Task-Sensitive Reconfiguration of Corticocortical 6–20 Hz Oscillatory Coherence in Naturalistic Human Performance

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Abstract: Electrophysiological oscillatory coherence between brain regions has been proposed to facilitate functional long-range connectivity within neurocognitive networks. This notion is supported by intracortical recordings of coherence in single-out corticocortical connections in the primate cortex. However, the manner in which this operational principle manifests in the task-sensitive connectivity that supports human naturalistic performance remains undercharacterized. Here, we demonstrate task-sensitive reconfiguration of global patterns of coherent connectivity in association with a set of easier and more demanding naturalistic tasks, ranging from picture comparison to speech comprehension and object manipulation. Based on whole-cortex neuromagnetic recording in healthy behaving individuals, the task-sensitive component of long-range corticocortical coherence was mapped at spectrally narrow-band oscillatory frequencies between 6 and 20 Hz (theta to alpha and low-beta bands). This data-driven cortical mapping unveiled markedly distinct and topologically task-relevant spatiotemporal connectivity patterns for the different tasks. The results demonstrate semistable oscillatory states relevant for neurocognitive processing. The present findings decisively link human behavior to corticocortical coherence at oscillatory frequencies that are widely thought to convey long-range, feedback-type neural interaction in cortical functional networks. *Correspondence to: Timo Saarinen; Brain Research Unit, O.V. Lounasmaa Laboratory, Aalto University, P.O. Box 15100, FI-00076 AALTO, Finland. E-mail: timo.p.saarinen@aalto.fi Received for publication 2 July 2014; Revised 24 February 2015; Accepted 24 February 2015. DOI: 10.1002/hbm.22784

Key words: task performance; behavior; cognition; functional neuroimaging; cerebral cortex; electrophysiology; connectome; functional brain network

INTRODUCTION

Neurocognition has been thought to rely on electrophysiological interactions and information transfer within large-scale functional brain networks [Friston, 2008; Mesulam, 2008; Siegel et al., 2012; Varela et al., 2001]. Oscillatory activity is a prominent feature of neuronal signaling [Buzsáki and Draguhn, 2004], and coherent oscillations between brain regions have been implicated in formation of functional connections and network structures [Bressler and Kelso, 2001; Fries, 2005; Siegel et al., 2012]. Coherence...
describes the mutual dependency of oscillatory components at the operational millisecond timescale of neural activity. Empirically, coherence has been demonstrated in dynamic episodes of interregional neuronal communication [Bosman et al., 2012; Bressler et al., 1993; Buschman and Miller, 2007; Pesaran et al., 2008; Salazar et al., 2012], and it is thought to promote functional coupling through periodically occurring electrical potentiation of signaling and temporal coordination of synchronous activity [Bressler and Kelso, 2001; Fries, 2005; Izhikevich, 2001; Siegel et al., 2012].

The primate cerebral cortex has been shown to utilize interregional coherence to support behavior [Bosman et al., 2012; Bressler et al., 1993; Buschman and Miller, 2007; Pesaran et al., 2008; Salazar et al., 2012]. Intracortical recordings in monkeys have found task-sensitive coherent activity between selected pairs of functionally relevant and anatomically connected but distant cortical sites [Buschman and Miller, 2007; Pesaran et al., 2008; Salazar et al., 2012]. Such long-range coherence can operate at several frequencies, including the prominently oscillatory range across the theta, alpha and beta bands (approximately 4–30 Hz) [Bastos et al., 2015; Bosman et al., 2012; Bressler et al., 1993; Buschman and Miller, 2007; Pesaran et al., 2008; Salazar et al., 2012; von Stein and Sarnthein, 2000]. Long-range coherence has also been demonstrated to link subpopulations of cortical neurons that contain information relevant for the task at hand [Bosman et al., 2012; Buschman and Miller, 2007; Pesaran et al., 2008; Salazar et al., 2012]. Coherence may thus facilitate task-sensitive routing of functional connections and inclusion of relevant information processing in the large-scale corticocortical network.

The present, noninvasive magnetoencephalography (MEG) coherence imaging study on human naturalistic performance builds up on these intracortical descriptions of coherent connections. Here, in a data-driven search across the cortex, we sought to characterize task-sensitive reconfiguration of the global patterns of coherent oscillatory connectivity. We studied performance within a set of naturalistic and low-constrained tasks [cf. Felsen and Dan 2005; Hasson et al. 2010; Ingram and Wolpert, 2011; Pesaran et al., 2008] that represent hallmarks of human behavior, including active vision, language ability and object manipulation. We applied a neuromagnetic imaging approach [Gross et al., 2001; Kujala et al., 2008] that allows direct sensor-to-cortex mapping of MEG coherence data into long-range corticocortical estimates, without resorting to any a priori seed regions, external reference signals or intermediate signal projections to the brain space [for a review on MEG connectivity, see Palva and Palva, 2012]. The present method of tracking coherent activity at the operational neural frequencies provides a whole-cortex imaging analog to the intracranial animal and patient coherence recordings that typically have more limited coverage [e.g., Buschman and Miller, 2007; Pesaran et al., 2008; Salazar et al., 2012; Sehatpour et al., 2008; Swann et al., 2012].

