Decreased auditory cortical activations and altered long-range alpha-band connectivity characterize passive auditory spatial attention in ASD

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Abstract

Autism Spectrum Disorder (ASD) is a developmental condition characterized by difficulties in social interaction, communication and sensory processing. The ability to orient towards sounds is a key component of social interactions, yet auditory spatial attention remains relatively understudied in ASD, despite some prior research indicating differences in this domain. Here, we investigate the neural signatures associated with passive auditory spatial attention in children with ASD (n = 21, ages 6-17) relative to age- and IQ-matched typically developing (TD) children (n = 31), using source-localized magnetoencephalography (MEG). Participants listened passively, while watching a silenced movie, to non-social auditory stimuli designed to either remain lateralized to one hemifield ("stay" trials), or to change in location from one side to the contralateral hemifield ("jump" trials). Linear mixed effect modeling showed decreased cortical activation in the auditory cortex in the ASD group in response to jump trials, relative to the TD group. Additionally, functional connectivity analyses showed increased alpha-band functional connectivity in the ASD group between auditory cortex seeds and frontoparietal regions known

to be recruited during auditory spatial attention. Right prefrontal alpha-band connectivity

estimates were associated with worse behaviorally assessed auditory processing scores and

increased ASD symptomatology. Our results align with the hypothesis that auditory spatial

attention generally, and specifically orientation to sounds even when experienced passively,

differs in ASD individuals.

Introduction

Autism Spectrum Disorder (ASD) is a developmental condition characterized by a complex set

of behavioral and sensory symptoms, the core of which is differences in social communication

(Ben-Sasson et al., 2019; Grzadzinski et al., 2013; Haesen et al., 2011; Hedger et al., 2020; O.

Miguel et al., 2017). Among its sensory manifestations, differences in processing of auditory

stimuli have been consistently documented among individuals with ASD (Alho et al., 2021;

Alho, Khan, et al., 2023; Alho, Samuelsson, et al., 2023; Bharadwaj et al., 2022; DePape et al.,

2012; Gonçalves & Monteiro, 2023; Haesen et al., 2011; Mamashli et al., 2017; Marco et al.,

2011; O'Connor, 2012; Port et al., 2015, 2017; Schafer et al., 2020).

Auditory spatial attention, i.e. our ability to perceive and track the location of sounds in

space, is a complex phenomenon involving several stages of sensory and cognitive processing.

For a sound to be successfully located in space, sensory encoding of acoustic features,

specifically the relative timing and amplitude differences between ears, needs to be integrated

with efficient selective attention and object recognition processes, for environmentally adequate

behavioral responses to be generated (Fritz et al., 2007; Kayser et al., 2005; van der Heijden et

al., 2019). In addition to primary and secondary auditory areas in the superior temporal gyrus

(STG), fMRI research has shown that locating a sound in space recruits a frontoparietal network

of cortical regions (Degerman et al., 2006; Deouell et al., 2007; Haist et al., 2005; Kong et al.,

2014; Smith et al., 2009; Zatorre et al., 1999). These regions include the ventrolateral prefrontal

cortex (vlPFC), the dorsolateral prefrontal cortex (dlPFC), the precentral sulcus and gyrus, the

central sulcus, the postcentral sulcus and gyrus, and the inferior and superior temporal lobes

(IPL/SPL). Unsurprisingly, these areas overlap with well-defined functional networks observed

in resting state fMRI data, including the ventral and dorsal attention networks, and the

frontoparietal executive control network (Hames et al., 2016; Sadaghiani & Kleinschmidt, 2016;

Supekar et al., 2013).

Research to date has shown that during auditory spatial attention tasks, individuals with

ASD show poorer behavioral performance, measured as higher reaction times and reduced

accuracy rates, when required to localize sounds in space (Soskey et al., 2017; Teder-Sälejärvi et

al., 2005). While previous studies using electroencephalography (EEG) and

magnetoencephalography (MEG) have established a range of differences in early sensory

encoding of auditory stimuli in ASD (Bruneau et al., 1999; Jansson-Verkasalo et al., 2005;

Orekhova et al., 2009; Roberts et al., 2010; Russo et al., 2009), these have been mostly

documented in the context of general auditory processing. One study that directly investigated

neural signatures in the context of auditory spatial attention in individuals with ASD showed that

poorer behavioral performance in ASD individuals during auditory spatial attention tasks is

accompanied by a reduced amplitude of the N1 and the P3 components, two Event-Related

Potentials (ERP) associated with early sensory encoding and attentional processing,

correspondingly (Teder-Sälejärvi et al., 2005). Another study that also focused on auditory

spatial attention in ASD, found that individuals with an ASD diagnosis showed an absence of the

Object-Related Negativity, an ERP component thought to reflect auditory segregation, as well as

a reduced amplitude of the P400, an attention-dependent ERP response thought to reflect

perceptual decision making (Lodhia et al., 2018). Notably, despite its importance for everyday

life and communication, auditory spatial attention remains understudied in ASD.

