Title: Alpha and SSVEP power outperforms Gamma power in capturing Attentional Modulation in Human EEG

Abbreviated title: Effect of spatial attention on different neural measures in human EEG

Authors: Aritra Das[1], Nilanjana Nandi[1], Supratim Ray[1*]

Affiliations:

[1]Centre for Neuroscience, Indian Institute of Science, Bangalore, India, 560012

Telephone: +91 80 2293 3437

*Correspondence should be addressed to: Supratim Ray, Centre for Neuroscience, Indian

Institute of Science, Old TIFR Building, India 560012. E-mail: sray@iisc.ac.in

Author Contributions: A.D. and S.R. conceived idea of research, A.D. and N.N. collected

data; A.D, N.N, and S.R. analyzed data; A.D. and S.R. wrote the paper.

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Abstract

The effect of visual attention has been extensively studied using various techniques such as macaque neurophysiology that yields spikes and local field potential (LFP), and human electro/magneto encephalogram (EEG/MEG). Attention typically suppresses power at low frequencies such as alpha band (8-12 Hz) and increases power in gamma band (>30 Hz) in brain signals. In addition, EEG studies often use flickering stimuli that produce a specific measure called steady-state visually evoked potential (SSVEP), whose power also increases with attention. However, effectiveness of these various neural measures in capturing attentional modulation is unknown since the stimuli and task paradigms vary widely across these studies. In a recent macaque neurophysiology study with flickering stimuli, we found that the effect of attention was more salient in the gamma band and beyond of the LFP, compared to alpha or SSVEP. To compare this with human EEG, we designed an orientation change detection task where we presented both static and counterphasing stimuli of matched difficulty levels to male (N=15) and female (N=11) subjects, allowing us to compare attentional modulation of various measures under similar conditions. We report two main results. First, attentional modulation was comparable for SSVEP and alpha. Second, non-foveal stimuli produced weak gamma despite various stimulus optimizations and showed a negligible effect of attention although full-screen gratings showed robust gamma activity. Our results are useful for brain-machine-interfacing studies where suitable features depending on recording modality are used for decoding attention, and also provide clues about the spatial scales of neural mechanisms underlying attention.

Significance Statement

Various neural measures such as alpha and gamma band power or SSVEP power capture signatures of visual attention. A systematic comparison of their effectiveness in capturing attentional modulation is important for understanding different neural computations underlying attention and for developing brain-machine-interfaces (BMIs) that can decode the focus of attention. Since stimulus configuration and task paradigms vary widely across visual attention studies, we recorded all the relevant neural signals during an attention task under similar stimulus and behavioral conditions in human EEG. In contrast to invasive recordings in which gamma outperforms other measures, our results show that in human EEG, attentional modulation of alpha and SSVEP are comparable, and non-foveal stimuli produce weak gamma which is not well modulated by attention.

Introduction

Humans routinely use covert spatial attention in daily situations like searching for

objects, playing sports, crossing the road, or driving. This mode of attention helps us overtly

direct our attention to behaviorally relevant stimuli after selectively filtering out irrelevant

information (Desimone and Duncan, 1995; Desimone, 1998). The neural correlates of covert

spatial attention have been extensively studied using several techniques that record neural

activity spanning different spatial scales, ranging from single/multi-unit spikes and local field

potential (LFP) recorded using microelectrodes in multiple macaque visual areas, to diffuse

population-level neural activity recorded using human electro/magnetoencephalogram

(EEG/MEG). Attention-mediated modulation has been observed in evoked responses as well

as in alpha (8-12 Hz) and gamma (>30 Hz) bands of the LFP activity in multiple visual areas

in response to static/drifting stimuli (Fries et al., 2001, 2008; Chalk et al., 2010; Das and Ray,

2018) as well as in human EEG studies (alpha: Worden et al., 2000; Thut et al., 2006; gamma:

Gruber et al., 1999; Müller et al., 2000) and MEG studies (Bauer et al., 2014; Magazzini and

Singh, 2018). In addition, to address the poor signal-to-noise ratio in EEG/MEG recordings,

human EEG studies often use flickering stimuli that produce a specific neural measure called

steady-state visually evoked potentials (SSVEPs) to study the effect of covert spatial attention

(Morgan et al., 1996; Kim et al., 2007; Itthipuripat et al., 2014b).

Since the stimuli and task paradigms vary widely across all these studies, previous

studies have not attempted a thorough comparison of the effectiveness of these various neural

measures in capturing attentional modulation. However, this comparison is important for two

reasons. First, although earlier brain-machine interfacing (BMI) studies mainly focused on

motor decoding (Bansal et al., 2012; Hwang and Andersen, 2013), recent studies have

attempted to decode the focus of attention (De Sousa et al., 2021; Prakash et al., 2021;

Chinchani et al., 2022) as well. For such applications, it is important to compare how attention

modulates various neural measures. Second, different neural signals reflect important

underlying neural mechanisms/computations operating over different spatial scales. For

example, gamma oscillations are generated due to excitatory-inhibitory interactions (Buzsáki

and Wang, 2012), especially involving soma-targeting parvalbumin positive interneurons

(Bartos et al., 2007; Sohal et al., 2009), which remain coherent over a few millimeters (Jia et

al., 2011; Murty et al., 2018). On the other hand, alpha oscillations could reflect thalamo-

cortical gating mechanisms (Contreras et al., 1996). Recently, alpha has been linked with inter-

areal feedback processing, while gamma is thought to reflect feedforward processing (van

Kerkoerle et al., 2014; Michalareas et al., 2016). How attention modulates these measures

therefore puts important constraints on the mechanisms underlying selective attention. For

example, in a recent attention study in macaque V4 area, we found that the effect of attention

was more salient in the gamma band and beyond of the LFP as compared to alpha or SSVEP

(Prakash et al., 2021). We also compared the decodability of attentional state using spikes and

LFP power at different frequency bands and found that gamma and high-gamma power

recorded using a single electrode performed better than spikes; however, the decoding accuracy

was comparable for high-frequency LFP (>30 Hz) and spikes when several electrodes were

used. These results suggest that the spatial spread of mechanisms underlying attention could

be comparable to the spatial spread of LFP (Prakash et al., 2022).

Here, we tested whether these results hold in human EEG, since it is a non-invasive

recording technique routinely used in human BMI studies. Further, recording from a larger

recording scale may also reveal important clues about the spatial scales of neural mechanisms

of attention. To address this, we designed an orientation change detection task where we

presented both static and counterphase flickering stimuli of matched difficulty levels, which

allowed us to compare the attentional modulation of various measures under similar stimulus

and behavioral conditions.

Materials and Methods

Human Subjects

A total of 28 subjects (16 Males and 12 Females, aged between 21-38 years, mean age of 26.10 years; standard deviation: 3.26 years) were recruited from the student community of the Indian Institute of Science for the EEG experiments. The Institute Human Ethics Committee of the Indian Institute of Science approved all the procedures before the start of the study. All the subjects reportedly had normal or corrected-to-normal vision, although none of the individuals were explicitly tested for visual acuity. The subjects voluntarily participated in the experiment after signing informed consent forms and were monetarily compensated after completing the experiments.

EEG Recordings

Raw EEG signals were recorded from 64 active electrodes (actiCAP) using the BrainAmp DC EEG acquisition system (Brain Products GmbH). Electrodes were placed according to the international 10-10 system on a CAC-64 Brain Products EasyCap Layout. Electrolyte-gel (SuperVisc High Viscosity Electrolyte-Gel manufactured by EASYCAP GmbH) was applied to the scalp at each electrode contact site to lower the electrode impedances. Impedances of all electrodes were kept below 25 kΩ. Raw signals were filtered online between 0.016 Hz (first-order filter) and 250 Hz (fifth-order Butterworth filter), sampled at 1000 Hz, and digitized at 16-bit resolution (0.1 µV/bit). EEG signals were referenced to the FC_z electrode during data acquisition. We selected unipolar referencing for alpha, gamma, and SSVEP power estimation. We found that the bipolar referencing scheme, in which electrodes are re-referenced with respect to a neighboring electrode (see Murty et al., 2020, for details), gave slightly better gamma response than the unipolar referencing scheme for full-field gratings, similar to what

132   has been reported by Murty and colleagues in earlier publications (Murty et al., 2018, 2020).

133   However, for our dataset with non-foveal peripheral stimuli sets, the gamma responses were

134   poor irrespective of the referencing scheme. Therefore, we used the unipolar referencing

135   scheme for all our analyses.

