Audiovisual Speech-In-Noise (SIN) Performance of Young Adults with ADHD

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ABSTRACT

Adolescents with Attention-deficit/hyperactivity disorder (ADHD) have difficulty processing speech with background noise due to reduced inhibitory control and working memory capacity (WMC). This paper presents a pilot study of an audiovisual Speech-In-Noise (SIN) task for young adults with ADHD compared to age-matched controls using eye-tracking measures. The audiovisual SIN task consists of varying six levels of background babble, accompanied by visual cues. A significant difference between ADHD and neurotypical (NT) groups was observed at 15 dB signal-to-noise ratio (SNR). These results contribute to the literature of young adults with ADHD.

KEYWORDS

ADHD, Eye-Tracking, Speech-In-Noise

1 INTRODUCTION

The recent estimated prevalence of diagnosed ADHD in children and adolescents has increased from 6.1% to 10.2% over the period of 1997 to 2016 in the U.S. [Xu et al. 2018]. Adolescents with ADHD have difficulty meeting time limits, controlling anger, inhibiting responses, and processing auditory information [Barkley 1997; Fields et al. 2017; Fostick 2017]. Processing speech in background noise requires fundamental language abilities, higher working memory, as well as a higher signal-to-noise ratio (SNR) [Schneider et al. 2007]. Since a person’s ability to process speech with background noise depends on that person’s auditory and cognitive system [Schneider et al. 2007], young adults with ADHD may experience difficulty processing auditory information in the presence of background noise due to reduced inhibitory control [Barkley 1997; Pazvantoglu et al. 2012; Woltering et al. 2013; Woods et al. 2002], and decreased working memory capacity (WMC) [Alderson et al. 2013; Banich et al. 2009; Michalek et al. 2014].

Unlike noise, which degrades listening conditions, the presence of external visual cues such as written, contextual information and facial movements, can enhance the processing of auditory information, especially when accompanied by noise [Fraser et al. 2010; Jääskeläinen 2010; Michalek et al. 2014; Mishra et al. 2013; Moradi et al. 2013; Rudner et al. 2009; Van Wassenhove et al. 2005; von Kriegstein et al. 2008]. At increased noise levels, semantically related visual cues have a positive impact on the perception of spoken sentences [Zekveld et al. 2011]. When increased noise is present during face-to-face conversation, adults tend to fixate more on the nose and mouth area of the speaker [Buchan et al. 2008], confirming that oral-motor movements of the speaker aide speech recognition [Bristow et al. 2008].

Neurotypical (NT) individuals are known to perceive audiovisual cues more accurately from the right visual field (RVF) than from the left visual field (LVF) [Kimura 1973]. Multiple studies on this [Carter et al. 1995; Heilman et al. 1991; Mitchell et al. 1990; Voeller and Heilman 1988] showed the presence of a lateralized deficit in the visual-spatial attention of ADHD subjects, which orient their attention to LVF targets.

Our work presents the performance of young adults with ADHD compared to age-matched controls using eye-tracking measures during an audiovisual SIN task. Our findings are consistent with the possibility that audiovisual cues, in general, are processed in such a way that WMC or cognitive load are not consistently impacted in increasing levels of background noise for NT adults [Michalek et al. 2018].

2 METHODOLOGY

2.1 Participants

Our pilot study consisted of five young adults (4 F, 1 M) with a prior diagnosis of ADHD, and six NT young adults (4 F, 2 M) as the control. All participants were aged between 18 - 30 years, with no history of psychotic symptoms and normal vision. Participants with a diagnosis of ADHD confirmed their diagnosis through medical documentation, including records from a physician or licensed psychiatrist. They were asked to remain medication-free for 12 hours prior to study participation. There were no participants who had been prescribed long lasting non-stimulants, so the 12-hour time frame was sufficient for all participants. Information on the risks of avoiding medication were provided prior to the experiment,
and participants acknowledged it by signing a consent form approved by University’s Institutional Review Board. Both ADHD and NT participants went through a hearing screening of 20 dB HL at frequencies 500 Hz, 1000 Hz, 2000 Hz, and 4000 Hz, bilaterally, to ensure their hearing was within normal limits.

