Age-related changes in neural synchronization and the sustained response to amplitude and frequency modulation of sound

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27   **Abstract**

Perception of speech requires sensitivity to features, such as amplitude and frequency modulations, which

are often temporally regular. Previous work suggests age-related changes in neural responses to temporally

regular features, but little work has focused on age differences for different types of modulations. We

recorded magnetoencephalography in younger (21–33 years) and older adults (53–73 years) to investigate

age differences in neural responses to slow (2-6 Hz sinusoidal and non-sinusoidal) modulations in amplitude,

frequency, or combined amplitude and frequency. Neural responses to sound onset (independent of

temporal modulations) were enhanced in older compared to younger adults, suggesting hyperresponsivity

and a loss of inhibition in the aged auditory system. Analyses of neural activity to modulations revealed

enhanced neural synchronization with amplitude, frequency, and combined amplitude-frequency

modulations for older compared to younger adults. This enhancement generalized across different degrees

of temporal regularity (sinusoidal and non-sinusoidal), although neural synchronization was generally lower

for non-sinusoidal modulation. Despite enhanced synchronization, sustained neural activity was reduced in

older compared to younger adults for sounds modulated both sinusoidally and non-sinusoidally in frequency.

Our results suggest substantial age differences in the sensitivity of the auditory system to features present in

speech and other natural sounds.

47   **Keywords:** Magnetoencephalography, aging, temporal regularity, sound processing, auditory cortex

Introduction

Many middle-aged or older adults live with some form of hearing impairment (Cruickshanks et al., 1998; Feder et al., 2015; Goman and Lin, 2016) and often experience difficulties understanding speech in the presence of background sound (Pichora-Fuller, 2003; Pichora-Fuller et al., 2016). Speech contains regular and quasi-regular low-frequency (<10 Hz) fluctuations in amplitude and frequency (Rosen, 1992; Varnet et al., 2017). Sensitivity to such amplitude and frequency fluctuations is thought to be important for segregating concurrent sound as separate streams (Schröger, 2005, 2007; Snyder and Alain, 2007; Winkler et al., 2009; Bendixen, 2014) and for recognizing and predicting features of the acoustic environment (Jones and Boltz, 1989; Barnes and Jones, 2000; Jones et al., 2002; Henry and Herrmann, 2014; Nobre and van Ede, 2018). The current study is concerned with how low-frequency amplitude and frequency modulations are processed in the brain and whether processing of amplitude and frequency modulations differ between younger and older adults.

Sensitivity of the auditory system to temporally regular structure in sounds has been examined by focusing on at least two neural signatures: neural synchronization and sustained neural activity. A large body of work has shown that neural activity recorded using electro- or magnetoencephalography (EEG/MEG) synchronizes with regular amplitude and frequency modulations in sounds (Maiste and Picton, 1989; John et al., 2001; Boettcher et al., 2002; Picton et al., 2003; Henry and Obleser, 2012; Henry et al., 2014; Goossens et al., 2016; Herrmann and Johnsrude, 2018; Goossens et al., 2019; Herrmann et al., 2019; Bauer et al., 2021) and that this neural synchronization increases with increasing coherence of a sound's temporal modulation (Herrmann and Johnsrude, 2018). Synchronization of neural activity with regular structure in sounds is thought to provide the means by which neural circuits support generating predictions about the timing and features of future sounds (Henry and Herrmann, 2014) and filter out irrelevant sound information with different temporal dynamics (Lakatos et al., 2008; Lakatos et al., 2013; O'Connell et al., 2014; Obleser and Kayser, 2019).

The second neural signature of temporal regularity processing is a sustained response in EEG/MEG recordings (Barascud et al., 2016; Southwell et al., 2017; Herrmann et al., 2021). The sustained response is an ultralow-frequency response (i.e., a constant or slowly drifting offset) elicited by the occurrence of a temporally regular pattern in a sound. A sustained response has been demonstrated in response to repeated sets of pure tones in longer sequences of tones (Barascud et al., 2016; Southwell et al., 2017; Herrmann and Johnsrude, 2018; Southwell and Chait, 2018; Herrmann et al., 2022), coherent frequency changes in multi-tone chords (Teki et al., 2016; Dheerendra et al., 2021), and complex sounds made of isochronous tone sequences (Sohoglu and Chait, 2016). A sustained response has also been observed for amplitude (Ross et

al., 2002; Herrmann et al., 2019) and frequency modulations in sounds (Herrmann and Johnsrude, 2018),

and for isochronously presented clicks or vowels compared to irregular presentations (Gutschalk et al., 2002;

Gutschalk and Uppenkamp, 2011). A sustained response elicited by temporally regular patterns in sounds

has been suggested as a possible indicator of prediction-related processes (Heilbron and Chait, 2018).

Aging and hearing loss are associated with a loss of inhibition and an increase in excitation in the

auditory system (Caspary et al., 1995; Caspary et al., 2008; Takesian et al., 2012; Knipper et al., 2013;

Auerbach et al., 2014; Ouellet and de Villers-Sidani, 2014; Resnik and Polley, 2017; Salvi et al., 2017;

Herrmann and Butler, 2021b) that are thought to result from damage to the auditory periphery and the

associated deprived input to the auditory neural pathway (Zhao et al., 2016; Herrmann and Butler, 2021a).

The consequence of reduced inhibition and increased excitation is hyperactivity and hyperresponsivity to

suprathreshold sounds, particularly at the level of auditory cortex (Auerbach et al., 2014; Salvi et al., 2017;

Herrmann and Butler, 2021a). In EEG/MEG recordings, hyperresponsivity is prominently represented as

larger auditory cortex responses to sound onset for older compared to younger adults or for adults with

hearing loss compared to adults without (Ross and Tremblay, 2009; Sörös et al., 2009; Lister et al., 2011;

Alain et al., 2014; Bidelman et al., 2014; Parry et al., 2019; Irsik et al., 2021; Herrmann et al., 2022).

Consistent with hyperresponsivity to sound, neural synchronization with the low-frequency

amplitude modulations of narrow-band sounds has been shown to be larger in older compared to younger

people (Purcell et al., 2004; Goossens et al., 2016; Herrmann et al., 2019). Synchronization with the amplitude

envelope of speech also appears to be enhanced in older compared to younger people (Presacco et al., 2016b,

a; Decruy et al., 2020; Broderick et al., 2021) and in older adults with hearing loss compared to those without

(Decruy et al., 2020; Fuglsang et al., 2020; but see Presacco et al., 2019). Synchronization seems particularly

enhanced in older adults when the amplitude modulation envelope is characterized by rapid onsets and slow

offsets compared to the reverse (Irsik et al., 2021). Age differences in neural synchronization to frequency-

modulated, or amplitude- and frequency-modulated, sounds have been less investigated. In one study, neural

synchronization in older people appeared to be increased when sounds were frequency modulated,

consistent with hyperresponsivity (Boettcher et al., 2002). In another, no age difference in synchronization

to a frequency-modulated sound was observed under passive listening conditions, whereas synchronization

was reduced for older adults under active listening conditions (Henry et al., 2017). However, the sounds

used in the Henry et al (2017) study contained short gaps that briefly interrupted the frequency modulation.