137   **Experimental Design and Behavioral Tasks**
138

Each subject was seated comfortably in a dark room in front of an LCD screen. Their heads

were supported comfortably by a chin rest. The screen (BenQ XL2411) was gamma-corrected

with a resolution of 1280 x 720 pixels and a refresh rate of 100 Hz. The screen was placed at a

distance of 57 cm from the eye level of the subject. This arrangement allowed us to present

full-screen gratings which subtended a width of ~52° and height of ~30° of visual field at an

individual subject's retina. All the visual stimuli were calibrated to the viewing distance of 57

cm. The stimuli were presented by custom software running on the MAC operating system,

which controlled the task flow based on eye-tracking. The different types of behavioral tasks

are summarized as follows:

Protocol 1: Gamma Tuning Protocol using Fixation Task

The goal of this protocol was to find the orientation and spatial frequency of a full-screen

grating stimulus that produced the strongest gamma response. Every trial started with the onset

of a white circular fixation spot (0.1°), presented at the center of the screen. The subjects were

instructed to hold and maintain their gaze on that spot. After an initial blank period of 1000ms

since the start of fixation, a series of stimuli (2-3) was presented for 800 ms each, with an

interstimulus interval of 700 ms. These stimuli were full-screen sinusoidal luminance gratings

presented in full (100% luminance w.r.t to gray background) contrast. During stimulus

presentation, the gratings were either static or flickering at 16 Hz counterphase. The spatial

frequency (SF) of the gratings could take either of the two values: 2 or 4 cycles per degree

(cpd), whereas the orientation of the gratings could be any of the four values: 0°, 45°, 90°, and

135°. We chose these stimulus parameters as they were shown to produce robust gamma

responses previously (Murty et al., 2018, 2020). Stimuli were presented pseudo-randomly to

avoid any possible adaptation effects. Subjects performed this task for 15-20 minutes with each

of the eight combinations (2 SFs x 4 Orientations) presented pseudo-randomly 20-30 times

(Refer to Figure 1A). The subject's fixation was monitored online by EyeLink 1000 (SR

Research Ltd, sampled at 500 Hz) during the entire trial period. Characteristic auditory

feedback was provided to the subject for trial onset, fixation onset, trial abortion due to fixation

break, or correctly completing a trial by maintaining fixation. Twenty-eight subjects (16 Males

and 12 Females) performed this task at the beginning of a recording session. We determined

the spatial frequency and orientation of grating stimuli which generated the maximum gamma

band power response (25-70 Hz; see details of power estimation below). We used this spatial

frequency and orientation for the remaining protocols.

Protocol 2: Stimulus Size and Eccentricity Mapping using Fixation Task

Since covert spatial attention tasks typically use two stimuli presented equidistant from a

central fixation spot, we had to find the optimal stimulus eccentricity and size to elicit robust

neural responses in EEG. The size and eccentricity of non-foveal stimuli can profoundly affect

behavior and neural responses. If the stimuli are too big, it would be easier to detect orientation

changes in an attention protocol and subjects may only attend to the portion of the stimulus;

therefore, the attentional modulation of neural signals may be lower. On the other hand, if the

stimuli are small, they may not elicit a robust neural response in EEG. Similarly, the covert

attention task would become easier if the stimuli are presented near the fixation spot (lower

eccentricity). But if stimuli are presented too far from the fixation spot, they may not produce

a robust response, or, in our attention task design (described below), it could be difficult for

subjects to make eye movements toward the target within a specified time.

We therefore tested several possible combinations of stimulus size and eccentricity to

find a combination that produced robust neural signals in EEG. Every trial started with the

onset of a white fixation spot (0.1°). We presented two achromatic grating stimuli located

equidistant from the central fixation spot in the lower visual hemifield with the azimuth and

elevation of the center of the two stimuli at (-X°, -Y°) and (X°, -Y°), with the diameter being

D° of visual angle. In four consecutive experiments, we varied the values of X, Y, and D set at

2°, 4°, 6°, or 8° (refer to Figure 1B). The eccentricities of the four non-foveal stimuli sets, i.e.,

the distance between the center of the stimulus and the center of vision (point of fixation or

fovea), were 2.8, 5.6, 8.5, and 11.3°, respectively. After an initial blank period of 1000 ms after

the start of fixation, a series of stimuli (2-3) was presented for 800 ms each, with an

interstimulus interval of 700 ms. These stimuli were full-screen sinusoidal luminance gratings

presented in full (100% luminance w.r.t to gray background) contrast. During each stimulus

presentation, either both the gratings were static, or the left and right gratings were

counterphase flickering at 12 Hz and 16 Hz, respectively, to mimic the experimental paradigm

for the attentional task later. We presented both the stimuli in the lower visual hemifield

because the occipital electrodes placed on the scalp are directly below the region of the primary

visual cortex above the calcarine fissure in the occipital cortex of the human brain, and they

are known to be retinotopically mapped to the visual cortex from the lower visual field. The

subjects performed each of the four protocols for 5-10 minutes with 25-30 repeats each for

static and flickering stimuli presentations.

The first two protocols were run on 11 subjects (7 Males and 4 Females) before any

attention task (Protocols 3 and 4; see below) were done. We found that X, Y, and D of 4° gave

the best response for SSVEPs, while the responses for alpha and gamma did not depend on the

eccentricity (Fig 3). Therefore, we fixed this stimulus configuration for all subjects. Attention

task on these 11 subjects was done on a different day. For the remaining subjects, Protocol 1

was followed by Protocols 3 and 4 and all three protocols were done on the same day.

Protocols 3 and 4: Attention Task

We finally used the spatial frequency, orientation, size, and eccentricity optimized for the

attention task, where we presented both static and flickering stimuli to measure all the relevant

neural measures. Subjects were required to maintain their gaze within 2° of a small central dot

(diameter of 0.10°) during the task while covertly attending to one of the two achromatic

grating stimuli presented in the lower hemifield equidistant from the fixation spot. The subjects

were cued to one of the two stimulus locations (left or right) in different blocks of trials by

presenting an instruction trial at the start of the block, in which there was only a single stimulus

either on the left or right of the fixation spot, indicating the location to be attended in subsequent

trials of that block. During each trial, either both the stimuli were static, or both were

counterphase flickering, one at 12 Hz and the other at 16 Hz. We used 12 Hz and 16 Hz

frequencies since these had 3 and 4 complete cycles during a 250ms interval. In our protocol,

one "stimulus" was considered to be of 250 ms, and a sequence of such stimuli was presented

without any interstimulus interval, and therefore appeared as a continuous stimulus flickering

at 12 or 16 Hz. The contrasts of both the attended and ignored locations were at 100% in each

trial. There was an initial time period of 1000ms at the beginning of each trial when the subject

maintained their fixation, followed by the stimulus onset. Following stimulus onset, the

orientation of the stimulus at the cued location changed (target). The target appeared at an

unstipulated time drawn from an exponential distribution (mean 2000 ms, range 250-5000 ms)

following the initial stimulus lead period of 1000 ms. The initial stimulus lead period was used

to avoid transient activity generated by the event-related potential during the 250 ms after the

233 first stimulus onset. An exponential distribution was used for target appearance to minimize

234 the expectation of target appearance and to keep the attentional state uniform during a trial

235 since the hazard function is flat for an exponentially distributed target onset time. Subjects were

236 required to make a saccadic eye movement towards the target within 800 ms of the target onset.

237 Once the target appeared, it remained at the changed orientation for 800 ms. To account for

238 saccade latency and minimize guessing, saccades beginning at least 100 ms after the target

239 appearance were considered correct trials. We also introduced catch trials in which a trial was

240 terminated at 6250 ms if the target had not appeared till then (15-20% of all trials).

241 Characteristic auditory feedback was provided to the subjects for either correctly detecting

242 targets in a trial (Hit) or for maintaining fixation throughout catch trials. A different auditory

243 sound was provided as feedback for failing to detect the target (Miss) or fixation breaks. For

244 the non-catch trials, targets appeared at the cue location (cue validity 100%; no invalid trials).

245 This task design was used to rule out any distributed attention strategy the subjects may use,

246 which can influence the neural responses.

247   Note that our task design made the detection of a target for the static case much easier

248 than the flickering case. For counterphase flickering stimuli, the contrast varied sinusoidally to

249 complete 3 or 4 cycles within a 250 ms "stimulus period", so the target stimulus (with a

250 different orientation) appeared initially at 0% contrast at the beginning of the target stimulus

251 onset (in other words, the orientation change occurred when the stimulus contrast was 0%).

252 This was not the case for static stimuli, making them much easier to detect. Therefore, to match

253 the difficulty levels of the static and flickering stimuli, we performed 2-3 blocks of attention

254 task with five different orientation changes for each of the static, 12 Hz, and 16 Hz stimuli for

255 each of the two attended locations (left and right). Each block comprised of 10 trial repeats for

256 both hit and miss trials combined. We plotted the behavioral accuracy for the two attended

257 locations for all five orientation changes. We fitted the combined behavioral performance for

both left and right locations with a Weibull function for psychometric analysis. We determined

the fitted orientation change value for each temporal frequency (TF; static, 12, and 16 Hz), for

which the behavioral accuracy was at 50%.

We then set the single orientation change value obtained from the above manual

behavioral staircase procedure for each of the TFs for our main attention task. The subjects

again performed the main attention task for six blocks for each of the two attended locations

where either both static or both flickering stimuli were presented in each trial. Each block

comprised of 10 trial repeats for both hit and miss trials combined. We had at least 100-120

trial repeats for each TF with hits and misses conditions combined for each subject. Two

subjects (one male and one female) were unable to perform the attention task. Therefore, the

attention dataset had 15 male and 11 female subjects.