2.2 Speech-in-Noise Task

We used QuickSIN [Killion et al. 2004] software to simultaneously present a sentence repetition task with background noise (i.e. speech babble) at six SNRs: 25 dB, 20 dB, 15 dB, 10 dB, 5 dB, and 0 dB. Each SNR represents the ratio of the dB level of speech to dB level of noise. The level of background noise increases as the SNR decreases. The audiovisual QuickSIN setup is presented in Figure 1a. Participants were asked to listen to the sentences while simultaneously viewing the speaker’s face and then repeat each sentence verbally. Participants were presented with nine sentence sets, each having six sentences representing all background noise levels. Each sentence had an average of 8-13 words including five keywords (e.g., The weight of the package was seen on the high scale). Participants were scored based on the number of keywords accurately repeated per sentence. The presentation of the nine sentence blocks was randomized and counterbalanced across participants.

2.3 Eye-tracking Setup

We used Tobii Pro X2-60 computer screen-based eye tracker (60 Hz, 0.4° accuracy) to record the eye movements of participants during the QuickSIN task. Prior to the experiment, each participant was calibrated using Tobii’s standard calibration methods. We used Tobii Studio analysis software to pre-process gaze metrics using the I-VT filter (velocity threshold set to 30°/second) to extract eye movement metrics recorded throughout the study.

We specified four areas of interest (AOIs): 1) left eye, 2) right eye, 3) nose, and 4) mouth of the eye-tracking stimulus to analyze the eye-movements of participants (see Figure 1b).

2.4 Analysis

To observe how the eye-tracking measurements change with audiovisual cues, we used our RAEMAP [Jayawardena 2020] eye movement processing pipeline, which is a modified version of gaze analytics pipeline [Duchowski 2017]. Upon correct mapping of variables, the original gaze analytics pipeline has the capability of extracting raw gaze data from various eye trackers [Duchowski 2017]. After extracting raw gaze data, the gaze analytics pipeline: (1) classify raw gaze points into fixations, and (2) aggregate fixations related information for statistical analysis. The gaze analytics pipeline facilitates computation of numerous eye movement metrics. Also, it has the capability of generating visualizations of gaze points, fixations within AOIs, heat maps, ambient/focal fixations, and microsaccades per scan path. The current implementation of the gaze analytics pipeline handles eye-tracking data recorded during each task of each person sequentially. This process is computationally expensive, where the split and merge approach generates large number of intermediate files along the way of eye gaze metrics calculations.

RAEMAP is developed such that calculations of eye gaze metrics utilize distributed computing resources as illustrated in Figure 2. RAEMAP facilitates computation of traditional positional gaze metrics such as fixation count and fixation duration, as well as advanced metrics such as gaze transition entropy [Krejtz et al. 2015], and complex pupilometry measurements such as index of pupillary activity (IPA) [Duchowski et al. 2018] which indicate cognitive load. RAEMAP also has the capability of generating visualizations of gaze points, AOIs, scan paths, and fixations on AOIs (see Figure 1). The architecture of RAEMAP is shown in Figure 2.

In RAEMAP, the calculations of eye gaze metrics of subjects are done in separate processes utilizing distributed computing resources as illustrated in Figure 2 since they are independent of one another to enhance the efficiency. The aggregation of calculated eye gaze metrics of all participants in each task is done using Message Passing Interface (MPI). In addition, RAEMAP have the stream processing capability to calculate eye gaze metrics and visualize the scan path as data is being streamed by the eye tracker.

We applied RAEMAP to calculate gaze points, AOIs, scan paths, and fixations on AOIs per each sentence of the QuickSIN task for each participant. Figure 1c shows a visualization of fixations of one participant while watching one sentence in the QuickSIN task. Figure 1d shows a visualization of fixations on pre-defined AOIs.

We generated gaze transition matrices and corresponding gaze transition entropies for both participant groups. We also calculated the IPA counts for participants in both groups.

3 RESULTS

We first report the performance of ADHD and NT participants during the QuickSIN task. Next we analyze changes in eye movements in relation to the six SNRs. A mixed, repeated measures ANOVA using a 2x6 design with main factors of group (ADHD or NT) and SNR (0 dB to 25 dB with 5dB increments) was carried out on the performance of QuickSIN task and the eye-tracking measures.