To address this gap in the literature, we investigated cortical responses in ASD and

Typically Developing (TD) individuals, ages 6 to 17, to direct manipulations of spatial

lateralization cues in sound. Using source-localized MEG signals, we focused on evoked

responses and functional connectivity from the auditory cortex to characterize the cortical

signatures of differences in passive auditory spatial orienting in ASD. We also conducted

correlation analyses to investigate possible associations between brain function and clinically

relevant behavioral measures of ASD.

Methods

Participants

Thirty-one typically developing (TD, mean age = 13.07, SD = 3.43, range = 6-17), and twenty-

one children diagnosed with autism spectrum disorder (ASD, mean age = 13.50, SD = 2.73,

range 7-17) participated in this study (table 1). All participants met an IQ cutoff of full-scale IQ

$\geq$ 70, as measured by the Kaufman Brief Intelligence Test II (KBIT, Kaufman & Kaufman,

2004). Participants were right-handed except two ASD individuals, determined using the Dean

Questionnaire (Dean, 1988). All participants had normal hearing, defined as pure tone thresholds

better than or equal to 15 dB in both ears at octave frequencies between 250 Hz and 8 kHz, tested

in a soundproof room using an audiometer. Informed consent was provided by parents or legal

guardians, in addition to participants' assent, according to protocols approved by the Institutional

Review Board of the Massachusetts General Hospital.

All ASD individuals had a formal clinical diagnosis and met criteria on the Autism

Diagnosis Observation Schedule, Version 2, (ADOS-2, Lord et al., 2000), administered by

trained research personnel with demonstrated inter-rater reliability, the Social Communication

Questionnaire, Lifetime Version (SCQ Lifetime, Rutter et al., 2003), and the Social

Responsiveness Scale (SRS, (Constantino & Gruber, 2012)). All participants were also

administered the Sensory Profile Questionnaire (ASPS) (Brown & Dunn, 2002). ASD

participants who met ADOS-2 criteria but had a borderline ADOS-2 score, did not meet a cutoff

of ≥ 15 on the SCQ, or a total T score < 59 on the SRS, were further evaluated by expert

clinician (RMJ) to confirm the ASD diagnosis. A diagnosis of Fragile X syndrome, tuberous

sclerosis, and other known risk factors, such as gestation for less than 36 weeks, were criteria for

exclusion from participation in the study.

All TD subjects were below SCQ thresholds. Parental report was used to rule out the

possibility of any neurological/psychiatric condition (including psychotic episodes, seizures,

clinical EEG abnormalities, Tourette syndrome, or any other tics), or substance abuse in the last

six months. Participants with Tourette syndrome or tic disorders were excluded from both

groups. Due to high comorbidity with ASD, participants with a diagnosis of ADHD, depression,

anxiety or language delay were excluded from the TD group, but not from the ASD group.

	ASD (N = 21, 4 females)			TD (N = 31, 4 females)			
	N	*Mean (SD)*	*Range*	*N*	*Mean (SD)*	*Range*	**p-value**
Age	21	13.50 (2.73)	7-17	31	13.07 (3.43)	6-17	0.640
NVIQ	17	102.76 (15.14)	59-130	20	112.35 (16.44)	83-147	0.083
VIQ	17	98.82 (19.13)	61-139	20	114.20 (13.84)	86-139	0.009
ADOS	21	11.19 (3.06)	7-18	-	-	-	-
SCQ	19	17.97 (8.86)	0-31	30	3.42 (3.21)	0-12	< 0.001
SRS	20	71.85 (16.02)	40-95	29	43.41 (6.03)	34-62	< 0.001
ASPS	20	12.15 (4.08)	5-20	30	17.23 (2.49)	12-20	< 0.001

Table 1. Characterization of participants. P-values are shown for independent samples t-tests for
the difference of means across groups. NVIQ (non-verbal IQ), VIQ (Verbal IQ), ADOS (Autism
Diagnostic Observation Schedule), SCQ (Social Communication Questionnaire), SRS (Social
Responsiveness Scale), ASPS (Auditory Sensory Profile Score).

