddball paradigm, in which the subjects are presented with linguistic contrasts between frequent repetitive stimuli and unexpected rare ones without having to perform an overt task (for reviews, see Pettigrew, Murdoch, Chenery, & Kei, 2004; Pulvermüller & Shtyrov, 2006; Pulvermüller & Shtyrov, 2007b). As no attention or stimulus-related task is required (in fact, the volunteers are distracted from the auditory input), the recorded brain activation is considered to be automatic (Näätänen, Paavilainen, Rinne, & Alho, 2007). Furthermore, because the oddball response is elicited by contrasts between the frequent and rare stimuli, this allows for full control over acoustic factors, as the same acoustic contrasts can be incorporated into linguistically different contexts (Shtyrov & Pulvermüller, 2007b).

Indeed, a series of studies using this approach established it as a sensitive tool to study the neural correlates of linguistic access including a wide range of information types: phonological, lexical, semantic and syntactic (Shtyrov & Pulvermüller, 2007b). These have shown that linguistic processing can be activated in the brain – and, importantly, this activation can be recorded in MEG/EEG – without an explicit task or an instruction to focus on the speech input. This raises the possibility of assessing neural linguistic processes non-invasively in subjects who are unable to carry out an active experimental task.

One such obvious linguistic process, utilised routinely in human communication, is so-called lexical access, i.e., access to information(6,10),(995,993)

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It has been long known that unexpected sounds that strongly differ from their environment cause an automatic attention switch, which is what we suggest to use here to allocate attention to linguistic input without an overt instruction. Our suggested design is therefore based on presenting the subjects with a variety of non-linguistic stimuli (tones), sometimes replaced by spoken items, which, due to their marked acoustic differences from the tone stimuli, are bound to draw attention automatically. Such an automatic shift of attention is known to be reflected in brain activity in the form of the so-called P3a deflection (also known as ‘novelty P3’) in electro- or magnetoencephalogram, originating from stimulus-driven attention mechanisms (Friedman, Cycowicz, & Gaeta, 2001; Polich, 2007). The presence or absence of the P3a deflection could therefore be used to probe, on the one hand, stimulus salience in terms of automatic attention shifts, and on the other hand – and more importantly – the nervous system’s ability to process such a novel auditory event.

Still, once the putative attention shift has occurred, how could one test whether the linguistic stimuli are processed as such, not just as a striking acoustic deviance? Here, our previous studies mentioned above make a clear prediction: under conditions when attention is allocated to speech, a rare meaningless pseudoword elicits a larger negative-going1 response than a meaningful word as early as 100–200 ms, a phenomenon that is hypothesised to reflect the intensified processes of lexical search (for a review, see Shtyrov, 2010). This divergence is also predicted by the already mentioned neural model of word memory traces realised as strongly linked cell assemblies distributed across auditory perceptual and motor articulatory cortices that become active when presented with linguistic perceptual input (Garagnani et al., 2008). A similar pattern, albeit with a later latency (> 250 ms), is seen in active N400 studies that involve, for example, a lexical decision task on word/pseudoword stimuli (Friedrich et al., 2006). Therefore, the presence of such a lexical distinction in the brain response pattern could indicate language-specific processing of stimuli, not merely acoustic deviance aspects.

Obviously, the moment of word/pseudoword divergence must be distinct from the point of speech/non-speech divergence, to allow for attention shifts to occur, on the one hand, and to disentangle the initial effect of auditory novelty from the lexical effect of word-pseudoword contrast, on the other hand. We therefore suggest a stimulation design, in which the attention-modulating speech/non-speech contrast occurs in the very beginning of the stimulus, whereas the lexical contrast is postponed until the very end, to the last phoneme of the speech stimuli which can render the whole stimulus to be a meaningful word or a meaningless pseudoword.

In sum, our suggested paradigm amounts to presenting the subjects with a majority of non-speech stimuli, occasionally replaced (in a modification of an oddball design) with spoken items. Spoken items, in turn, consist of words and pseudowords that only become distinct in the very end, at their last phoneme. We predict: (i) a P3a deflection specific to speech stimuli, at latencies of ~250 ms relative to the stimulus onset, and (ii) a larger pseudoword than word response at 100–200 ms relative to the onset of the last phoneme. Such effects could be used to probe, on the one hand, the neural function of automatic auditory attention, and, on the other hand, the presence of neural processing of lexical information in the incoming speech. Additionally, (iii) all infrequent stimuli could elicit a mismatch negativity (MMN) response, indicative of automatic auditory discrimination (Nätänen et al., 2007). Aimed in principle at assessing language function in subjects/patients who are not able to cooperate with a more active experimental task, this novel paradigm is for the first time tested here, using high-density multi-channel MEG (generally considered as the most patient-friendly brain imaging technique), in healthy volunteers.