Task-sensitive coherence was determined for both easier (E) and more demanding (D) variants of visual picture comparison (easy distinct, more demanding nondistinct differences), speech comprehension (easy native, more demanding non-native language) and bimanual object manipulation (easy handling, more demanding solving of a Rubik’s cube) tasks. These tasks were performed in a freely unfolding, subject-driven manner and they were designed to elicit distributed neurocognitive information processing that would tax differentially the perceptual, motor, attentional-executive, and memory abilities. The central goal was to determine whether the global connectivity patterns within each of the three behavioral modalities would show specific topology compared to a generic task-non-specific pattern. Building on the concept that elevated coherence indicates efficient interregional coupling and relevant connections for a task [Fries, 2005; Siegel et al., 2012], we focused on functional connections that would be more engaged in a particular task, rather than suppressed, compared to the task-non-specific baseline [cf. “task-positive” pattern; Fox et al., 2005]. Task-sensitivity was determined as increase of interregional coherence with respect to the mean pattern (task-mean) calculated across all the tasks. Importantly, the resulting connectivity estimates represent the modulated components of the coherent connectivity among the tasks and not a contrast to, for example, less well defined rest data [Morcom and Fletcher, 2007].

Task-related functional networks are a current and important theme in neuroscience [see Park and Friston, 2013]. By taking a step from the spontaneous rest activity and restricted, single-domain experimental protocols, we aimed for a data-driven view on task-sensitivity of functional oscillatory connectivity for a set of naturalistic human behaviors. In general, a proof-of-concept for functional large-scale networks as the basis of neurocognition is to uncover flexible functional connectivity patterns that feature both task-sensitivity and topological task-relevance. To achieve this, we quantified functional corticocortical connections using direct mapping of electrophysiological coherence at spectrally narrow-band, oscillatory frequencies. We hypothesized that neural oscillatory coherence would provide a key mechanism for the task-related network function. We further hypothesized that to fulfill the different neurocognitive requirements of each naturalistic task, the spatiotemporal connectivity pattern, mediated by coherence, would show distinct task-specific modulation [task-sensitivity; see Engel et al., 2013; Siegel et al., 2012].

The topological pattern should not be sporadic but the cortical regions involved should cover task-associated regions (task-relevance) while overlapping in multidomain areas, such as frontoparietal cortex [Barbey et al., 2012; Duncan, 2010; Fox et al., 2005]. This study was conducted with an explorative approach to highlight the most prominent components of oscillatory connectivity that support...
naturalistic performance. We successfully uncovered task-sensitive 6–20 Hz coherent connectivity patterns for each naturalistic task.

MATERIALS AND METHODS

Participants and Tasks

The participants were 10 right-handed, Finnish-speaking healthy adults (5 females; 5 males mean age 32 years, range 21–40 years). They gave their informed consent for the experiment, in agreement with a prior approval of the local Ethics Committee (Hospital District of Helsinki and Uusimaa; in compliance with Declaration of Helsinki). The subjects performed naturalistic tasks in three different modalities: visual, auditory, and sensorimotor. In each modality, task design was implemented as both an easy and a more demanding variant. Continuous task performance was monitored with video feed from the measurement room and questions (during pauses, see below) but no further quantification was made. The subjects were actively engaged with the tasks.

In the naturalistic visual tasks, the participants compared two color pictures of vegetation with differences in either highly distinct features (color modification in simplified photograph-based images with fine details removed; easier variant) or more subtle features (fine structural details present or not in photographs; more demanding variant). Comparison of these perceptually meaningful images required visual search and varying levels of visual attention and working memory. The stimuli were presented on a projector screen ~1 m away from the subject’s eyes. The two pictures (width × height, ~12 × 8.5 cm per picture) were presented simultaneously, one above the other, for 30 s (altogether six picture pairs, 3 min of data in total). During a 7-s pause between stimulus presentations, the subjects were asked to indicate how many differences they found between the pictures, with the response given as finger lifts.

The naturalistic auditory conditions were speech comprehension tasks with sound track clips from fairly unknown Finnish-speaking movies (easier variant) and English-speaking movies (more demanding variant). The stimuli were delivered binaurally through nonmagnetic earplugs, at comfortable listening level. These speech comprehension tasks required concentration on verbal material embedded in natural environmental sounds and with a varying degree of linguistic challenge. All the Finnish-speaking subjects had studied English as part of their basic education (~10 years). There were three ~1-min clips, each followed by a 10-s pause during which the participants answered a question in Finnish about the content.

The naturalistic sensorimotor tasks were bimanual free manipulation of a Rubik’s cube (without twisting it) and solving the Rubik’s cube task, respectively, for 3 min. The Rubik’s cube was a small (3 × 3 cm) nonmagnetic version, with the original 3-by-3 grid. Both tasks required visuomotor processing in object manipulation, and solving the Rubik’s cube additionally demanded effortful visuospatial reasoning to guide the movements. None of the participants reported to be an expert in solving the Rubik’s cube; to maintain the difficulty at a reasonable level, they were instructed to first try to align the color of only one side of the cube.

MEG Recordings and Procedure

MEG signals were recorded in a magnetically shielded room at the MEG Core, Aalto University, using a whole-head 306-channel system (102 triple sensor elements composed of two orthogonal planar gradiometers and one magnetometer; Elekta Oy, Helsinki, Finland). MEG data were passband filtered at 0.1–200 Hz and sampled at 600 Hz. The MEG recording was linked to the head anatomy with the help of four head position indicator coils that were attached to the participant’s scalp: The locations of the coils with respect to three anatomical landmarks (nasion and preauricular points) were determined using a 3-D digitizer (Isotrak; Polhemus, Colchester, VT), and with respect to the MEG array by briefly energizing the coils before each recording session. Identification of the three anatomical landmarks on structural magnetic resonance images (MRIs) further allowed coregistration of MEG signals and the individual brain anatomy. The MRIs were acquired at the Advanced Magnetic Imaging Center, Aalto University, with a 3T Signa EXCITE MRI scanner (GE Healthcare). For behavioral monitoring, electrooculography and the left and right arm electromyography were collected simultaneously with MEG.