Stimuli and procedures

The auditory stimuli consisted of 1200 ms broadband (0.5-2.5 kHz) noise bursts (see Figure 1a),

presented in 200 trials. Stimuli were designed with a 400 microsecond Interaural Time Delay

(ITD) between the stereo channels, thus generating a left- or right-lateralized percept depending

on the direction of the ITD. In each trial, the broadband noise was amplitude modulated at either

25 Hz or 43 Hz (Ahlfors et al., 2024). To manipulate the ITD, discontinuities of ±400

microseconds were introduced in the noise carriers in each channel 550 ms after stimulus onset.

Crucially, the discontinuities were designed to coincide with a trough of the amplitude

modulation to render them monaurally imperceptible. In half the trials, the 400-microsecond

discontinuities occurred in the same direction in both ears, thus holding the ITD constant for the

full duration of the trial, as shown in Figure 1b. For the other half of the trials, carrier

discontinuities were introduced in opposite directions, thus switching the ITD polarity and

creating a percept of a spatial jump in the location of the sound from one hemifield to the

contralateral side midway through the trial, as shown in Figure 1c.

Figure 1. Stimuli and schematic representation of conditions. **a.** Spectrogram (top) and soundwave (bottom) of the auditory stimuli in the stay condition. The shaded area in the bottom panel represents the time window from which panels in b and c (top) were extracted. **b.** For stay events, the directionality of the ITD was kept constant between the two stereo channels (top). Stimuli were presented with either left- or right-leading ITD (bottom), with equal probability. **c.** For jump events, the polarity of the ITD was reversed at 550ms (dotted line, top). The sound was initially presented with either left- or right-leading ITD, with equal probability. The ITD reversal created a percept of a spatial jump in the lateralization of the sound to the contralateral side (bottom).

Note that subjects listened passively to the auditory stimuli while watching a silent movie

with no subtitles. Participants were explicitly asked to ignore the auditory stimuli. After rejection

of bad epochs (see below for details), the mean total number of trials across participant was 177

per participant. The mean number of trials across groups did not differ significantly (ASD jump

= 181.57, SD = 20.23; ASD stay = 180.00, SD = 22.62; TD jump = 175.48, SD = 19.78, TD stay

= 176.16, SD = 19.18). The interstimulus interval was uniformly distributed between 300 and

400ms. Stimuli were presented at 65 dB-SPL via MEG compatible earphones.

MEG Data acquisition

MEG data was acquired using a 306-channel Vector-View Magnetoencephalography system

(MEGIN Oy, Finland) inside a magnetically shielded room (IMEDCO, Switzerland). Signals

were bandpass filtered between 0.1 and 330 Hz at a sampling rate of 1,000 or 3,000 Hz. Head

position was recorded during data acquisition using four head position indicator (HPI) coils

attached to the scalp. The location of the HPI coils, the nasion, preauricular points, and additional

points in the scalp surface were digitized using a Fastrack digitizer (Polhemus) system. The

digitization information was used to co-register the MEG and MRI data for source modeling

analyses. Electrocardiography (ECG) and electro-oculography (EOG) electrodes were used to

detect heartbeat, saccades, and blinks. Finally, five minutes of empty-room data were recorded at

the end of the session to measure the instrumentation and environmental noise levels detected by

the MEG sensor on the data acquisition date.

Structural MRI

T1-weigthed magnetization-prepared rapid gradient echo (MPRAGE) structural images were

acquired using a 3T Siemens MRI scanner and a 32-channel head coil. Reconstruction of cortical

surface, cortical volume and anatomical parcellation was conducted using FreeSurfer (Dale et al.,

1999; Fischl, Sereno, & Dale, 1999).

MEG data preprocessing

MEG data was preprocessed and analyzed using MNE-Python (Gramfort et al., 2013). Noisy

MEG channels were visually identified and excluded from further analyses. Temporal Signal

Space Separation (tSSS, Taulu & Simola, 2006) was implemented to compensate for head

movements and minimize noise from sources outside of the virtual sphere circumscribing the

MEG sensors. Next, data was bandpass filtered between 0.1 and 144 Hz, and notch filters were

applied at 60 Hz and 120 Hz to remove line noise and its harmonics. Signal Space Projection

(SSP, Tesche et al., 1995; Uusitalo & Ilmoniemi, 1997) was used to remove cardiac and

oculomotor artifacts by computing projections from the bandpass filtered (1-35 Hz) signal. The

data were then epoched from -400 ms to 1400 ms with respect to sound onset and detrended to

remove slow drifts. Finally, epochs were detrended to remove slow drifts in the MEG Signal and

removed if they exceeded a peak-to-peak threshold of 4000 fT for magnetometer channels and

5000 fT/cm for planar gradiometer channels.