Response to gaps may differ between younger and older adults (Ross et al., 2010; Harris et al., 2012) and

this may affect measured neural synchronization. Given this conflicting evidence, here we examine how the

brains of younger (aged 21-33 years) and older (aged 53-73 years) people respond to sounds with low-
frequency modulation in amplitude, frequency, and both combined.

Most studies that have investigated neural synchronization, and age differences therein, have focused
on sinusoidal amplitude or frequency modulations (Boettcher et al., 2002; Goossens et al., 2016; Henry et
al., 2017; Goossens et al., 2019; Herrmann et al., 2019) which are not very naturalistic. Natural speech
contains low-frequency, non-sinusoidal fluctuations in amplitude and frequency; these include the amplitude
changes associated with uttering syllables/words and frequency changes associated with the prosodic
contour (Rosen, 1992; Varnet et al., 2017). Neural synchronization to the amplitude envelope of speech –
reflecting non-sinusoidal amplitude fluctuations – has been shown to be enhanced for older compared to
younger adults (Presacco et al., 2016b, a; Decruy et al., 2020; Broderick et al., 2021). Age differences in neural
synchronization with non-sinusoidal frequency modulations have not yet been investigated.

Whether the sustained neural response differs between age groups has been investigated much less
compared to age differences in neural synchronization. The existing work points to an age-related reduction
in the sustained response. For example, short pure tones elicited a smaller sustained response in older
compared to younger adults (Pfefferbaum et al., 1979). The sustained neural response elicited by repeated
sets of tones in longer tone sequences has also been shown to be reduced for older relative to younger adults
(Al Jaja et al., 2020; Herrmann et al., 2022). A similar age-related reduction in the sustained response has
been reported for low-frequency amplitude modulations (Herrmann et al., 2019), although the effect of age
group (younger vs older) was only marginally significant in this study. Whether the sustained response for
frequency-modulated sounds or sounds with concurrent amplitude and frequency modulations – sinusoidal
or non-sinusoidal – differs between younger and older adults has not been examined.

In the current study, MEG was recorded in two experimental sessions while younger and older adults
listened to narrow-band noise sounds that were modulated in amplitude, frequency, or both concurrently.
Amplitude and frequency were modulated either sinusoidally at 4 Hz or non-sinusoidally between 2 and 6
Hz. We examined the effects of modulation type (amplitude, frequency, both) and regularity (sinusoidal,
non-sinusoidal) on neural synchronization and the sustained neural response, and whether these neural
signatures differ between younger and older adults.

<p style="text-align:center">**Methods and Materials**</p>

Participants

Twenty-six younger (mean: 26.7 years; range: 21–33 years; 13 males and 13 females) and twenty-five older adults (mean: 63.9 years; range: 53–73 years; 11 males and 14 females) were recruited for the current study. Participants provided written informed consent. None of the participants reported having a neurological disease, hearing impairment, or a hearing-aid prescription. Data were acquired in two sessions separated by at least one and maximal 43 days (median: 7 days). There was no age-group difference in the number of days between sessions ($t_{49} = 0.99$, p = 0.327). The study was conducted in accordance with the Declaration of Helsinki and the Canadian Tri-Council Policy Statement on Ethical Conduct for Research Involving Humans, and was approved by the Nonmedical Research Ethics Board of the University of Western Ontario. Additional data (~12 min) for a different project (Herrmann et al., 2022) were recorded during each of the two each sessions. Apart from audiometric hearing thresholds, none of the data reported here have been published previously.

Audiometric hearing assessment and hearing thresholds

For each participant, pure-tone audiometric thresholds were estimated for pure tones at frequencies of 0.25, 0.5, 1, 2, 4, and 8 kHz (Figure 1). Mean thresholds averaged across 0.25, 0.5, 1, 2, and 4 kHz (i.e., pure-tone average) were higher for older compared to younger adults ($t_{49} = 7.79$, p = 4×10^{-10}, d = 2.182; Figure 1, right; Herrmann et al., 2022). This is consistent with mild-to-moderate hearing loss in our sample of older adults, as would be expected (Moore, 2007; Plack, 2014; Presacco et al., 2016b; Herrmann et al., 2018).

Figure 1: Audiograms and pure-tone average thresholds. Left: Audiometric thresholds: The thin lines reflect data from each individual participant. Thick lines reflect the mean across participants. **Right:** Pure-tone average hearing threshold (mean across 0.25, 0.5, 1, 2, and 4 kHz). Gray dots reflect the pure-tone average threshold for individual participants. Error bars show the standard error of the mean. *p<0.05

In order to obtain a reference threshold in MATLAB software for sound presentation during the main experimental procedures, we also estimated the sensation level for a narrow-band noise that was

generated similarly to the sounds used in the current study (see below), using a method-of-limits procedure
(Herrmann et al., 2022). Participants listened to a 15-s narrow-band noise (~0.84–2.6 kHz) that changed
continuously in intensity at a rate of 4 dB/s (either decreased [i.e., starting at suprathreshold levels] or
increased [i.e., starting at subthreshold levels]). Participants pressed a button when they could no longer hear
the tone (intensity decrease) or when they started to hear the tone (intensity increase); the sound stopped
after the button press. The sound intensity at the time of the button press was noted for 6 decreasing sounds
and 6 increasing sounds (decreasing and increasing sounds alternated), and these were averaged to determine
the sensation level. As expected, given the audiometric pure-tone average thresholds (Figure 1), sensation
levels were elevated for older compared to younger adults (t_{49} = 4.593, p = 3.07 × 10^{-5}, d = 1.287, mean
difference: 6.5 dB).

To approximately match audibility across age groups, all sound stimuli presented during the main
experiment (described below) were presented at 55 dB above each individual's sensation level (Herrmann et
al., 2022). Since sensation levels and audiometric thresholds were on average larger for older compared to
younger adults, sounds presented during the main experiment were on average more intense in sound-
pressure level in older compared to younger adults. Previous research shows that brain responses to sound
increase with increasing sound intensity (Picton et al., 1974; Picton et al., 1978; Pfefferbaum et al., 1979;
Polich et al., 1988; Schadow et al., 2007; Herrmann et al., 2018). Because we predicted regularity-related
sustained activity to be reduced in older compared to younger adults (Pfefferbaum et al., 1979; Al Jaja et al.,
2020; Herrmann et al., 2022), the higher sound levels used with older adults works against this hypothesis.
However, we are predicting that responses to sound onset and neural synchronization with temporal
regularity in sounds to be larger in older compared to younger adults (Sörös et al., 2009; Alain et al., 2012;
Goossens et al., 2016; Herrmann et al., 2019). A higher sound level could thus bias our data towards our
hypothesized outcome. To mitigate this issue, data analysis for sound-onset responses and neural
synchronization was conducted for a subset of 17 participants per age group for which sensation levels and
sound-presentation levels did not differ (t_{32} = 1.331, p = 0.193, d = 0.456; see also Herrmann et al., 2022;
for this, we selected the 17 younger adults with the sensation levels and the 17 older adults with the lowest
sensation levels).