The subjects were first introduced to the study and prepared for the measurement. Thereafter, they were seated under the MEG helmet and the different tasks were performed in separate blocks. These blocks were grouped by modality (visual, auditory, sensorimotor). The order of the three modalities, and of the tasks within each modality, was pseudorandomized across subjects.

MEG Data Processing and Connectivity Analysis

Preprocessing of the MEG data included application of the Spatiotemporal Signal Space Separation method (tSSS) to suppress noncerebral, artifactual signals (external noise, movement-related signals) [Taulu and Hari, 2009; Taulu and Simola, 2006]. This approach of separating and removing signals that do not originate in the brain/head space is based on physical properties of the MEG measure, that is, sensor geometry and electrodynamics, as well as statistical separation in the time domain.

For the characterization of corticocortical connectivity, we focused on frequencies that displayed the strongest coherence. This was estimated from the long-range (>10 cm) sensor–sensor coherence spectra, averaged across sensors for each task (Fig. 1). The frequencies below 20 Hz showed the
A frequency-domain spatial filter approach, Dynamic Imaging of Coherent Sources (DICS) [Gross et al., 2001; Kujala et al., 2008], was applied to the MEG sensor-level data to obtain a brain-level estimate of long-range cortico-cortical coherence as a function of neural oscillatory frequencies. This source reconstruction method has two interesting properties. First, it performs a direct cortico-cortical mapping of sensor-level coherent activity, thus requiring no precursory signal projection to the brain-space and subsequent separate pair-wise connectivity analysis of the resulting source-level time courses. Second, it is based on temporally resolved signal cross-correlation, that is, linear dependence of oscillatory components in two time series as they evolve [Gross et al., 2001; Kujala et al., 2008; cf. Hipp et al., 2012]. Like any other M/EEG source reconstruction [Palva and Palva, 2012], the DICS estimate is nonunique and faces the problem of signal field spread which prevents reliable quantification of local (short-range) connectivity [Schoffelen and Gross, 2009]. The spatially smooth record of cortical activity provided by MEG is thus better suited for long-range connectivity analysis. DICS spatial filter can readily distinguish local maxima that are at least 2 cm apart [Liljeström et al., 2005]. We sought to suppress spurious connectivity [Schoffelen and Gross, 2009] by requiring a minimum distance of 4 cm between the connection start and end points and by focusing on contrasts between experimental conditions rather than absolute coherence values.

In the DICS analysis, first, a common grid of 9-mm side length, determined in one example brain, was transformed to cover cortical surface of the individual MRIs at corresponding grid sites (MNE Suite software; Hämäläinen, Martinos Center for Biomedical Imaging, Massachusetts General Hospital, MA). To ensure best possible signal-to-noise ratio for the DICS estimation, we retained for further analysis 3,600 cortical grid points whose activity could be reliably picked up by the MEG helmet (i.e., grid points closest to the MEG sensors). The included grid points provided a good coverage of the cortical sheet (fissural and gyral). Second, cross-spectral density matrices across all planar-gradiometer sensor signals were computed for each task condition using Welch’s averaged periodogram method (Hanning 2048-point window; 0.3 Hz resolution; 75% window overlap). These matrices contained information of linear oscillatory dependencies in the sensor-level data. Third, in a global all-to-all search, long-range coherence estimates (4 cm minimum distance) were computed for each cortical grid point with the other grid points in the frequency bands of interest using a DICS spatial filter.

To extract possible task-sensitive components of cortico-cortical connectivity, per frequency band, the global search results for each task were contrasted against the mean pattern calculated across all the tasks (task > task-mean). In addition, we tested pairwise comparisons between the easy and demanding variants of the tasks (E > D and D > E) within each task modality. To preclude changes of signal strength as an explanatory factor in the coherence-

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Figure 1.
Coherence as a function of frequency. Sensor–sensor coherence spectra, averaged across all sensors, with all task conditions overlaid (light gray lines, picture comparison; medium gray lines, speech comprehension; black lines, object manipulation; thicker lines for the easy and thinner lines for the demanding variants of the tasks). The top 20% of the maximum coherence, across tasks and subjects, is reached in the area above the gray bar. Remnants of the power line 50-Hz artifact stand out as a narrow coherence peak.

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Coherent Connectivity in Naturalistic Tasks

Based connectivity estimates [Schoffelen and Gross, 2009], we verified that the signal power was not significantly modulated by the task (task vs. task-mean; E vs. D) at any MEG sensor in a given contrast \( n = 10 \); paired \( t \)-statistics, \( P > 0.05 \), false discovery rate (FDR) corrected. If modulation was found, the frequency band in question was excluded from the further analysis of this particular task contrast.

Statistical significance of the coherent connections \( (n = 10; \text{paired } t \text{-statistics}, P < 0.05, \text{FDR corrected}) \) was determined on parcellated cortex (Fig. 2). The clustering of the grid points into <100 regions reduces the spatial sampling of the source-level data to a resolution that facilitates relative independence of the measurements. The regions of interest were generated in the common subject space, based on the Desikan-Killiany parcellation scheme [Desikan et al., 2006]. The Desikan-Killiany regions within 8.25 cm from the closest MEG sensor (Euclidean distance, average of grid points included in the region) and containing at least 10 grid points were retained in the analysis, resulting in a total of 75 regions (37 left; 38 right; Fig. 2).

