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Age-related differences in the neural network interactions underlying the predictability gain

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1 **Abstract**

2 Speech comprehension is often challenged by increased background noise, but can be
3 facilitated via the semantic context of a sentence. This predictability gain relies on an interplay
4 of language-specific semantic and domain-general brain regions. However, age-related
5 differences in the interactions within and between semantic and domain-general networks
6 remain poorly understood. Using functional neuroimaging, we investigated commonalities and
7 differences in network interactions enabling processing of degraded speech in healthy young
8 and old participants. Participants performed a sentence repetition task while listening to
9 sentences with high and low predictable endings and varying intelligibility. Stimulus intelligibility
10 was adjusted to individual hearing abilities. Older adults showed an undiminished behavioural
11 predictability gain. Likewise, both groups recruited a similar set of semantic and cingulo-
12 opercular brain regions. However, we observed age-related differences in effective
13 connectivity for high predictable speech of increasing intelligibility. Young adults exhibited
14 stronger connectivity between regions of the cingulo-opercular network and between left insula
15 and the posterior middle temporal gyrus. Moreover, these interactions were excitatory in young
16 adults but inhibitory in old adults. Finally, the degree of the inhibitory influence between cingulo-
17 opercular regions was predictive of the behavioural sensitivity towards changes in intelligibility
18 for high predictable sentences in older adults only. Our results demonstrate that the
19 predictability gain is relatively preserved in older adults when stimulus intelligibility is
20 individually adjusted. While young and old participants recruit similar brain regions, differences
21 manifest in underlying network interactions. Together, these results suggest that ageing affects
22 the network configuration rather than regional activity during successful speech
23 comprehension under challenging listening conditions.

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28 **Keywords:** speech in noise, semantic predictability, cingulo-opercular network, effective
29 connectivity, ageing

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1 Introduction

2 In everyday conversations, speech comprehension is often hampered by challenging listening
3 conditions, such as environmental noise. To maintain successful communication, listeners
4 exploit facilitating factors, such as semantic context. The semantic context of a sentence
5 provides information that can be used to predict upcoming words (“predictability gain”). This
6 predictability gain has its maximum impact on comprehension at intermediate levels of
7 intelligibility, that is, when the acoustic speech stream is neither completely intelligible nor
8 completely unintelligible (Hartwigsen, Golombek, & Obleser, 2015; Pichora-Fuller, Schneider,
9 & Daneman, 1995).

10 With respect to the neural correlates underlying the predictability gain, several
11 neuroimaging studies found consistent activation in left (and less consistently right) angular
12 gyrus (AG), left posterior middle temporal gyrus (pMTG) and left inferior frontal gyrus (IFG;
13 Adank, 2012; Golestani, Hervais-Adelman, Obleser, & Scott, 2013; Obleser & Kotz, 2010;
14 Obleser, Wise, Dresner, & Scott, 2007; Rysop, Schmitt, Obleser, & Hartwigsen, 2021). These
15 regions have been suggested to constitute key regions of semantic processing (Binder, Desai,
16 Graves, & Conant, 2009; Jefferies, 2013; Seghier, 2013). In addition, some studies also
17 showed increased connectivity of left angular gyrus with other nodes of the semantic network
18 when listening to high versus low predictable sentences at intermediate levels of intelligibility
19 (Golestani et al., 2013; Obleser & Kotz, 2010; Obleser et al., 2007). Other work demonstrated
20 the functional relevance of left angular gyrus for the predictability gain with inhibitory
21 neurostimulation (Hartwigsen et al., 2015). Together, these studies provide evidence for a key
22 role of semantic regions, particularly left angular gyrus, in the successful integration of
23 semantic information into sentential context under challenging listening conditions in healthy
24 young adults.

25 Aside from the contribution of semantic regions, sentence comprehension under
26 challenging listening conditions results in increased executive demands (Fitzhugh, Schaefer,
27 Baxter, & Rogalsky, 2021), leading to a recruitment of domain-general networks. Specifically,
28 the cingulo-opercular network (Dosenbach, Fair, Cohen, Schlaggar, & Petersen, 2008; Uddin,
29 Yeo, & Spreng, 2019), encompassing bilateral anterior insulae as well as a portion of the dorsal
30 anterior cingulate and pre-supplementary motor area (pre-SMA), is the most frequently
31 reported network in degraded speech processing (Alain, Du, Bernstein, Barten, & Banai, 2018;
32 Erb & Obleser, 2013; Rogers & Peelle, 2022; Vaden Jr., Kuchinsky, Ahlstrom, Dubno, & Eckert,
33 2015; Vaden Jr. et al., 2013). Prior studies investigated the contribution of cingulo-opercular
34 regions to challenging listening conditions either by using manipulations of the auditory signal
35 itself (e.g., noise vocoding, where spectral information of an acoustic signal is degraded while

1 temporal information is preserved, see Shannon, Zeng, Kamath, Wygonski, & Ekelid, 1995) or
2 by masking the acoustic signal with broadband noise or multi-talker babble (Darwin, 2008;
3 Miller, 1947). Although increased activity in cingulo-opercular regions has been linked to error
4 monitoring, executive functions and maintenance of auditory attention (Wilsch, Henry,
5 Herrmann, Maess, & Obleser, 2015), its exact role in acoustically-demanding speech
6 perception remains vaguely defined. In one study, increased activity in cingulo-opercular
7 regions during challenging listening conditions was predictive of successful word recognition
8 in the following trial (Vaden et al. 2013), thereby indicating that activity in these regions is
9 directly relevant to behaviour. In line with this observation, Rysop et al. (2021) investigated
10 context-dependent changes in effective connectivity between nodes of the cingulo-opercular
11 network during adverse listening conditions. They found a stronger inhibitory influence on the
12 connections within the cingulo-opercular network, when semantic context was highly predictive
13 and intelligibility increased. Crucially, the degree of inhibitory influence at intermediate levels
14 of intelligibility was associated with an increased behavioural predictability gain. This finding
15 suggests that inhibition of the cingulo-opercular network is beneficial in young adults, as long
16 as semantic cues can be efficiently used to support comprehension of the distorted signal.
17 Otherwise, increased activity in cingulo-opercular regions might help to maintain task
18 performance. Yet, the contribution of the cingulo-opercular network to challenging listening
19 conditions in the ageing brain is unclear.

20 Ageing is associated with a decline across sensory and cognitive domains such as hearing
21 acuity, processing speed or working memory capacity (Salthouse, Atkinson, & Berish, 2003;
22 Wingfield, Amichetti, & Lash, 2015). However, semantic abilities are thought to remain
23 relatively stable across the lifespan (Verhaeghen, 2003; Wingfield & Stine-Morrow, 2000).
24 While listening to degraded speech is already demanding for young listeners, older adults
25 frequently report enhanced difficulties in understanding speech in challenging listening
26 situations, even if they show no signs of age-related hearing loss or difficulties in ideal listening
27 conditions (Gordon-Salant, 2005; Gordon-Salant & Fitzgibbons, 1995; Humes, 1996; Pichora-
28 Fuller, Alain, & Schneider, 2017). With regard to the predictability gain, behavioural studies
29 point towards either an increased reliance (Desjardins & Doherty, 2013; Goy, Pelletier, Coletta,
30 & Kathleen Pichora-Fuller, 2013; Pichora-Fuller et al., 1995; Sheldon, Pichora-Fuller, &
31 Schneider, 2008) or similar use of contextual information in older compared to younger adults
32 (Dubno, Ahlstrom, & Horwitz, 2000; Sheldon et al., 2008). For example, in a semantic priming
33 paradigm with an acoustically degraded prime, older participants benefitted from contextual
34 information as much as young participants (Sheldon et al. 2008).

35 While the beneficial effect of semantic context generally seems to be preserved in older
36 adulthood, changes at the neural level remain debated. Some authors argue that the reduced

1 magnitude of electrophysiological measures to high predictable sentences in old versus young
2 adults might reflect less efficient use of sentential context (Federmeier, 2007; Payne &
3 Federmeier, 2018; Wlotko, Federmeier, & Kutas, 2012). Likewise, a recent meta-analysis
4 showed less activation in semantic key regions for older participants but increased recruitment
5 of domain-general regions during semantic tasks (Hoffman & Morcom, 2018). Yet, these
6 changes were mainly observed when task performance was poor. Thus, it remains unclear
7 whether increased recruitment of domain-general networks in the ageing brain reflects a
8 (compensatory) attempt to maintain performance during speech comprehension under
9 challenging listening conditions (Eckert et al., 2008; Peelle, Troiani, Grossman, & Wingfield,
10 2011) or rather a pattern of dedifferentiation, indicating less functional specialization (Li,
11 Lindenberger, & Sikström, 2001). Some authors suggest that older adults already recruit such
12 domain-general resources under relatively easy conditions (Erb & Obleser, 2013; Fitzhugh,
13 Braden, Sabbagh, Rogalsky, & Baxter, 2019; Peelle, 2018; Reuter-Lorenz & Cappell, 2008).
14 Accordingly, stronger activity in domain-general regions was associated with better
15 comprehension of degraded auditory sentences selectively in old adults (Erb & Obleser, 2013).
16 These findings suggest that the (potentially compensatory) upregulation of activity in domain-
17 general areas is more pronounced in older listeners.

18 The additional upregulation of domain-general regions is accompanied by altered patterns
19 of functional connectivity (Sala-Llonch, Bartres-Faz, & Junque, 2015). It has been consistently
20 reported that older adults exhibit decreased levels of connectivity within domain-general
21 networks including the cingulo-opercular network (Geerligs, Renken, Saliasi, Maurits, & Lorist,
22 2015), while connectivity between networks was increased at rest (Setton et al., 2022;
23 Zonneveld et al., 2019) and during cognitive tasks (Zhang, Gertel, Cosgrove, & Diaz, 2021).
24 However, studies addressing age-differences in task-related connectivity during language
25 processing are sparse. Hence, it remains unclear how ageing affects functional interactions
26 within and between networks during speech comprehension under adverse listening
27 conditions.