Source localization

Gradiometer signals were source localized to the FreeSurfer cortical reconstruction of each

individual. The forward solution was computed using the single-compartment boundary element

model (BEM, Hamalainen & Sarvas, 1989). Inner skull surface triangulations were generated

from MRI structural data using the watershed algorithm. The cortical current distributions were

obtained using the dynamic Statistical Parametric Mapping (dSPM, Dale et al., 2000) with a

loose orientation constraint of 0.2 and depth weighting of 0.8 (Lin et al., 2006). To reduce the

contribution of instrumentation and environmental artifacts to MEG signals, a noise covariance

matrix obtained from empty room recordings. The noise covariance matrix was then used to

construct the inverse operator (regularization parameter = 0.1), thus noise-normalizing the source

estimates (Dale et al., 2000).

Regions of interest

Cortical labels in primary auditory regions were identified for each subject's anatomical surface

reconstruction by visually examining the time course activations in response to ITD reversals.

Labels corresponding to peak earliest activations were then manually delineated for each

individual participant, in both the right and the left hemisphere. Figure 2 shows the overlap

probability for left and right auditory labels across all subjects. Label size was not significantly

different across groups (ASD mean = 63.07, TD mean = 64.45, t = -0.21, p = 0.83. Table s1,

supplemental materials). Evoked responses were computed by extracting the mean dSPM

activation across all vertices within each auditory label.

In addition to auditory areas, studies of spatial auditory attention using fMRI reported the

recruitment of a frontoparietal network during auditory spatial attention tasks. Areas within this

network include the dorsolateral Prefrontal Cortex (dlPFC), the ventrolateral Prefrontal Cortex

(vlPFC), the precentral/central sulci, the pre/postcentral gyri, the Superior Parietal Lobe (SPL)

and the Inferior Parietal Lobe (IFL). These target regions were therefore grouped into frontal,

central and parietal labels for subsequent functional connectivity analyses (table s2, supplemental

materials), using Freesurfer cortical parcellations (Destrieux et al., 2010).

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212 **Figure 2.** Anatomical overlap across all subjects for the manually drawn,
     left and right auditory labels. Labels were identified on the basis of cortical
213 activations in response to ITD reversals in the Jumps condition. Overlap is
     presented in a common cortical space (Freesurfer's fsaverage cortical
     surface).

Functional connectivity analyses

We computed seed-based functional connectivity for each condition (stay and jump) from the left

and right auditory labels with the regions of interest outlined above, including the dlPFC and

vlPFC (superior, middle and inferior prefrontal sulci, middle and inferior prefrontal gyri), central

regions (precentral and central sulci, precentral and postcentral gyri) and parietal regions

(postcentral and intraparietal sulci, superior and inferior parietal sulci), bilaterally. The seed time

courses for functional connectivity analyses corresponded to the auditory cortical activations in

response to jump trials. Functional connectivity was measured using the Phase Lag Index (PLI),

a measure of phase synchronization that penalizes zero-lag estimates, thus effectively controlling

for spurious connectivity due to field spread of MEG signals (Stam et al., 2007). Data was

resampled to 250Hz, and the PLI was computed for each trial between the right and left seed

ROI time course and the time course at each vertex, using the wavelet method (spanning four

cycles) between 4Hz and 30Hz, for a 2000ms window between -400ms and 1400ms. PLI

estimates in the jump condition were then normalized with respect to the stay condition to obtain

a z-scored PLI value, hereafter referred to as zPLI (see supplementary materials). The zPLI

estimate is a z value which reflects the extent to which connectivity during jump trials deviates

with respect to the baseline condition (i.e. stay), while additionally controlling for the non-

normal distribution of connectivity estimates and the difference in trial number across subjects

and conditions (Khan et al., 2013; Maris et al., 2007).