3.1 QuickSIN performance

We first analyze the performance of both ADHD and NT participants at each SNR. Each participant was assigned a score for every sentence, based on the number of keywords accurately repeated out of five. There was no main effect of the participant group for QuickSIN performance \( F(1,9) = 1.97, p > 0.05 \), indicating that performance was similar between ADHD and NT participants. There was a significant main effect of the SNR on QuickSIN performance, \( F(1,23,11.11) = 127.78, p < 0.001 \), indicating that performance was different among SNRs. There was no significant interaction effect between SNR and participant group, \( F(1,23,11.11) < 1, p > 0.05 \). To further evaluate the main effect of the SNR, we conducted a t-test for each SNR, identifying a significant difference of QuickSIN performance between the two groups at 15 dB SNR, \( p < 0.05 \).

The performance of NT participants was best at 15 dB SNR whereas, the performance of participants with ADHD was best at 20 dB SNR. In general, when the task’s difficulty level was easy (SNR>15 dB), both ADHD and NT participants performed well by recalling 4.7 keywords out of 5 on average per sentence. In contrast, when the task was difficult (0 dB SNR), both ADHD and NT participants did not perform well by recalling 2.3 keywords out of 5 on average per sentence. At 15 dB SNR, participants with ADHD recalled 4.7 keywords on average and NT participants recalled 4.9 keywords on average.
Figure 1: The Audiovisual QuickSIN Setup. (a) Speaker’s face as viewed by the participants during the audiovisual SIN task, (b) Four AOIs created for the eye-movement analysis: left eye, right eye, nose, and mouth, (c) Sample scan-path with fixations, and (d) fixations on the AOIs of a participant while listening to one sentence.

Figure 2: The architecture of RAEMAP which could process eye-tracking data as being streamed by an eye-tracker. RAEMAP API distributes tasks among the nodes using MPI. Each node hosts an instance of the RAEMAP providing the functionality raw to extract raw gaze data, along with parallel processing of process and graph steps. Process step calculate fixations, fixations in AOIs, saccade amplitudes, saccade duration, and IPA, whereas graph step generate visualizations. MPI gather function facilitates the aggregation of calculated eye gaze metrics in collate step, which provides data for statistical analysis in stats step.

3.2 Analysis of Fixation Count

Fixation count indicates the number of times eyes fixated on an AOI. We observed that participants with ADHD fixate more on left eye whereas NT participants fixate more on right eye. At SNRs 20 dB, 15 dB, and 10 dB, participants with ADHD fixated mostly on the left eye region.

We conducted repeated measures 2x6 two-way ANOVA with main factors of group, and SNR on fixation counts on each AOI. We observed a significant main effect of the SNR, $F(2.5, 22.5) = 22.14, p < 0.001$ as well as group, $F(1, 9) = 12.27, p < 0.008$ on fixation counts on left eye indicating that number of fixations differed among ADHD and NT participants as well as different SNRs. There was a significant interaction effect between SNR and group, $F(2.5, 22.5) = 2.958, p < 0.05$, indicating that fixation counts on the left eye on different listening conditions differed depending on the ADHD diagnosis.

We observed a significant main effect of the SNR for fixations on right eye, nose, and mouth, all $p < 0.02$, but no main effect of the group, all $F(1, 9) < 2.6, p > 0.05$. Also, there was no significant interaction effect between SNR and group for right eye, nose, and mouth, all $p > 0.05$. Contrasts of the SNR revealed that the number of fixations on the nose significantly differed when compared 25 dB, 20 dB, and 15 dB SNRs against 0 dB, all $F(1, 9) > 5.3, p < 0.05$ among the two groups. The number of fixations on left eye, right eye, and mouth on all SNRs significantly differed when compared 0 dB, $p < 0.05$ (see Table 1).
whereas NT participants tend to make gaze transition from any AOI to mouth region regardless the difficulty level of the task. In-between SNR and group, $F$ counts, indicating that cognitive load did not differ among participants.

Gaze transition matrices for different listening conditions suggest that, in general, participants with ADHD tend to make unpredictable gaze transitions at different difficulty levels of the task whereas NT participants tend to make gaze transition from any AOI to mouth region regardless the difficulty level of the task. Interestingly, it can be observed that participants with ADHD tend to re-fixate on the left eye region at 20 dB SNR, where the task is relatively easy.