Mean coherence values for each of the regions were computed from the grid-based data per each subject, task and frequency band. Coherent connections between regions with a Euclidean distance of 1 cm or less between any of their vertex points were discarded; in practice, this meant that connections to border adjacent regions were excluded from the analysis. As a result, 418 connections were evaluated within the left hemisphere, 447 connections within the right hemisphere and 2221 connections when the whole cortex was considered at once. Significant coherence values for each of the regions were computed from the grid-based data per each subject, task and frequency band. Coherent connections between regions with a Euclidean distance of 1 cm or less between any of their vertex points were discarded; in practice, this meant that connections to border adjacent regions were excluded from the analysis. As a result, 418 connections were evaluated within the left hemisphere, 447 connections within the right hemisphere and 2221 connections when the whole cortex was considered at once. Significant coherence modulations were visualized on circular connectograms using a hierarchical bundling approach [Holten, 2006]. The statistical testing was conducted both within each hemisphere and also in the whole cortex, allowing interhemispheric connections.

Independence between the factors frequency band and task modality, as well as between frequency band and cortical area (frontal, central, cingulate, temporal, parietal, occipital), were tested with \( \chi^2 \)-statistics, using adjusted standardized residuals as post hoc indicators of which bands in which task modality or cortical area differed from the expected pattern (if independence between the factors does not hold). \( \chi^2 \)-statistics was also used to evaluate whether the observed distribution of significant coherent connections (pooled across the frequency bands) displayed a specific structure that differed from an equal distribution of connections (null hypothesis) per naturalistic task. For the \( \chi^2 \)-test the missing values due to the

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**Figure 2.**
Cortical parcellation scheme. Color codes for each cortical region (75 regions based on a Desikan-Killiany scheme [Desikan et al., 2006]) displayed on the surface views (lateral and medial) and on a circular connectogram. Numbering in the circular plot (running clock-wise: 1–38 right, 39–75 left hemisphere) corresponds to the regions indicated on the surface views. Region titles are listed below the figure. Note that there were a small number of regions that passed the criterion for selection in one hemisphere only: for 3 regions in the left and 2 regions in the right hemisphere, the homologous area in the other hemisphere was not included in the analysis. Abbreviations: LH, left hemisphere; RH, right hemisphere; Fr, frontal; Cen, central; Tpl, temporal; Ptl, parietal; Occ, occipital.
power level difference at certain frequency bands were substituted with estimates based on cross-tabulation of factors and proportional values of connections in the conditions (tasks, bands) with no excluded data. Hemispheric lateralization of connectivity was quantified with laterality index $LI = \frac{(left - right)}{(left + right)}$, where positive values indicate left lateralization [Jansen et al., 2006]. In the whole-cortex analysis, interhemispheric connections were categorized and counted as nonhomologous (connections to nonhomologous and nonadjacent regions), near homologous (connections to adjacent regions) and homologous connections (connections to homologous regions).

These connectivity results were studied further on a coarser lobar level where eight lobar areas were collapsed from the 75 parcelled regions. These eight lobar areas were: inferior frontal (regions left 69, 72, 73; right 3, 4, 6, 7, 10), superior frontal (regions left 70, 71, 74, 75; right 1, 2, 5, 8, 9), interior parietal (regions left 47–50, 66, 67; right 12, 13, 28–31), superior parietal (regions left 51, 52, 68; right 11, 26, 27), inferior temporal (regions left 44, 54–56, 58, 59; . . . )

Figure 3.
Circular connectograms of significant task-sensitive coherent connectivity. (A) Within-hemisphere analysis for the three different modalities of naturalistic tasks (visual, auditory and sensorimotor; easy and demanding variants pooled). The lines indicate significant connections (task > task-mean), with the different oscillatory frequency bands (6–9, 9–13, 13–20 Hz) plotted in different colors. Note that the 13–20 Hz band was excluded in the auditory task due to a significant difference in signal power between the task and task-mean level. Below the circular connectograms, the connected regions are displayed on the brain surface views. For the color key of the cortical parcellation scheme, see Figure 2. (B) Overlap of the connected cortical regions between the task modalities (within-hemisphere analysis): all three tasks overlap (dark red), two tasks overlap (orange), present in one task only (light orange). (C) Whole-cortex analysis for the three different task modalities.
right 19, 20, 22–24, 33), superior temporal (regions left 57, 60–63; right 15–18, 21), occipital (regions left 39–43; right 34–38), and posteromedial cortex (regions left 45, 46, 53, 64, 65; right 14, 25, 32). At the lobar level, the coherent connections were characterized with graph network measures by calculating degree centrality (unweighted edges; frequency bands pooled) for each lobar area as well as edge degree that counts the interregional significant connections between the lobes [https://sites.google.com/site/bctnet/Home; Sporns et al., 2007]. The lobar nodes with the highest degree centrality were identified as hub regions.

RESULTS

Coherent Connectivity within the Oscillatory 6–20 Hz Range Forms Distinct Patterns for Different Naturalistic Tasks

Coherent oscillatory connections were mapped from the MEG sensor data into the human cortex (see Materials and Methods; [Gross et al., 2001; Kujala et al., 2008]) in the frequency bands of 6–9 Hz (theta/low-alpha), 9–13 Hz (high-alpha) and 13–20 Hz (low-beta). This frequency range (6–20 Hz) featured distinct peaks in the sensor-level coherence spectra, consistently across the studied tasks, indicating oscillatory (narrow-band) signaling that was distinguishable from the spectrally more diffuse high-frequency activity (>30 Hz) (Fig. 1; see Materials and Methods).