Data Analysis
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272   All data were analyzed using custom codes written in MATLAB (The Mathworks, RRID:

273   SCR_001622). Individual data analysis methods are briefly summarized below.

274   ***EEG Artifact Rejection and Electrode Selection***
275

We followed a standard set of procedures for electrode selection and artifact rejection in human

EEG experiments reported previously in detail (Murty et al., 2020; Murty and Ray, 2022). We

briefly describe the procedure here with a minor modification. To find bad stimulus repeats for

each electrode, we applied a thresholding procedure on raw waveforms (high pass filtered the

signal at 1.6 Hz to eliminate slow drifts) and multi-taper PSD in the analysis period (Fixation

Protocols: -500 ms to 750 ms of stimulus onset; Attention Protocol: combined periods of -1000

ms to 1250 of the first stimulus onset in a trial and -1000 to 0 ms of target onset) using the

Chronux toolbox in MATLAB (Bokil et al., 2010). We labeled a particular stimulus repeat as

bad if the RMS value of the time-series signal was beyond a pre-specified threshold (min: 1.5

µV, max: 35 µV) or PSD of that stimulus repeat deviated by six times the standard deviation

from the mean at any frequency bin between 0 and 200 Hz for a given electrode. The only

difference in EEG artifact rejection in this study compared to previous studies is the usage of

RMS threshold for the time-series signal instead of the standard deviation-based threshold

around the mean of the time-series signal reported in our previous EEG studies (Murty et al.,

2020; Murty and Ray, 2022). If an electrode had more than 30% bad repeats, we discarded that

electrode. We then created a set of common bad stimulus repeats across the remaining

electrodes if a particular repeat was labeled as bad in more than 10% of those remaining

electrodes. Finally, any repeat that was labeled bad in any of the ten electrodes (P3, P1, P2, P4,

PO3, POz, PO4, O1, Oz, and O2) was added to the pool of common bad repeats, thus finalizing

the list of common bad stimulus repeats. In addition, we removed electrodes with baseline PSD

slopes less than zero in the gamma range of 56-84 Hz (for details, please refer Murty et al.,

2020). Using this set of stringent criteria, we rejected ~20% of data for fixation protocols,

similar to previous reports by Murty and colleagues. However, this increased to ~30% when

we considered attention datasets where we checked bad trials for the combined periods of -

1000 ms to 1250 of the first stimulus onset in a trial and -1000 to 0 ms of target onset, since

two epochs were considered instead of one. For all the experiments, we analyzed occipital eight

electrodes (P1, P3, PO3, and O1 on the left hemisphere and P2, P4, PO4, and O2 on the right

hemisphere). Previous EEG studies from our group have also used the similar set of occipital

electrodes to characterize alpha and gamma oscillations (Murty et al., 2018, 2020) and SSVEP

response (Murty et al., 2020; Liza and Ray, 2022).

EEG data analysis

Power spectral densities (PSDs) for different stimulus or attention conditions were computed

using the multi-taper method with one Slepian taper using the Chronux toolbox (Bokil et al.,

2010; https://chronux.org/, RRID: SCR_005547).

For Figures 2B and 2D, the analysis period was selected between 250 ms to 750 ms after

stimulus onset to avoid stimulus onset-related transients and compared against a 'baseline

period' between -500 and 0 ms of stimulus onset (a gray screen). This yielded a frequency

resolution of 2 Hz for the PSDs. Absolute PSDs were expressed in logarithmic units (base 10),

and the changes in PSDs were plotted with respect to the baseline response:

$$\Delta PSD = 10 \left(log_{10}(ST) - log_{10}(BL) \right) \qquad \text{(Eq. 1)}$$

where ST is stimulus power spectra (across frequency f) averaged across repeats for all eight

SF-Ori combinations (two SFs and four orientations) or stimulus combination that produced

the maximum gamma response and all analyzable electrodes. BL denotes the baseline power

spectra averaged across repeats for all stimulus combinations and analyzable electrodes.

We calculated the change in alpha and gamma power (Figures 2 and 3) as follows:

$$\Delta Power = 10 \left(\log_{10} \frac{\sum_f ST(f)}{\sum_f BL(f)} \right) \qquad \text{(Eq. 2)}$$

where ST (all/max gamma stimulus combination) and BL (all) are stimulus and baseline power

spectra (across frequency f) averaged across repeats and analyzable electrodes. For alpha, $f \in$

[8 12] Hz, and for gamma, $f \in$ [25 70] Hz. For changes in SSVEP power, we calculated the

power at the 2nd harmonic of the counterphasing frequency. The fundamental frequency

response (the first harmonic of the flickering frequency) in the PSD, generated due to activity

of cortical simple cells, mostly averages out due to the random spatial distribution of the ON

and OFF sub-regions of their receptive fields with respect to the symmetric luminance flickers

of counterphase flickering gratings (Campbell and Maffei, 1970; De Valois et al., 1982; Di

Russo et al., 2001; Hou et al., 2003). The SSVEP response at second harmonic ($f_2 = 2f_1$) arises

due to activity of the frequency doubling cortical complex cells (De Valois et al., 1982; Di

Russo et al., 2001; Kim et al., 2007) whose responses are largely spatial-phase invariant.

For attention results in figures 5 (static gratings), and 7 (flickering gratings), the

analysis period was selected for the pre-target period between -1000 ms to 0 ms of target onset

and compared against a 'baseline period' between -1000 and 0 ms of first stimulus onset (a

gray screen). This yielded a 1 Hz frequency resolution in PSDs. The change in PSDs (violet

traces in figures 5C, and 7B) between attended and ignored conditions were computed as

follows:

$$340 \qquad \Delta PSD = 10 \left(\log_{10} \frac{pTG_{Att}}{pTG_{Ign}} \right) \qquad \text{(Eq. 3)}$$

341   Where pTG_{att} and pTG_{Ign} are pre-target power spectra for attended and ignored conditions,

342   respectively (across frequency f), averaged across trials (static and flickering trials separately),

343   attention conditions, and analyzable electrodes.

344   We calculated the change in power in attended and ignored conditions w.r.t. to baseline period

345   using the following equation:

$$346 \qquad \Delta Power = 10 \left(\log_{10} \frac{\sum_f pTG_{Att\ or\ Ign}(f)}{\sum_f BL_{Ign}(f)} \right) \qquad \text{(Eq. 4)}$$

Where pTG and BL are pre-target, and baseline power spectra (across frequency f) were

averaged across either static grating conditions (Figure 5A-B 'Attend Left' and 'Ignore Left'

scalp maps) or flickering grating conditions (Figure 7A 'Attend Left' and 'Ignore Left' scalp

maps) for alpha, $f \in [8\ 12]$ Hz, and for gamma, $f \in [25\ 70]$ Hz ranges. For flickering gratings,

the power at SSVEP harmonics were excluded during computation of gamma band power. For

changes in SSVEP power, we calculated the power at the frequency twice that of the

counterphasing frequency (i.e., 24 Hz or 32 Hz depending on the attended or ignored grating

flickering at 12Hz or 16 Hz, respectively). We use equation 4 to calculate power change in bar

plots, with pTG and BL being the pre-target and baseline power spectra (across frequency f),

which were averaged across all trials (static and flickering trials separately), attention

conditions, and analysis electrodes. For delta PSDs (Figure 5C and Figure 7B) and bar plots

(Figure 5D and 7C), we computed the change in PSD/power between attend-left/ignore-left

conditions for the four right-hemispheric occipital electrodes (P2, P4, PO4, and O2) and for

attend-right/ignore-right conditions for the four left-hemispheric occipital electrodes (P1, P3,

PO3, and O1). Please note that the same baseline power is subtracted from both attention

conditions to rule out the possibility that the results are due to changes in baseline power.

All the absolute PSDs, relative PSDs, and change in power values computed for an

individual subject (equations 1-4) were subsequently averaged across subjects, whereas, for

scalp maps of 64 unipolar electrodes, each analyzable electrode data was averaged across

analyzable repeats/trials, attention conditions and subjects (Figures 2, 3, 5-7). Please note that

we also removed the three frequency bins centered around the line noise peak (52 Hz; Figures

2 and 3) and five frequency bins centered around the line noise peak (51 Hz; Figures 5 and 7)

while calculating gamma power to remove the line noise artifact.

Scalp Maps (Figure 2, 3, 5-7)

Unlike the scalp maps for alpha, gamma, and SSVEP in the gamma tuning protocol (Figure 2)

and alpha and gamma in the size-eccentricity protocol (Figure 3), we consolidated the SSVEP

responses for our size-eccentricity protocol (Figure 3A; bottom two rows) and attention

protocol (Figure 7A; bottom two rows) for a simple and effective comparison across different

neural measures by presenting SSVEP responses on the right hemispheric occipital electrodes

in the consolidated scalp maps. To achieve this, we mirror-flipped the scalp map for the 32 Hz

SSVEP response for 16 Hz counterphasing grating presented on the lower right visual

hemifield in our size-eccentricity mapping protocols with respect to the midline set of

electrodes (z-line of ActiCap-64 Cap) and averaged with the 24 Hz SSVEP response for the 12

Hz counterphasing grating presented on the lower left visual hemifield. Similarly, for the

attention protocol, the SSVEP responses for all four attention conditions (two attended

locations and two TFs; figure 6A-D) were averaged after mirror-flipping the scalp maps for

two attend-right (12 Hz/ 16 Hz) conditions. For the alpha and gamma scalp maps for flickering

gratings, the scalp maps were averaged across only attend-left 12 Hz and attend-left 16 Hz

conditions.

SSVEP power calculation

We computed SSVEP power in two different ways, which are as follows:

Method 1: We computed the power for each trial and subsequently computed the mean of the

power across all trials. We termed this the **trial-by-trial SSVEP power estimate** (Itthipuripat

et al., 2014b; Murty et al., 2020; Salelkar and Ray, 2020)**.**

Method 2: Averaged the EEG signal across all trials and then computed the power from the

averaged signal. We called this the **trial-averaged SSVEP power** (Morgan et al., 1996; Liza

and Ray, 2022).