Sound stimulation and procedure
Acoustic stimuli were 4-s long narrow-band noises (see also Herrmann and Johnsrude, 2018; Herrmann et
al., 2019). Narrow-band noises were created by adding 100 pure tones (components) with different carrier
frequencies (Cf) that were randomly selected between 1200 and 2000 Hz, with random starting phase. Each

of the 100 pure-tone components was amplitude and frequency modulated. The amplitude modulation

depth was 100%. The range of the modulation in carrier frequency was defined as Cf ±Cf×0.3 (e.g., for a

component with a mean carrier frequency of 1200 Hz, the carrier frequency modulation ranged between

840 Hz and 1560 Hz). The rate at which the amplitude and the carrier frequency was modulated was not

fixed but changed randomly between 2 Hz and 6 Hz over time for each pure-tone component. The onset

phase of the amplitude and frequency modulations as well as the randomization of amplitude- and

frequency-modulation rate differed between components and between amplitude and frequency

modulations. The 100 components were summed to obtain a narrow-band noise sound. We henceforth refer

to this stimulus as the 'random' condition, because the signal resulting from the sum of the 100 pure-tone

components does not contain any low-frequency (2–6 Hz) amplitude or frequency modulations (Figure 2

top).

Figure 2: Stimulus conditions. Spectrograms for samples of narrow-band noise sounds for each of the seven stimulus conditions. The 'random' condition did not contain low-frequency amplitude or frequency modulations and had a relatively consistent spectrum over time. The other six conditions smoothly transitioned to sinusoidally (left) or non-sinusoidally (right) amplitude, frequency, or amplitude and frequency-modulated noises at around 1 s after sound onset. Abbreviations: sinusoidal amplitude modulation (AMsin), sinusoidal frequency modulation (FMsin), sinusoidal amplitude and frequency modulation (AMFMsin), non-sinusoidal amplitude modulation (AMnosin), non-sinusoidal frequency modulation (FMnosin), non-sinusoidal amplitude and frequency modulation (AMFMnosin).

202   Six additional conditions were created in which either the amplitude modulation (AM), frequency

modulation (FM), or both the amplitude and frequency modulations (AMFM) of each of the 100 pure-tone

components became synchronized (phase-aligned) approximately 1 second after sound onset. There was no

transient change at the transition from non-aligned to AM/FM phase-aligned pure-tone components (see

also Herrmann and Johnsrude, 2018; Herrmann et al., 2019; Figure 2). The modulation rate of the AM, FM,

or AMFM after phase-alignment was either constant and sinusoidal at 4 Hz (Figure 2 left) or changed

randomly and non-sinusoidally between 2 and 6 Hz for the rest of the sound (from ~1 s to 4 s; Figure 2

right). Averaging the 100 pure-tone components resulted in a narrow-band noise sound that was non-

modulated (random) for about 1 second, after which it transitioned to a sinusoidal or non-sinusoidal AM,

FM, or AMFM. Henceforth, we refer to the six conditions as 'sinusoidal amplitude modulation' (AMsin),

'sinusoidal frequency modulation' (FMsin), 'sinusoidal amplitude and frequency modulation' (AMFMsin),

'non-sinusoidal amplitude modulation' (AMnosin), 'non-sinusoidal frequency modulation' (FMnosin), 'non-

sinusoidal amplitude and frequency modulation' (AMFMnosin).

In each of the two recording sessions, participants were presented with five ~10-min blocks of

stimulation. Participants listened passively to 70 trials of each condition per session (5 blocks × 14 trials per

condition = 70 trials), and thus to 140 trial per condition across the study. Trials of the seven conditions

were presented pseudo-randomly throughout a block, such that each condition could occur maximally three

times in direct succession. Trials were separated by a 2-s inter-stimulus interval. During sound presentation,

participants watched a muted, subtitled, movie of their choice that was projected into the electromagnetically

shielded room via a mirror system.

Recordings and initial preprocessing of magnetoencephalographic data

Methods of data recording and preprocessing were similar to our previous work (Herrmann et al., 2022).

For each participant, magnetoencephalography was recorded in an electromagnetically shielded room

(AK3b, Vacuumschmelze, Hanau, Germany) using a 306-channel Neuromag Vectorview MEG (MEGIN

Oy, Helsinki, Finland) at the Max Planck Institute for Human Cognitive and Brain Sciences in Leipzig,

Germany. Data were recorded with a sampling rate of 1000 Hz and an online low-pass filter of 330 Hz (no

online high-pass filter). The signal space separation (SSS) method (maxfilter© version 2.2.15; default

parameter setting $L_{in} = 8$; $L_{out} = 3$) was used to suppress external interference, interpolate bad channels

(identified through visual inspection), and transform each person's individual data to the sensor space of the

first block of the first session to ensure the data are in a common space (Taulu et al., 2004; Taulu et al.,

2005).

Combination of magnetometer and gradiometer channels

Magnetic fields were recorded by one magnetometer channel and two gradiometer channels in each of 102

spatial locations distributed around a participant's head. Magnetometers measure magnetic fields in Tesla

(T), whereas gradiometers (a coupled pair of magnetometers) measure differences in the same magnetic

fields over a distance of 0.0168 m in Tesla per meter (T/m). Combining the two MEG channel types for

data analysis requires accounting for their different units. All gradiometer channels were transformed to

magnetometer channels, because this transformation only requires a linear interpolation that results in the

same unit for all channels (Herrmann et al., 2018, 2022). To this end, a scaling matrix **S** was multiplied with

the data matrix **D** to obtain scaled data according to the following equation:

$$D_{scaled} = S \times D,$$

243   where **D** is a 3 × n matrix. The three rows of **D** refer to the two gradiometers and one magnetometer

244   of one MEG sensor triplet (i.e., in one spatial location) and n to the number of data samples over time. The

245   scaling matrix **S** is a 5 × 3 matrix with the following elements:

$$S = \begin{matrix} -0.0084 & 0 & 1 \\ 0.0084 & 0 & 1 \\ 0 & -0.0084 & 1 \\ 0 & 0.0084 & 1 \\ 0 & 0 & 1 \end{matrix}$$

The value 0.0084 reflects half of the distance between the two gradiometer loops measured in meters,

and the transformation constitutes a linear approximation of the magnetic field at each of the triplets. The

transformation replaces the sensor triplet by a sensor quintet of magnetometers. The columns of **S** refer to

the triplet of two gradiometers and one magnetometer and the rows of **S** refer to the resulting five

magnetometers. This procedure resulted in signals from 510 magnetometer channels centered on and around

102 locations around a participant's head (Herrmann et al., 2018, 2022).