2. Methods

2.1. Subjects

Twenty healthy right-handed (handedness assessed according to Oldfield (1971)) native English-speaking volunteers (age 19–43, mean age 29.8, 5 females) with normal hearing and no record of neurological diseases were presented with non-speech tones and spoken English language stimuli. All subjects were paid for their participation. Ethics permission for the experiments was granted by the Cambridgeshire Local Research Ethics Committee (LREC, Cambridge, UK).

2.2. Stimuli

For stimulus presentation, we adapted the so-called ‘optimum’ or ‘multi-feature’ version of the oddball paradigm (Kujala, Tervaniemi, & Scherg, 2007; Näätänen, Pakarinen, Rinne, & Takegata, 2004), which, unlike traditional oddball designs, can accommodate multiple types of rare (“deviant”) sounds in order to use experimental time more efficiently. In this design, every second stimulus is recognised to be the frequent (“standard”) sound, while the remaining stimulus locations are randomly distributed between the different types of deviants. Our choice of stimuli was determined by the requirements to (1) have an overall majority of non-speech stimuli, (2) ensure the linguistic word and pseudoword spoken stimuli only differ in their end, and (3) balance acoustic and phonetic/phonological contrasts across spoken stimuli to control for low-level perceptual effects. This led to the following stimulation set-up (see also Fig. 1):

1/2 of all stimuli were frequent (standard) 330-ms tones composed of a 275 Hz sine-wave combined, in equal proportion, with its two first harmonics (i.e., 550 and 875 Hz, similar to spoken language, where the main, so-called “fundamental” frequency is complemented by further formants). This sound linearly increased in amplitude over the first 40 ms to reach 50 dB above hearing threshold, and similarly faded out in the last 40 ms, to avoid abrupt onset and offset. A further 1/8 of the stimuli were two non-speech deviants: a 20% longer duration deviant and a 20% higher-pitch frequency deviant, distributed equiprobably. Finally, a minority of stimuli were spoken items: two word (bite, pipe) and two pseudoword (“bipe,” “pise”) deviants, each having a probability of 1/12. Note that for the bi-starting stimuli, a [t]-ending results in a word, while the final [p]-sound leads to a pseudoword. The opposite is true for the p-starting stimuli. Consequently, acoustic-phonetic features are fully inter-balanced with respect to their contribution to the word-pseudoword distinction across the two pairs of spoken stimuli, meaning that any resultant lexicality effects are not subject to acoustic or phonological confounds. Such stimuli are known to generate robust lexical effects in event-related brain responses (Garagnani et al., 2009; Shtyrov & Pulvermüller, 2002; Shtyrov, Nikulin, & Pulvermüller, 2010b).

Furthermore, in stimulus production, physically identical recordings of initial consonant-diphthong CV fragments (bi- or pi-), obtained from digitally recorded speech of a native British English speaker and matched acoustically, were used in each combination of [p][t]-ending items. This design guaranteed that, whilst all spoken items were clearly different from the tones at the onset, they could only be fully identified at their last phoneme ([t] vs. [p]). These final phonemes were also made identical across the two combinations, using cross-splicing, to remove any associated acoustic confounds. To avoid co-articulation, they were taken from recordings of acoustically similar but not identical items (hype, high). This technique allowed us to control exactly the point in time when the phonetic contrast occurred and, consequently, when each spoken item could be identified. The initial consonant-diphthong parts of all spoken stimuli were equal to the standard tone both in length (330 ms) and in fundamental frequency (275 Hz). Together with their stop-consonant endings, all spoken stimuli were 485 ms in duration. The word-pseudoword divergence point was at 410 ms following an 80-ms silent closure typical of English stop-consonants and providing (1) an ideal place for cross-splicing the stop-consonants across stimuli for their optimisation, and (2) a silent interval that can be used for baseline-correction of neurophysiological responses related to the divergence point. To complete the matching process, all stimuli had identical peak amplitude. The analysis of the naturally-spoken stimuli  

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1 Negative polarity of this response is observed in EEG at midline electrodes (e.g., Shtyrov et al., 2010) when using average, nose or mastoid reference. In MEG, the notion of voltage polarity is absent, but the pattern can be well-observed as a larger field gradient in event-related fields (ERF) over the temporal lobes (Garagnani, Shtyrov, & Pulvermüller, 2009).

2 Here and throughout we use a linguistic convention an marking pseudo-words with an asterisk (*).
and their modifications were implemented in Cool Edit 2000 program (Syntrillum Software Corp., AZ, USA).