Statistical Analysis

Statistical tests were performed for each of the three neural measures to check whether attention

significantly modulated these measures or whether the magnitude of these modulations differed

between different pairs of neural measures. We used pairwise t-tests to compare the magnitude

of attentional modulation on different neural measures (Figures 5D and 7C). Since gamma and

SSVEP showed an increase in attentional modulation whereas alpha showed suppression with

attention, for statistical comparisons of magnitude of attentional modulations between alpha

and other neural measures, we changed the signs of alpha power changes and then performed

pairwise t-tests. We also performed t-tests for each neural measure to check whether they were

significantly modulated by attention.

Results

Full-field stimuli produce robust alpha, gamma, and SSVEP responses in human EEG

We first analyzed the three neural measures (alpha, gamma, and SSVEP power) by recording EEG signals for full-field stimuli presented to subjects passively fixating at the central fixation dot. We recorded full-field stimulus responses for both static as well as flickering stimuli for 28 subjects and analyzed the results for eight electrodes (Electrodes P3, P1, PO3, and O1 on the left hemisphere and P2, P4, PO4, and O2 on the right hemisphere) referenced to the FCz electrode. We computed the raw power separately between 250-750 ms after stimulus onset and a baseline period of -500 ms to 0 ms of stimulus onset for all the electrodes of an individual subject. We averaged the power across the occipital electrodes and computed the power changes in decibels for an individual subject, followed by averaging the power changes across all the subjects.

Figure 2A left column shows the scalp maps separately for alpha, gamma, and SSVEP responses for all SF-Ori combinations for all subjects. All sixty-four electrodes are shown as dots in scalp maps, with the analysis electrodes highlighted as bold dots. The full-field stimuli produced strong alpha suppression (~ -2 dB) and robust gamma power (~1 dB) for static stimuli in the parietal, parieto-occipital, and occipital electrodes. We also found strong SSVEP power (trial-by-trial estimate; see Methods for details) at 32 Hz (~8 dB), the second harmonic of the 16 Hz counterphase flickering stimuli. The scalp maps in Figure 2A right column show the alpha, gamma, and SSVEP responses for the spatial frequency and orientation that produced the maximum gamma response for the eight analysis electrodes. As expected, we observed an increase in gamma power in this stimulus combination. Alpha and SSVEP responses expectedly remained almost similar for the two conditions (left and right panels in A) since the stimulus combination was chosen to maximize the gamma response.

We computed the PSDs of the eight analysis electrodes in Figure 2B for static stimuli, which

show suppression in the alpha band for the stimulus period (black trace) with respect to the

baseline period (green trace). The blue trace indicates the change in PSDs. The PSDs further

show an increase in gamma band power for all SF-Ori and max Gamma SF-Ori combinations

for static gratings (Figure 2C) and an increase in SSVEP power for flickering gratings (Figure

2D). We quantified alpha and gamma power for static stimuli and SSVEP for flickering stimuli

in Figures 2C and 2E, respectively. Even though we observed an increase in gamma power for

flickering stimuli, the PSDs show that the counterphase stimuli produced strong SSVEP signals

at the second harmonic and all subsequent even harmonics, thus interfering with gamma

response. Therefore, for the gamma power calculation in Panel E, we removed the power at

SSVEP harmonics. Taken together, the magnitude of SSVEP power was much higher (~8-9

dB) compared to the magnitude of alpha suppression (~2 dB) and gamma (~1-1.5 dB). These

results were consistent with previous EEG studies (Murty et al., 2018, 2020).

Comparison of alpha, gamma, and SSVEP responses in human EEG for varying size and

eccentricity of non-foveal stimuli

Next, we used the particular spatial frequency and orientation that produced maximum gamma

response for the size and eccentricity mapping protocols in 11 subjects. We systematically

varied both the azimuth and elevation of the stimulus center as well as the diameter of the

stimulus from 2° to 8° in steps of 2 degrees of visual angles in four consecutive experiments.

We show the scalp maps for the alpha, gamma, and SSVEP power for these different

experiments in Figure 3A, along with the scalp maps for full-screen stimuli in the leftmost

column for comparison. We found robust alpha suppression for all sets of non-foveal stimuli,

although it was reduced compared to full-screen gratings in the occipital electrodes. The alpha

suppression decreased in the non-foveal gratings by about 0.4-0.6 dB (Figure 3B).

For gamma power, non-foveal stimuli produced a negligible response despite showing

a robust gamma response of ~1 dB in full-field stimuli (Figure 3A gamma scalp maps and

Figure 3C). In an earlier study on humans with size tuning protocols, Murty and colleagues had

varied the diameter of foveal gratings from 1°-32° in steps of 2° in addition to full-screen

gratings (>32°). They found the magnitude of gamma response progressively increased with

increasing stimulus size, and they also found two distinct bands, slow (20-40 Hz) and fast (40-

70 Hz), from a diameter of 4° onwards (Murty et al., 2018). However, we did not observe

gamma responses in either band when we separately analyzed both slow and fast gamma bands

for our non-foveal stimuli (data not shown).

Next, we computed the SSVEP power for each electrode of an individual subject in two

ways: 1) trial-by-trial SSVEP power computed for each stimulus repeat and subsequently

averaged across repeats, and 2) trial-averaged SSVEP power computed by averaging the signal

across stimulus repeats and then computing the power (see Methods). The trial-averaged

SSVEP is often used in EEG studies to estimate SSVEP power better because averaging the

trials before power calculation enhances the signal-to-noise ratio at the SSVEP frequency

(second harmonic and all subsequent even harmonics). For the trial-by-trial SSVEP power

estimate, the best SSVEP response was ~1 dB in the analysis electrodes for stimuli presented

for location and diameter of 4° compared to other non-foveal stimuli sets, although it was much

smaller than the full-field response (Figure 3D). The SSVEP response, however, increased to

about ~6 dB for diameter and location of 4° when computed by the trial-averaged method. Note

that we combined the SSVEP responses at 24 Hz from the right-side analyzable electrodes for

the 12 Hz flickering stimuli on the left and SSVEP responses at 32 Hz from the left-side

analyzable electrodes for the 16 Hz flickering stimuli on the right. For scalp maps of SSVEP

response, we mirror-flipped the scalp maps for 32 Hz along the midline of our cap layout to

show on the right hemisphere, which is why the SSVEP response appears lateralized to the

right side. Based on these results from 11 subjects, we selected the grating location and

diameter of 4° for attention tasks because it showed the maximum SSVEP response and robust

alpha suppression.

Behavioral Results for Attention Task
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The difficulty levels for static and flickering stimuli in the attention task were matched by

fitting psychometric curves for behavioral performances for a 2-3 block pilot attention

experiment for each of the attended sides (left or right) with five possible values of orientation

change. A single orientation change value for each of the temporal frequencies (static, 12 Hz,

and 16 Hz) of the cued location was obtained from the fitted values for which the behavioral

performance was at 50%. We used these orientation changes for static and flickering stimuli in

the main attention task (see methods section). Subjects performed at least five to six blocks on

each attended side (left or right). In Figure 4, we present the behavioral performances of all 26

subjects who performed the main attention task. Behavioral accuracies were similar in all

attended locations and temporal frequencies, except for the att-left 16 Hz condition (Figure

4A). The behavioral accuracy for att-Left 16 Hz, when compared to other attention conditions,

was significant (all the p values < 0.05) whereas the behavioral accuracy was not significantly

different between the other attention conditions (all p values > 0.18). The orientation thresholds

(Figure 4C) were set to yield a behavioral performance of 50%, but the actual performance was

slightly better (~70% on average), potentially due to practice and familiarization with the task

paradigm. Response times for static gratings (both left and right attended locations) were lower

than for flickering gratings (Figure 4B). The pairwise t-tests between the response time

between any static and flickering stimuli were significant (all the p values $< 1.12 \times 10^{-5}$).

Orientation change values (target) were considerably smaller (~0.6-1°) for static as compared

to flickering stimuli (~10°), as described in the Methods section.

Comparison of attentional modulation on alpha and gamma response for static stimuli

We first analyzed the pre-target alpha and gamma responses (-1000 ms to 0 ms w.r.t the target

onset) for static stimuli. We used the baseline (-1000 ms to 0 ms w.r.t the stimulus onset in a

trial) alpha power of the attend-right condition to subtract from both the attend-left and attend-

right (which also means ignored-left) pre-target period. The scalp maps for change in alpha

power w.r.t baseline are shown in Figure 5A for attend-left and attend-right conditions. The

elliptical disks in front of each scalp map show the attended location (left or right) as filled,

and the ignored location is shown as blank disks. Alpha power was suppressed in the analyzable

electrodes contralateral to the attended location. For the attend-left condition, we found alpha

suppression in the contralateral analysis electrodes on the right hemisphere and vice versa for

the attend-right condition. However, the alpha suppression was not lateralized in attend-right

conditions like attend-left conditions. However, when we computed the scalp map of the

difference between attend-left and attend-right (i.e., ignored-left), we observed alpha

suppression on the right-side analysis electrodes (contralateral to attended side) compared to

the left electrodes (ipsilateral to the attended side), consistent with previous findings on the

attention-mediated alpha suppression in the occipital regions in the contralateral hemisphere in

both human EEG studies (Foxe et al., 1998; Sauseng et al., 2005; Rihs et al., 2007) and monkey

LFP studies (Chalk et al., 2010; Das and Ray, 2018; Prakash et al., 2021).