Preprocessing of magnetoencephalographic data

A high-pass (0.7 Hz; 2391 points, Hann window) and a low-pass filter (20.3 Hz, 119 points, Kaiser window)

were applied to the data. Data were downsampled from the 1000-Hz to a 250-Hz sampling frequency and

epochs ranging from –1 s pre-stimulus onset to 5 s post-stimulus onset were extracted. Independent

components analysis was calculated for epoched data using the Fieldtrip MATLAB toolbox (v20130727,

Oostenveld et al., 2011) with the runica method (Makeig et al., 1996) and the logistic infomax algorithm (Bell

and Sejnowski, 1995) to identify and remove activity related to blinks, horizontal eye movements, muscle

activity, and noisy channels. Finally, epochs that contained a signal change larger than 8 picotesla (pT) in any

channel were excluded. These data were used to investigate condition and age differences in evoked

responses to sound onset and neural synchronization.

The same analysis pipeline was separately run with slightly different parameters to investigate the

sustained response. The sustained neural response is a very low-frequency signal reflecting an offset shift

(Barascud et al., 2016; Southwell et al., 2017; Herrmann and Johnsrude, 2018). Application of a high-pass

filter would remove this signal and was thus omitted for the analysis of the sustained response. Moreover,

sounds contained temporal modulations (AM, FM, AMFM) at 4 Hz (sinusoidal) and 2–6 Hz (non-sinusoidal)

that drive neural activity in the corresponding neural frequency bands. In order to avoid that the measured

sustained response is influenced by neural-activity changes in the 2–6 Hz frequency band, data were filtered

with a 1.2 Hz low-pass filter (3001 points, Kaiser window). The analysis pipeline was otherwise similar. The

0-1.2 Hz band data were downsampled to 250 Hz and divided into 6-s epochs. Activity related to blinks,

horizontal eye movements, muscle activity, and noisy channels was removed using the identified

components from the 0.7-20.3 Hz band data. Epochs in which a signal change larger than 8 pT occurred in

any channel were excluded.

Analysis of responses to sound onset

Responses to sound onset were analyzed using the 0.7-20.3 Hz band data. To investigate whether older

adults exhibit hyperresponsivity to sound, and thus possibly a loss of inhibition, we investigate the M50 and

M100 responses to sound onset (cf. Sörös et al., 2009; Alain et al., 2012; Herrmann et al., 2018). Sounds

from different conditions were similar for the first second after sound onset, and data for this onset period

were thus averaged across all epochs (ignoring conditions). Absolute values were calculated for signals of

each channel to account for opposite polarities of the magnetic fields in directions perpendicular to the

tangential orientation of the underlying neural source. Separately for each channel, the mean signal from the

0.2 s period immediately prior to sound onset was subtracted from the signal at each time point (baseline

correction). In order to focus the analysis on the most sensitive channels, signals were averaged across the

10% of channels (N=51) at which the onset responses were maximal within the M50/M100 time window

(0.03–0.12 s), calculated separately for each participant. This resulted in one response time course per

participant.

For the statistical analysis, responses were separately averaged in the 0.03–0.07 s and the 0.08–0.12

s time windows for the M50 and M100, respectively. Separate independent samples t-tests were used to

compare M50 and M100 responses between age groups. We focused analyses on the 17 participants from

each age group for which sensation level – and thus the level at which sounds were delivered – did not differ

between age groups, but data for all participants are provided as well.

Analysis of neural synchronization: Cross-coherence

Neural synchronization was analyzed using the cross-coherence measure (Nolte et al., 2004; Keitel et al.,

2016). Cross-coherence is related to the more common measures of neural synchronization, namely inter-

trial phase coherence or evoked amplitude/power (Lachaux et al., 1999; Picton et al., 2003; Nozaradan et

al., 2011; Anderson et al., 2012; Herrmann et al., 2013; Herrmann and Johnsrude, 2018). Inter-trial phase

coherence and evoked amplitude/power require that neural activity is temporally consistent across trials. In

the current study, the temporal modulations of sounds with a non-sinusoidal stimulus rhythm differed from

trial to trial. Inter-trial phase coherence and evoked amplitude/power are thus not suitable for the analysis

of neural activity in the current study. Cross-coherence provides an alternative that is similarly interpreted

as inter-trial phase coherence (Nolte et al., 2004; Keitel et al., 2016).

Calculation of cross coherence focused on the 1–4 s time window following sound onset (the first

second did not contain any low-frequency AM, FM, or AMFM) and involved the following steps. A fast

Fourier transform (FFT; Hann window; zero-padding) was calculated for each single-trial stimulus time

course and for each single-trial neural time course (separately for each MEG channel). This resulted in a

stimulus and a neural spectrum of complex numbers for each trial and channel. For each MEG channel and

neural-frequency bin in the FFT spectrum, coherence was then calculated between the stimulus spectrum

and the neural spectrum according to the following equation (Nolte et al., 2004):

$$coh = \left| \frac{\langle x \cdot \bar{y} \rangle}{(\langle x \cdot \bar{x} \rangle \times \langle y \cdot \bar{y} \rangle)^{0.5}} \right|,$$

where x and y reflect the complex numbers and $\bar{x}$ and $\bar{y}$ the complex conjugates of the stimulus and

the neural spectrum, respectively. The dot refers to the multiplication operation. The bracket notation $\langle \rangle$

refers to the expected value, that is, the mean across trials. The notation $||$ reflects the magnitude of the

complex number.

To analyze cross-coherence for the 4-Hz sinusoidal stimulation, cross-coherence was averaged in

the 3.95 to 4.05 Hz frequency window. To analyze cross-coherence for the 2–6 Hz non-sinusoidal

stimulation, coherence was averaged in the 2–6 Hz frequency window. In order to focus the analysis on the

most sensitive channels, coherence was averaged across the 10% of channels (N=51) yielding the largest

coherence; channels were identified independently for each participant and condition. Note that this

selection is orthogonal to comparisons between conditions or age groups.