The present data-driven, all-to-all interregional mapping (in a parcelled cortex; Fig. 2) showed distinct spatiotemporal patterns of coherent connectivity for the different naturalistic tasks, as illustrated in Figure 3 for all three modalities (visual, auditory, sensorimotor). Task-sensitive connectivity was quantified as enhanced coherence (unweighted edges; frequency bands pooled) for each lobar area as well as edge degree that counts the interregional significant connections between the lobes [https://sites.google.com/site/bctnet/Home; Sporns et al., 2007]. The lobar nodes with the highest degree centrality were identified as hub regions.

The Sub-bands of the 6–20 Hz Range Are Emphasized Differently in Different Tasks and Cortical Areas

The task modality (visual, auditory, sensorimotor) and frequency band (6–9, 9–13, 13–20 Hz) were not independent factors as regards the distribution of coherent connections \(x^2(4, N=163) = 33.4, P < 0.001\); see Fig. 3A). Evaluated with the adjusted standardized residuals \((-1.96 < z < 1.96\), the visual modality showed a higher than expected number of connections in the 6–9 Hz range (z-score 5.1; observed/expected 15/5) and a lower than expected number in the 13–20 Hz range (z-score −3.4; 16/25). Reversely, the sensorimotor task displayed a higher than expected number of connections in the 13–20 Hz band (z-score 3.0; 66/57) and a lower than expected number in the 6–9 Hz band (z-score −5.2; 1/12). The auditory
modality did not differ from the expected pattern (note that the 13–20 Hz band was excluded and missing values were estimated: z-score 0; estimated/expected 20/20).

Across the tasks, the cortical area (frontal, central, cingulate, temporal, parietal, occipital; see Fig. 3A) and frequency band were not independent factors \[v^2(10, N = 328) = 26.0, P < 0.005; \text{missing values estimated}\]. The 6–9 Hz band was utilized less than expected by the central regions \(z\)-score \(-2.3; 1/6\). The 9–13 Hz band was utilized more than expected by the temporal cortex \(z\)-score \(2.4; 19/12\) and less than expected by the frontal \(z\)-score \(-2.2; 6/12\) and occipital cortices \(z\)-score \(-2.7; 4/11\). The 13–20 Hz band was utilized less than expected by the temporal cortex \(z\)-score \(-3.4; 22/33\). Spectral patterning in the parietal and cingulate cortices did not deviate from the expected pattern.

The Easy and Demanding Tasks Recruit Functional Networks with Different Emphasis

The task-sensitive connectivity results are illustrated in Figure 4, separately for the easy (E) and demanding (D) variants of the naturalistic tasks (contrasts to the task-mean pattern). Direct pair-wise comparisons of the connections in the easy and demanding tasks did not generally reach significance \([\text{paired } t(9)-\text{tests}, E > D \text{ or } D > E, \ldots]\).

| TABLE I. Number of significant coherent connections for each task modality and difficulty level in three frequency bands of interest (within-hemisphere analysis). |
|---------------------------------------------------------------|
| Frequency (Hz) | Visual | | Auditory | | Sensorimotor |
|----------------|--------|-----------------|--------|-----------------|--------|-----------------|
| E | D | E | D | E | D |
| 6–9 | 14 | 1 | 0 | 6 | 0 | 1 |
| 9–13 | 9 | 0 | 6 | 0 | 0 | 24 |
| 13–20 | 9 | 7 | – | – | 4 | 62 |

Abbreviations: E: easy; D: demanding; –: excluded due to significant power difference.
P > 0.05, FDR corrected; only a single connection showed significantly stronger coherence in the Rubik’s cube solution than cube manipulation task. However, the number of significant coherent connections (task > task-mean; Table I) deviated from an equal pattern across the tasks [pooled across the frequency bands; $\chi^2(5, N = 163) = 177.7, P < 0.001;$ missing values estimated]. The demanding sensorimotor (Rubik’s) task showed the most numerous connections ($z$-score 11.5; observed/expected 87/27), whereas the demanding visual ($z$-score 2$3.7; 8/27), easy auditory ($z$-score $23.3; 10/27) and easy sensorimotor tasks ($z$-score $24.4; 4/27) showed less than expected amount of connections (see Table I for each frequency band).

There were core patterns of connectivity for the visual ( fusiformal-parietal connectivity) and sensorimotor (frontoparietal connectivity) tasks that were highlighted in both the easy and demanding variants of the tasks. In the speech comprehension task, the significant connections in the easy and demanding conditions emphasized different components of the network: the left perisylvian connectivity at 9–13 Hz did not reach significance in the demanding (non-native) task but, instead, there was connectivity at 6–9 Hz involving markedly the occipital regions. The connectivity pattern observed in the demanding speech condition was salient also in the whole-cortex connectivity analysis (Fig. 3C). In fact, all the significant connections found in the whole-cortex analysis were related to the demanding task in the speech comprehension and object manipulation tasks, whereas the easy condition dominated in the visual task (85% of all connections within the modality).