We calculated the gaze transition entropy to determine the overall distribution of attention over AOIs. Small entropy values indicate predictable gaze transitions among AOIs, while large entropy values indicate less predictable gaze transitions among AOIs when transitioning from any source AOI to any destination AOI with similar probabilities [Krejtz et al., 2015].

Corresponding transition entropies of computed gaze transition matrices are shown in Table 2. There was no significant main effect of the SNR, $F(1.9, 17.18) = 2.3, p > 0.05$, or the group, $F(1.9) = 0.00, p > 0.9$ on transition entropies, indicating no difference among participant groups or SNRs. Also, there was no significant interaction effect between SNR and participant group, $F(1.9, 17.18) = 0.96, p > 0.3$. Table 2 shows a tendency of higher entropy for both ADHD and NT participants during the most difficult listening condition (0 dB), indicating less predictability in gaze transitions. Also, $t$-tests on transition entropies of participants at each SNR (i.e. without aggregating per participant) showed a significant effect for the NT group, at 0 dB compared to the other listening conditions (all $p < 0.03$).

### 3.3 Gaze Transition Matrices

The gaze transition matrices [Krejtz et al., 2015] indicate the probability of transition of gaze between two AOIs. Figure 3 shows the computed gaze transition matrices for ADHD and NT participants at gradually increasing levels of background noise.

Table 1: Fixation counts on AOIs of ADHD and NT Participants.

| SNR | Left Eye | Right Eye | Nose | Mouth |
|-----|---------|-----------|------|-------|
|     | ADHD    | NT        | ADHD | NT    | ADHD   | NT    | ADHD | NT    |
| 25 dB | 93.6 ± 14.5 | 54.0 ± 13.2 | 93.4 ± 23.1 | 125.7 ± 21.1 | 94.4 ± 13.5 | 44.3 ± 12.3 | 115.6 ± 23.7 | 150.0 ± 21.7 |
| 20 dB | 105.4 ± 10.9 | 52.3 ± 10.0 | 74.0 ± 18.7 | 96.7 ± 17.1 | 85.4 ± 11.4 | 51.7 ± 10.4 | 71.0 ± 19.2 | 116.5 ± 17.5 |
| 15 dB | 83.6 ± 9.20 | 68.7 ± 8.40 | 60.6 ± 9.70 | 76.2 ± 8.80 | 68.4 ± 12.6 | 48.5 ± 11.5 | 76.0 ± 15.3 | 85.2 ± 13.9 |
| 10 dB | 60.2 ± 6.10 | 42.3 ± 5.60 | 42.2 ± 3.70 | 51.5 ± 3.40 | 52.4 ± 10.0 | 44.7 ± 9.10 | 46.4 ± 8.80 | 66.7 ± 8.10 |
| 5 dB | 33.8 ± 3.90 | 30.0 ± 3.60 | 34.2 ± 2.70 | 27.3 ± 2.40 | 42.6 ± 12.9 | 29.5 ± 11.8 | 36.2 ± 4.50 | 35.7 ± 4.10 |
| 0 dB | 13.8 ± 5.80 | 10.2 ± 5.30 | 19.0 ± 6.90 | 5.30 ± 6.30 | 41.6 ± 16.9 | 34.8 ± 15.5 | 18.8 ± 6.80 | 16.0 ± 6.20 |

#### Table 2: Gaze Transition Entropy and IPA of ADHD and NT Participants.

| SNR | Entropy | IPA |
|-----|---------|-----|
|     | ADHD    | NT  | ADHD | NT |
| 25 dB | 0.53 ± 0.02 | 0.59 ± 0.02 | 0.29 ± 0.03 | 0.36 ± 0.02 |
| 20 dB | 0.59 ± 0.02 | 0.56 ± 0.02 | 0.29 ± 0.02 | 0.39 ± 0.02 |
| 15 dB | 0.59 ± 0.05 | 0.60 ± 0.04 | 0.33 ± 0.02 | 0.31 ± 0.02 |
| 10 dB | 0.62 ± 0.04 | 0.58 ± 0.04 | 0.35 ± 0.03 | 0.34 ± 0.03 |
| 5 dB | 0.61 ± 0.05 | 0.56 ± 0.04 | 0.36 ± 0.02 | 0.33 ± 0.02 |
| 0 dB | 0.66 ± 0.07 | 0.69 ± 0.06 | 0.30 ± 0.02 | 0.30 ± 0.02 |