For the statistical analyses, cross-coherence for the random condition was subtracted from the cross-

coherence of each of the other six conditions (AMsin, FMsin, AMFMsin, AMnosin, FMnosin,

AMFMnosin). Statistical analyses were conducted separately for sinusoidal and non-sinusoidal stimulus

conditions, because the different stimulation frequencies resulted in qualitatively different coherence spectra

(see Figure 4). Hence, two separate repeated-measures ANOVAs were calculated (sinusoidal; non-

sinusoidal) using cross-coherence as the dependent measure. The within-subjects factor was Modulation

Type (AM, FM, AMFM) and the between-subjects factor was Age Group (younger, older). The Greenhouse

and Geisser correction was used when Mauchly's test indicated that sphericity was violated (Greenhouse

and Geisser, 1959). The original degrees of freedom are reported, along with the epsilon coefficient, and the

corrected p-value. Analyses focused on the 17 participants from each age group for which sensation level

did not differ between age groups, but data for all participants are provided as well.

Analysis of neural synchronization: Relationship between stimulus phase and neural response

Cross-coherence provides a spectrum that can be interpreted similar to inter-trial phase coherence (Lachaux

et al., 1999; Nolte et al., 2004; Keitel et al., 2016). However, non-sinusoidal responses in the time domain

lead to distributed energy across the fundamental frequency and its harmonics in the cross-coherence

spectrum (due to the sinusoidal assumption of the Fourier transform). Moreover, cross-coherence spectra

cannot be compared between sinusoidal and non-sinusoidal modulations due to their different

characteristics. To account for these disadvantages associated with coherence spectra, we analyzed neural

response amplitude as a function of stimulus phase (Irsik et al., 2021). To this end, the stimulus AM, FM, or

AMFM of each trial was extracted from the 1–4 s time window after stimulus onset and downsampled to

match the sampling of the MEG signal. Because the random condition did not contain a coherent low-

frequency temporal modulation, we randomly selected, for each random-condition trial, a stimulus AM, FM,

or AMFM from another trial within the same block. The analytic Hilbert transform was calculated for each

trial-specific stimulus modulation (AM, FM, or AMFM), and the resulting complex numbers were used to

calculate the phase angle for each time sample within 1–4 s. As a result, data for each trial comprised a

stimulus phase series and the corresponding neural amplitudes (for each MEG channel). The neural

amplitude data were binned according to the stimulus phase values (number of bins: 100; window width:

0.314 radians; with overlap), such that neural amplitude was represented as a function of stimulus phase

ranging from $-\pi$ to π (binning accounted for circularity). This resulted in a phase-amplitude series for each

MEG channel, reflecting the neural response in a stimulation cycle.

In the phase-amplitude series, the amplitude range between positive and negative deflections

captures the magnitude of the synchronized response. In order to calculate whether neural responses differed

between conditions and age groups, the amplitude range was calculated separately for each stimulus

condition (random, AMsin, FMsin, AMFMsin, AMnosin, FMnosin, AMFMnosin), age group (younger,

older), and MEG channel as the difference between the maximum and the minimum amplitude in the phase-

amplitude series. The amplitude range was averaged across the 10% of channels (N=51) yielding the largest

amplitude (channels were independently identified for each participant and condition).

For statistical analyses, the amplitude range of the random condition was separately subtracted from

the amplitude range of each of the other six conditions (AMsin, FMsin, AMFMsin, AMnosin, FMnosin,

AMFMnosin). A rmANOVA was calculated using amplitude range as the dependent measure. Within-

subjects factors were Modulation Type (AM, FM, AMFM) and Regularity (sinusoidal, non-sinusoidal), and

the between-subjects factor was Age Group (younger, older). The Greenhouse and Geisser correction was

used when Mauchly's test indicated that sphericity was violated (Greenhouse and Geisser, 1959). The original

degrees of freedom are reported, along with the epsilon coefficient, and the corrected p-value. Analyses

focused on the 17 participants from each age group for which sensation level did not differ between age

groups, but bar graphs and individual data points are also provided.

Analysis of the sustained response

The sustained neural responses was analyzed using the 0-1.2 Hz band data. Single-trial time courses ranging

from -1 to 5 s time-locked to sound onset were averaged, separately for each condition (random, AMsin,

FMsin, AMFMsin, AMnosin, FMnosin, AMFMnosin). The absolute values of the averaged time courses

were calculated for each channel. The mean signal in the 1-s time window prior to sound onset was averaged

and subtracted from the signal at each time point, separately for each channel (baseline correction). Signals

were averaged across the 10% of channels (N=51) that yielded the highest amplitude of the sustained

response (channels were independently identified for each participant and condition), resulting in one

response time course per condition and per participant.

For statistical analyses, responses for the random condition were subtracted from the responses of

each of the other six conditions. Statistical comparisons focused on the 2–3.5 s time window. The start of

the time window at 2 s is similar to our previous work were sounds changed to a regular pattern

approximately 1 s after sound onset (Herrmann et al., 2022) and allows for the auditory system to be fully

responsive to the temporal modulations (see also Barascud et al., 2016; Teki et al., 2016; Herrmann and

Johnsrude, 2018). The very low-frequency low-pass filter can potentially smear neural responses in time. To

reduce the possibility that the response to the sound offset at 4 s affects our measurement of the sustained

response, we chose to limit the analysis time window to 3.5 s instead 4 s.

An rmANOVA was calculated using the sustained response as the dependent measure. Modulation

Type (AM, FM, AMFM) and Regularity (sinusoidal, non-sinusoidal) were within-subjects factors, and Age

Group (younger, older) was a between-subjects factor. The Greenhouse-Geisser correction was used when

Mauchly's test indicated that sphericity was violated (Greenhouse and Geisser, 1959). The original degrees

of freedom, epsilon coefficient, and corrected p-value are reported. All participants were included in the

analysis of the sustained response, since the more intense stimuli presented to older people would lead to

larger responses, if anything, and we hypothesized that sustained activity would be reduced in older

compared to younger adults (Pfefferbaum et al., 1979; Al Jaja et al., 2020; Herrmann et al., 2022).

392   **Results**

Enhanced response to sound onset in older adults

Time courses of neural responses to sound onset are displayed in Figure 3A. Responses were larger for older

compared to younger adults for both the M50 time window (0.03–0.07 s: t_{32} = 5.693, p = 2.64 × 10^{-6}, d =

1.953; Figure 3B) and the M100 time window (0.08–0.12 s: t_{32} = 5.123, p = 1.39 × 10^{-5}, d = 1.757; Figure

3C). This is consistent with previous work showing hyperresponsivity to sound in the auditory cortex of

older compared to younger adults (Anderer et al., 1996; Sörös et al., 2009; Alain et al., 2012; Herrmann et

al., 2022). Topographical distributions indicate neural generators in auditory cortex (Herrmann et al., 2022).

Figure 3: Results for sound-onset responses. A: Time courses in response to sound onset (from the 10% of channels with the highest response magnitude). **B:** M50 topographies and mean responses (0.03–0.07 s). **C:** M100 topographies and mean responses (0.8–0.12 s). Gray dots in panels B and C reflect data from individual participants. Data in panel A and topographies in panels B and C reflect the averaged data of the seventeen participants in each age group for which sensation levels did not differ statistically. Error bars reflect the standard error of the mean.