Cluster-based permutation statistics within anatomical ROIs

Clusters of vertices showing a significant difference in connectivity across groups within the

target regions of interest were identified using cluster-based permutation statistics (Maris &

Oostenveld, 2007). This test was conducted for the alpha band (8Hz-12Hz) by averaging FC

estimates over the time dimension between 550ms and 850ms, a time window corresponding to

the onset and offset of brain activation to jump trials. This statistical test was implemented for

each seed (left and right) and target ROI (prefrontal, central and parietal) separately, using a one-

way ANOVA with a cluster-forming threshold of $F = 10.54$ (corresponding to a p value = 1e-06)

and 1000 permutations. Clusters were identified based on spatial adjacency. To achieve this, data

from each participant was morphed to a FreeSurfer average cortical representation with 10,242

vertices per hemisphere (Fischl, Sereno, Tootell, et al., 1999). Cluster size was then estimated for

each surviving cluster and results were filtered to keep only the top 5% biggest clusters. Finally,

clusters were sorted by their statistical significance, which was obtained by summing F values at

each vertex within the formed cluster, and only the most significant cluster per seed and

anatomical ROI was considered for further analyses. As a post-hoc control analysis, we repeated

the FC analysis pipeline using Coherence. This was done to examine the robustness of cluster

identification across the two connectivity methods. The two clusters reported here correspond to

those which showed overlap across the two methods. Cluster based permutation statistics were

implemented using MNE-python (Gramfort et al., 2013).

Linear mixed effect modelling

Linear mixed effect (LME) modeling was used to investigate brain responses to jump trials,

using the LME4 package (Bates et al., 2015) in R version 2023.6.1. The predicted variable was

the brain response to jump trials, quantified as the Area Under the Curve (AUC) at the activation

post jump onset. For this, the onset and offset latencies of brain activations in response to jump

trials were visually identified and recorded for each subject. This time window did not

significantly differ across groups (t=1.70, p =0.091. Table s3, supplementary materials). The

AUC for stay trials was obtained using the same subject-specific time window for jump

activations. Fixed effects included group (ASD, TD) condition (stay, jump), and hemisphere

(left, right). To control for within-subject nature of the task, a random effect for subject as

included. Maximum Likelihood Estimation analyses confirmed that the model outperformed a

null (intercept-only) model in predicting evoked responses to jumps $\chi^2(7) = 62.78$, p = 4.19e-12,

BIC $= 3209.95$, $R^2_{conditional} = 0.32$, $R^2_{marginal} = 0.24$).

Brain-behavior correlations

Spearman correlation was estimated between normalized FC estimates extracted from the two

surviving statistically significant clusters and two behavioral measures: the Social

Responsiveness Scale (SRS) total t score, a measure of Autism severity, and the Auditory

Sensory Profile Score (ASPS), a measure of auditory sensory processing in ASD. The ASPS was

quantified as the average of the first five questions of the auditory section of the Sensory Profile

(parent report), which measure auditory processing atypicality in terms of sensitivity to and

distraction by noise. These questions are categorically different from questions 6-8, which are

related to social communication (e.g., ignoring what parents say or not responding to their name

being called, question 6-7) and to enjoying or seeking to produce strange noises (question 8),

rather than to more basic auditory processing. Multiple comparisons were addressed by

correcting p values using the False Discovery Rate (FDR).

Results

Evoked responses in the auditory cortex

Both ASD and TD individuals showed increased activations during jump trials compared

to stay trials (i.e., a jump-evoked response). AUC values between onset and offset of peak

activations during jump trials for both conditions are presented in Table 1.

	AUC - Jump				AUC - Stay			
	Left		Right		Left		Right	
	Mean	*SD*	*Mean*	*SD*	*Mean*	*SD*	*Mean*	*SD*
TD	1267.09	835.24	1321.34	864.35	538.77	451.13	532.25	382.52
ASD	845.09	730.18	950.64	434.52	430.63	326.76	503.19	301.30

Table 1. Area Under the Curve (AUC) for TD and ASD individuals at peak between onset and
offset of brain activations to jump trials.

Figure 3 shows the mean source time course for ASD and TD individuals in left and right

auditory labels. Mixed effect modeling confirmed a significant effect of group (F (194.10) =

7.84, p = 0.006), condition (F (165.53) = 52.31, p = 1.67-11) as well as a statistically significant

interaction between group and condition (F (165.53) = 3.97, p = 0.049). There was no significant

effect of hemisphere in predicting the AUC in brain activations (F (165.53) = 0.47, p = 0.49).