The number of task-sensitive connections (pooled across frequency bands) showed distinct lateralization for the different tasks (see Fig. 4). The right hemisphere predominated in the visual task (laterality index LI = $-0.7$ easy, LI = $-1.0$ demanding). In the auditory task, the pattern changed from left lateralized (LI = $+1.0$) in the easy condition to right lateralized in the demanding condition (LI = $-0.7$). In the sensorimotor task, the left-lateralized pattern (LI = $+1.0$) in the easy condition became bilateral (LI = $+0.1$) in the demanding condition. In the more bilateral maps, the parieto-occipital connections in the visual task and the frontoparietal connections in the sensorimotor task were mirrored in the nondominant hemisphere.

Lobar level graph plots (Fig. 5A; see Materials and Methods) illustrate the coherent networks for the easy and demanding tasks within the hemisphere that is dominant for the task. In picture comparison, there was a main hub in the right inferior temporal cortex. In the easier variant, the highest edge degree was found between this infero-temporal hub region and the posterior superior parietal cortex. The demanding variant displayed a generally lower degree of connectivity. In speech comprehension, the main hub in the easy condition was located in the left inferior parietal cortex that had a high-degree linkage to the left superior temporal cortex. The demanding speech comprehension highlighted right-hemisphere connection between
anterior temporal and superior frontal cortices, along with the midline cingulo-occipital connectivity (Figs. 3C, 4). In object manipulation, the main hub was located in the left inferior parietal cortex that connected to the inferior and superior frontal cortices. The edge degree for the frontoparietal connectivity was notably high in the demanding task variant, and connectivity emerged between the occipital, temporal, and parietal cortices. Across the tasks (Fig. 5B), the postero medial cortex showed higher degree and edge degree for the demanding than easier variants, with coherent connections formed particularly with the bilateral temporal and the left inferior parietal cortices.

DISCUSSION

The key objective of the present human MEG study was to gain a cortex-wide view on task-sensitivity of coherent oscillatory connectivity in association with naturalistic performance. As hypothesized, the data-driven mapping of long-range coherence revealed task-sensitive reconfiguration of global connectivity patterns for a set of naturalistic tasks. The elevated 6–20 Hz long-range coherence was differentially routed for each task, in comparison with the generic task-mean pattern. The connectivity patterns were well formed as each type of naturalistic behavior was associated with topologically task-relevant connectivity patterns. The different corticocortical task networks utilized the studied oscillatory bands (6–9, 9–13, 13–20 Hz) to a varying degree. The networks engaged both nodal regions that were distinct for each task modality and others that were more overlapping between modalities (including frontoparietal cortex). Overall, the present study uncovered flexible spatiotemporal patterning of coherent connectivity that forms well-defined functional network structures to support different naturalistic behaviors.

Task-Sensitive 6–20 Hz Coherence

The task-sensitive coherent connectivity was mapped within the 6–20 Hz oscillatory range which has been eminently linked with long-range interregional coupling and neurocommunication [Kopell et al., 2000; von Stein and Sarnthein, 2000]. Spectrally narrow-band oscillatory activity is prominent at the <30 Hz neural frequencies, whereas the higher frequencies (>30 Hz; low/high-gamma band) encompass both oscillatory and arrhythmic broadband signaling [e.g., Bressler et al., 1993; Buzsáki and Draguhn, 2004; Crone et al., 2006; Hari and Salmelin, 1997]. The lower-frequency oscillations may coordinate and interact with the high-frequency activity [Buschman et al., 2012; Lisman and Jensen, 2013] and, in some instances, emerge as concatenation of higher frequency signals [Kopell et al., 2011]. The spectral content, or fingerprint, of connectivity may index specific computation performed in the cortical microcircuitry [Bastos et al., 2012; Siegel et al., 2012]. The oscillatory frequencies studied here have been associated with maintained, semistable states and feedback-type corticocortical interactions [Cannon et al., 2014; Siegel et al., 2012; van Kerkove et al., 2014; von Stein et al., 2000] rather than more transient, feedforward-type responses [associated recently with frequencies >30 Hz and <6 Hz; Bastos et al., 2015]. Oscilatory interaction at 6–20 Hz has been proposed to play a role in various neurocognitive processes, such as memory retention, multimodal processing, rule-switching and decisional performance [Buschman et al., 2012; Cannon et al., 2014; Fell and Axmacher, 2011; Kopell et al., 2011; Pesaran et al., 2008], all important for naturalistic behavior.

The subbands of the 6–20 Hz range were differently emphasized in the tasks: whereas the picture comparison task relied heavily on the lowest (6–9 Hz) band, the object manipulation task utilized particularly the highest band (13–20 Hz), instead. The pronounced theta/low-alpha coherence (6–9 Hz) in the naturalistic picture comparison task agrees with earlier primate and human findings on visuo-attentional and working memory performance [Lindau and Fries, 2012; Liebe et al., 2012; Watrous et al., 2013]. Interregional theta coupling has been linked with memory maintenance of visual material [Fell and Axmacher, 2011; Liebe et al., 2012] as well as with attentional reorienting to unexpected stimuli in the frontoparietal network [Daitch et al., 2013]. The prominent low-beta coherence (13–20 Hz) for the sensorimotor task has been found also in the monkey frontoparietal cortex [Pesaran et al., 2008]. The wider beta band (15–30 Hz) has been associated with motor function [Brovelli et al., 2004; Sale-nius and Hari, 2003]. The frontoparietal low-beta band interaction can be related to cognitive aspects of active motor performance [Cannon et al., 2014; Pesaran et al., 2008] rather than pure movement patterning [cf. Gross et al., 2002].