28 In the present study, we assessed age-related differences in the behavioural predictability
29 gain and its neural correlates and network configuration while participants listened to
30 sentences of varying acoustic intelligibility and semantic predictability. In contrast to previous
31 studies, we used an optimized adaptive tracking paradigm that considered individual
32 differences in auditory perception by adjusting stimulus intelligibility to individual hearing
33 abilities. This assures comparable task performance across participants, which is particularly
34 important in light of variable hearing performance in old participants. Using dynamic causal
35 modelling (DCM) we analysed age-related differences in connectivity within the semantic and

1 cingulo-opercular networks, respectively (i.e. within-network connectivity) as well as between
2 nodes of both networks (i.e. between-network connectivity).

3 We expected a beneficial effect of semantic context when speech is acoustically degraded
4 in both age groups, as intelligibility levels were adjusted to individual hearing abilities. At the
5 neural level, we expected potentially weaker engagement of semantic regions and relatively
6 stronger recruitment of cingulo-opercular regions in older listeners. Age-related differences
7 should also be reflected in decreased connectivity within the cingulo-opercular and semantic
8 network and increased connectivity between the semantic and cingulo-opercular network in
9 older listeners, reflecting increased recruitment of domain-general resources.

10 To foreshadow our results, when controlling for speech-to-noise reception thresholds as
11 done in the present study, no age-related differences in the behavioural performance or task-
12 related activity were observed. However, relative to young listeners, older listeners showed
13 decreased effective connectivity within the cingulo-opercular network and between left anterior
14 insula and posterior middle temporal gyrus when predictability was high and intelligibility
15 increased. Stronger inhibitory influence between cingulo-opercular regions was associated
16 with better task performance selectively in older listeners, demonstrating its behavioural
17 relevance.

18

19

1 **Materials and Methods**

2 **Participants**

3 Thirty healthy middle-aged to old German native speakers were recruited for the fMRI
4 experiment. The sample size was determined based on our previous study (Rysop et al.,
5 2021). Inclusion criteria comprised age-normal hearing (tested with pure-tone audiometry; see
6 below for further details) and right-handedness according to the German version of the
7 Edinburgh Handedness Inventory (Oldfield, 1971). Participants were excluded if they showed
8 signs of cognitive impairment (Mini Mental State Examination score < 27; MMSE; Folstein,
9 Folstein, & McHugh, 1975) or reported a history of neurological or psychiatric disorders. Data
10 from three participants were excluded from the analyses due to excessive head movement
11 during fMRI (movement parameter exceeded 1.5 times the voxel size), and data from one
12 participant because of a ceiling effect in the behavioural performance.

13 The final sample size was $n = 26$ ($M = 62$ years, range: 50–77 years, 19 female, see
14 Supplementary Table 1 for further information). This group was compared to 26 healthy young
15 participants ($M = 25$ years, range: 19–29 years, 15 female), who had performed the same in-
16 scanner task under the exact same procedures in an earlier experiment (see Rysop et al.,
17 2021 for a detailed analysis of the neural predictability gain in this younger group). Participants
18 were reimbursed with 10 € per hour and gave written informed consent in accordance to the
19 declaration of Helsinki prior to participation. The study was approved by the local ethics
20 committee (University of Leipzig).

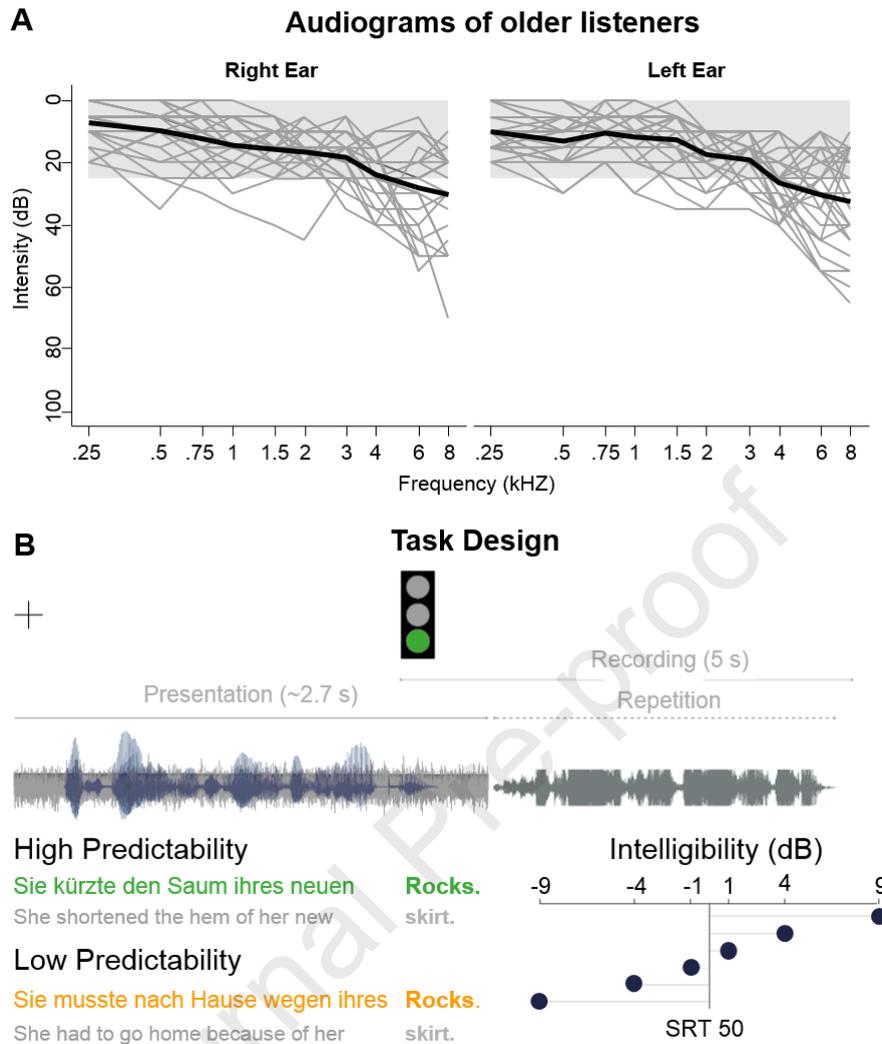
21 **Experimental Procedure**

22 The experiment consisted of two sessions. In the first session, each participant's hearing status
23 and cognitive abilities were tested using pure-tone audiometry and MMSE. Only participants
24 with relatively well-preserved peripheral hearing abilities (pure-tone average < 25 dB HL in the
25 listener's better ear) were included. The hearing test was conducted in a soundproof chamber
26 using an audiometer (Oscilla SM910-B Screening Audiometer). Pure-tone averages were
27 computed across frequencies from 250 kHz to 8000 kHz in the right and left ear separately
28 (see Figure 1A for an illustration of the peripheral hearing status of included participants). Six
29 participants had a PTA > 25 dB in one ear, but met the criterion of PTA < 25 dB in the other
30 ear and were therefore included. For included participants, the first session proceeded with
31 further neuropsychological assessments. Working memory was tested using the auditory
32 forward and backward version of the Digit Span Task (Wechsler, 1955). Measures of executive
33 functions (processing speed and cognitive flexibility) were assessed with both versions of the
34 Trail Making Test (Reitan, 1958). Further, we tested verbal fluency and flexibility using two
35 subtests of a German word fluency test (Regensburger Wortflüssigkeitstest; Aschenbrenner,

1 Tucha, & Lange, 2000). In this test, participants had to generate as many nouns as possible
2 either for one specific semantic category (subtest A, category *food*) or two alternating semantic
3 categories (subtest B, categories *fruits* and *sports*) within two minutes. The first session had a
4 duration of approximately one hour.

5 The fMRI experiment was conducted in a second session on a separate day. Participants
6 were placed in the scanner and equipped with MR compatible in-ear headphones (MR Confon;
7 Magdeburg, Germany). The volume was initially set to a comfortable intensity. To determine
8 the participants' in-scanner speech reception threshold (SRT; i.e., the signal-to-noise ratio
9 (SNR) required to correctly repeat a sentence with a probability of 50 %), we conducted an
10 adaptive up-down staircase experiment (Kollmeier, Gilkey, & Sieben, 1988; for a detailed
11 description of the procedure see Rysop et al., 2021). The SRT was used as a reference to
12 adjust stimulus intelligibility for each participant individually in the main experiment, thereby
13 correcting for differences in peripheral hearing and ensuring comparable task performance
14 across participants.

15 In the main experiment, participants listened to sentences while viewing a white fixation
16 cross on a black screen. Participants were instructed to overtly repeat the whole sentence as
17 accurately as possible to prevent listening strategies such as focussing on the last word. If they
18 had not understood anything, they were asked to say so. Responses were cued by a green
19 traffic light that appeared on the screen (see Figure 1B for an overview of the experimental
20 design). Overt responses were recorded for a period of 5 seconds using a FOMRI-III
21 microphone (Optoacoustics, Yehuda, Israel). The intertrial interval varied between 2000 and
22 7000 ms and served as an implicit baseline for the fMRI analysis. Stimuli were presented in
23 six blocks using Presentation (version 18.0, Neurobehavioral Systems, Berkeley, USA,
24 www.neurobs.com). The fMRI experiment had a duration of 50 minutes, the total duration of
25 the second session was approximately 90 minutes.