Post-hoc pairwise comparisons showed that, within each group, mean AUC values for jump trials

were significantly higher than stay trials (ASD: t(171) = 3.39, p = 9.0e-04, TD: t(171) = 7.26, p <

1.0e-4, insets in figures 3a and 3b). Across groups, AUC values during jump trials were

significantly higher for the TD group compared to the ASD group (t (185) = 3.39, p = 9.0e-05) in

both left (figure 3c) and right (figure 3d) auditory labels.

Activations in response to ITD reversals peaked on average at 734ms in the left auditory

cortex (SD = 52ms) and at 724ms in the right auditory cortex (SD = 48ms) in the ASD group. In

the TD group, the left hemisphere responses peaked at 742ms (SD = 50ms) whereas and the right

hemisphere responses peaked at 739ms (SD = 51ms). There was no statistically significant

difference in peak latency across the two groups.

Figure 3. Evoked activations in auditory cortical labels. **a.** and **b.** Activations in the left (**a**) and right (**b**) auditory cortices for the TD (green) and ASD (purple) groups in response to jump (solid lines) and stay (dotted lines) trials. Solid vertical line at time zero represents sound onset. Dotted vertical line at 0.55 s represents the onset of and ITD reversal in the jump condition. Gray shades indicate the 95% confidence interval. **c.** and **d**. Mean AUC values for jump trials for the TD (green) and ASD (purple) groups. Whiskers indicate the Standard Error of the Mean (SEM). *** p < 0.001, ** p < 0.01, * p < 0.05.

Functional connectivity between auditory areas and frontoparietal ROIs

Next, we investigated whether the spatial jump was also associated with different alpha-band

functional connectivity dynamics, using zPLI, which corresponds to the PLI estimates for the

Jumps condition normalized by PLI for the Stay condition. The seed time courses were the

bilateral activations observed in left and right auditory cortices within the manually outlines

labels, and the target regions were the prefrontal, central and parietal anatomical ROIs defined on

the basis of previously reported fMRI findings (see methods and supplemental materials). Based

on our prior studies where across-group differences were modulated by functional connectivity

in the alpha (8-12Hz) band (Alho, Khan, et al., 2023; Alho, Samuelsson, et al., 2023; Khan et al.,

2013; Mamashli et al., 2021), here too we chose to focus on zPLI estimates in the alpha

frequency between 550ms and 850ms, a time window corresponding to the onset-to-offset of the

evoked response to the jump event.

Non-parametric cluster-based permutations showed two statistically significant clusters where

alpha-band connectivity differed between the two groups. The first cluster ($n_{vertices}$ = 16, F =

44.29, p = 0.0010, figure 4a, left panel) showed significant connectivity differences across

groups between the left auditory seed and dorsolateral portions of the prefrontal cortex (dlPFC).

Within this cluster, alpha-band connectivity was significantly lower during jump trials compared

to stay trials (median zPLI = -0.037, SD = 0.022) for TD individuals (t (30) = -10.28, p = 1.23e-

10, figure 4b, left),), but higher during jump trials compared to stay trials (median zPLI = 0.021,

SD = 0.027) for the ASD group (t (20) = -2.71, p = 0.015, figure 4b, right). Additionally,

prefrontal zPLI estimates significantly differed across groups (t(50) = -8.25, p = 1.23e-10, figure

4b).

The second cluster ($n_{vertices}$ = 17, F = 67.25, p = 0.0010, figure 4c, left panel) showed significant

connectivity differences across groups between the left auditory seed and portions of the left

superior postcentral sulcus and superior parietal gyrus. Within this cluster, alpha-band

connectivity was significantly lower during jump trials compared to stay trials (median zPLI = -

0.06, SD = 0.02) for TD individuals (t (30) = -16.38, p = 1.49e-15, figure 4d, left), but zPLI

estimates during jump trials did not significantly differ from the stay condition (t(20) = 1.34, p =

0.19) for the ASD group (median zPLI = 0.01, SD = 0.02, figure 4d, right).). Parietal functional

connectivity within this parietal cluster was also significantly different across groups (t(50) = -

11.19, p = 1.44e-14, figure 4d). All reported p values are FDR-corrected.