The spectral subbands also differed in their topological emphasis. The temporal regions relied less on the highest (13–20 Hz) band and the frontocentral regions less on the lowest bands (6–9 Hz/9–13 Hz). This finding parallels the task-sensitive pattern where the picture comparison task, involving the temporal visual regions, relied relatively more on the theta/low-alpha band compared to the low-beta band. The object manipulation task, involving the frontocentral motor regions, showed a reversed pattern. This indicates that cortical topology and task-related spectral patterning are intertwined. Conversely, when network nodes were found in regions that are known to generate prominent spontaneous rhythms [at rest; Hari and Salmelin, 1997], such nodes did not seem to consistently utilize the spontaneous oscillatory bands for their coherent connectivity: the occipital regions did not primarily link to other areas at the high-alpha band of the posterior spontaneous rhythm [Hari and Salmelin, 1997], and the temporal regions relied on the high-alpha band [cf. Bonte et al., 2009] rather than on the theta/low-alpha band activity [spontaneous “tau” rhythm; Lehtelä et al., 1997] to connect with other areas.
Reconfiguration of Coherent Connectivity for Naturalistic Tasks

Previous imaging studies on electrophysiological connectivity have largely focused on spontaneous rest activity [e.g., Brookes et al., 2011; Hipp et al., 2012] or on fairly restricted, single-domain experimental paradigms in visual [e.g., Gross et al., 2004; Palva et al., 2010; Siegel et al., 2008], language [e.g., Alho et al. 2014; Kujala et al., 2007, 2012; Liljestrom et al., 2015] and motor domains [e.g., Gross et al., 2002; Jerbi et al., 2007; Pollok et al., 2005]. A naturalistic approach has since emerged to complement these results, as exemplified by research on natural vision [Betti et al., 2013; Watrous et al., 2013].

In the present study, the focus was on active, continuous performance of low-constrained tasks, with meaningful stimuli. In primates, long-range coherence has been enhanced by low-constrained tasks that require active decisional performance [Pesaran et al., 2008]. In ecological settings, neurocognition adapts to serve active behavior [Kingstone et al., 2008]. For understanding the brain–behavior relationship, naturalistic experimental approaches can importantly complement the more constrained designs that are often, unavoidably, rather artificial [see Felsen and Dan, 2005; Hasson et al., 2010; Ingram and Wolpert, 2011].

The naturalistic approach is likely to highlight behaviorally significant patterns of functional connectivity.

The present data-driven analysis of enhanced 6–20 Hz coherence in naturalistic tasks uncovered task-relevant hubs and coherent connections that are in good harmony with previous activation-, stimulation-, and lesion-based descriptions of human cortical organization. In the picture comparison task, coherent connections emerged between the temporoparietal–occipital hub region in the ventral (object/color-related) visual stream and the higher-order nodes of the dorsal (spatial/motor-related) stream [Goodale and Milner, 1992; Zeki and Marini, 1998]. The right-hemispheric frontoparietal coherence connectivity agrees with the attention-related cortical networks [Corbetta and Shulman, 2002, 2011]. The easy speech comprehension task highlighted coherent connectivity of central cortical regions with the left perisylvian cortex [Hickok and Poeppel, 2007; Turken and Dronkers, 2011], in line with the suggested ventral temporoparietal (speech recognition/meaning-related) and dorsal temporoparietal-frontal (articulation-related) cortical streams for speech perception [Hickok and Poeppel, 2007; Rauschecker and Scott, 2009]. In the object manipulation task, the main hub was observed in the left inferior parietal cortex that connected with the frontal cortex. This pattern is in good agreement with activation and stimulation studies on intentionally guided manual behavior [Binkofski et al., 1999; Desmurget et al., 2009; Haaland and Harrington, 1996; Rizzolatti and Sinaglia, 2010].

The demanding variants of the tasks, as contrasted to the task-mean pattern, emphasized partly different components of the network, possibly reflecting differential pre-dominance of distinct neurocognitive states in the demanding versus easy task. In the demanding picture comparison task, the relative emphasis was on the fronto-frontal low-beta coherence [cf. Womelsdorf et al., 2014] rather than the frontoparietal theta/low-alpha connectivity [cf. Daitch et al., 2013]. This may highlight the difference in strategy between the pop-up-type, stimulus-driven search in the easy comparison task and the more top-down-driven processing in the demanding task [cf. Buschman and Miller, 2007]. In the non-native speech comprehension, there was prominent connectivity involving the occipital visual regions (particularly the whole-cortex analysis) that may relate to visual imagery [Kosslyn et al., 1999] elicited by the movie soundtrack clips, without direct visual stimulation. The anterior temporal to frontal connectivity may reflect higher-order semantic processing [Visser et al., 2010]. The demanding Rubik’s task that required both object manipulation and visuospatial reasoning strongly engaged the frontoparietal connections [Binkofski et al., 1999; Corbetta and Shulman, 2002], as well as the visual system [Goodale and Milner, 1992]. The most numerous connectivity (at 6–20 Hz) observed in the Rubik’s task may be partly associated with the task requirement of active reasoning rather than automatic, rule-based performance [cf. Buschman et al., 2012].