Following the recommendations for the optimal multi-feature stimulus sequence (Naätänen et al., 2004), each consecutive group of 12 stimuli contained all 6 deviants randomised in a different order within each group occupying, between them, every second position, the remaining positions being taken by the standard stimuli. A total of ~1350 stimuli were presented in such a pseudo-random sequence, with a jittered stimulus onset-to-onset interval (stimulus onset asynchrony, SOA) of 840–940 ms (mean 890 ms).

2.3. Magnetoencephalographic recording

The subjects were seated in light-weight magnetically-shielded room (IMEDECO AG, Hägendorf, Switzerland) and asked to ignore the auditory stimulation and to watch a self-selected silent video. The stimuli were delivered through non-magnetic ear-pieces connected via plastic tubes to acoustic transducers (Etymotic Research Inc, Elk Grove Village, IL, USA) outside the room. Throughout the experiment, the brain’s magnetic activity was continuously recorded using a 306-channel (102 magnetometer and 204 planar gradiometer sensors) Vector-view MEG system (Elekta Neuromag, Helsinki, Finland) with passband 0.10–330 Hz and 1 KHz sampling rate (i.e., sampling rate was ~50% higher than required by Nyquist–Shannon–Kotelnikov theorem, effectively meaning the data were oversampled for better anti-aliasing). Four magnetic coils were attached to the head and their position was digitized using the Polhemus Isotak digital tracker system (Polhemus, Colchester, VT, USA). To assist in off-line reconstruction of the head model, an additional set of points randomly distributed over the scalp was also digitized. During the recording, the position of the magnetic coils was continuously tracked (continuous head position identification, CHPI, taken every 200 ms), providing information on the exact position of the head in the dewar. To control for eye-movement artifacts, horizontal and vertical eye movements were recorded using electrooculogram (EOG) electrodes.

2.4. Data analysis

Raw MEG recordings were pre-processed using the temporal extension of the signal source separation (SSS) technique (Taulu & Simola, 2006; Taulu, Kajola, & Simola, 2004), as implemented in MaxFilter 2.0 software (Elekta Neuromag), in order to remove externally-generated noise from the data. Furthermore, we employed the same software to correct for head movements within the dewar and non-magnetic ear-pieces connected via plastic tubes to acoustic transducers (Etymotic Research Inc, Elk Grove Village, IL, USA) outside the room. To this end, 3D (2D sensor x 1D time) images were computed for each participant’s averaged MEG data by projecting the sensor positions onto a 2D grid of 32 × 32 pixels, and tiling across a third dimension corresponding to each time sample of 4 ms (see also Fig. 3). The data for each voxel within these images were fit by a General Linear Model that included all condition and subject effects. Several linear contrasts of the resulting parameter estimates were estimated across conditions, using a single pooled error whose non-sphericity was estimated by restricted maximum likelihood estimation (Friston et al., 2002). Each contrast resulted in a statistical parametric map (SPM), in which clusters of contiguous field-intensity variation exceeding 3000 femtotesla per centimeter, fT/cm) were identified for each condition. The analysis proceeded independently on the magnetometer and gradiometer data. For gradiometers, root mean squared (RMS) values were calculated for planar gradiometer pairs at each spatial location (204 gradiometers are grouped into 102 pairs, in which 2 sensors are orthogonal to each other), giving a scalar value representing field gradient magnitude at that location. RMS is computed as a square root of the sum of squared amplitudes recorded in the two sensors.

Significant differences between the standard and deviant tones, as well as between words and pseudowords were identified using a 3-dimensional sensor x time statistic parametric map using SPM5 software (Fil, London, UK). To this end, 3D (2D sensor x 1D time) images were computed for each participant’s averaged MEG data by projecting the sensor positions onto a 2D grid of 32 × 32 pixels, and tiling across a third dimension corresponding to each time sample of 4 ms (see also Fig. 3). The data for each voxel within these images were fit by a General Linear Model that included all condition and subject effects. Several linear contrasts of the resulting parameter estimates were estimated across conditions, using a single pooled error whose non-sphericity was estimated by restricted maximum likelihood estimation (Friston et al., 2002). Each contrast resulted in a statistical parametric map (SPM), in which clusters of contiguous suprathreshold voxels (using threshold corresponding to p < 0.001 uncorrected) were identified for each condition. The analysis proceeded independently on the magnetometer and gradiometer data. For gradiometers, root mean squared (RMS) values were calculated for planar gradiometer pairs at each spatial location (204 gradiometers are grouped into 102 pairs, in which 2 sensors are orthogonal to each other), giving a scalar value representing field gradient magnitude at that location. RMS is computed as a square root of the sum of squared amplitudes recorded in the two sensors.