### 3.4 The index of pupillary activity (IPA)

The IPA is calculated using a wavelet-based algorithm that relies on wavelet decomposition of the pupil diameter signal, and its wavelet analysis. For the IPA calculation, we used Daubechies-4 wavelet for a 60 Hz signal as suggested in [Duchowski et al., 2018]. Low IPA counts reflect little cognitive load whereas high IPA counts indicate strong cognitive load [Duchowski et al., 2018].

There was no significant main effect of the SNR, $F(5, 45) = 1.371, p > 0.05$, or of the group, $F(1.9) = 30.7, p > 0.05$ on IPA counts, indicating that cognitive load did not differ among participant groups or SNRs. We observed a significant interaction effect between SNR and group, $F(5, 45) = 3.265, p < 0.02$ on IPA counts such that cognitive load on different SNRs differed in ADHD and NT groups. Contrasts revealed significant interactions when comparing SNRs 25 dB and 0 dB, $F(1, 9) = 5.26, p < 0.05$, and SNRs 20 dB and 0 dB, $F(1, 9) = 7.27, p < 0.03$. These effects reflect that the cognitive load differed significantly among easiest and hardest listening conditions between the two groups (see Table 2). The remaining contrasts revealed no significant interaction when comparing two groups to different listening conditions, $p > 0.05$.

The hardest listening condition, we expect listening demands to be greater for participants with ADHD, yielding a significant difference in IPA counts, because their innate WMC is lower compared to the NT participants [Alderson et al., 2013; Banich et al., 2009; Michalek et al., 2014], thus perform significantly different on QuickSIN task. But, $t$-tests on IPA counts of participants at sentence level (i.e. without aggregating per participant) yielded no significant difference between the ADHD and NT groups, $p > 0.08$ for all SNRs, except 15 dB SNR. Interestingly, at 15 dB SNR, IPA counts of participants at sentence level yielded a significant difference between the ADHD and NT groups, $p < 0.04$ indicating a significant difference in cognitive load between the two groups. Since cognitive load inherently reduces WMC [Chandler and Sweller, 1991; Sweller et al., 1990], we expected participants with ADHD to do worse at 15 dB SNR, as their cognitive load is high and their WMC does not commensurate with NT participants. The expected behavior is confirmed by the significant performance difference observed in the evaluation of the number of keywords recalled between ADHD and NT groups at 15 dB SNR.

### 4 DISCUSSION

Our results indicate ADHD, and NT adolescents perform equally likely in the SIN task where audiovisual cues are present when the task difficulty is very high or very low. However, significant differences of QuickSIN performance between participants groups were observed at 15 dB SNR where ADHD and NT participants
with 15 dB SNR, we observed a significant difference in cognitive load as well as performance between the two groups.

In the future we expect to explore eye movement behavior when scanning the speaker’s face in terms of advanced eye movement metrics such as coefficient $\kappa$ of measurement of focal or ambient viewing, and a larger representation of ADHD and NT adolescents.

**REFERENCES**

R Matt Alderson, Lisa J Kasper, Kristen L Hudec, and Connor HG Patros. 2013. Attention-deficit/hyperactivity disorder (ADHD) and working memory in adults: a meta-analytic review. *Neuropsychology 27*, 3 (2013), 287.

Marie T Banich, Gregory C Burgess, Brendan E Depue, Luka Ruzic, I Cinnamon Budwell, Sena Hitt-Laustsen, Yiping P Du, and Erik G Wällcutt. 2009. The neural basis of sustained and transient attentional control in young adults with ADHD. *Neuropsychologia 47*, 14 (2009), 3095–3104.

Russell A Barkley. 1997. Behavioral inhibition, sustained attention, and executive functions: constructing a unifying theory of ADHD. *Psychological bulletin 121*, 1 (1997), 65.