Figure 4. Functional connectivity or the two statistically significant clusters. **a**. and **c**.
Anatomical location of the statistically significant clusters in prefrontal (**a**) and the superior
parietal regions (**c**). **b**. and **d**. Median alpha-band connectivity estimates obtained from the
dlPFC (**b**) and SPL clusters (**d**) for the TD (green) and ASD (purple) groups. Whiskers
represent the Standard Error (SE). Asterisks on top of horizontal lines denote significant
differences across groups. Asterisks below horizontal lines denote a significant difference

Correlations between brain measures and behavioral measures

To test whether the MEG measures correspond to relevant ASD traits, we correlated the

magnitude of the evoked responses, as well as estimates of normalized functional connectivity

(zPLI) with behaviorally assessed phenotypic measures of ASD severity (SRS t-score) and

auditory processing, (the ASPS score). We did not find significant associations between any of

the behavioral measures and the evoked responses. Similarly, neither behavioral measure was

significantly correlated with zPLI in the TD group

In contrast, for the ASD group, higher alpha band functional connectivity in the prefrontal cluster

was positively correlated with autism severity, as measured by the SRS t-score (r(18) = 0.51, p =

0.021, p = 0.08 corrected), although this correlation did not survive multiple comparison

correction. Prefrontal connectivity in the alpha band was also negatively correlated with the

auditory processing component of the Sensory Profile Questionnaire in the ASD group (r(18) = -

0.63, p = 0.003, p $_{FDR-adjusted}$ = 0.025). The correlations between the behavioral measures and

zPLI are shown in Figure 5. Lastly, to check for any possible effects of age, we computed post-

hoc correlations between age and FC estimates, and between age and cortical activations. No

significant correlation was found between age and these two MEG measures of brain function,

meaning age was not a factor in the response patterns and observed group differences.

Figure 5. Brain-behavior correlations. **a**. and **b**. Correlation between alpha-band connectivity estimates obtained from the prefrontal cluster and scores in the Social Responsiveness Scale, SRS (**a**) and Auditory Sensory Profile Scores, ASPS (**b**) in the TD group (green). **c**. and **d**. Correlation between alpha-band connectivity in the prefrontal cluster and SRS (**c**) and ASPS scores (**d**) in the ASD group (purple).

Discussion

The aim of this study was to investigate passive auditory spatial attention in ASD. More specifically, we examined cortical activations and alpha-band functional connectivity in individuals with ASD and TD individuals in response to changes in sound location (i.e., jump trials) under passive listening conditions. We found that ASD individuals showed decreased cortical activations in cortical auditory regions and altered alpha frequency band functional connectivity dynamics between auditory and frontal areas. Additionally, functional connectivity estimates for the ASD group were significantly correlated with behavioral measures of ASD severity. Our findings provide insight into the differential neural processing of spatial auditory information in ASD and TD groups.

ASD individuals show reduced auditory activation during jump trials compared to TDs

The fact that both groups show cortical responses to jump trials confirms that both ASD and TD individuals are sensitive to acoustic cues indicative of a change in the location of a sound in space. This would be in line with findings which suggest that despite relative differences in performance, both groups can behaviorally locate sounds in space (Lodhia et al., 2018; Soskey et al., 2017; Teder-Sälejärvi et al., 2005). While both groups detected changes in localization, auditory cortical activations were significantly lower in the ASD group, which may be indicative of altered encoding of perceived acoustic features that are important to map sounds in space. These results are in line with previous electrophysiological research, which show a smaller amplitude of early and mid-latency auditory components during spatial (Lodhia et al., 2018; Teder-Sälejärvi et al., 2005) and non-spatial auditory processing tasks (Bruneau et al., 1999;

Jansson-Verkasalo et al., 2005; Orekhova et al., 2009; Port et al., 2015; Russo et al., 2009;

Soskey et al., 2017). They are also in line with postmortem human evidence in ASD of

malformed brainstem circuits (i.e., superior olive) that are thought to be first nuclei along the

ascending auditory pathway to process binaural cues, such as interaural time delays (Kulesza et

al., 2011). The significant group-by-condition interaction, where the ASD group exhibited a

smaller magnitude of activation during jump trials, further supports the notion that individuals

with ASD exhibit atypical auditory processing, potentially resulting from altered encoding or

binding of relevant acoustic features (Bharadwaj et al., 2022; Gonçalves & Monteiro, 2023;

Hames et al., 2016; Keehn et al., 2010, 2017; Lodhia et al., 2018; Marco et al., 2011).