400   *Enhanced neural synchronization in older adults*

401   Neural synchronization was first analyzed using cross-coherence. Repeated-measures ANOVAs were

402   conducted separately for stimuli with sinusoidal and non-sinusoidal modulations, because different

403   frequency bands were used to capture synchronized activity: 4 Hz for sinusoidal stimulation and 2–6 Hz for

404   non-sinusoidal stimulation, corresponding to the respective stimulation frequencies.

Figure 4: Results for the cross-coherence (XCOH) analysis. A: Cross-coherence spectra for sinusoidal sound stimulation (from the 10% of channels with the highest cross-coherence). **B:** Topographical distributions for cross-coherence at 4 Hz for sinusoidal stimulation. **C:** Mean cross-coherence at 4 Hz for sinusoidal stimulation. The asterisk indicates a significant difference between the each condition relative to the random condition (p ≤ 0.05; FDR-corrected). **D-F:** Same as for panels A-C for non-sinusoidal sound stimulation and 2–6 Hz cross-coherence. Gray dots in panels C and F reflect data from individual participants. Data in panels A, B, D, and E reflect the averaged data of the seventeen participants in each age group for which sensation levels did not differ statistically. Error bars reflect the standard error of the mean.

We observed larger cross-coherence for 4-Hz sinusoidally modulated sounds in older compared to
younger adults, but this was only marginally significant (effect of Age Group: $F_{1,32} = 3.47$, $p = 0.072$, $\omega^2 =$
0.036). Cross-coherence was also larger for FM ($t_{33} = 2.538$, $p = 0.016$, $d = 0.435$) and AMFM sounds (t_{33}
$= 2.706$, $p = 0.011$, $d = 0.464$) compared to AM sounds (effect of Modulation Type: $F_{2,64} = 4.653$, $p = 0.013$,
$\omega^2 = 0.033$), but coherence for FM and AMFM did not differ ($t_{33} = 0.165$, $p = 0.87$, $d = 0.028$). The
Modulation Type × Age Group interaction was not significant ($F_{2,64} = 1.866$, $p = 0.163$, $\omega^2 = 0.008$; Figure
4A-C). Results for sounds with a 2–6 Hz non-sinusoidal modulation were similar: We observed larger, but
only marginally significant cross-coherence for older compared to younger adults (effect of Age Group: $F_{1,32}$
$= 3.916$, $p = 0.056$, $\omega^2 = 0.042$). Cross-coherence was larger for FM ($t_{33} = 2.032$, $p = 0.050$, $d = 0.348$) and
AMFM sounds ($t_{33} = 3.258$, $p = 0.003$, $d = 0.559$) compared to AM sounds (effect of Modulation Type: $F_{2,64}$
$= 5.328$, $p = 0.007$, $\omega^2 = 0.044$), but did not differ between FM and AMFM ($t_{33} = 1.243$, $p = 0.223$, $d =$
0.213). The Modulation Type × Age Group interaction was not significant ($F_{2,64} = 0.497$, $p = 0.611$, $\omega^2 <$
0.001; Figure 4D-F).

Figure 5: Results for the phase-amplitude analysis. A: Neural amplitude as a function of stimulus phase for sinusoidal sound stimulation (data averaged across the 10% of channels with the largest amplitude range). **B:** Topographical distributions for amplitude ranges for sinusoidal stimulation. **C:** Mean amplitude range for sinusoidal stimulation. The asterisk indicates a significant difference between the each condition relative to the random condition (p ≤ 0.05; FDR-corrected). **D-F:** Same as for panels A-C for non-sinusoidal sound stimulation. Gray dots in panels C and F reflect data from individual participants. Data in panels A, B, D, and E reflect the averaged data of the seventeen participants in each age group for which sensation levels did not differ statistically. Error bars reflect the standard error of the mean.

Neural synchronization was also quantified by calculating the relation between stimulus phase and

neural response amplitude. The rmANOVA for the amplitude range in the phase-amplitude response curves

showed larger response ranges for FMs than AMs (t_{33} = 2.44, p = 0.02, d = 0.419), for AMFMs than AMs

(t_{33} = 5.924, p = 1.21 × 10^{-6}, d = 1.016), and for AMFMs than FMs (t_{33} = 2.846, p = 0.008, d = 0.488; effect

of Modulation Type: $F_{2,64}$ = 15.352, p = 3.58 × 10^{-6}, ω^2 = 0.117). Sinusoidal stimulation also led to larger

amplitude ranges compared to non-sinusoidal stimulation (effect of Regularity: $F_{1,32}$ = 84.452, p = 1.72 ×

10^{-10}, $\omega^2 = 0.212$). The Modulation Type × Regularity interaction was marginally significant ($F_{1,64} = 3.373$, p

$= 0.051$, $\varepsilon = 0.824$, $\omega^2 = 0.008$). We confirmed for each of the different modulation types (AM, FM, AMFM)

that the sinusoidally modulated sounds elicited larger amplitude ranges than non-sinusoidally modulated

sounds (for all: $t_{33} > 5.5$, $p < 0.001$, $d > 0.95$).

Critically, we also observed larger response ranges for older compared to younger adults (effect of

Age Group: $F_{1,32} = 10.493$, $p = 0.003$, $\omega^2 = 0.126$), but no interaction involving Age Group (Modulation

Type × Age Group: $F_{2,64} = 1.318$, $p = 0.275$, $\omega^2 = 0.003$; Regularity × Age Group: $F_{1,32} = 3.243$, $p = 0.081$,

$\omega^2 = 0.007$; Modulation Type × Regularity × Age Group: $F_{2,64} = 1.337$, $p = 0.268$, $\varepsilon = 0.824$, $\omega^2 = 0.001$).

However, given that the Regularity × Age Group interaction was marginally significant, we confirmed that

larger amplitude ranges for older compared to younger adults were observed for sinusoidal ($t_{32} = 3.052$, $p = $

0.005, $d = 1.047$) and non-sinusoidal sound stimulation ($t_{32} = 3.279$, $p = 0.003$, $d = 1.125$; Figure 5).

Topographical distributions of cross-coherence (Figure 4B, 4E) and responses in phase-amplitude

curves (Figure 5B, 5E) suggest that auditory cortex underlies neural synchronization activity.