1

2 **Figure 1. Experimental procedures A.** Pure-tone audiograms for the older participants, plotted for the right and
 3 left ear separately. Grey lines represent the hearing curve of single participants across each tested frequency; the
 4 bold black line represents the average across participants. The intensity range for normal hearing is shaded in grey.
 5 **B.** In the sentence repetition task, each trial consisted of a listening period in which a short sentence (blue waveform)
 6 was embedded in speech-shaped noise (grey waveform) and a fixation cross was presented on the screen. With
 7 the onset of the sentence-final word (i.e., keyword), a green traffic light appeared and marked the recording period,
 8 in which participants were asked to repeat the whole sentence as accurately as possible. The sentence context
 9 was either highly (green) or lowly (orange) predictive of the sentence-final keyword. The signal-to-noise ratio of the
 10 sentences against the speech-shaped background noise was adjusted to individual speech reception thresholds
 11 (SRT 50).

12 Stimulus Material

13 Experimental sentences were taken from the German Speech Intelligibility in Noise (G-SPIN)
 14 material (Kalikow, Stevens, & Elliott, 1977; for a detailed description of the German version,
 15 see Erb, Henry, Eisner, & Obleser, 2012). The 216 sentences consisted of pairs with the same
 16 sentence-final words (i.e., keyword) but different preceding sentence frames: While the frame
 17 of one of the sentences was predictive of the keyword (high predictability: “*She shortened the*
 18 *hem of her new skirt.*”), the other one was not (low predictability: “*She had to go home because*

1 *of her skirt*). Twenty additional G-SPIN sentences with highly predictable keywords were used
2 for the adaptive tracking procedure.

3 The intelligibility of the sentences was manipulated by varying the sentence intensity
4 relative to constant speech-shaped noise in the background. Sentence intensity was varied
5 symmetrically and relative to the previously determined SRT in six steps (-9 dB, -4 dB, -1 dB,
6 +1 dB, +4 dB, +9 dB SNR relative to the individual SRT; Rysop & Schmitt et al. 2021 for
7 details). These intelligibility levels were chosen to cover the full range of behavioural
8 performance in each participant while sampling those (intermediate) intelligibility levels with
9 the largest predictability gain more densely.

10 Experimental conditions yielded a factorial design with the factors *predictability* (high,
11 low) and *intelligibility* (-9 dB, -4 dB, -1 dB, +1 dB, +4 dB, +9 dB SNR relative to SRT), with 18
12 sentences per condition. Sentences were presented in a pseudorandomized order with the
13 restriction that each intelligibility level was not presented more than three times in a row to
14 prevent adaptation.

15 **MRI acquisition**

16 Functional MRI data were collected on a 3 Tesla Siemens Prisma Scanner with a 32-channel
17 head coil. We used a dual gradient-echo planar imaging multiband sequence (Feinberg et al.,
18 2010). The scanning parameters were: TR = 2,000 ms; TE = 12 ms, 33 ms; flip angle = 90°;
19 voxel size = 2.5 x 2.5 x 2.5 mm with an interslice gap of 0.25 mm; FOV = 204 mm; multiband
20 acceleration factor = 2. For each participant, 1,500 volumes à 60 slices were acquired in axial
21 direction and interleaved order. Slices were tilted by 10° off the AC-PC line to increase
22 coverage of anterior temporal lobe regions. Field maps were acquired for later distortion
23 correction (TR = 620 ms; TE = 4 ms, 6.46 ms). Additionally, high-resolution T1-weighted
24 images were either obtained from the in-house database if available or were acquired in the
25 second session using an MPRAGE sequence (whole brain coverage, TR = 1300 ms, TE =
26 2.98 ms, voxel size = 1 x 1 x 1 mm, matrix size = 256 x 240 mm, flip angle = 9°).

27 **Data Analysis**

28 **Behavioural Analyses**

29 All participants' spoken response recordings were cleaned from scanner noise using Audacity
30 (version 2.2.2, <https://www.audacityteam.org/>) and transcribed by one rater, who additionally
31 determined speech onset times and speech durations. Repetitions of the keywords were rated
32 either as correct or incorrect before calculating the proportion of correctly repeated keywords
33 per participant and condition. Thereafter, psychometric curves were fitted to keyword repetition
34 accuracies across intelligibility levels, separately for sentences with low and high predictability.

1 More specifically, we employed cumulative Gaussian sigmoid functions using the Psignifit
2 toolbox (Fründ, Haenel, & Wichmann, 2011) in MATLAB (version R2018b, MathWorks; see
3 Supplementary Material for further details).

4 All empirical deviances between the fitted and a saturated model fell within the 97.5%
5 confidence interval of their respective reference distribution, indicating that psychometric
6 curves properly represented the observed behavioural data in every participant. There was no
7 significant difference between the deviance of fits for sentences with low vs. high predictability
8 ($t_{24} = 0.11$, $p = .916$, $r = 0.06$, $BF_{10} = 0.21$). To illustrate goodness of fit, we additionally
9 calculated the R^2_{KL} , which is based on the information-theoretic measure of Kullback-Leibler
10 divergence and represents the reduction of uncertainty by the fitted model relative to a constant
11 model (Cameron & Windmeijer, 1997). On average, psychometric curves of sentences with
12 high predictability yielded an R^2_{KL} of 0.96 (SD = 0.04, range: 0.86–0.995), sentences with low
13 predictability an R^2_{KL} of 0.94 (SD = 0.05, range: 0.81–0.996).

14 To quantify the predictability gain, we compared threshold and slope parameters of the
15 psychometric functions for sentences with low versus high predictability. The threshold
16 parameter represents the intelligibility level at which participants correctly repeat half of the
17 keywords. A smaller threshold parameter for sentences with high compared to low
18 predictability would indicate a stronger benefit from semantic context already at levels of lower
19 intelligibility. The slope parameter describes the steepness of the curve at a proportion correct
20 of 0.5 and denotes the sensitivity to changes in intelligibility at intermediate levels. Parameter
21 estimates of the two psychometric curves were analysed using an ANOVA with the factors *age*
22 and *predictability*. Furthermore, we analysed speech onset times (SOTs) as a measure of
23 response speed. SOTs were log-transformed and submitted to a three-way ANOVA with the
24 factors *age*, *predictability* and *intelligibility*. Statistical analyses were performed with RStudio
25 (version 4.0.2; R Core Team, 2021) and JASP (version 0.9.1; <https://jasp-stats.org/>;
26 Wagenmakers et al., 2018).

27 **Functional MRI analyses**

28 Functional MRI data were preprocessed and analysed in SPM12 (version 7219, Wellcome
29 Department of Imaging Neuroscience, London, UK) and MATLAB (version R2018b). As a first
30 step, the images of the two echo times were combined based on a weighted average of the
31 voxel-wise temporal signal-to-noise ratio maps of the first 30 volumes and realigned using a
32 custom Matlab script. The rationale behind this approach is to improve signal quality in brain
33 regions that typically suffer from signal loss (e.g., anterior temporal lobes; see Halai,
34 Welbourne, Embleton, & Parkes, 2014 for a similar approach). Further, fMRI data were
35 distortion corrected, co-registered to individual high-resolution structural T1-scans and

1 spatially normalized to the standard template by the Montreal Neurological Institute (MNI). We
2 preserved the original voxel size and finally smoothed the data, using a 5 mm³ FWHM
3 Gaussian kernel to allow statistical inference based on Gaussian random-field theory (Friston,
4 Worsley, Frackowiak, Mazziotta, & Evans, 1994).

5 Preprocessed data were submitted to a general linear model (GLM). Sentence
6 presentation times were modelled with stick functions and convolved with SPM's canonical
7 hemodynamic response function. We included one regressor of no interest capturing the verbal
8 responses using participant- and trial-specific speech onset times and durations. As additional
9 nuisance regressors we included the six motion parameters obtained from the rigid-body
10 transformation of the realignment step (see Supplementary Figure 1) and a vector for each
11 volume that exceeded a framewise displacement of 0.9 mm (Siegel et al., 2014). A high-pass
12 filter with a cut-off at 128 s was applied to the data.

13 At the single-participant level of both age groups, we computed direct contrasts between
14 experimental conditions. These contrasts encompassed the main effect of intelligibility (i.e.,
15 linear increase in intelligibility, $-9 < -4 < -1 < +1 < +4 < +9$ dB SNR), the main effect of
16 predictability (high predictable > low predictable sentences and vice versa) and the interaction
17 between predictability and intelligibility (high predictable > low predictable sentences, with a
18 linear increase in intelligibility and vice versa). The single-participant interaction contrasts were
19 further used to extract the timeseries for the effective connectivity analysis.

20 The resulting contrast maps were submitted to one-sample t-tests, to investigate effects
21 **within** each age group. Two-sample t-tests were used to investigate differences **between**
22 groups (Poldrack, Nichols, & Mumford, 2011). To characterize commonly activated regions by
23 both groups, the respective contrast maps were submitted to a conjunction analysis based on
24 the minimum statistic (Nichols, Brett, Andersson, Wager, & Poline, 2005). To correct for
25 multiple comparisons, we used the family-wise error (FWE) technique applying a cluster-level
26 threshold of $p < 0.05$ and a voxel-wise threshold of $p < 0.001$). Statistical maps were visualized
27 with BrainNet Viewer (Xia, Wang, & He, 2013). Anatomical locations were specified using the
28 SPM anatomy toolbox (version 3.0; Eickhoff et al., 2005, 2007) and the Harvard-Oxford atlas
29 (<https://fsl.fmrib.ox.ac.uk/fsl/fslwiki/>).