However, we found negligible effect of attention on gamma power in the scalp maps

and the analysis electrodes (Figure 5B). We quantified the attention-modulated band power

response for alpha and gamma for the occipital analysis electrodes (see method section). We

found significant alpha suppression with attention (t-test, N= 26, p = 3.3×10^{-3}) but negligible

gamma modulation (t-test, N= 26, p = 0.88), with significant difference in attentional

modulation between alpha and gamma magnitudes (Figure 5D; t-test, N= 26, p = 0.03).

Comparison of attentional modulation of alpha, gamma, and SSVEP response for

flickering stimuli

For the SSVEP response for the flickering condition, four conditions had to be combined since

the attention task had two attended locations (left or right) and two different frequency tags (12

and 16 Hz). Figure 6A-D shows the scalp maps for all these four conditions. The filled and

blank elliptical disks in each scalp map indicate the attended location (left or right), and the

ignored location, respectively, with the flickering frequency tags mentioned on top of the disks.

As expected, the SSVEP responses for all combinations were higher for the attended condition

than the ignored condition for the contralateral occipital electrodes, consistent with an earlier

study (Joon Kim et al., 2007; see their Figure 3). We consolidated all the SSVEP responses for

a simple and effective comparison across different neural measures by presenting SSVEP

responses on the right hemispheric occipital electrodes for a consolidated scalp map as if the

attended location was always on the left. To achieve this, we mirror-flipped the scalp maps

with respect to the midline set of electrodes (z-line of ActiCap-64 Cap) for the attend right

conditions (Figure 6B and Figure 6D), and then averaged the four conditions (Figure 6E).

Finally, we compared the attentional modulation of alpha, gamma, and SSVEP responses for

flickering stimuli. We combined the alpha responses for both 12 Hz and 16 Hz flickering

stimuli. Similar to the static stimuli, we found significant alpha suppression in the analyzable

occipital electrodes in the contralateral hemisphere to the attended location (Figure 7A, top

row; t-test, N= 26, p = 2.32×10^{-6}). Interestingly, we found more substantial alpha suppression

for flickering stimuli (mean $\pm$ SEM: -0.69 $\pm$ 0.11 dB across 26 subjects) compared to the static

stimuli (-0.43 ± 0.13 dB), potentially because the flickering stimuli produced stronger and more

sustained activation of the cortex compared to the static stimuli. Again, attentional modulation

of gamma was negligible for flickering stimuli, as can be observed in the scalp maps (Figure

7A, second row) and bar plots (Figure 7C, t-test, N= 26, p = 0.09). We computed SSVEPs

using two methods (please refer to the Methods section) and combined different conditions (as

discussed previously). SSVEP responses were enhanced in the trial-averaged estimate (Figure

7A, the bottom row) compared to the trial-by-trial estimate (Figure 7A, third row). This can be

better appreciated by comparing the SSVEP power at 24 and 32 Hz in the trial-by-trial PSD

estimates (Figure 7B, top) with the trial-averaged PSD (Figure 7B, bottom). We used data from

19 subjects for this panel, since for seven subjects, the SSVEP response appeared at 23 Hz and

31 Hz instead of the expected 24 and 32 Hz because of an extra stimulus frame during the

presentation of flickering stimuli. Overall trends for these seven subjects were similar except

the SSVEP peaks were at 23 and 31 Hz. For scalp map and bar plot analyses, we pooled the

responses at 23 and 31 Hz for seven subjects and 24 and 32 Hz for the remaining 19 subjects.

The SSVEP response at 24 and 32 Hz (cyan lines) improved when we computed the trial-

averaged SSVEP, although the standard deviation was also higher.

We found a strong modulation in SSVEP response with attention (t-test, N= 26, p = 2.43×10^{-3})

which was enhanced even more (t-test, N= 26, p = 4.48×10^{-4}) if we computed SSVEP using

the 'trial-averaged estimate.' There was a significant difference between attention-mediated

alpha and gamma modulation (Figure 7C t-test, N= 26, p =1.61×10^{-4}; note that alpha values

were sign reversed before statistical analysis) as well as for gamma and trial-by-trial SSVEP

modulation (t-test, N= 26, p = 6×10^{-3}). The modulation in alpha was comparable to the trial-

by-trial SSVEP estimate (t-test, N= 26, p = 0.96) but significantly different from the trial-

averaged SSVEP estimate (t-test, N= 26, p = 6.6×10^{-4}). Taken together, SSVEP and alpha

outperformed gamma in capturing attentional modulation in human EEG studies.

Discussion

We used both static and flickering stimuli to record three relevant neural measures: alpha

power, gamma power, and SSVEP power in human EEG experiments under similar behavioral

conditions and compared the magnitude of attentional modulation of those measures. Non-

foveal stimuli did not produce robust gamma response in spite of several stimulus

optimizations. Alpha and SSVEP were comparable and outperformed gamma in capturing the

attentional effect in human EEG. Finally, the effect of attention was more prominent for trial-

averaged SSVEP compared to trial-by-trial SSVEP.

Attention studies at different recording scales - macaque LFP vs. human EEG studies

Alpha and Gamma Power

The effect of attention on alpha and gamma power in LFP signals has been studied extensively

in multiple visual areas along the visual pathway, such as in V1 (Chalk et al., 2010; Das and

Ray, 2018) and V4 (Fries et al., 2001; Prakash et al., 2021), as well as in human EEG studies

(Klimesch et al., 1998; Worden et al., 2000; Sauseng et al., 2005; Thut et al., 2006). A few

studies found attentional modulation of stimulus-induced gamma band activity in human EEG

by introducing a unique stimulus and task paradigm where subjects had to attend to a moving

bar or other moving stimuli on the left or right visual field (Gruber et al., 1999; Müller et al.,

2000), which were different from Posner-spatial cueing task paradigm. We however used non-

foveal grating stimuli in our task paradigm to remain consistent with classical spatial attention

studies in macaques. Previous macaque LFP and human EEG studies have reported that gamma

rhythms are stimulus-induced and depend on various stimulus properties such as spatial

frequency, orientation, size of the stimuli (Ray and Maunsell, 2010; Jia et al., 2013; Murty et

al., 2018) as well as eccentricity of the stimulus in human EEG/MEG studies

(Muthukumaraswamy and Singh, 2013; van Pelt and Fries, 2013). We designed a systematic

approach to find the optimum size and eccentricity of grating stimuli that would produce robust

gamma response without compromising the difficulty of our covert spatial attention task but

failed to elicit strong gamma response or gamma modulation with attention. This could be due

to three reasons. First, since we used peripheral gratings, we expected gamma response to be

reduced due to reduced cortical magnification factor for peripheral stimuli (Daniel and

Whitteridge, 1961; Slotnick et al., 2001). Indeed, we found negligible stimuli-induced gamma

response in all four size-eccentricity mapping protocols studies (Figure 3), although we used

peripheral gratings of similar size and eccentricities used by van Pelt and Fries in their MEG

study (2013). Second, gamma is a "local" signal which is coherent only over a short spatial

scale (see Mechanisms section below for more discussion) and therefore may not be well

represented in the diffuse neural activity recorded in EEG. Third, the effect of attention on

gamma power is variable across visual cortex, with reduction in power in V1 and enhancement

in V4 (Chalk et al., 2010). These effects may cancel out in an aggregate signal such as EEG,

leading to negligible overall attentional modulation.

SSVEP Power

A recent macaque LFP study (Prakash et al., 2021) in area V4 found that alpha and gamma

outperformed SSVEP in capturing attentional modulation, possibly due to the SSVEP

frequency (10 Hz counterphase) being close to the alpha range, as shown in previous studies

(Gulbinaite et al., 2019). Attention studies using human EEG have specifically used SSVEP

power to study attentional modulation (Morgan et al., 1996), to understand sensory gain in the

context of attention (Kim et al., 2007; Itthipuripat et al., 2014a), and to test out predictions of

the normalization model of attention on neural gain mechanisms (Itthipuripat et al., 2014b).

We also found an increase in SSVEP power with attention which was higher if SSVEP was

computed on the trial-averaged signal compared to trial-by-trial basis. While alpha suppression

and SSVEP enhancement are routinely observed in EEG studies, to our knowledge, no prior

study has systematically measured the relative modulation by attention under matching

stimulus and behavioral conditions.

Neural mechanisms of alpha and gamma

Our results, together with our previous report using LFP (Prakash et al., 2021) in which

attentional modulation was strongest for gamma frequencies and beyond, provide important

clues about the spatial scales over which mechanisms related to various oscillations operate.

For example, previous studies have pointed out a thalamo-cortical origin of occipital alpha

oscillations with conflicting evidence of alpha pacemakers being present either in the thalamus

(Saalmann et al., 2012; Vijayan and Kopell, 2012) or the cortex (Bollimunta et al., 2008; van

Kerkoerle et al., 2014). For example, macaque LFP studies show higher spike-field coherence

in the alpha range in the deeper layers compared to the superficial layers during attention tasks

with attentional suppression of alpha power and coherence being stronger in deeper layers

compared to superficial layers (Buffalo et al., 2011), which points towards the layer 5 alpha

pacemakers. Since EEG signals capture a spatiotemporally smoothed version of the LFP

signals, integrated over an area of 10 cm^2 (Buzsáki and Wang, 2012) or more which is several

orders of magnitude higher than that of LFP, the alpha pacemakers in the visual cortex and

thalamus must be coherent over a long range to yield a prominent occipital alpha in EEG. In

addition, there is evidence of alpha acting as a short-range inter-areal feedback signal among

visual areas, as found from ECoG recordings (Halgren et al., 2019), human MEG recordings

(Michalareas et al., 2016) as well as laminar recordings from multiple macaque visual areas

(van Kerkoerle et al., 2014). Top-down attention signals from higher-order areas can therefore

modulate the gain of lower-order areas by influencing the alpha feedback signals. Studies of

spatial attention have demonstrated that when covert attention is directed to one hemifield,

alpha power in the contralateral hemisphere is suppressed in response to behaviorally relevant

stimuli whereas alpha power increases in the ipsilateral hemisphere encoding for irrelevant

stimuli as evidenced by human EEG studies (Worden et al., 2000; Thut et al., 2006; Rihs et al.,

2007) and macaque LFP studies that show alpha suppression for attend-inside RF versus

attend-outside RF conditions (Fries et al., 2001; Chalk et al., 2010). Together, these results

suggest that the alpha rhythm is coherent over a relatively large cortical area, spanning perhaps

a hemifield, such that robust alpha suppression can be observed even in an aggregate signal

such as EEG.