Our main predictions are based on the amplitude changes in the MEG signal, which could be located in space and time using ERFs as recorded by MEG sensors. Further to this, in an attempt to delineate possible neural origins of ERP effects, their cortical sources were estimated using L2 minimum-norm current estimates that models the unique constellation of active neuronal current elements that models the recorded magnetic field distribution with the smallest amount of overall activity (Hämäläinen & Ilmoniemi, 1984; Hämäläinen & Ilmoniemi, 1994; Ilmoniemi, 1993). The L2 MNE does this by determining the combination of sources that have a minimum total power (calculated as an integral of squared current amplitudes).

MNE solutions were calculated for grand-average data, which has a benefit of substantially reduced noise and therefore an improved signal-to-noise ratio (SNR) which MNE solutions are highly sensitive to. A three-compartment (skull-skin-brain) boundary element model (BEM) was used for computational geometry and spherical coordinates. Cortical grey matter surface was triangalised (~8000 triangles, i.e., ~4000 vertices) using an MRI image of a standardised brain (Montreal Neurological Institute), as individual subjects’ MRIs were not available. The currents were restricted to the grey matter with a requirement of normal orientation of current dipoles to the surface (i.e., a single dipole at each vertex). CURRY 6 software (Compumedics Neuroscan, Hamburg, Germany) was used for these procedures. The drawback of such a grand-average source approach is the lack of individual source statistics; in this case, however, a solid statistical support was provided by the signal-level analysis whereas source solutions only have illustratory purposes.
Finally, to test potential applicability of our paradigm to neurocognitive deficits, we repeated this experiment in 7 healthy volunteers who were sleep-deprived for a minimum of 24 h prior to MEG recording. Sleep deprivation is often an associated feature, as well as a possible cause, of psychosis (Gove, 1970; Rimon, Fujita, & Takahata, 1986), which provides an opportunity to model a psychotic condition in healthy individuals by depriving them of normal night-time sleep (Kahn-Greene, Killgore, Kamimori, Balkin, & Killgore, 2007). Given the small number of subjects, their ERF data are provided here for demonstration only; no statistical or source analyses were possible at this stage.

3. Results

The recordings were carried out successfully, and all stimuli elicited magnetic auditory evoked responses (see Fig. 2). The differences between frequent standard and infrequent deviant stimuli were assessed using the 3D SPM analysis (Fig. 3). For the frequency deviant, we found significant effects \((p < 0.001\) uncorrected for height, \(p < 0.05\) corrected for extent) of frequency deviants that were maximal at a number of points. This mismatch-negativity-like effect (larger deviant than standard ERF) peaked at \(\sim 140\) ms after the onset; there also appeared to be a second phase of this response with a maximum at \(\sim 460\) ms. The difference between the frequency deviant and the standard response surpassed the significance threshold at 65–186 ms and 387–960 ms in magnetometers and 100–194 ms and 396–666 ms in gradiometers (see Fig. 3 for statistical parametric maps). No obvious P3-like deflection was found for the frequency deviant (Figs. 2 and 3).

Spoken deviant stimuli (pooled together) elicited ERFs that similarly diverged from the standard tone with the earliest peak approximately at 160 ms. Their mean ERF also had a second peak at \(\sim 540\) ms (see Fig. 2), which was unsurprisingly later than that of the frequency deviant given the stop consonant plosion release in the end of all spoken items. These differences between the spoken deviants and the standard reached significance, according to SPMs, at 120–200 ms and 470–800 ms in gradiometers and...
120–190 ms and 390–910 ms in magnetometers (see Fig. 3). However, in addition to the bi-phasic mismatch response that generally followed the same pattern as the frequency deviant, spoken stimuli caused an additional deflection of the opposite polarity which peaked at 240 ms (Fig. 2) and proved to be significantly different from the standard tone both in magnetometers (at 190–360 ms) and in gradiometers (230–370 ms). This deflection (which was also significant, with small variations in time, in all 4 individual spoken conditions) seems to correspond best to the P3a wave known from the earlier ERP research. As noted above, no corresponding deflection could be located for the frequency deviant.

Finally, we compared the activation between the rare deviant words and pseudowords. Their activation significantly diverged at 546–855 ms (this effect was seen almost exclusively in the gradiometers, Fig. 4). The peak of this difference, which clearly showed larger pseudoword than word activity, occurred in left temporal channels at 570 ms after the sound onset (Fig. 5), which corresponds to 160 ms after the divergence point (the onset of the last consonant when words and pseudowords could be uniquely identified).