30 **Dynamic Causal Modelling (DCM)**

31 We used Dynamic Causal Modelling (DCM; version 12.5; Friston, Harrison, & Penny, 2003)
32 and Parametric Empirical Bayes (PEB; Zeidman, Jafarian, Corbin, et al., 2019; Zeidman,
33 Jafarian, Seghier, et al., 2019) to investigate age-related differences in effective connectivity.
34 DCM is a method that uses a biophysically informed generative model to estimate effective
35 connectivity between a set of brain regions. The estimated signal is compared to the measured

1 fMRI BOLD signal and quantified in terms of negative variational free energy. Three sets of
2 parameters have to be defined: (1) intrinsic parameters, reflecting endogenous connectivity
3 between regions in the absence of experimental manipulation (A-Parameters), (2) modulatory
4 parameters that encode the rates of change in connectivity due to an experimental
5 manipulation (B-Parameters) and (3) driving input parameters, reflecting the influence of an
6 experimental task on single nodes within the network (C-Parameters).

7 While the classical DCM approach allows to investigate the network architecture (or
8 configuration) underlying an experimental effect, PEB spans a hierarchical model over the
9 connectivity parameters, allowing to quantify differences and commonalities in coupling
10 strengths among groups of participants (Friston et al., 2016; Zeidman, Jafarian, Corbin, et al.,
11 2019). PEB provides a Bayesian GLM-like approach that divides inter-subject variability into
12 regressor effects and unexplained random effects (Zeidman, Jafarian, Seghier, et al., 2019).
13 In contrast to classical inference over parameter estimates, the PEB framework does not only
14 take the mean but also the uncertainty of individual parameters into account. Consequently,
15 participants with more uncertain parameters will influence the group estimates less than
16 participants with more certain parameters (Zeidman, Jafarian, Seghier, et al., 2019).

17 We were interested in the context-dependent effective connectivity *within* the semantic
18 and cingulo-opercular networks as well as *between* nodes of both networks. We approached
19 these questions in a two-step procedure. First, we performed a standard DCM analysis within
20 the cingulo-opercular network, motivated by the findings of our previous study, to test the
21 goodness of fit of the winning model that was identified in young participants (Rysop et al.,
22 2021). In that study, the data of young participants was best explained by a model in which
23 intelligibility and predictability of the sentences jointly modulated the connections from pre-
24 SMA and right anterior insula to the left anterior insula. This model had a relatively high
25 exceedance probability of 0.73. Here, we aimed at testing whether the optimal model for young
26 participants also succeeds in explaining the older participants' data (see Supplementary
27 Methods for a detailed description of the standard DCM approach).

28 Second, we employed the DCM PEB approach which enabled us to test for age-related
29 effects in a larger model space encompassing nodes of the semantic and cingulo-opercular
30 network identified by the conjunction analysis. Specifically, we asked how predictability and
31 intelligibility of sentences affects the coupling (1) within the semantic and cingulo-opercular
32 network, (2) between both networks, as well as (3) whether coupling strengths differ across
33 groups and (4) whether age-related differences in coupling strength are related to behavioural
34 differences.

35 **Seed Region Selection**

1 For the standard DCM approach, we defined the seed regions of the cingulo-opercular network
2 based on the interaction contrast in the older subgroup. The seed regions comprised the pre-
3 SMA [-4 20 42] and anterior insulae [left: -30 23 -2, right: 33 26 -2]. For the PEB DCM approach,
4 we defined the seed regions based on the results of the conjunction analyses and on our
5 previous findings in young participants (Rysop et al., 2021). Seed regions comprised bilateral
6 AG [left: -42 -70 35; right: 46 -64 35] and left pMTG [-57 -60 -2] of the semantic network, as
7 well as pre-SMA [-7 13 48] and anterior insulae [left: -33 23 -2, right: 33 23 0] of the cingulo-
8 opercular network. The first principal component (eigenvariate) was extracted from all voxels
9 within a spherical region of interest (6 mm radius) that passed a liberal threshold of $p_{\text{uncorrected}} < .05$
10 for each seed region. The sphere was initially centred on the group coordinate and was
11 allowed to move within a spherical search space of 10 mm around the group coordinate to
12 account for variation in the exact participant-specific maximum. The search space for the
13 bilateral insulae was constrained by a 5 mm sphere centred on the group coordinate to ensure
14 that the individual peak fell within the anterior insula and not the surrounding regions. The
15 extracted time courses were adjusted by an F -contrast spanning across the experimental
16 conditions to regress out effects of no interest (i.e., overt speech or head movement).

17 **PEB DCM Setup**

18 The participant-level design matrix for the PEB DCM analysis consisted of three regressors.
19 The first regressor coded the onsets of all experimental stimuli and served as driving input
20 regressor. The ensuing regressors served as modulatory input and coded the effect of 1) high
21 and 2) low predictability with the level of intelligibility added as parametric modulation (i.e., the
22 effect of predictability, scaled linearly with intelligibility).

23 At the first-level, we specified and inverted one “full” DCM for each participant (see Figure
24 6A for a schematic full model). This model included all possible reciprocal connections
25 between seed regions as well as self-connections. Because we did not include a primary
26 sensory region in the model, we set the driving input to every seed region (C-Matrix). Each
27 connection, except for the self-connections, could receive modulation by experimental inputs
28 (B-Matrix). The inputs were not mean-centered, so that the intrinsic parameter estimates (A-
29 matrix) can be interpreted as baseline connectivity in the absence of a task (Dijkstra, Zeidman,
30 Ondobaka, Van Gerven, & Friston, 2017; Marreiros, Kiebel, & Friston, 2008). Note that
31 parameter estimates represent rates of change of the influence from one region to another
32 (intrinsic connectivity) or of an experimental manipulation on the connection between regions
33 (modulatory connectivity). Positive parameters translate to a positive influence of one region
34 onto another region or an experimental manipulation on a connection and can be interpreted
35 as excitatory. Negative parameter estimates can be interpreted as inhibitory influence (Dijkstra
36 et al., 2017; Zeidman, Jafarian, Corbin, et al., 2019; Zeidman, Jafarian, Seghier, et al., 2019).

1 All full models were inverted using variational Laplace. We extracted the percent variance
2 explained to obtain a parameter for the model fit.

3 At the second-level, we set up the PEB model of between-participant effects to investigate
4 commonalities and differences in intrinsic and modulatory parameters. The PEB model
5 included a group-level design matrix with one regressor encoding the mean across both groups
6 (or commonalities across participants) whereas the second regressor encoded age group, to
7 identify differences in connectivity between younger and older participants. In contrast to
8 traditional DCM, this procedure does not compare the model evidence of different network
9 architectures, but the effect of the presence or absence of an experimental manipulation on a
10 connection. Thus, each connection, opposed to the whole network architecture, receives a
11 posterior probability. To only keep those parameters with a high probability of contributing to
12 the model evidence, we performed an automatic search over reduced models using Bayesian
13 Model Comparison and Bayesian Model Reduction (BMR; Friston et al., 2016). In this
14 procedure, parameters are systematically switched on and off (i.e., nested models) and
15 compared against the full model, pruning away those parameters that do not contribute to the
16 model evidence (negative variational free energy).

17 Lastly, we computed a Bayesian Model Average (BMA) over the parameters of the final
18 iteration of the automatic search. Here, parameters were weighted by their posterior
19 probabilities (Zeidman, Jafarian, Seghier, et al., 2019). To retain those parameters with 'strong'
20 evidence of being non-zero (Kass & Raftery, 1995), we applied a threshold at a posterior
21 probability of > 95 % based on free energy. Results were considered as within-network
22 connectivity when task-dependent modulations were found between nodes of the semantic or
23 the cingulo-opercular network, respectively. In case the task modulated the connection
24 between one node of the semantic and one node of the cingulo-opercular network, this result
25 was interpreted as between-network connectivity. Further, we probed whether the effect sizes
26 were large enough to predict a left-out participant's age group using leave-one-out cross-
27 validation on the estimated parameters that had a significant age-related difference. Finally, to
28 relate differences in connectivity to the behavioural performance, we extracted participant-
29 specific parameters from the connections that showed a significant group effect and correlated
30 these with the respective behavioural parameters of the psychometric curves.