On the other hand, gamma oscillations are generated due to excitatory-inhibitory

interactions (Buzsáki and Wang, 2012), especially involving soma-targeting parvalbumin

positive interneurons (Bartos et al., 2007; Cardin et al., 2009; Sohal et al., 2009), which remain

coherent only over a few millimeters (Jia et al., 2011; Murty et al., 2018), implying that gamma

oscillations are local in nature. Unlike alpha, gamma is shown to act as feedforward signals in

visual cortex (van Kerkoerle et al., 2014; Michalareas et al., 2016) and gamma generated in

local neural networks can interact with each other through synchronization to propel the signals

from lower-order visual areas to higher-order visual areas. Cortical spread of LFP signals can

capture gamma in the local population and the effect of attention can also be captured in the

LFP signals possibly due to comparable spatial spread of LFP (~0.5-1 mm; Katzner et al., 2009;

Xing et al., 2009; Dubey and Ray, 2016) to that of attentional field (~0.5-2.5 mm; Connor et

al., 1997; Womelsdorf et al., 2008) in the visual cortex (see Prakash et al., 2022 for further

discussion). The gamma generated in the local population in response to peripheral stimuli

could have possibly been lost, as observed in our study (Figure 3), due to spatial averaging and

reduced cortical magnification factor. However, robust gamma response has been reported in

LFP gamma studies (Ray and Maunsell, 2010; Jia et al., 2013) or attention studies (Fries et al.,

2001, 2008; Chalk et al., 2010) when non-foveal stimuli presented inside receptive field of

microelectrodes. MEG signals, with better spatiotemporal resolution (~2-3 mm; Buzsáki and

Wang, 2012), can also capture the gamma generated by peripheral stimuli (van Pelt and Fries,

2013) and attentional modulation of gamma (Bauer et al., 2014; Magazzini and Singh, 2018).

Taken together, this implies that mechanisms responsible for the generation of gamma

oscillations are too local to be well captured in EEG.

BMI applications for decoding focus of attention
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Previous LFP studies in macaque visual areas have shown spikes and LFP can be reliably used

to decode attentional state or attended location (De Sousa et al., 2021; Prakash et al., 2021).

On the other hand, our study shows that the effect of attention was equally prominent in alpha

and SSVEP power, making them more useful measures in decoding attention using human

EEG. In fact, the effect of covert spatial attention on occipital alpha power is so salient in EEG

across contralateral versus ipsilateral hemisphere, even for single trials, that it has earlier been

used in brain-machine interfaces (van Gerven and Jensen, 2009; Bahramisharif et al., 2010).

SSVEPs, on the other hand, have been used extensively in BMI applications to decode the

focus of attention based on stimulus location or feature using human EEG (Min et al., 2016;

Chinchani et al., 2022). Although the underlying neural mechanisms might be different; since

alpha and SSVEP both show comparable attentional modulation in EEG, a combination of both

the neural measures may be able to decode attended location better than a single measure.

Therefore, our findings can have important implications for development of better BMIs to

decode attentional state.

References

Bahramisharif A, Van Gerven M, Heskes T, Jensen O (2010) Covert attention allows for continuous control of brain–computer interfaces. Eur J Neurosci 31:1501–1508.

Bansal AK, Truccolo W, Vargas-Irwin CE, Donoghue JP (2012) Decoding 3D reach and grasp from hybrid signals in motor and premotor cortices: spikes, multiunit activity, and local field potentials. J Neurophysiol 107:1337–1355.

Bartos M, Vida I, Jonas P (2007) Synaptic mechanisms of synchronized gamma oscillations in inhibitory interneuron networks. Nat Rev Neurosci 8:45–56.

Bauer M, Stenner MP, Friston KJ, Dolan RJ (2014) Attentional Modulation of Alpha/Beta and Gamma Oscillations Reflect Functionally Distinct Processes. J Neurosci 34:16117–16125.

Bokil H, Andrews P, Kulkarni JE, Mehta S, Mitra PP (2010) Chronux: a platform for analyzing neural signals. J Neurosci Methods 192:146–151.

Bollimunta A, Chen Y, Schroeder CE, Ding M (2008) Neuronal Mechanisms of Cortical Alpha Oscillations in Awake-Behaving Macaques. J Neurosci 28:9976–9988.

Buffalo EA, Fries P, Landman R, Buschman TJ, Desimone R (2011) Laminar differences in gamma and alpha coherence in the ventral stream. Proc Natl Acad Sci 108:11262–11267.

Buzsáki G, Wang X-J (2012) Mechanisms of Gamma Oscillations. Annu Rev Neurosci 35:203–225.

Campbell FW, Maffei L (1970) Electrophysiological evidence for the existence of orientation and size detectors in the human visual system. J Physiol 207:635–652.

Cardin JA, Carlén M, Meletis K, Knoblich U, Zhang F, Deisseroth K, Tsai L-H, Moore CI (2009) Driving fast-spiking cells induces gamma rhythm and controls sensory responses. Nature 459:663–667.

Chalk M, Herrero JL, Gieselmann MA, Delicato LS, Gotthardt S, Thiele A (2010) Attention reduces stimulus-driven gamma frequency oscillations and spike field coherence in V1. Neuron 66:114–125.

Chinchani AM, Paliwal S, Ganesh S, Chandrasekhar V, Yu BM, Sridharan D (2022) Tracking momentary fluctuations in human attention with a cognitive brain-machine interface. Commun Biol 5:1–17.

Connor CE, Preddie DC, Gallant JL, Essen DCV (1997) Spatial Attention Effects in Macaque Area V4. J Neurosci 17:3201–3214.

Contreras D, Destexhe A, Sejnowski TJ, Steriade M (1996) Control of Spatiotemporal Coherence of a Thalamic Oscillation by Corticothalamic Feedback. Science 274:771–774.

Daniel PM, Whitteridge D (1961) The representation of the visual field on the cerebral cortex
in monkeys. J Physiol 159:203–221.

Das A, Ray S (2018) Effect of Stimulus Contrast and Visual Attention on Spike-Gamma Phase
Relationship in Macaque Primary Visual Cortex. Front Comput Neurosci 12:66–66.

De Sousa C, Gaillard C, Di Bello F, Ben Hadj Hassen S, Ben Hamed S (2021) Behavioral
validation of novel high resolution attention decoding method from multi-units & local
field potentials. NeuroImage 231:117853.

De Valois RL, Albrecht DG, Thorell LG (1982) Spatial frequency selectivity of cells in
macaque visual cortex. Vision Res 22:545–559.

Desimone R (1998) Visual attention mediated by biased competition in extrastriate visual
cortex. Philos Trans R Soc Lond B Biol Sci 353:1245–1255.

Desimone R, Duncan J (1995) Neural mechanisms of selective visual attention. Annu Rev
Neurosci 18:193–222.

Di Russo F, Spinelli D, Morrone MC (2001) Automatic gain control contrast mechanisms are
modulated by attention in humans: evidence from visual evoked potentials. Vision Res
41:2435–2447.

Dubey A, Ray S (2016) Spatial spread of local field potential is band-pass in the primary visual
cortex. J Neurophysiol 116:1986–1999.

Foxe JJ, Simpson GV, Ahlfors SP (1998) Parieto-occipital approximately 10 Hz activity
reflects anticipatory state of visual attention mechanisms. Neuroreport 9:3929–3933.

Fries P, Reynolds JH, Rorie AE, Desimone R (2001) Modulation of Oscillatory Neuronal
Synchronization by Selective Visual Attention. Science 291:1560–1563.

Fries P, Womelsdorf T, Oostenveld R, Desimone R (2008) The effects of visual stimulation
and selective visual attention on rhythmic neuronal synchronization in macaque area
V4. J Neurosci Off J Soc Neurosci 28:4823–4835.

Gruber T, Müller MM, Keil A, Elbert T (1999) Selective visual-spatial attention alters induced
gamma band responses in the human EEG. Clin Neurophysiol 110:2074–2085.

Gulbinaite R, Roozendaal DHM, VanRullen R (2019) Attention differentially modulates the
amplitude of resonance frequencies in the visual cortex. NeuroImage 203:116146–
116146.

Halgren M, Ulbert I, Bastuji H, Fabó D, Erőss L, Rey M, Devinsky O, Doyle WK, Mak-
McCully R, Halgren E, Wittner L, Chauvel P, Heit G, Eskandar E, Mandell A, Cash SS
(2019) The generation and propagation of the human alpha rhythm. Proc Natl Acad Sci
116:23772–23782.

Hou C, Pettet MW, Sampath V, Candy TR, Norcia AM (2003) Development of the spatial
organization and dynamics of lateral interactions in the human visual system. J
Neurosci 23:8630–8640.