Reduced sustained neural response in older adults

Response time courses for sustained neural responses are displayed in Figure 6A. The mean sustained

response in the 2–3.5-s time window was analyzed using an rmANOVA. The sustained response was larger

for FMs ($t_{50} = 5.121$, $p = 4.89 \times 10^{-6}$, $d = 0.717$) and AMFMs ($t_{50} = 6.401$, $p = 5.19 \times 10^{-8}$, $d = 0.896$)

compared to AMs (effect of Modulation Type: $F_{2,98} = 27.616$, $p = 5.35 \times 10^{-9}$, $\varepsilon = 0.845$, $\omega^2 = 0.096$), but

did not differ between FM and AMFM ($t_{50} = 0.149$, $p = 0.882$, $d = 0.021$). The sustained response was also

larger for non-sinusoidal compared to sinusoidal sound stimulation (effect of Regularity: $F_{1,49} = 10.328$, $p = $

0.002, $\omega^2 = 0.012$). The Modulation Type × Regularity interaction was not significant ($F_{2,98} = 0.559$, $p = $

0.574, $\omega^2 < 0.001$).

Figure 6: Results for the sustained response. A: Time courses for sinusoidal and non-sinusoidal sound stimulation. **B:** Topographical distributions for mean sustained activity in the 2–3.5-s time window. **C:** Mean sustained activity in the 2–3.5-s time window. The asterisk indicates a significant difference between each condition and the random condition (p ≤ 0.05; FDR-corrected). Error bars reflect the standard error of the mean. Gray dots in panels C and F reflect data from individual participants.

Critically, the sustained response was larger for younger compared to older adults (effect of Age

Group: $F_{1,49} = 8.566$, $p = 0.005$, $\omega^2 = 0.07$), but only for FM ($t_{49} = 3.515$, $p = 9.57 \times 10^{-4}$, $d = 0.985$) and

AMFM stimulation ($t_{49} = 2.59$, $p = 0.013$, $d = 0.725$), and not for AM stimulation ($t_{49} = 1.532$, $p = 0.132$, d

$= 0.429$; Modulation Type × Age Group interaction: $F_{2,98} = 5.701$, $p = 0.005$, $\varepsilon = 0.845$, $\omega^2 = 0.018$; Figure

6). The difference between age groups was larger for non-sinusoidal compared to sinusoidal stimulation

(Regularity × Age Group interaction: $F_{1,49} = 10.050$, $p = 0.003$, $\omega^2 = 0.012$), although the effect of age group

was significant for both (sinusoidal: $t_{49} = 2.173$, $p = 0.035$, $d = 0.609$; non-sinusoidal: $t_{49} = 3.42$, $p = 0.001$,

$d = 0.958$). Moreover, the increase the sustained response for non-sinusoidal compared to sinusoidal

modulations was only significant for younger ($t_{25} = 4.507$, $p = 1.34 \times 10^{-4}$, $d = 0.884$), but not for older

adults (t_{24} = 0.031, p = 0.976, d = 0.006). The Modulation Type × Regularity × Age Group interaction was

not significant ($F_{2,98}$ = 2.267, p = 0.109, ω^2 = 0.003).

Discussion

The current study investigated whether neural synchronization and sustained responses to low-frequency

modulations in a sound's amplitude, frequency, or both differ between younger and older adults. The MEG

data demonstrate that neural synchronization was reduced for non-sinusoidal compared to sinusoidal

modulation, but both were stronger, across all modulation types, in older compared to younger adults. The

sustained neural response elicited by the frequency and combined amplitude-frequency modulations was

reduced for older compared to younger adults. The sustained response was also reduced for sinusoidal

compared to non-sinusoidal sound stimulation in younger, but not in older adults. The results indicate that

the auditory systems of older and younger adults differ in their response to amplitude, frequency, and

combined amplitude-frequency modulations in sounds.

Neural activity is sensitive to low-frequency modulations in amplitude, frequency, and both combined

Synchronization with low-frequency amplitude and frequency modulations has been reported previously

(Maiste and Picton, 1989; Boettcher et al., 2002; Picton et al., 2003; Goossens et al., 2016; Henry et al., 2017;

Irsik et al., 2021), although not all studies focus on this speech-relevant frequency region of ~4 Hz (Edwards

and Chang, 2013). Here, neural synchronization was investigated using cross-coherence and a phase-

amplitude analysis. Cross-coherence (Nolte et al., 2004) is an alternative metric to inter-trial phase coherence

and allows the investigation of neural synchronization to non-sinusoidal modulations (Keitel et al., 2016).

Frequency modulations and amplitude-frequency modulations led to larger synchronization than did

amplitude modulations (Figure 4). The phase-amplitude analysis (Irsik et al., 2021) appeared to be somewhat

more sensitive, showing a larger response for combined amplitude-frequency modulations compared to

amplitude or frequency modulation alone (Figure 5). In the current study amplitude and frequency

modulation were not perceptually matched in, for example, perceived fluctuation strength (Moore et al.,

1996). Hence, the frequency modulation may have been perceived as fluctuating more than the amplitude

modulation, potentially explaining the larger synchronization.

The sustained response was also sensitive to amplitude and frequency modulations. This is consistent

with previous work demonstrating a sustained response to low-frequency amplitude and frequency

modulations in sounds (Herrmann and Johnsrude, 2018; Herrmann et al., 2019), and more generally with an

increasing body of work showing that temporally regular patterns elicit a sustained neural response (Barascud

et al., 2016; Sohoglu and Chait, 2016; Teki et al., 2016; Southwell et al., 2017; Southwell and Chait, 2018;

Dheerendra et al., 2021; Herrmann et al., 2021). Similar to neural synchronization, the sustained response

was larger for frequency modulations and amplitude-frequency modulations compared to amplitude

modulations (Figure 6). This suggests that frequency modulations compared to amplitude modulations are

especially potent drivers of the sustained response and, as mentioned above, this may reflect perceived

fluctuation strength.

Sinusoidal vs non-sinusoidal temporal modulations differently affect neural activity

Quasi-rhythmic stimulation has been shown to drive synchronized activity (Stefanics et al., 2010; Keitel et

al., 2016), including the tracking of non-sinusoidal amplitude fluctuations in speech (Giraud and Poeppel,

2012; Obleser et al., 2012; Peelle and Davis, 2013; Peelle et al., 2013; Brodbeck and Simon, 2020). Consistent

with this, we observed that neural synchronization was significant for non-sinusoidal modulation in sound,

although it was reduced compared to that for sinusoidal modulation (Figure 5).

In contrast, the sustained neural response was increased for non-sinusoidal compared to sinusoidal

modulations in sounds, in younger adults at least. Some previous work has suggested that the sustained

response increases with the degree of temporal regularity in sounds (Gutschalk et al., 2002; Herrmann and

Johnsrude, 2018), and that the degree of predictability in regular repeating sounds may increase the sustained

response (Heilbron and Chait, 2018). However, we have previously shown that the repetition of a regular

auditory pattern can reduce the sustained response, possibly because novelty/interestingness is reduced

(Herrmann et al., 2021). The invariant sinusoidal modulation at 4 Hz may have become less interesting over

time, compared to the non-sinusoidal modulation which was more variable (although still somewhat regular).