31

1 Results

2 Behavioural Results

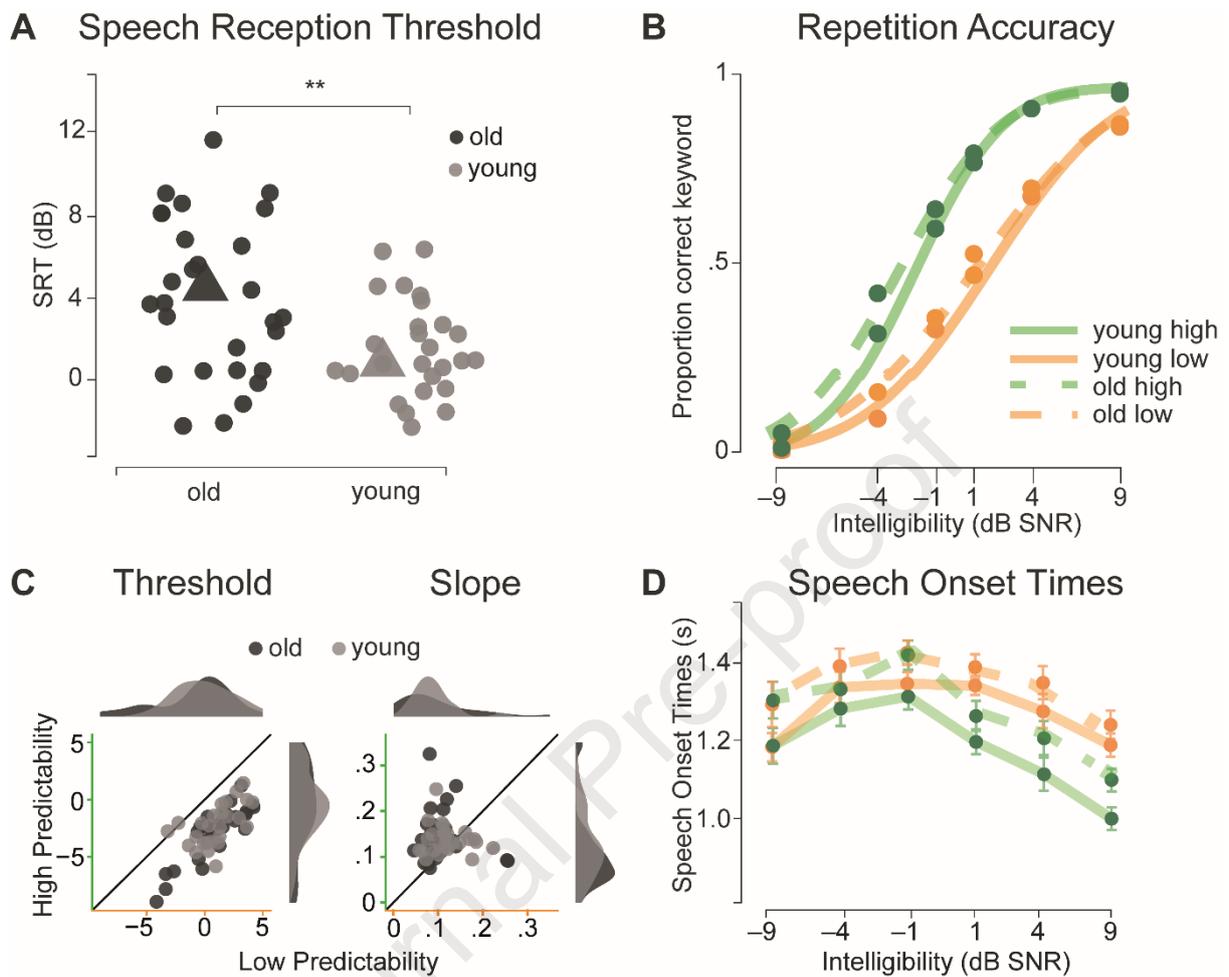
3 Semantic context benefits comprehension under challenging listening conditions 4 independent of age

5 The average speech reception threshold obtained during the adaptive tracking procedure was
6 significantly higher for older (3.75 dB) than for younger participants (1.68 dB; $t(41.12) = 2.36$,
7 $p = 0.023$; Figure 2A). This finding is in line with a previous report that SRT's are approximately
8 2 dB higher in older adults (Schneider, Pichora-Fuller, & Daneman, 2010).

9 On average, keyword repetition accuracies were comparable between young ($M = 52.33$
10 %, $SD = 8.09$ %) and old participants ($M = 54.59$ %, $SD = 10.56$ %). To explore effects of age
11 on speech comprehension, we fitted psychometric curves to keyword repetition accuracies
12 across intelligibility levels separately for both age groups and predictability levels (Figure 2B).
13 We found a significant main effect of predictability for the threshold parameter of psychometric
14 curves ($F(1,50) = 302.3$, $p < 0.001$, $\eta^2 = 0.41$; Figure 2C), indicating that high predictability
15 benefits comprehension already at lower levels of intelligibility. Further, a significant main effect
16 of predictability on the slope parameters ($F(1,50) = 7.46$, $p = 0.009$, $\eta^2 = 0.08$; Figure 2C)
17 indicates that an increase in intelligibility benefits comprehension more strongly when speech
18 is highly predictable. Together these results demonstrate that speech comprehension at
19 intermediate levels of intelligibility generally benefits from high predictability. Most central to
20 the aims of the present study, psychometric curves were strikingly similar between both age
21 groups. Accordingly, the main effect of *age* and the interaction of *age* and *predictability* was
22 not significant for any parameter of the psychometric curves (Supplementary Table 2 for all
23 results).

24 The analysis of log-transformed speech onset times revealed a significant main effect of
25 predictability ($F(1,50) = 163.51$, $p < 0.001$, $\eta = 0.038$), a significant main effect of intelligibility
26 ($F(5,250) = 36$, $p < 0.001$, $\eta = 0.08$) and a significant interaction between predictability and
27 intelligibility ($F(5,250) = 23.29$, $p < 0.001$, $\eta = 0.02$). Post-hoc pairwise comparisons showed
28 that highly predictable sentences were repeated significantly faster than lowly predictable
29 sentences at the three highest intelligibility levels (+1 dB SNR: $t(101.54) = -2.997$, $p_{\text{corr}} = 0.003$;
30 +4 dB SNR: $t(99.3) = -3.57$, $p_{\text{corr}} < 0.001$; +9 dB SNR: $t(101.62) = -3.63$, $p_{\text{corr}} < 0.001$). There
31 was no main effect of age ($F(1,50) = 0.92$, $p = 0.34$, $\eta = 0.014$) and no significant interaction
32 of age with intelligibility or predictability (Figure 2D).

1



2

3 **Figure 2. Behavioural results.** **A** Speech reception thresholds (SRT) for old and young participants. Single dots
 4 represent individual participants; triangles represent the mean SRT per age group $** = p < 0.01$. **B** Psychometric
 5 curves fitted to the proportion of correctly repeated keywords (dots) across intelligibility levels for sentences of high
 6 (green) and low (orange) predictability, separately for each age group (solid lines = young, dotted lines = old
 7 participants). **C** Individual threshold (left) and slope (right) parameters extracted from the psychometric curves at
 8 50 % correct for high (y-axis) and lowly (x-axis) predictable keywords. Dark grey = old participants, light grey =
 9 young participants. **D**. Mean speech onset times for highly (green) and lowly (orange) predictable sentences across
 10 intelligibility levels.

11

12 **fMRI Results – Task-related activity**

13 **Intelligibility effects in bilateral superior temporal cortex are independent of age**

14 First, we were interested in task-related activity sensitive to increasingly intelligible speech,
 15 irrespective of its predictability. A conjunction analysis revealed increased activity in bilateral
 16 primary auditory regions in the superior temporal gyri for increasingly intelligible speech
 17 comprehension independent of age (Supplementary Figure 2A and Supplementary Table 3).
 18 The direct comparison between age groups revealed increased activation in left and right
 19 precentral gyrus as well as left and right cerebellum in older participants. Contrary, older
 20 participants exhibited less activation in medial frontal regions than young listeners.

1 **Semantic regions are commonly recruited when speech is highly predictable**

2 Next, we explored brain regions that support the processing of semantic information,
3 irrespective of intelligibility. Here we identified that activation common to both age groups
4 during highly predictable speech was located in left AG, pMTG and precuneus/paracingulate
5 gyrus (Supplementary Figure 2B and Supplementary Table 4). These regions are considered
6 key regions (pMTG and AG) or extended regions (precuneus) of the semantic system in young
7 and older adults (Binder et al., 2009; Hoffman & Morcom, 2018). Direct comparisons between
8 age groups revealed increased activation in a large cluster in the precuneus in young
9 compared to older participants during the processing of highly predictable speech. No other
10 differences were found. The reversed contrast revealed that the pre-SMA was the only region
11 commonly recruited by both groups for low predictable sentences irrespective of intelligibility
12 (Supplementary Figure 2C).

13 **The recruitment of semantic versus cingulo-opercular regions depends on the** 14 **predictability of speech when intelligibility is increasing but not on age**

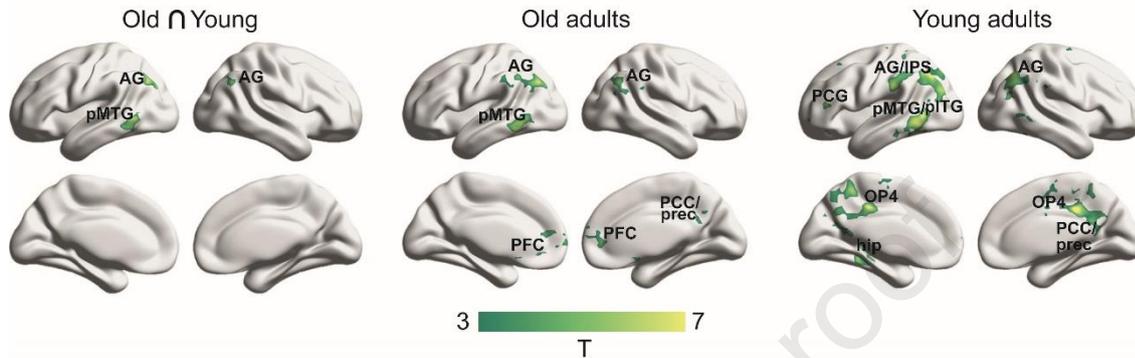
15 Finally, we were interested in brain regions that are differentially affected by predictability when
16 intelligibility is varied parametrically (i.e., the interaction between intelligibility and
17 predictability). To this end, we modelled intelligibility as linear parametric modulation of high
18 and low predictable sentences. In both groups, we found higher activation for high versus low
19 predictable sentences under increasing intelligibility in several large left- and right-hemispheric
20 clusters (Figure 3). Overall, the activation pattern was broadly similar in both age groups, while
21 cluster sizes were considerably larger in young adults (Figure 3). The conjunction analysis of
22 the interaction contrast revealed that left angular gyrus (PGa & PGp), right angular gyrus (PGp)
23 as well as left posterior middle temporal gyrus were recruited by both age groups (Figure 3A,
24 left, Supplementary Table 5). Note that, opposed to the main effect of predictability, the
25 homologous right angular gyrus seems to play a role in both age groups when sentence
26 intelligibility is varied. Young adults showed increased activity in a cluster comprising the
27 postcentral gyrus in the right hemisphere. No other significant differences between age groups
28 were found.