Hwang EJ, Andersen RA (2013) The utility of multichannel local field potentials for brain–machine interfaces. J Neural Eng 10:046005.

Itthipuripat S, Ester EF, Deering S, Serences JT (2014a) Sensory Gain Outperforms Efficient Readout Mechanisms in Predicting Attention-Related Improvements in Behavior. J Neurosci 34:13384–13398.

Itthipuripat S, Garcia JO, Rungratsameetaweemana N, Sprague TC, Serences JT (2014b) Changing the spatial scope of attention alters patterns of neural gain in human cortex. J Neurosci Off J Soc Neurosci 34:112–123.

Jia X, Smith MA, Kohn A (2011) Stimulus selectivity and spatial coherence of gamma components of the local field potential. J Neurosci 31:9390–9403.

Jia X, Xing D, Kohn A (2013) No Consistent Relationship between Gamma Power and Peak Frequency in Macaque Primary Visual Cortex. J Neurosci 33:17–25.

Katzner S, Nauhaus I, Benucci A, Bonin V, Ringach DL, Carandini M (2009) Local Origin of Field Potentials in Visual Cortex. Neuron 61:35–41.

Kim YJ, Grabowecky M, Paller KA, Muthu K, Suzuki S (2007) Attention induces synchronization-based response gain in steady-state visual evoked potentials. Nat Neurosci 10:117–125.

Klimesch W, Doppelmayr M, Russegger H, Pachinger T, Schwaiger J (1998) Induced alpha band power changes in the human EEG and attention. Neurosci Lett 244:73–76.

Liza K, Ray S (2022) Local Interactions between Steady-State Visually Evoked Potentials at Nearby Flickering Frequencies. J Neurosci 42:3965–3974.

Magazzini L, Singh KD (2018) Spatial attention modulates visual gamma oscillations across the human ventral stream. NeuroImage 166:219–229.

Michalareas G, Vezoli J, van Pelt S, Schoffelen J-M, Kennedy H, Fries P (2016) Alpha-Beta and Gamma Rhythms Subserve Feedback and Feedforward Influences among Human Visual Cortical Areas. Neuron 89:384–397.

Min B-K, Dähne S, Ahn M-H, Noh Y-K, Müller K-R (2016) Decoding of top-down cognitive processing for SSVEP-controlled BMI. Sci Rep 6:36267.

Morgan ST, Hansen JC, Hillyard SA (1996) Selective attention to stimulus location modulates the steady-state visual evoked potential. Proc Natl Acad Sci U S A 93:4770–4774.

Müller MM, Gruber T, Keil A (2000) Modulation of induced gamma band activity in the human EEG by attention and visual information processing. Int J Psychophysiol 38:283–299.

Murty DVPS, Manikandan K, Kumar WS, Ramesh RG, Purokayastha S, Javali M, Rao NP, Ray S (2020) Gamma oscillations weaken with age in healthy elderly in human EEG. NeuroImage 215:116826–116826.

810   Murty DVPS, Ray S (2022) Stimulus-induced Robust Narrow-band Gamma Oscillations in
811   Human EEG Using Cartesian Gratings. Bio-Protoc 12:e4379–e4379.

812   Murty DVPS, Shirhatti V, Ravishankar P, Ray S (2018) Large visual stimuli induce two distinct
813   gamma oscillations in primate visual cortex. J Neurosci Off J Soc Neurosci 38:2730–
814   2744.

815   Muthukumaraswamy SD, Singh KD (2013) Visual gamma oscillations: The effects of stimulus
816   type, visual field coverage and stimulus motion on MEG and EEG recordings.
817   NeuroImage 69:223–230.

818   Prakash SS, Das A, Kanth ST, Mayo JP, Ray S (2021) Decoding of Attentional State Using
819   High-Frequency Local Field Potential Is As Accurate As Using Spikes. Cereb Cortex
820   31:1–15.

821   Prakash SS, Mayo JP, Ray S (2022) Decoding of attentional state using local field potentials.
822   Curr Opin Neurobiol 76:102589–102589.

823   Ray S, Maunsell JHR (2010) Differences in gamma frequencies across visual cortex restrict
824   their possible use in computation. Neuron 67:885–896.

825   Rihs TA, Michel CM, Thut G (2007) Mechanisms of selective inhibition in visual spatial
826   attention are indexed by α-band EEG synchronization. Eur J Neurosci 25:603–610.

827   Saalmann YB, Pinsk MA, Wang L, Li X, Kastner S (2012) The Pulvinar Regulates Information
828   Transmission Between Cortical Areas Based on Attention Demands. Science 337:753–
829   756.

830   Salelkar S, Ray S (2020) Interaction between steady-state visually evoked potentials at nearby
831   flicker frequencies. Sci Rep 10:1–16.

832   Sauseng P, Klimesch W, Stadler W, Schabus M, Doppelmayr M, Hanslmayr S, Gruber WR,
833   Birbaumer N (2005) A shift of visual spatial attention is selectively associated with
834   human EEG alpha activity. Eur J Neurosci 22:2917–2926.

835   Slotnick SD, Klein SA, Carney T, Sutter EE (2001) Electrophysiological estimate of human
836   cortical magnification. Clin Neurophysiol 112:1349–1356.

837   Sohal VS, Zhang F, Yizhar O, Deisseroth K (2009) Parvalbumin neurons and gamma rhythms
838   enhance cortical circuit performance. Nature 459:698–702.

839   Thut G, Nietzel A, Brandt SA, Pascual-Leone A (2006) α-Band Electroencephalographic
840   Activity over Occipital Cortex Indexes Visuospatial Attention Bias and Predicts Visual
841   Target Detection. J Neurosci 26:9494–9502.

842   van Gerven M, Jensen O (2009) Attention modulations of posterior alpha as a control signal
843   for two-dimensional brain–computer interfaces. J Neurosci Methods 179:78–84.

844   van Kerkoerle T, Self MW, Dagnino B, Gariel-Mathis M-A, Poort J, van der Togt C, Roelfsema
845   PR (2014) Alpha and gamma oscillations characterize feedback and feedforward
846   processing in monkey visual cortex. Proc Natl Acad Sci 111:14332–14341.

van Pelt S, Fries P (2013) Visual stimulus eccentricity affects human gamma peak frequency.
NeuroImage 78:439–447.

Vijayan S, Kopell NJ (2012) Thalamic model of awake alpha oscillations and implications for
stimulus processing. Proc Natl Acad Sci 109:18553–18558.

Womelsdorf T, Anton-Erxleben K, Treue S (2008) Receptive Field Shift and Shrinkage in
Macaque Middle Temporal Area through Attentional Gain Modulation. J Neurosci
28:8934–8944.

Worden MS, Foxe JJ, Wang N, Simpson GV (2000) Anticipatory Biasing of Visuospatial
Attention Indexed by Retinotopically Specific α-Bank Electroencephalography
Increases over Occipital Cortex. J Neurosci 20:RC63–RC63.

Xing D, Yeh C-I, Shapley RM (2009) Spatial Spread of the Local Field Potential and its
Laminar Variation in Visual Cortex. J Neurosci 29:11540–11549.

Figure Legends

Figure 1: Task Paradigms for human EEG experiments

(A) Gamma Tuning Protocol: While the subjects maintained passive fixation, full-screen grating stimuli were presented at full contrast for 800ms with an inter-stimulus interval of 700 ms, with 2-3 stimuli shown in each trial. During a stimulus presentation, stimuli could either be static or counterphase flickering at 16 Hz. The values of spatial frequency (SF) were either 2 or 4 cpd, and orientation values could take four possible values (0°, 45°, 90°, and 135°).

(B) Stimulus Size and Eccentricity Mapping Protocol: While the subjects maintained passive fixation, two equidistant non-foveal achromatic gratings were presented peripherally in the lower visual field while the diameter and eccentricity of the gratings were varied systematically in separate experiments (protocol 2a-2d). The azimuth and elevation of the center of the gratings and the diameter were varied from 2°-8° in steps of 2° in protocols 2a-2d. During each trial in any of the protocols, the pair of stimuli were presented at full contrast for 800 ms with an inter-stimulus interval of 700 ms, with 2-3 stimuli shown in each trial. During stimulus presentation, either both the gratings were static or the left grating counterphase flickering at 12 Hz and the right grating flickering at 16 Hz. The spatial frequency and orientation values for an individual subject were determined based on the SF-Ori combination that produced the maximum gamma response (25-70 Hz) for static gratings in the gamma tuning protocol described in A.