Age-related enhancement of neural responses to sound onset

Sound onsets elicited larger responses in the older compared to younger group, consistent with a large body

of work (Ross and Tremblay, 2009; Sörös et al., 2009; Lister et al., 2011; Alain et al., 2014; Bidelman et al.,

2014; Parry et al., 2019; Irsik et al., 2021; Herrmann et al., 2022). A loss of inhibition and an increase in

excitation in auditory cortex resulting from peripheral hearing loss and aging processes are thought to

underlie hyperresponsivity to sound (Caspary et al., 1995; Caspary et al., 2008; Takesian et al., 2012; Knipper

et al., 2013; Auerbach et al., 2014; Ouellet and de Villers-Sidani, 2014; Zhao et al., 2016; Resnik and Polley,

2017; Salvi et al., 2017; Herrmann and Butler, 2021b). Our older group did have pure-tone average

audiograms typical of individuals with presbycusis, and the age-related enhancement in sound-onset
responses is consistent with reduced inhibition in auditory cortex resulting from presbycusic deafferentation.

Enhanced neural synchronization in older compared to younger adults

We observed enhanced neural synchronization in the older compared to the younger age group for both
sinusoidal and non-sinusoidal modulations in sounds (Figure 5). The age-related enhancement appears to
generalize across the different modulation types (AM, FM, AMFM), since we did not find related interactions
with age group. The age-related enhancement in synchronization with sound modulations is consistent with
the hyperresponsivity observed for sound-onset responses (Figure 3; see also Anderer et al., 1996; Sörös et
al., 2009; Alain et al., 2012), suggesting a loss of inhibition in the auditory system (Caspary et al., 2008; Zhao
et al., 2016; Herrmann and Butler, 2021a).

Previous studies investigating age-related differences in neural synchronization have mainly focused
on amplitude modulations in sounds, and have demonstrated enhanced synchronization in older compared
to younger listeners to narrow-band noises (Purcell et al., 2004; Goossens et al., 2016; Herrmann et al., 2019)
and speech stimuli (Presacco et al., 2016b, a; Decruy et al., 2020; Broderick et al., 2021), consistent with the
results reported here. Age differences in neural synchronization with frequency modulations in sounds have
been less investigated, and the studies that did focus on slow frequency modulations have led to
contradictory results (Boettcher et al., 2002; Henry et al., 2017). Our investigation demonstrates enhanced
synchronization in older adults compared to younger adults for AM, FM, and AMFM sounds and that the
enhancement is present regardless of the shape – sinusoidal and non-sinusoidal – of the modulation.

Given our data, it is interesting to ask whether the age-related enhancement in synchronization with
FMs of narrow-band noise sounds would generalize to more naturalistic FMs, such as the low-frequency
prosodic contour in speech or musical melodies. The current frequency modulation spanned ~1.5 octaves
(here: 1760 Hz), which is a much wider frequency range than the range observed in natural prosody or even
musical melodies. Some previous work suggests that smaller frequency ranges may still lead to larger
synchronization with sinusoidal frequency modulations in older adults (Boettcher et al., 2002). Investigating
possible generalization of age-related differences in neural synchronization to these more naturalistic stimuli
will be an important next research step.

Reduced sustained neural response in older compared to younger adults

In the current study, frequency modulations and amplitude-frequency modulations in sounds resulted in a
smaller sustained response in the older compared to the younger group, but was not observed for AM-only

stimuli. The age-related reduction in the sustained response to FMs and AMFMs was observed for both
sinusoidal and non-sinusoidal modulations.

Age-related reductions in the sustained response have been observed previously for short sine tones
(Pfefferbaum et al., 1979) and for regular patterns in sequences of short tone pips (Al Jaja et al., 2020;
Herrmann et al., 2022). A previous study using AM sounds (without FM), observed a marginally significant
reduction in the sustained response for older compared to younger adults (Herrmann et al., 2019). This
weak, marginal group effect is consistent with the absence of an age-group effect in the current study (Figure
6). AM-only stimuli elicited smaller sustained responses (Figure 6) and less synchronization (Figure 5)
compared to FM and AMFM sounds. Age-group differences to AM-only sounds may be obscured because
of the lower response magnitudes (i.e., floor effect).

The difference between age groups in the magnitude of the sustained response was larger for non-
sinusoidal than sinusoidal modulations. Since the magnitude of the sustained response did not differ between
non-sinusoidal and sinusoidal modulations in older adults, this interaction is likely due to an enhanced
sustained response for non-sinusoidal relative to sinusoidal modulations in younger adults. The apparent
lack of sensitivity in older adults to the difference between sinusoidal and non-sinusoidal modulation further
highlights how auditory-system sensitivity to low-frequency temporal structure in sounds is altered in older
adulthood.

The sustained response has been suggested to reflect prediction-related processes (Heilbron and
Chait, 2018), such that response magnitude increases with increasing predictability of the structure in sounds.
A reduced sustained response would then suggest a reduction in prediction-related processes in older adults.
However, in younger adults, the sustained response was bigger for less predictable (non-sinusoidal)
compared to more predictable (sinusoidal) temporal modulations (Figure 6). Moreover, the sustained
response decreases when the same regular pattern is presented several times, that is, for more predictable
(and less novel) patterns (Herrmann et al., 2021). An age-related reduction in the sustained response may
indicate that the auditory system of older adults treated the trial-by-trial repetition of temporal modulations
as less novel compared to the auditory system of younger adults.

Conclusions

We recorded magnetoencephalography to investigate age differences in neural responses to low-frequency
modulations in amplitude, frequency, or combined amplitude and frequency. We observed hyperresponsivity
to sound onset (independent of modulation types) in older compared to younger adults, consistent with a

loss of inhibition in the aged auditory system. Neural synchronization with amplitude, frequency, and
combined amplitude-frequency modulations were also enhanced for older compared to younger adults. This
enhancement generalized across sinusoidally and non-sinusoidally (less regular) modulation types. Despite
enhanced synchronization, the sustained neural response was reduced in older compared to younger adults
for frequency-modulated sounds, regardless of whether modulation was sinusoidal or non-sinusoidal. Our
results thus demonstrate that the aged auditory system exhibits both hyperresponsiveness and
hyporesponsiveness to sound features as reflected in different neural signatures. The data may help explain
difficulties that people in late middle age and at older ages experience when tracking speech in the presence
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Author contributions

591 BH conceptualized and designed the study, recorded data, analyzed the data, interpreted the results, and
592 wrote the manuscript. BM analyzed the data, interpreted the results, and edited the manuscript. ISJ
593 conceptualized and designed the study, interpreted the results, and edited the manuscript.

Declaration of conflicts of interest

595 None.

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