ASD and TD Groups show opposite connectivity patterns in response to jump trials

Our functional connectivity analyses revealed significant differences in temporo-frontal and

temporo-parietal alpha-band connectivity between ASD and TD groups, with the ASD group

showing increased connectivity during jump trials relative to the TD group. These results suggest

that the ASD group may lack the typical modulation of connectivity in response to sensory

changes, potentially reflecting a disruption in the dynamic adjustment of attentional networks

required for processing auditory spatial shifts, especially given that this is a passive paradigm,

where attention was presumably directed away from the sounds. Atypical functional connectivity

involving the auditory cortex, measured with EEG or MEG, has been extensively documented in

ASD, by our group and many others (Alho, Khan, et al., 2023; Alho, Samuelsson, et al., 2023;

Demopoulos et al., 2024; Mamashli et al., 2017). Long-range alpha-band functional connectivity

has been previously proposed as a neurophysiological marker of attentional modulation via

coordination of temporal windows of neural excitability across cortical space (Klimesch et al.,

2007; Palva & Palva, 2011; Sadaghiani & Kleinschmidt, 2016). Increased alpha-band

connectivity in the ASD group could therefore be associated with atypical synchronization of

attentional networks involved in sensory integration and cognitive control, which are often

reported as dysregulated in ASD. The lower connectivity observed in the TD group could reflect

a more typical response to the auditory shift, suggesting more efficient attentional allocation and

neural processing. Although not related to functional connectivity, previous studies on auditory

spatial attention in ASD have reported lower amplitude of mid-lantency ERP components

associated with attention and executive function, namely the P300 and P400 (Lodhia et al., 2018;

Teder-Sälejärvi et al., 2005). This lends additional support to the idea that, on top of sensory

encoding, attentional processes also differ in ASD during execution of auditory spatial attention

tasks.

Prefrontal alpha connectivity in ASD correlates with auditory processing and autism severity

We found that higher temporo-parietal alpha band connectivity was positively correlated with

higher manifestations of ASD symptoms as measured by the Social Responsiveness Scale (SRS).

While this correlation did not survive correction for multiple comparisons, the trend was very

strong, and is consistent with other studies showing a correlation between functional connectivity

of auditory cortex with both cerebral and cerebellar ROIs and ASD symptoms severity (Alho,

Samuelsson, et al., 2023; Sharda et al., 2018; Wilson et al., 2022).

Additionally, higher alpha-band connectivity between auditory and the prefrontal regions

was negatively correlated with behaviorally assessed auditory processing symptoms, as

measured by the Auditory Sensory Profile Score (ASPS) in the ASD group, potentially indicating

that higher connectivity is linked to poorer auditory integration of spatial cues. This is also in

line with previous work in our group, showing that increased connectivity between auditory and

parietal regions is associated with poorer auditory processing skills, as measured by the ASPS

(Alho, Khan, et al., 2023).

The finding of increased alpha-band functional connectivity in ASD during spatial

attention tasks is in line with similar findings observed in resting state which propose alpha-band

hyperconnectivity as a characteristic feature of ASD (Demopoulos et al., 2024; Orekhova et al.,

2014; Wang et al., 2020). These findings highlight the potential impact of prefrontal and parietal

regions on auditory processing in ASD.

Limitations

This study has two limitations that should be addressed in future research. First, the relatively

small sample sizes of the ASD and TD groups limit the power of the findings. While the findings

align well with prior studies, future studies with larger cohorts will be important for further

confirming the observed group differences in cortical activations and connectivity patterns, as

well as to potentially uncover additional differences. Second, because our results were obtained

during passive listening conditions, we lack task-relevant behavioral correlates that allow us to

rule out the possibility that participants with ASD actively engaged auditory attention differently

from the TD participants. Future studies with an active spatial auditory attention task and

behavioral assessments of spatial auditory attention will be needed to elucidate any behavioral

differences between the groups, and links between these differences and cortical measures.

Conclusion

The results presented here provide novel insights into the neural mechanisms underlying auditory

spatial attention in ASD. We found that ASD was associated with reduced auditory cortical

activations and altered alpha-band functional connectivity between auditory cortex and frontal

and parietal regions, in response to shifts in the spatial location of sounds. The

neurophysiological measures were correlated with measures of ASD symptoms extent and

behaviorally assessed auditory processing, in the ASD group, thus confirming their relevance to

the ASD phenotype. Future directions include the evaluation of these metrics with an active,

rather than passive, auditory spatial attention paradigm, and extending it to larger samples as

well as to related conditions, so as to elucidate the specificity of the results to ASD.

Conflict of interest

The authors declare no conflict of interest

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