(C) Attention Task: While the subjects maintained fixation, two achromatic gratings were presented at equidistant spatial locations from the fixation spot in the lower visual field. The subjects were cued to covertly attend to one of the two locations in different blocks of trials (indicated by a black dotted circle for clarity; not visible to the subjects). Grating parameters (orientation and spatial frequency) were determined for each subject based on

A. The diameter, azimuth, and elevation for all the subjects were set to 4° (Protocol 2b),

based on results of B on a group of 11 subjects before performing any attention

experiments. During each trial, either both the stimuli were static, or both were

counterphase flickering, one at 12 Hz and the other at 16 Hz. The contrasts of both the

attended and ignored locations were at 100% during each trial. There was an initial time

period of 1000ms at the beginning of each trial when the subject maintained their fixation,

followed by an unstipulated time before the orientation of the stimulus at the cued location

changed (target). The stimuli were considered to be 250 ms long, which contained 3 and 4

full cycles of the 12 and 16 Hz stimuli, presented without any inter-stimulus interval so that

they appeared to be flickering continuously. The target appeared at an unstipulated time

drawn from an exponential distribution (mean 2000 ms, range 250-5000 ms) following the

initial stimulus lead period of 1000 ms. The subjects had to make a saccade to the target

location within 800 ms of the target onset for a correct trial (hit). If there was no change

during a trial (catch trial), the subjects had to maintain their fixation. In protocol 3, each

subject performed this attention task for 2-3 blocks on each side where the orientation

change of the cued gratings (target) could occur randomly from any of 5 preset values. The

preset orientation change values were different for each of the temporal frequencies (static,

12, and 16 Hz). The psychometric curves for each static and flickering gratings were fitted

separately with Weibull functions after pooling the behavioral accuracy values from both

sides. The orientation change values for 50% accuracy for each of the static and flickering

gratings were determined from the fits. This single target orientation (separate for static, 12

and 16 Hz) was used for protocol 4 in which the subjects performed 5-6 blocks on each

side.

Figure 2: alpha, gamma, and SSVEP response in subjects for full-field grating stimuli

(A) Scalp maps showing 64 electrodes (represented as dots). Scalp maps show the change in

alpha power (top row) in the stimulus analysis period (250 to 750 ms) w.r.t baseline (-500

to 0 ms; zero denotes stimulus onset), gamma (middle) for static gratings, and SSVEP

power (bottom) computed by the trial-by-trial power estimation method averaged across

28 subjects. The power changes are separately shown for all SF-Ori combinations (left

column) and the SF-Ori combination that produced maximum gamma response (right

column). Eight analysis electrodes used in Panels B-E are marked in bold dots.

(B) Left ordinate (black) shows raw power spectral densities (PSDs) vs. frequency for stimulus

(250 to 750 ms, black trace), baseline (-500 to 0 ms, green trace) analysis periods averaged

across all 28 subjects for the eight electrodes highlighted in panel A for all SF-Ori

combinations (left), and max SF-Ori combination (right) in response to static gratings; right

ordinate (blue) shows the same for change in PSD (in dB, solid blue trace). Solid black and

red bars on the X-axis indicate alpha (8-12 Hz) and gamma bands (25-70 Hz), respectively.

(C) Change in alpha (black bar), gamma (red) power in the stimulus analysis period (250 to

750 ms, black trace) w.r.t baseline (-500 to 0 ms, green trace) analysis period averaged

across all 28 subjects for the eight electrodes highlighted in panel A for all SF-Ori

combinations (left), and max SF-Ori combination (right)) in response to static gratings;

The respective powers were calculated in the frequency ranges highlighted in panel B.

Error bars indicate the standard error of the mean across subjects (N=28).

(D) Same layout as in B in response to 16 Hz counterphase flickering gratings. The SSVEP

response at 32 Hz (twice that of the flickering frequency) is highlighted by a solid cyan

vertical line.

(E) Same layout as in C in response to 16 Hz counterphase flickering gratings with the addition

of change in SSVEP power (cyan bar).

Figure 3: alpha, gamma, and SSVEP response in subjects in response to non-foveal

stimuli pair

(A) Scalp maps showing 64 electrodes (represented as dots). Color maps show the change in

power in the stimulus analysis period (250 to 750 ms) w.r.t baseline (-500 to 0 ms) (top

row) for alpha, gamma, SSVEP power using trial-by-trial SSVEP power estimate, and by

the trial-averaged SSVEP power estimate from top to bottom rows averaged across 11

subjects. The columns indicate the diameter and stimulus center locations of the gratings

used in separate experiments in protocol 1 and protocols 2a-2d mentioned in Figure 1.

Eight electrodes used in Panels B-D are marked as bold dots. Note that while all 8

electrodes are used for alpha and gamma calculation, the calculation is done differently for

SSVEPs (protocols 2a-2d). Here, since the two stimuli flicker at two different frequencies

(12 and 16 Hz), the SSVEP response for a particular frequency is measured only for the

contralateral 4 electrodes. The SSVEP plots shown here are obtained by first mirror-

flipping the scalp map for the stimulus on the right side and then averaging with the scalp

map for the left stimulus (as if the stimulus is always on the left side), which is why the

SSVEP responses are lateralized on the right side. Consequently, only the 4 contralateral

electrodes are highlighted in these plots. Note, however, that these 4 electrodes represent

averaged data from all 8 electrodes, since data from the 4 left electrodes are mirror-flipped

to the right side.

(B-D) Change in power in alpha (B), gamma (C), and SSVEP (D) for stimulus (250 to 750 ms,

black trace), baseline (-500 to 0 ms, green trace) analysis periods averaged across all 11

subjects for the electrodes highlighted in panel A. Note that, the ordinate of panel B is flipped

upside down to highlight the magnitude of alpha suppression. Panel D shows trial-by-trial

SSVEP power estimates (solid cyan lines) and trial-averaged power estimates (dotted cyan

lines).

Figure 4: Behavioral Results for Attention Task

(A) Behavioral accuracy (in percent) for different attended locations (left-white bars, right-

gray bars) and different temporal frequencies of the cued/attended stimuli.

(B) Response time (in milliseconds) for different attention locations and temporal

frequencies of the cued/attended stimuli.

(C) Change in orientation values (target) for different attention locations and temporal

frequencies of the cued/attended stimuli.

Figure 5: Attentional modulation of alpha and gamma response in attention tasks for

static stimuli

(A) Scalp maps showing the change in power in the pre-target analysis period (-1000 to 0 ms

of target onset) w.r.t ignore left (i.e., attend-right) baseline (-1000 to 0 ms of first stimulus

onset in a trial) for alpha power (8-12 Hz) averaged across 26 subjects. The left column

represents the attend-left condition, the middle column represents ignore-left (attend-

right), and the right column indicates the difference between attend-left and attend-right

conditions. Analysis electrodes used in Panels C and D are marked as bold dots.

(B) Layout same as A for gamma power (25-70 Hz).

(C) Change in PSD vs. frequency for the attended condition versus ignored condition in the

pre-target analysis period (-1000 to 0 ms of target onset averaged across all 26 subjects for

the eight occipital analysis electrodes (difference between the 4 electrodes on the right and

4 electrodes on the left) highlighted in panel A. Solid black and red bars on the X-axis

indicate alpha and gamma bands, respectively.

(D) Change in alpha (black bar) and gamma (red) power between the attended condition and

the ignored condition in the pre-target analysis period (-1000 to 0 ms of target onset),

averaged across all 26 subjects for the eight occipital analysis electrodes highlighted in

panel A. The respective powers were calculated in the frequency ranges highlighted in

panel B. For statistical comparison of the magnitude of attentional modulation between

alpha and gamma power, we sign-reversed the alpha power values (negative values to

positive and vice versa) and performed a pairwise t-test. Error bars indicate the standard

error of the mean across subjects (N=26).

Figure 6: Scalp maps for attention-mediated SSVEP modulation for different attended

locations and temporal frequencies of stimuli

(A-D) The attended location in each condition is presented as filled elliptical disks and the

ignored location as blank elliptical disks. The flickering frequencies are mentioned on top of

the disks. The analysis electrodes are highlighted in bold.

(E) Consolidated scalp map after averaging responses in panels A-D. For panels B and D, the

scalp maps were mirror-flipped w.r.t the midline of the cap to yield SSVEP responses in the

right hemisphere prior to averaging.

1000

Figure 7: Modulation of alpha, gamma, and SSVEP response in attention tasks for

flickering stimuli

(A) Scalp maps showing the change in alpha (8-12 Hz), gamma (25-70Hz), trial-by-trial

SSVEP power, and trial-averaged SSVEP power (combined for 24 and 32 Hz) (top to

bottom row) in the pre-target analysis period (-1000 to 0 ms of target onset) w.r.t attend-

right (i.e., ignored left) baseline (-1000 to 0 ms of first stimulus onset in a trial), averaged

across 26 subjects. The left column represents the attend-left condition, the middle column

represents ignored-left, and the right column indicates the difference between attend-left

and ignored-left conditions. Analysis electrodes are highlighted. As explained in Figure 3,

alpha/gamma and SSVEP responses are computed using different approaches, which is

reflected by showing only the contralateral 4 electrodes for the SSVEP responses.

(B) (top) Trial-by-trial estimate of change in power spectral density (PSD) vs. frequency for

the attended condition and ignored condition in the pre-target analysis period (-1000 to 0

ms of target onset), averaged across 19 subjects for the eight occipital analysis electrodes;

(bottom) shows the same for trial-averaged signal estimates. Solid black and red bars on

the X-axis indicate alpha and gamma bands, respectively whereas the SSVEP responses at

24 Hz and 32 Hz are highlighted by solid cyan vertical lines.

(C) Change in alpha (black bar), gamma (red), and SSVEPs for both estimates (cyan) power

between the attended and ignored condition in the pre-target analysis period (-1000 to 0

ms of target onset), averaged across all 26 subjects for the eight occipital analysis

electrodes. The respective powers were calculated in the frequency ranges highlighted in

panel B. To statistically compare the magnitude of attentional modulation between alpha

power and other neural measures, we sign- reversed the alpha power values (negative

values to positive and vice versa) and performed a pairwise t-test. Error bars indicate the

standard error of the mean across subjects (N=26).

Attend Left **Ignore Left** **Attend Left - Ignore Left**

A

Alpha
(8-12 Hz)

B

Gamma
(25-70 Hz)

C

D