A magnetic MMN-like deflection that uniformly differentiated between infrequent deviants and frequent standard tones (similar for frequency and spoken deviants) has been repeatedly shown to predominantly originate from temporal sources (Alho et al., 1996; Picton, Alain, Otten, Ritter, & Achim, 2000; Shtryov et al., 2005); the current data confirmed these previous findings (Fig. 6). However, sources of the P3a-effect elicited by rare spoken stimuli and of the larger pseudoword than word activity at 160 ms post-

identification have been scarcely investigated. To explore these effects further, we attempted to reconstruct their cortical sources using distributed minimum-norm source estimates. The P3a-like effect appeared to be generated by distributed bilateral neuronal sources in predominantly temporal and inferior-frontal areas (Fig. 7), whilst the attention-caused pseudoword advantage over word stimuli appeared to be predominantly sustained by activity in the left middle and posterior temporal cortex (Fig. 8).

In an attempt to test our paradigm in a model pathological neurocognitive state, we also recorded MEG data in small set of sleep-deprived individuals. These results (Fig. 9), albeit not subject to statistical analysis, show a stark contrast with our main testing group. Whilst a residual MMN pattern could be observed, visual inspection indicates neither P3a-like activity nor word-pseudoword differences after the disambiguation point.

4. Discussion

In this study, we have suggested and tested a novel paradigm which, in the absence of focussed attention on the auditory input, can shift attention to the spoken stimulus thereby affecting its lexical processing. Lexical differences (word-pseudoword distinction) were postponed to the very end of the stimuli to disentangle their effects from those of the earlier acoustic deviation. We registered the brain’s activity elicited by non-speech and speech stimuli in this paradigm using high-density whole-head MEG equipment and analysed resulting event-related fields. Three main results are as follows:

1. Non-speech (frequency) and speech deviants that diverge from the standard non-speech tone at the sound onset elicit a bi-phasic response showing increased amplitude over that to the standard tone.
2. In addition, rare speech sounds elicit a P3a-like deflection that has the opposite polarity to the main mismatch peaks and reaches maximum at ~240 ms after the stimulus onset.
3. At ~160 ms after the spoken stimuli can be uniquely identified (at their last phoneme), their responses diverge showing larger neural activation for meaningless pseudowords as opposed to meaningful words.

Below, we will briefly consider our findings in more detail:

4.1. Deviant-standard differences

Deviant stimuli, both non-speech and spoken, produced a response that was markedly different from that to the standard tone: a prominent peak at ~150 ms, followed by a second wave which peaked somewhat earlier (460 ms) for non-speech and later (540 ms) for speech sounds. This amplitude increase for rare auditory stimuli as opposed to more frequent ones is best described as the magnetic equivalent for the mismatch negativity (MMN) response, first described by Näätänen, Gaillard, and Mäntytsalo (1978). MMN is an evoked brain response elicited by rare unexpected deviant acoustic stimuli occasionally presented in a sequence of frequent standard stimuli (Alho, 1995; Näätänen, 1995). Importantly, MMN (and its magnetic counterpart, MMNm) can be elicited in the absence of the subject’s attention to the auditory input (Schröger, 1996; Tiitinen, May, Reinkainen, & Näätänen, 1994) whilst its magnitude and latency are dependent on the acoustic discrepancy between standard and deviant stimuli. It is therefore thought to reflect the brain’s automatic discrimination of changes in the auditory sensory input and thus to be an indicator of automatic cerebral processing of acoustic events (Näätänen, 1995). The current paradigm is different from the ‘classical’ MMN design, in which stimulus order is fully...
random, and the standards constitute the absolute majority (e.g., > 80%); here, standard stimuli take up 50% and occupy every second position. Still, this ‘multi-feature’ paradigm appears to generate results very similar to classical MMN designs (Kujala et al., 2007; Näätänen et al., 2004). Appearance of the MMN(m) in non-attend electrophysiological recordings is usually taken as a sign of functional (if not intact) neural mechanisms of auditory discrimination, which can be used as an indicator of the brain’s neurological status, e.g., in brain-damaged patients (Fischer, Morlet, & Giard 2000). From this point, the ability of the current paradigm to evoke an MMNm could be beneficial for using it to assess basic auditory functions, in addition to our originally set goals concerned with attentional and linguistic processes.