The distinct task-sensitive patterns were complemented by a subset of network nodes that participated in the coherent networks regardless of the naturalistic task modality, thus reflecting regional one-to-many organization [cf. Mesulam, 2008]. There was prominent overlap of nodal regions in the frontoparietal cortex that is thought to form a multitask core structure in distributed neurocognitive architecture [Barbey et al., 2012; Duncan, 2010; Fox et al., 2005]. A major region of this overlap emerged in the (left) inferior parietal cortex, an associative region identified as an important hub of the human lateral cortex [Sepulcre et al., 2012]. Furthermore, task-independent nodes were found in the medial frontal and cingulate cortices, regions previously linked to a rich-club-type backbone of the cortical network [van den Heuvel et al., 2012]. The posterioromedial cortex, including the cingulate cortex, constitutes the major cortical hub of the resting-state activity [Greicius et al., 2003; Hagmann et al., 2008; Raichle, 2010]. The significant role for the posterioromedial cortex in effortful task performance in the present data supports the view that this central hub is not merely involved in purely intrinsic, task-negative function [Fox et al., 2005; Greicius et al., 2003; Raichle, 2010].

Cortex-Wide Mapping of Task-Sensitive Coherence

The all-to-all MEG mapping of interregional coherence in the human cortex extends the intracortical primate studies on singular task-sensitive coherent connections. Noninvasive electromagnetic brain imaging facilitates reliable
estimation of long-range corticocortical connectivity in the intact human brain [when method-inherent confounds and artifactual sources of signals (see Materials and Methods) are properly taken into account; Hipp and Siegel, 2013; Kujala et al., 2008, 2012; Schoffelen and Gross, 2009]; estimation of short-range connectivity is problematic because of the field spread of the electromagnetic signal [Schoffelen and Gross, 2009]. MEG reflects population-level dendritic activity in the cortical pyramidal cells [Hämäläinen et al., 1993; Okada et al., 1997], a measure that has a cellular-level counterpart in the local field potential (LFP) recordings [Buzsákí et al., 2012]. LFP signals in the primate cortex show interregional coherence between task-sensitive subpopulations of neurons [Buschman and Miller, 2007; Salazar et al., 2012] and are coupled with the spiking activity of the cells in the functionally connected cortical regions [Gregoriou et al., 2009; Pesaran et al., 2008]. In descriptive terms, neural coherence is a dynamic property of interregional signaling but, theoretically, it can also be regarded as a key mechanism for neuronal information processing.

The present results can be taken to support the hypothesis that oscillatory coherence plays a key role in the cortical implementation of neurocognition [Bressler and Kelso, 2001; Engel et al., 2013; Fries, 2005; Siegel et al., 2012]. Mechanistically, electrophysiological oscillations in dendrites can provide periodically opening temporal windows for effective neuronal communication [Fries, 2005; Izhikevich, 2001], and long-range coherence of these oscillatory signals can facilitate temporal coordination and interregional routing in the network signaling [Akam and Kullmann, 2010; Bosman et al., 2012; Engel et al., 2013; Siegel et al., 2012; Varela et al., 2001]. The exact functional interpretation of coherence depends on anatomical scaffolding, as tightly and bidirectionally connected regions, such as the frontal and parietal cortex, can engage in reciprocal hand-shake-type oscillatory interaction [Pesaran et al., 2008]; however, coherence can also reflect indirect, cortically, or subcortically mediated connections and thus more widespread synchrony in the brain network [Haider and McCormick, 2009; Saalmann et al., 2012; Siegel et al., 2012; Varela et al., 2001]. Here, the results do not dissociate between such different network configurations but, instead, they offer an overview of elevated coherent episodes occurring in the cortical networks in association with various naturalistic tasks.

The coherence mapping uncovered real-time task-sensitive oscillatory components of functional connectivity, instead of focusing on connectivity patterns that persist across task and rest periods alike [Power et al., 2011; Smith et al., 2009; Yeo et al., 2011]. The current approach agrees with electrophysiological studies that track oscillatory interaction (coherence, phase stability etc.) at the millisecond scale [Bressler and Kelso, 2001; Palva and Palva, 2012; Siegel et al., 2012] and differ from the slower-scale characterizations of band-passed power envelope correlation [Brookes et al., 2011; Hipp et al., 2012] and hemodynamic comodulation [Cole et al., 2013; Power et al., 2011; Yeo et al., 2011]. The millisecond-scale connectivity has been linked to on-going, task-related neuronal processing, whereas the slower-scale modulations presumably represent an intrinsic coupling mode related to the more general state or history of the network [Engel et al., 2013; Siegel et al., 2012]. A correspondence between such faster- and slower-scale characterizations is not straightforward. For example, electrophysiological power envelopes may be correlated at the low-beta band [e.g. Hipp et al., 2012] but such correlations are suppressed by task [Betti et al., 2013], unlike the millisecond-scale coherence that shows salient task-dependent enhancement. Conversely, the electrophysiological activity <20 Hz activity may correlate particularly well with the hemodynamic measures of functional connectivity [Wang et al., 2012].

Neurocognitive networks are widely considered an important concept in understanding brain function [Friston, 2008; Mesulam, 2008]. In the future, it will be important to deepen the understanding of the relationships between the intracranial and noninvasive electrophysiological and hemodynamic measures of functional connectivity, and of the way they capture network function in different behaviorally relevant tasks [see Betti et al., 2013; Park and Friston, 2013]. The present distinct connectivity patterns in varied naturalistic tasks provide an essential data-driven reference set for that important endeavor.

CONCLUSIONS

The present noninvasive mapping of task-sensitive long-range coherence shows that the human cortex self-organizes into distinct spatiotemporal patterns of interregional coherence to support naturalistic behavior. These findings provide empirical evidence that coherent oscillatory interaction at the narrow-band (<30 Hz) neuronal frequencies plays a significant role in the cortical implementation of neurocognition.

ACKNOWLEDGMENTS

The authors thank Hannu Laaksonen and Mia Liljestrom for comments on the manuscript. The authors declare no competing financial interests.

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