29 The opposite direction of the interaction contrast targeted brain regions sensitive to
30 sentences of low predictability under increasing intelligibility. There were no significant
31 differences between age groups. Both groups showed significant activity in the pre-SMA and
32 bilateral anterior insulae, extending into the frontal operculum (Figure 3B, Supplementary
33 Table 6). These regions are associated with the domain-general cingulo-opercular network
34 (Dosenbach et al., 2008). Note that in contrast to the main effect of predictability reported
35 above, older adults here indeed showed increased activity in the anterior insular regions. The

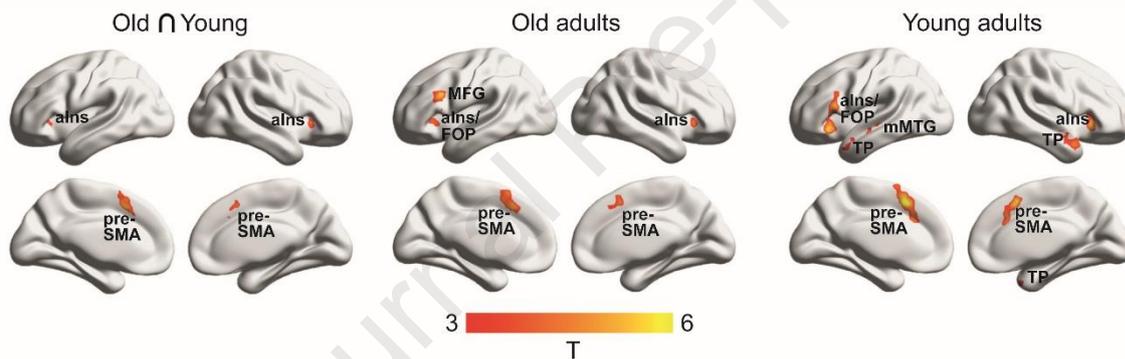
1 clusters in pre-SMA and bilateral insulae were identified as common activation across both
 2 age groups via a conjunction analysis and were used as seed regions for the effective
 3 connectivity analysis, together with left pMTG and bilateral AG.

Interaction Predictability x Intelligibility

A High > low predictable, increasing intelligibility



B Low > high predictable, increasing intelligibility



4
 5 **Figure 3. fMRI results showing the interaction effect of intelligibility and predictability.** **A** Brain regions that
 6 show significant effects for the interaction contrast (high x increasing intelligibility > low x increasing intelligibility).
 7 The left column shows regions commonly activated by both groups. Activation maps of the interaction contrast are
 8 displayed separately for old (middle) and young participants (right) for visualization. **B** Brain regions that show
 9 significant effects in the opposite direction of the interaction (low x increasing intelligibility > high x increasing
 10 intelligibility). All activation maps are thresholded at cluster-level $p_{FWE} < 0.05$, with a voxel-wise threshold of $p <$
 11 0.001 , AG = angular gyrus, pMTG/pITG = posterior middle/inferior temporal gyrus, PFC = prefrontal cortex,
 12 PCC/prec = posterior cingulate cortex/precuneus, PCG = precentral gyrus, IPS = inferior parietal sulcus, OP4 =
 13 parietal operculum, hip = hippocampus, prec = precuneus, FOP = frontal operculum, alns = anterior insula, pre-
 14 SMA = pre-supplementary motor area, TP = temporal pole, mMTG = mid-middle temporal gyrus.

15 fMRI Results – Effective connectivity

16 At the network level, we investigated commonalities and age-related differences in effective
 17 connectivity driven by the interactive effect of predictability and intelligibility within the semantic
 18 and cingulo-opercular network and between both networks. We followed a dual-step procedure
 19 comparing (1) age-related differences in the network architecture of the cingulo-opercular
 20 network based on previous findings (Rysop et al. 2021) and (2) age-related differences in a
 21 larger network configuration containing seed regions based on the exact coordinates from the
 22 univariate conjunction analyses reported above (Figure 4A).

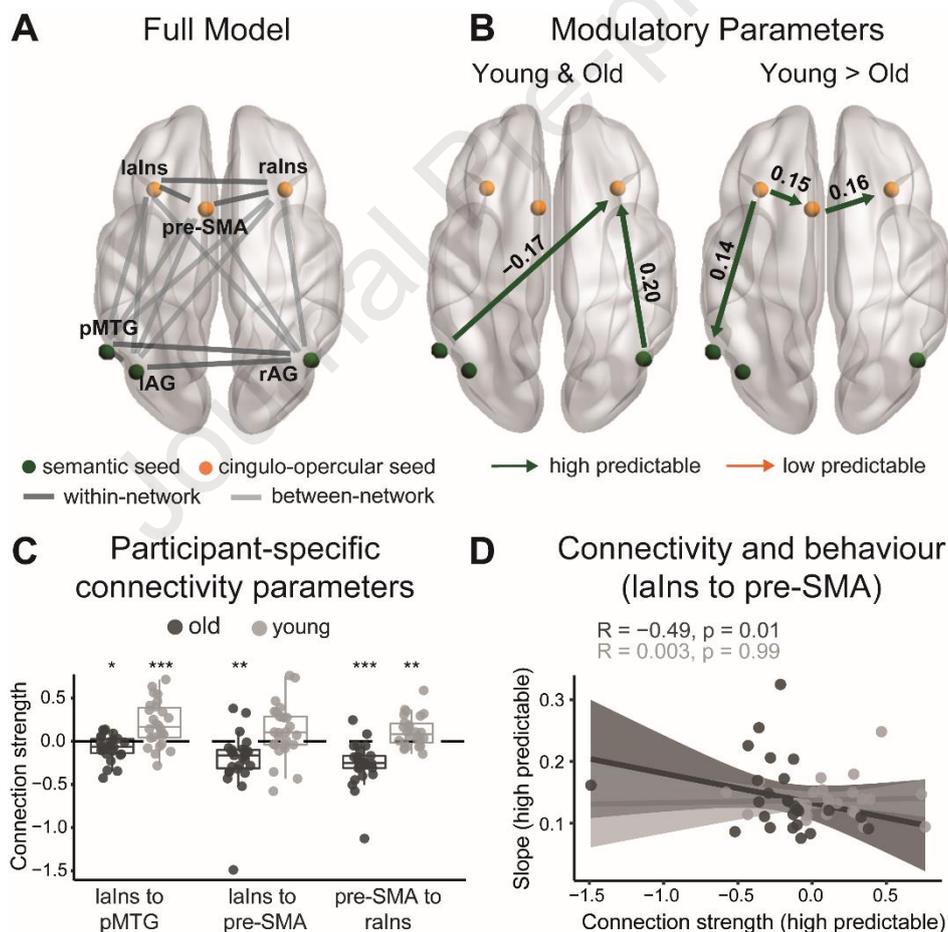
1 First, we asked whether the cingulo-opercular model that was identified as the best
2 model explaining the data of young participants would also succeed in explaining the data of
3 older participants. Overall, the exceedance probabilities for all families of models included in
4 the DCM analysis were remarkably low in the older subgroup, with the highest exceedance
5 probability of 0.35 (opposed to 0.73 in young participants). No clear winning family of models
6 could be identified. The winning model that was identified in young participants corresponded
7 to the second probable model in the older group, with an exceedance probability of 0.24 (see
8 Supplementary Figure 3) and an average of 14.36 % explained variance (opposed to 22.13 %
9 in young participants), thus indicating that the winning model of young participants must not
10 necessarily serve as an optimal model for older participants as well. Independent-samples t-
11 tests conducted on the parameter estimates that were extracted from the modulated
12 connections of that model (i.e., winning model in young participants) indicated no significant
13 age-related differences (see Supplementary Figure 3).

14 **High predictability modulates connectivity from semantic regions to the right anterior** 15 **insula in both age groups**

16 As a next step, we conducted a PEB DCM analysis that is more powerful in the analysis of
17 age-related differences in effective connectivity as it allows a larger network space
18 encompassing seed regions from both networks of interest across both age groups. The full
19 model yielded an average of 22.78 % explained variance across both groups (22.75 % for old
20 participants, 22.82 % for young participants). First, we examined which connections were
21 modulated by either high or low predictable sentences under increasing intelligibility in both
22 groups (e.g., commonalities across participants). Here we found that high predictability had an
23 inhibitory influence on the connection from left pMTG to right anterior insula (Figure 4B, left;
24 see also Table 1 and Supplementary Figure 5 for individual parameters of common modulatory
25 effects). There was no intrinsic connection between these regions, thus these regions form a
26 functional interaction during the task, but do not interact with each other during rest (see
27 Supplementary Tables 7 and 8 and Supplementary Figure 4 for intrinsic connections).
28 Moreover, high predictability had an excitatory influence on the connection from right AG to
29 right anterior insula, further strengthening the positive intrinsic connection between these
30 regions. Low predictable sentences did not have a significant effect on any of the connections,
31 aside from weak evidence for the inhibitory influence from pMTG to right anterior insula
32 (posterior probability = 0.64).