Here, we are reporting a bi-phasic MMN response. Although most previous work concentrated on the first MMN peak that normally occurs before 200 ms, it appears that the second wave is often present even in those experiments that do not discuss it explicitly (see e.g., Näätänen et al., 2004); we have certainly observed this bi-phasic nature in our previous studies (Pulvermüller, Shtyrov, Hasting, & Carlyon, 2008; Shtyrov et al., 2010a). Consistent with the current data, a suggestion has been made to separate the MMN into early and late components (Korpilahti et al., 2001). Some previous research interpreted the early peak as a reflection of fully automatic attention-independent processing, whilst the later activation is more subject to top-down attention-dependent control, at least for linguistic stimuli (Shtyrov, 2010).

4.2. P3a to spoken stimuli

Although the MMNm here is elicited by spoken and frequency deviants alike, speech-evoked ERPs demonstrate a more complex profile: in addition to the MMNm peaks, they generate another

![Fig. 6. Time course and cortical topography of the mismatch negativity response elicited by frequency and speech deviants: time-amplitude dynamics of magnetic field gradients (deviant minus standard subtraction) at the sensor 0242 which exhibited maximum response magnitude (middle), grand-mean MEG field topographies and distributed source models (L2 minimum-norm current estimates) for the early and late MMN deflection in the two hemispheres. ERF and source maps shown are taken at the response peak latency. For scaling, the maximum source activation at this time is taken as 100%.](image-url)
Whilst the MMN has been linked to detection of when they are not attended (Paavilainen, Karlsson, Reinikainen, & Alho, 1998). This is in line with the aims of the current study, where we attempt to estimate the nervous system’s ability to re-orient its attention automatically to such ecologically important and uniquely human information input as spoken language. The important difference between the current study and previous research is that we have used speech stimuli for P3a elicitation in a passive task whereas the majority of previous research utilised non-linguistic novel stimuli and frequently required active discrimination tasks. Clearly, automatic attention shifts to speech successfully take place in the current setting, where healthy volunteers are under investigation; testing this paradigm with neurological patients whose speech function is impaired (e.g., in aphasia) would be an important next step.

One potential limitation of the results we present is that there is no unattended speech condition to use as a baseline against which the P3a (and other effects) could be tested. This is precluded by the design of the current study which intentionally avoids a fully non-attend condition, concentrating instead on the effects of involuntary attention shift to spoken stimuli as the main target. However, there is solid body of evidence available from previous research using words and pseudowords in fully non-attend designs. Those studies, as reviewed in the Introduction, consistently reported a lexical response enhancement for words as opposed to pseudowords in the absence of attention, as well as the reversal of this effect when attention is paid to the stimuli. More importantly in the context of this discussion, none of the numerous previous studies using passive non-attend designs with spoken words found a P3a effect. At the same time, such an effect was reported in a direct comparison of attend and non-attend conditions as arising from voluntary attention allocation to speech, achieved through an overt experimental instruction and stimulus-related target detection task (Shtyrov et al., 2010a). Thus, the current data exhibit the same pattern as those with an overt instruction to attend to spoken material, which provide a direct support for our interpretation of the current effects as stemming from an attention shift. The important addition to the previous literature is that we control attention allocation here without an overt instruction or behavioural task by causing it to occur involuntarily.

Another potential criticism that can be made is that we compare acoustically very different stimuli – speech and tones – and the pattern of P3a responses to them may not be the same, with a possibly reduced P3a effect for tones related to their acoustic properties. Also, the SNR for tone stimuli could potentially be lower since they were fewer in number. However, on the one hand, also separately analysed individual speech stimuli could be seen here as evoking P3a. On the other hand, and more importantly, previous studies have established a solid P3 to non-speech stimuli when they are attended to (or when attention is forced to them by a strong deviance), and its absence in passively presented unattended tonal contrasts (Alho et al., 1998; Escera et al., 2001; Nätänen, 1992; Paavilainen et al., 1989). This is in line with our interpretation of the current design as not drawing attention to the tones, thus preventing the appearance of a P3a, which can, however, be seen for the spoken items. Future studies could use further modifications of our design in combination with
More generally, such an attention shift effect as reported here need not be limited to speech stimuli. We would predict that any strongly deviant stimulus – not necessarily speech – should attract attention, which would in turn generate a P3 response. However, the critical advantage of using speech is that, as a result of attention allocation, we also obtain a specific pattern of lexical effects which will be discussed below.