Davina Bristow, Ghislaine Delaume-Lambertz, Jeremie Mattout, Catherine Soares, Teodora Gliga, Sylvain Baillet, and Jean-François Mangin. 2008. Hearing faces: how the infant brain matches the face it sees with the speech it hears. *Journal of cognitive neuroscience 21*, 5 (2008), 905–921.

Julie N Buchanan, Martin Paré, and Kevin G Munhall. 2008. The effect of varying talker identity and listening conditions on gaze behavior during audiovisual speech perception. *Brain research 1242* (2008), 162–171.

Cameron S Carter, Penelope Krener, Marc Chaderjian, Chersie Northcutt, and Virginia Wolfe. 1995. Asymmetrical visual-spatial attentional performance in ADHD: evidence for a right hemispheric deficit. *Biological psychiatry 37*, 11 (1995), 789–797.

Paul Chandler and John Sweller. 1991. Cognitive load theory and the format of instruction. *Cognition and instruction 8*, 4 (1991), 293–332.

Andrew T Duchowski. 2017. The Gaze Analytics Pipeline: In Eye Tracking Methodology. *Springer, New York, NY*, 175–191.

Andrew T Duchowski, Krysztof Krejtz, Izabela Krejtz, Cesary Biele, Anna Niedzielak, Peter Kirer, Martin Raubal, and Ioannis Giannopoulos. 2009. The index of pupillary activity: measuring cognitive load vis-à-vis task difficulty with pupil oscillation. In *Proceedings of the 2008 CHI Conference on Human Factors in Computing Systems*. ACM, Montréal, QC, Canada, 282.

Scott A Fields, William Michael Johnson, and Madison B Hassig. 2017. Adult ADHD: Addressing a unique set of challenges. *The journal of family practice 66*, 6 (2017), 68–74.

Leah Fostick. 2017. The effect of attention-deficit/hyperactivity disorder and methylphenidate treatment on the adult auditory temporal order judgment threshold. *Journal of Speech, Language, and Hearing Research 60*, 7 (2017), 2124–2134.

Sarah Fraser, Jean-Pierre Gagné, Majdolaine Alepis, and Pascale Dubois. 2010. Evaluating the effort expended to understand speech in noise using a dual-task paradigm: The effects of providing visual speech cues. *Journal of Speech, Language, and Hearing Research 53*, 1 (2010), 18–33. https://doi.org/10.1044/1092-4388(2009/08-0140)

5 CONCLUSION

Our work presents an analysis of audiovisual SIN performance for young adults with ADHD compared to age-matched controls using eye-tracking measures. We analyzed the performance of the participants and eye-movement parameters such as fixation count on AOIs, gaze transition entropy, and JPA. We observed that participants with ADHD primarily fixated on the left eye of the speaker whereas NT group fixated on the right eye, supporting the literature that ADHD orient attention to the LVF whereas NT individuals orient attention to the RVF. When the task difficulty was at a medium level

Figure 3: Gaze transition matrices of ADHD and NT participants, at varying levels of background noise yielding six SNR levels: 0 to 25 dB with 5dB increments.
Kenneth M Heilman, Kytja KS Voeller, and Stephen E Nadeau. 1991. A possible pathophysiologic substrate of attention deficit hyperactivity disorder. *Journal of Child Neurology* 6, 1 (suppl) (1991), S76–S81.

Liro P Jaukkelainen. 2018. The role of speech production system in audiovisual speech perception. *The open neuroimaging journal* 4 (2010), 30.

Gavindya Jayawardena. 2020. RAEMAP: Real-Time Advanced Eye Movements Analysis Pipeline. In Symposium on Eye Tracking Research and Applications 2020. ACM, Stuttgart, Germany.

Mead C Killion, Patricia A Niquette, Gail I Gudmundsen, Lawrence J Revit, and Shilpi Gavindya Jayawardena. 2020. RAEMAP: Real-Time Advanced Eye Movements Analysis Pipeline. In Symposium on Eye Tracking Research and Applications 2020. ACM, Stuttgart, Germany.

Krzysztof Krejtz, Andrew Duchowski, Tomasz Samidt, Izabela Krejtz, Fernando González Perrelli, Ana Pires, Anna Vilaro, and Natalia Villalobos. 2015. Gaze transition entropy. *ACM Transactions on Applied Perception (TAP)* 13, 1 (2015), 4.