33 **Young adults exhibit stronger connectivity within cingulo-opercular regions and** 34 **between left insula and pMTG when sentences are highly predictable and increasingly** 35 **intelligible**

1 Next, we looked at connections that showed age-related differences in the modulatory effect
 2 of high and low predictable sentences. Age-related differences were found in the modulatory
 3 influence of high predictability on three connections: from left anterior insula to pMTG, from left
 4 anterior insula to pre-SMA and from pre-SMA to right anterior insula (Figure 4B, right; see also
 5 Table 2). Coupling between these connections was stronger for younger adults. Testing
 6 connectivity values for each group against zero, revealed that the modulatory effect was
 7 negative (i.e., inhibitory) in older adults for all three connections ($p < 0.05$; see Figure 4C for
 8 individual parameters of connections that showed age-related differences and Supplementary
 9 Figure 5 for individual parameters of connections with a similar modulatory effects). In contrast,
 10 all modulatory parameters of young participants were positive (i.e., excitatory), except for the
 11 non-significant modulation of the connectivity from left anterior insula to pre-SMA ($p = 0.06$).
 12 None of the connections had a stronger coupling for older compared to younger adults. Low
 13 predictable sentences did not have a significant modulatory influence on any connection.



14

15 **Figure 4. DCM PEB results.** **A** Illustration of the full DCM model; semantic seeds are coloured in green, cingulo-
 16 opercular seeds are coloured in orange, within-network connections are shown in dark grey, between-network
 17 connections are shown in light grey; lalns/ralns = left/right anterior insula, pMTG = posterior middle temporal gyrus,
 18 l/rAG = left/right angular gyrus, pre-SMA = pre-supplementary motor area. **B** Group results of the Bayesian model
 19 average (BMA) of the modulatory parameters (B-parameters), thresholded at a posterior probability of > 0.95 .
 20 Modulatory effects of high (green arrow) and low (orange arrow) predictable sentences with parametrically
 21 modulated intelligibility. Left: Modulatory effects of the intelligibility x predictability interaction that are common for
 22 both groups. Right: Age-related differences (young > old adults) for the modulatory effects of high predictable
 23 sentences. **C** Individual modulatory parameters for high predictable sentences at each connection showing an age-

1 related difference. Asterisks indicate connections that significantly differ from zero, * = $p < 0.05$, ** = $p < 0.01$, *** =
 2 $p < 0.001$. **D** Negative correlation between connectivity (modulation from left anterior insula to pre-SMA) and the
 3 slope parameter from the psychometric curve for high predictable sentences, shaded area represents 95 %
 4 confidence intervals.

5

6 To test the predictive validity of the age-related connectivity differences, we performed
 7 a leave-one-out cross-validation on the three connectivity parameters that differed between
 8 age groups. The out-of-samples Pearson correlation between the predicted and actual group
 9 membership was significant for all three connections (left anterior insula → pMTG: $r = 0.29$, p
 10 = 0.028; left anterior insula → pre-SMA: $r = 0.28$, $p = 0.023$; pre-SMA → right anterior insula: r
 11 = 0.34, $p = 0.006$). Thus, the effect sizes were large enough to predict a left-out participant's
 12 age group based on their connectivity.

13

14 **Table 1.** Modulatory parameter estimates from the Bayesian model average (BMA) reflecting
 15 commonalities in connectivity across groups.

Commonalities across age groups				
Connection	Modulatory connectivity			
	High * SNR	Pp	Low * SNR	Pp
<i>Between Network Connections</i>				
rAG → ralns	0.20	1	0.00	0
pMTG → ralns	-0.17	1	-0.09	0.64

16 Note: Only connections with at least one parameter exceeding a posterior probability (Pp) > .95 are shown and
 17 highlighted in bold. See Supplementary Table 7 for commonalities in intrinsic connectivity and self-connections.

18

19 **Table 2.** Modulatory parameter estimates from the Bayesian model average (BMA) reflecting
 20 differences in connectivity between groups.

Differences between age groups (young > old)				
Connection	Modulatory connectivity			
	High * SNR	Pp	Low * SNR	Pp
<i>Within Cingulo-Opercular Network</i>				
lalns → pre-SMA	0.15	1	0.00	0.00
pre-SMA → ralns	0.16	1	0.00	0.00
<i>Between Network Connections</i>				
lalns → pMTG	0.14	1	0.00	0.00

21 Note: Only connections with at least one parameter exceeding a posterior probability (Pp) > .95 are shown and
 22 highlighted in bold. For the differences between age groups positive values indicate stronger connectivity for
 23 young adults. See Supplementary Table 8 for differences in intrinsic connectivity.

24

1 Finally, we were interested whether age differences in connectivity strength were associated
2 with behavioural performance. To this end, we correlated connectivity parameters that showed
3 significant group effects with the respective parameters from the psychometric curves,
4 representing the behavioural predictability gain (threshold and slope parameter). We found a
5 significant negative relationship between the slope parameter of high predictable sentences
6 (representing the sensitivity to changes in intelligibility) and the coupling strength for the
7 connection from left anterior insula to the pre-SMA (i.e., *within* the cingulo-opercular network)
8 in old participants ($\rho = -0.49$, $p = 0.012$; Figure 4D). Those older participants with a more
9 negative connectivity were more sensitive to changes in intelligibility of highly predictable
10 sentences at the behavioural level. Put differently, increased inhibition between these two
11 domain-general regions was associated with higher accuracy at the behavioural level in old
12 participants. This connectivity-behaviour association was not significant in young participants
13 ($\rho = 0.046$, $p = 0.83$). No other connectivity-behaviour relationships were significant.

14 Together, these results demonstrate that connectivity within the semantic and the
15 cingulo-opercular network, as well as between both networks, was mainly modulated by high
16 predictable sentences. Younger participants exhibited stronger coupling between subregions
17 of the cingulo-opercular network and between left anterior insula and pMTG when speech was
18 increasingly intelligible and highly predictive compared to their older counterparts. Decreased
19 connectivity within the cingulo-opercular network was associated with increased task
20 performance selectively in older adults.

21

22

1 **Discussion**

2 In this study, we investigated commonalities and differences during speech comprehension
3 under challenging listening conditions in young and middle-aged to old participants. First, we
4 found that both age groups showed a comparable behavioural predictability gain, reflected in
5 overlapping psychometric curves. Second, depending on the semantic context of increasingly
6 intelligible speech, both groups recruited distributed left-hemispheric semantic and cingulo-
7 opercular brain regions. Third, young and old listeners differed in effective connectivity during
8 highly predictable and intelligible speech: Young adults exhibited overall stronger connectivity
9 between regions of the cingulo-opercular network and between left anterior insula and pMTG.
10 Moreover, these interactions were excitatory in young adults but inhibitory in old adults. Finally,
11 the degree of the inhibitory influence from left anterior insula to pre-SMA was predictive of the
12 behavioural sensitivity towards changes in intelligibility for high predictable sentences in older
13 adults only. Our results demonstrate that the predictability gain is relatively preserved in
14 middle-aged to older adults with good peripheral hearing when stimulus intelligibility is adjusted
15 to individual hearing abilities. Accordingly, age-related differences at the neural level are subtle
16 and manifest in effective connectivity, where inhibition between cingulo-opercular regions is
17 associated with better comprehension when speech is highly predictive.

18

19 **Old listeners benefit from semantic context as much as young listeners**

20 Our experiment was designed to assure comparable task difficulty for young and old listeners,
21 as older listeners are known to have increased difficulties with hearing speech in challenging
22 listening conditions even if they have relatively normal audiograms (Pichora-Fuller et al., 2017).
23 Although we only included older participants with relatively good peripheral hearing as
24 measured by an audiogram, they had significantly higher speech reception thresholds than
25 young listeners. This apparent mismatch converges with the notion that hearing acuity in quiet
26 (as measured by pure-tone audiometry) is not a good predictor of hearing abilities in noise
27 (Schneider et al., 2010). Moreover, this stresses the importance of taking individual differences
28 in hearing abilities into account when conducting cross-sectional experiments in the auditory
29 domain, for instance by carefully adjusting stimulus material.

30 Indeed, when controlling for differences in peripheral hearing abilities, both age groups
31 showed a comparable benefit from semantic context, suggesting that younger and older
32 listeners did not differ in their ability to exploit contextual cues. In earlier studies, older adults
33 showed a similar or even slightly enhanced benefit from semantic context (e.g., Pichora-Fuller,
34 2008; Pichora-Fuller et al., 1995). This was explained by a stronger recruitment of top-down
35 contextual information to compensate for the age-related decline in bottom-up sensory

1 information in older adults. If this was the case, we would have expected a relative “overuse”
2 of semantic cues when auditory degradation was adjusted to individual hearing acuity,
3 manifesting in a larger predictability gain in older adults. However, we did not observe age-
4 related differences at the behavioural level. In general, our finding that older adults show a
5 comparable beneficial effect from contextual cues is consistent with previous findings that
6 reported a comparable predictability gain for older listeners across a wide range of intelligibility
7 manipulations (e.g., Dubno et al., 2000; Goy et al., 2013; Humes, 1996; Sheldon et al., 2008).

8

9 **Age does not affect which brain regions are recruited in speech comprehension**

10 At the level of task-related brain activity, both age groups showed speech-sensitive
11 engagement of left and right auditory regions, encompassing Heschl’s gyrus and the superior
12 temporal cortex, as well as the right parietal operculum. The common recruitment of bilateral
13 primary auditory regions is well in line with the literature, where hypoactivation of the primary
14 auditory cortex was observed in listeners with age-related hearing loss (Peelle et al., 2011) or
15 listeners with normal hearing but worse task performance (Wong et al., 2009). With respect to
16 age-related differences, older adults exhibited less activity in frontal regions, including right
17 temporal pole and dorsal anterior cingulate cortex, which may reflect unspecific
18 dedifferentiation effects related to ageing. In contrast, stronger activity in the bilateral
19 precentral gyrus and the cerebellum for older relative to younger participants may reflect a
20 different listening strategy such as inner rehearsal in older participants (Guediche, Holt,
21 Laurent, Lim, & Fiez, 2015). Indeed, a stronger contribution of the cerebellum in older adults
22 has been previously reported during picture naming and was linked to articulatory and
23 phonological processes (Chen & Desmond, 2005; Ferré, Jarret, Brambati, Bellec, & Joannette,
24 2020).