### 4.3. Pseudoword-word difference

Finally, when analysing word and pseudoword responses separately, we found clear differences at \( > 100 \text{ ms} \) after these items could be identified: an elevated response to pseudowords. Acoustic/phonological features of the stimuli were fully balanced: both words and pseudowords equally included the initial segments bi- and pi- and final stop consonants p and t. Moreover, these segments were physically identical, but combined in a different order in order to create meaningful words or meaningless pseudowords. This makes a purely acoustic explanation rather unlikely, and suggests that the differences are caused by the linguistic features of the stimuli. Previous research has shown an increased response to pseudoword over word items, which is usually most apparent in a later time range (e.g., N400 as in Federmeier et al., 2000 or Carreiras, Vergara, & Barber, 2005) and could be linked to enhanced search in the mental lexicon for a representation that does not exist. Earlier effects of lexicality, including a range of word-pseudoword differences at latencies before 200 ms have also been found (for a review, see e.g., Shtyrov & Pulvermüller, 2007b), pointing to rapid access to the mental lexicon which has been long postulated by behavioural psycholinguistic studies (Marslen-Wilson, 1987, 1990).

Previous research using passive oddball paradigms has, however, consistently indicated a higher response to words than pseudowords (e.g., Korpilahti et al., 2001; Shtyrov & Pulvermüller, 2002), which is the opposite of the pattern observed here. The main difference between the current study and previous research is that here spoken items constitute a minority of stimuli, interspersed among sinusoidal standards and frequency and duration deviants, whereas in all previous studies words and pseudowords were presented among other spoken items. This specific feature of the current design, as we argued above, must have caused an automatic shift of auditory attention to them, which was evident in the generation of a magnetic P3a visible for spoken stimuli (but not others). This in turn means that, by the time the disambiguation point arrived in the end of speech stimuli, re-orienting of auditory attention had already taken place, and they were attended to. In conditions where spoken stimuli are attended, pseudoword-elicited activation is known to dominate the event-related response; this is true both in N400 (Friedrich et al., 2006) and in oddball MMN-type designs (Garagnani et al., 2009). More specifically, it was found that while early word responses are largely unchanged by attentional modulation, allocating attention to pseudoword input increases its corresponding event-related response amplitude in EEG and MEG (Garagnani et al., 2009; Shtyrov et al., 2010a). This enhanced auditory pseudoword response was also visible in active paradigms that have used fMRI, where it was located in the middle and posterior left temporal cortices (Newman & Twieg, 2001; Hugdahl, Thomsen, Erland, Rimol, & Niemi, 2003), which is well in line with the configuration of sources we found in the current study.

Although we chiefly speak of lexical processing here, we cannot rule out that some of the contribution to differential brain responses we observe is semantic in nature. In fact, reflections of automatic semantic access at similar early latencies were found in passive oddball designs (Menning et al., 2005; Pulvermüller, Shtyrov, & Ilmoniemi, 2005; Shtyrov & Pulvermüller, 2007a; Shtyrov, Hauk, & Pulvermüller, 2004), which suggests that the word-specific activation potentially includes a semantic element; it may therefore be more appropriate to speak of lexico-semantic access. In either case, the processes under consideration are related to linguistic memory trace activation. Since our paradigm was not specifically aimed at addressing semantic distinctions, we refrain from further speculations on this issue. Future modifications of this paradigm may incorporate semantic distinctions in the same way as lexical ones were implemented here.

Another word of caution must be mentioned regarding the generalisations that can be drawn from using a very limited number of stimuli to the entirety of language. By the virtue of the oddball design, only very few tokens, repeated multiple times, can be used in any one experiment. Whilst this has a number of methodological benefits which we mentioned before, it does pose a question of ecological validity of such repetitive oddball and multi-feature designs. On the one hand, results achieved with a small number of items cannot be refuted per se. On the other hand, the important role in establishing the reality of early lexical effects in such designs was played by replicating these studies with different settings. To date, automatic lexical effects in passive oddball presentation were found using different imaging methods (EEG, MEG, fMRI), languages (e.g., English, Finnish, German, Russian, Thai, Spanish) by different labs using various stimulation set-ups. Most importantly, with careful matching of word and pseudoword stimuli, nearly identical effects could also
be found in the absence of any stimulus repetition, by contrasting large groups of unique words and pseudowords (MacGregor et al., 2012). This supports the notion that the early effects we observe here are indeed linked to the lexical processing of the word and pseudoword stimuli which takes place automatically, regardless of the multiple repetition of the same tokens.