Anne MP Michalek, Ivan Ash, and Kathryn Schwartz. 2018. The independence of working memory capacity and audiovisual cues when listening in noise. *Scandinavian journal of psychology* 59, 6 (2018), 578–585.

Anne MP Michalek, Silvana M Watson, Ivan Ash, Stacie Ringleb, and Anastasia Raymer. 2014. Effects of noise and audiovisual cues on speech processing in adults with and without ADHD. *International journal of audiology* 53, 3 (2014), 145–152.

Sushmit Mishra, Thomas Lunner, Stefan Stenfelt, Jerker Rönnberg, and Mary Rudner. 2013. Seeing the talker’s face supports executive processing of speech in steady state noise. *Frontiers in Systems Neuroscience* 7 (2013), 96.

Wendy G Mitchell, John M Chavez, Sherryll A Baker, Bianca L Guzman, and Stanley P Azen. 1990. Reaction time, impulsivity, and attention in hyperactive children and controls: A video game technique. *Journal of Child Neurology* 5, 3 (1990), 195–204.

Shahram Moradi, Björn Lidestam, and Jerker Rönnberg. 2013. Gated audiovisual speech identification in silence vs. noise: Effects on time and accuracy. *Frontiers in Psychology* 4 (2013), 359.

Ozan Pazvantoğlu, Arzu Alptekin Aker, Koray Karabekiroğlu, Seher Akbaş, Gökhan Sarısoy, Salihay Baykal, Işıl Zabun Korkmaz, Emel Alkan Pazvantoğlu, Umer Boke, and Ahmet Eulfat Şahan. 2012. Neuropsychological weaknesses in adult ADHD; cognitive functions as core deficit and roles of them in persistence to adulthood. *Journal of the International Neuropsychological Society* 18, 5 (2012), 819–826.

Mary Rudner, Catharina Foo, Jerker Rönnberg, and Thomas Lunner. 2009. Cognition and aided speech recognition in noise: Specific role for cognitive factors following nine-week experience with adjusted compression settings in hearing aids. *Scandinavian Journal of Psychology* 50, 5 (2009), 405–418.

Bruce A Schneider, Liang Li, and Meredith Daneman. 2007. How competing speech interferes with speech comprehension in everyday listening situations. *Journal of the American Academy of Audiology* 18, 7 (2007), 559–572.

John Sweller, Paul Chandler, Paul Tierney, and Martin Cooper. 1990. Cognitive load as a factor in the structuring of technical material. *Journal of experimental psychology: general* 119, 2 (1990), 176.

Virginia Van Wassenhove, Ken W Grant, and David Poeppel. 2005. Visual speech speeds up the neural processing of auditory speech. *Proceedings of the National Academy of Sciences* 102, 4 (2005), 1181–1186.

Kytja KS Voeller and Kenneth M Heilman. 1988. Attention deficit disorder in children: A neglect syndrome? *Neurology* 38, 5 (1988), 806–806.

Katharina von Kriegstein, Özgür Dogan, Martina Grüter, Anne-Lise Giraud, Christian A Kell, Thomas Grüter, Andreas Kleinschmidt, and Stefan J Kirfel. 2008. Simulation of talking faces in the human brain improves auditory speech recognition. *Proceedings of the National Academy of Sciences* 105, 18 (2008), 6747–6752.

Steven Woltering, Zhongyu Liu, Alan Rokeach, and Rosemary Tannock. 2013. Neuropsychological characteristics of adults with ADHD: A comprehensive review of initial studies. *The Clinical Neuropsychologist* 16, 1 (2002), 12–34.

Guifeng Xu, Lane Strathearn, Buyun Liu, Binrung Yang, and Wei Bao. 2018. Twenty-year trends in diagnosed attention-deficit/hyperactivity disorder among US children and adolescents, 1997–2016. In *JAMA Network Open*. American Medical Association, Iowa City, IA, e181471–e181471.

Adriana A Zekveld, Mary Rudner, Ingrid S Johnsrude, Joost M Festen, Johannes HM Van Beek, and Jerker Rönnberg. 2011. The influence of semantically related and unrelated text cues on the intelligibility of sentences in noise. *Ear and hearing* 32, 6 (2011), e16–e25.