25 With respect to the interaction of predictability and intelligibility, older participants
26 showed broadly the same activation pattern as young adults, with generally smaller cluster
27 sizes. Common activity across age groups was mainly found in left angular gyrus and posterior
28 middle temporal gyrus regions associated with semantic (control) processes (Binder et al.,
29 2009; Jackson, 2021). This is further in line with a meta-analysis (Hoffman & Morcom, 2018)
30 reporting overall similar activation for young and old participants in semantic tasks. In that
31 meta-analysis, differences between age groups manifested as additional right-hemispheric
32 activity in older adults and were observed when task performance of older participants was
33 poor. We found no evidence for such additional domain-general activity in older adults, most
34 likely because we assured that the difficulty of the sentence repetition task was comparable
35 across participants. On the contrary, young listeners showed stronger activity in an additional

1 brain region, namely the precuneus, when contrasting high and low predictable sentences
2 irrespective of intelligibility. Increased activity in the precuneus/posterior cingulate is often
3 observed in semantic tasks and may serve as an interface between semantic and memory
4 systems (Binder et al., 2009). The exact role of the precuneus/posterior cingulate complex in
5 semantic cognition is not clear, as these regions are associated with numerous cognitive
6 functions (see Jung & Lambon Ralph, 2021; Smallwood et al., 2021). Age-related differences
7 in this region might reflect a consequence of neural dedifferentiation, manifesting as a change
8 in the relative balance of this region.

9 Aside from the expected association of left angular gyrus activity with the predictability
10 gain, we also found increased activity in right angular gyrus. The effective connectivity between
11 right angular gyrus and right anterior insula was strengthened in young and old listeners when
12 speech was highly predictable and increasingly intelligible. These findings point towards a
13 potential demand-related role of right angular gyrus in semantic processing. There is mounting
14 evidence that both left and right angular gyri are involved in (combinatorial) semantic
15 processes (Binder et al., 2009; Golestani et al., 2013; Graessner, Zaccarella, & Hartwigsen,
16 2021; Graves, Binder, Desai, Conant, & Seidenberg, 2010; Price, Bonner, Peelle, &
17 Grossman, 2015). During speech comprehension under challenging listening conditions, right
18 angular gyrus showed increased activity for related words at higher intelligibility in young
19 participants, but at a lower threshold than left angular gyrus (Golestani et al., 2013). This
20 supports the notion that right angular gyrus might indeed be specifically recruited during
21 semantic processing when task demands increase.

22 When semantic context was low, both age groups showed increased activity in regions
23 of the cingulo-opercular network for more intelligible speech, which was expected based on
24 the literature (Peelle, 2018; Vaden Jr. et al., 2015, 2013) and likely helps to maintain stable
25 performance under increased task demands (Peelle, 2018). However, contrary to our
26 hypothesis, we did not observe a stronger recruitment of cingulo-opercular regions in older
27 adults, which has been reported in previous work (Reuter-Lorenz & Cappell, 2008), and was
28 linked to successful performance in older listeners (Erb & Obleser, 2013). The difference
29 between the previous and present work is likely explained by our adaptive tracking paradigm
30 which calibrated noise levels to the individual hearing abilities. Our results imply that previously
31 reported differences between age groups may have resulted from increased task demands for
32 older participants due to differences in peripheral hearing.

33

34 **Age affects connectivity within and between networks recruited in degraded speech**
35 **comprehension**

1 Finally, at the neural network level, we found common interaction patterns across groups but
2 also age-related differences in effective connectivity. In both groups, task-related changes in
3 connectivity were observed selectively for high predictable sentences (with linearly increasing
4 intelligibility), but not for low predictable sentences. Independent of age, high predictable
5 sentences increased the inhibitory influence of left pMTG on the right anterior insula, as well
6 as the excitatory influence of right AG on the right anterior insula. This finding points towards
7 a complex interaction pattern of semantic-specific and domain-general areas during speech
8 comprehension under challenging listening conditions even when task-relevant semantic
9 information is available.

10 Although we did not find age-related differences at the level of activation, effective
11 connectivity patterns within the cingulo-opercular network (left anterior insula to pre-SMA, pre-
12 SMA to right anterior insula), and between cingulo-opercular and semantic regions (left anterior
13 insula to left pMTG) differed as a function of age. Connectivity between these areas was overall
14 stronger for young adults. Notably, the direction of the influence differed between both groups,
15 with excitatory connectivity in young and inhibitory connectivity in old listeners. This pattern of
16 overall reduced within-network connectivity in cognitive control networks in the ageing brain
17 converges with several previous findings, although it remains debated whether these changes
18 reflect processes of decline or compensation (Campbell, Grady, Ng, & Hasher, 2012; Madden
19 et al., 2010; O'Connell & Basak, 2018 but see Grady et al., 2010). In our study, the degree of
20 the inhibitory influence between cingulo-opercular regions was associated with a steeper slope
21 of the psychometric curve for high predictable speech in older adults only. This reflects higher
22 sensitivity towards intelligibility in highly predictable speech selectively in older adults and
23 suggests a potential beneficial effect of inhibition within the cingulo-opercular network when
24 semantic context can be used to predict upcoming speech. Further evidence supporting this
25 notion comes from a recent study demonstrating that increased connectivity within domain-
26 general networks was associated with less efficient behavioural performance in a semantic
27 task in older adults (Martin, Saur, & Hartwigsen, 2021). Importantly, the opposite pattern was
28 observed for young adults, which converges with our finding of increased connectivity within
29 the cingulo-opercular network in young adults. Together, the present results show that such
30 inhibition of the interactions in the domain-general cingulo-opercular network may constitute a
31 key mechanism underlying the predictability gain in the ageing brain.

32 It is noteworthy however, that younger listeners exhibited a pattern of inhibition within the
33 cingulo-opercular network in a previous investigation (Rysop et al., 2021). In the present study,
34 the same model appeared less suitable for older adults, as revealed by a low amount of
35 explained variance captured by the model alongside low exceedance probability. This
36 apparently conflicting finding demonstrates that effective connectivity results must be

1 interpreted with caution and with respect to the broader context of its setup. As the PEB DCM
2 approach yielded higher levels of explained variance, considers the (un)certainty of
3 participants' parameters and allows a larger model space enabling the investigation of within
4 and between network interactions, we believe that this approach is more suitable to investigate
5 age-related differences in effective connectivity.

6

7 **Limitations**

8 To our knowledge, this study represents the first investigation of age-related differences in the
9 effective connectivity underlying the predictability gain during speech processing under
10 challenging listening conditions. However, one drawback of the DCM methodology is that it
11 only allows to include a limited number of seed regions. Therefore, conclusions about
12 interactions with other regions and functional networks cannot be drawn based on our analysis.
13 Future studies may address how the presence of semantic context alters network
14 configurations between functional networks, such as the default mode, cingulo-opercular or
15 the fronto-parietal control network on a larger level, and how such configurations change with
16 age.

17

18 **Conclusions**

19 The present study shows that young and old adults recruit a similar set of semantic and
20 cingulo-opercular brain regions during speech comprehension under challenging listening
21 conditions when intelligibility was calibrated to the individual hearing thresholds. This highlights
22 the importance of carefully adjusting stimulus intensities to individual hearing levels in ageing
23 research. With such precautions, age-related differences in task-related recruitment are rather
24 subtle and manifest only at the network level. Older adults showed overall decreased
25 connectivity between semantic and cingulo-opercular regions as well as within the cingulo-
26 opercular network. Nevertheless, the degree of inhibition within the cingulo-opercular network
27 was positively related to speech comprehension in older adults only, demonstrating the
28 behavioural relevance of such interactions. These findings provide new insight into age-related
29 changes in the network configuration underlying the predictability gain.

30

1 **Author contributions**

2 **Anna Uta Rysop**: Investigation, Methodology, Data Curation, Formal analysis, Software,
3 Validation, Visualization, Writing – Original Draft, Writing – Review & Editing. **Lea-Maria**
4 **Schmitt**: Methodology, Software, Formal analysis, Writing – Review & Editing. **Jonas**
5 **Obleser**: Conceptualization, Funding Acquisition, Writing – Review & Editing, Resources,
6 Supervision. **Gesa Hartwigsen**: Conceptualization, Funding Acquisition, Writing – Review &
7 Editing, Resources, Supervision.

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17 **Transparency**

18 We report how we determined our sample size, all data exclusions, all inclusion/exclusion
19 criteria, whether inclusion/exclusion criteria were established prior to data analysis, all
20 manipulations, and all measures in the study. No parts of the study procedures and analyses
21 were pre-registered in a time-stamped, institutional registry prior to the research being
22 conducted.

23 Raw behavioural and neuroimaging data are protected under the General Data
24 Protection Regulation (EU). The conditions of our ethics approval do not permit public archiving
25 of the data. The data can be made available from the corresponding authors upon request and
26 on prior consultation with the ethics committee. The written version of the experimental stimuli
27 (German SPIN sentences) can be found in the appendix of Erb et al., 2012
28 (<https://www.sciencedirect.com/science/article/abs/pii/S0028393212002151?via%3Dihub>).
29 Auditory stimuli and custom scripts that were used for the adaptive tracking procedure and for
30 the analyses in addition to SPM and JASP are available at <https://osf.io/a8js7/>. Legal copyright
31 restrictions prevent public archiving of the Digit Span Task (Wechsler, 1955), Trail Making Test
32 (Reitan, 1958) and German word fluency test (Regensburger Wortflüssigkeitstest;

1 Aschenbrenner, Tucha, & Lange, 2000), which can be obtained commercially from the
2 copyright holders in the cited references.

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