What, however, could be the explanation for such a markedly different pseudoword behaviour in attend and non-attend conditions, and, more specifically, for the pseudoword advantage in the former? Whereas it appears that pseudowords, phonological in nature and phonotactically legal, may activate language areas in the brain (Mazoyer et al. 1993; Newman & Twieg, 2001; Petersen, Fox, Snyder, & Raichle, 1990; Posner, Abdullaev, McCandliss, & Sereno, 1996; Price, Wise, & Frackowiak, 1996; Shtyrov et al., 2005, 2008), the absence of long-term memory traces for such items implies that no activation of such traces per se is possible. Therefore, in non-attend conditions they produce a reduced response possibly reflecting acoustic and phonological aspects of their processing as well as partial activation of related lexical traces. The latter assumes a path of pseudoword perception compatible with Cohort (and some other) models of speech processing (Marslen-Wilson, 1987). That implies that all lexical neighbours become partially activated as soon as the relevant information is available. However, as additional information arrives, they become progressively excluded until there is only one candidate left (or none for pseudowords). In the context of this study, its design makes sure that this crucial information arrives at the same time in the end of the stimuli. At this time, the traces for lexical neighbours, activated partially and equally for words and pseudowords, already become extinguished as they are ruled out (once the stimulus is identified, its representation is hypothesised to inhibit other lexical neighbours/competitors). If there are no additional resources (attention) to develop and maintain their activity. The initial word activation seems to reach its maximum regardless of the level of attention due to strong links within the network subserving it. In the associative Hebbian logic of memory trace formation (Braitenberg & Schütz, 1992; Hebb, 1949; Pulvermüller, 1999), the strength of synaptic connections between subparts of such networks is dependent on the degree of co-activation between these parts. Thus, frequent use of words should lead to representations robust enough to such an extent that they are activated by the respective stimulus regardless of amount of attention paid to it. Pseudowords as such have no individual representations and thus produce a reduced response in non-attend conditions, as seen repeatedly in previous passive non-attend oddball experiments. However, with more additional resources becoming available, the early word activation still remains at its maximum, whereas partial but multiple neighbour activations appear to be able to benefit from these additional resources and thus be maintained at a higher level, leading to the pseudoword advantage effect (for review, see Shtyrov, 2010), which we also observed here. In other words, once the specific representation/lexical entry is selected in the case of a meaningful word, its neighbours are likely to be inhibited/suppressed, as posited in the majority of psycholinguistic theories of word recognition. In the attended pseudoword case, however, this selection point is never achieved, and thus competitors may not be fully ruled out. Furthermore, when attention is actively paid to such ill-shaped linguistic stimuli, they may prompt a more elaborate lexical search (and even reanalysis/repair) once the first attempt on automatic mapping of the input to a single entry fails. This leads to activation enhancement for pseudowords, whilst the initial word memory trace activation is robust and stable irrespective of attention and task demands.

These electrophysiological distinctions, predicted by a neurocomputational model of word representations in the brain (Garagnani et al., 2008), have been well documented in previous studies, particularly in passive MMN (word > pseudoword) and N400 (pseudoword > word) investigations, as reviewed above. The important difference between previous studies and the current one is that here we do not use an explicit instruction to focus the subjects’ attention to the stimulus stream. This is done by the design in which attention is drawn automatically to spoken input. The current paradigm therefore makes it possible to probe neural processes of automatic lexical access and selection in the absence of an overt instruction or task to carry out linguistic activity.

4.4. Effects of sleep deprivation

Lastly, we have pre-tested the performance of this paradigm in a set of sleep-deprived individuals. Sleep deprivation is often associated with psychotic-type pathologies (Gove, 1970; Kahn-Greene et al., 2007; Rimon et al., 1986). It therefore provides a relatively safe and easily-controlled model of psychotic conditions, which are some of potential future applications for the novel paradigm we suggest here, given many reports of attention and language disturbances in psychotic individuals (such as schizophrenia, see e.g., Braff, 1993; Mahler, 1991). While a residual MMN response could be seen in the sleep-deprived individuals, we observed a breakdown of the P3a component, indicating deficits in attention allocation processes, and, furthermore, no indication of lexicality effects (if anything, post-disambiguation responses to spoken stimuli are reduced to noise levels; Fig. 9). This suggests a deterioration of automatic lexical search and selection processes as a result of sleep deprivation. As only a small group of subjects was tested, meaning that the results cannot be verified statistically, these preliminary data must be treated with caution. Future studies are therefore necessary to investigate interactions between automatic lexical and attention processes in sleep deprivation and in clinical conditions.

5. Conclusions

In sum, we propose a paradigm that can test, in a single short non-invasive procedure: (1) neural discrimination of acoustic information, (2) neural mechanisms of automatic attention allocation and (3) neural correlates of lexical processing. It does not require explicit instructions and needs minimal cooperation from the subject, therefore having a great potential for assessing auditory and linguistic functions in subjects/patients who are unable to cooperate with a more active task. This paradigm has been tested here using a group of healthy volunteers. Future studies are necessary to investigate its potential implementation in clinical populations.

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