Neural Correlates of Statistical Learning in Developmental Dyslexia: An Electroencephalography Study

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- 5 Short title: Statistical Learning in Developmental Dyslexia
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- 7 Tatsuya Daikoku^{1,2*}, Sebastian Jentschke³, Vera Tsogli⁴, Kirstin Bergström⁵, Thomas
- 8 Lachmann^{5,6}, Merav Ahissar⁷, and Stefan Koelsch^{1,4}
- 9
- 10 ¹Department of Neuropsychology, Max Planck Institute for Human Cognitive and Brain
- 11 Sciences, Leipzig, Germany
- 12 ²International Research Center for Neurointelligence (WPI-IRCN), UTIAS, The
- 13 University of Tokyo, Tokyo, Japan
- 14 ³Department of Psychosocial Science, University of Bergen, Bergen, Norway
- 15 ⁴Department for Biological and Medical Psychology, University of Bergen, Bergen,
- 16 Norway
- 17 ⁵Center for Cognitive Science, University of Kaiserslautern, Kaiserslautern, Germany
- 18 ⁶ Centro de Investigación Nebrija en Cognición, Universidad Nebrija, Madrid (Spain)
- 19 ⁷Psychology Department, Hebrew University, Jerusalem, Israel
- 20
- 21 *Correspondence to: <u>daikoku.tatsuya@mail.u-tokyo.ac.jp</u>
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25 **Conflict of Interest**

26 The authors declare no competing financial interests.

27

28 Acknowledgements

- 29 This study was supported by the German-Israeli Foundation for Scientific Research and
- 30 Development and World Premier International Research Centre Initiative (WPI), MEXT,
- and JSPS KAKENHI Grant Number 20K22676, Japan. The sponsors played no role in the
- 32 study design; the collection, analysis, and interpretation of the data; or the writing of the
- 33 manuscript.

34 Abstract

35

36 The human brain extracts statistical regularities from the surrounding environment in a process referred to as statistical learning. Recent behavioural evidence suggests that 37 38 developmental dyslexia affects statistical learning. However, surprisingly few 39 neurophysiological studies have assessed how developmental dyslexia affects the neural processing underlying statistical learning. In this study, we used 40 41 electroencephalography to explore the neural correlates of an important aspect of 42 statistical learning - sensitivity to transitional probabilities - in individuals with 43 developmental dyslexia. Adults diagnosed with developmental dyslexia (n = 17) and 44 controls (n = 19) were exposed to a continuous stream of sound triplets in which a few triplet endings were location deviants (i.e., were presented from an unexpected speaker 45 direction) or statistical deviants (i.e., had a low transitional probability given the 46 triplet's first two sounds). Location deviants elicited a large location mismatch 47 48 negativity (MMN), which was larger in controls than dyslexics. Statistical deviants elicited a small, yet significant statistical MMN (sMMN) in controls, whereas the dyslexic 49 50 individuals did not exhibit a statistical MMN. These results suggest that the neural 51 mechanisms underlying statistical learning are impaired in developmental dyslexia.

52 Significance statement

53 We assessed the neural correlates of statistical learning in individuals with 54 developmental dyslexia. Statistical deviants, namely word endings with a low 55 transitional probability (as compared to high probability transitions) elicited a small, 56 yet significant statistical MMN in controls, whereas the dyslexic individuals did not 57 exhibit a statistical MMN. Location deviants elicited a MMN, which was larger in 58 controls than dyslexics. These results suggest that the neural mechanisms underlying 59 statistical learning are impaired in developmental dyslexia.

60 Introduction

61	The brain can identify statistical regularities in sequential information in a
62	process known as statistical learning (Saffran, Aslin, & Newport, 1996) or implicit
63	learning (Perruchet & Pacton, 2006). Statistical learning involves an implicit and innate
64	mechanism by which the brain calculates the transitional probability of sequential
65	information. This learning system is thought to play a crucial role in early language
66	acquisition. For example, 8-month-old infants (Saffran, Aslin, & Newport, 1996) and
67	neonates (Teinonen et al., 2009) can learn the probabilities of syllable transitions, thus
68	enabling them to detect word boundaries and isolate single words in natural speech.
69	
70	Developmental dyslexia impairs reading comprehension and spelling in
71	language, and is thought to arise mainly from phonological perceptual difficulties
72	(Snowling, 2000; Ramus et al., 2003; Vellutino et al., 2004). In addition, individuals with
73	developmental dyslexia may further have a wide range of other nonlinguistic as well as
74	linguistic impairments, including weakened auditory statistical learning abilities
75	(Arciuli & Simpson, 2012; Du & Kelly, 2013; Evans, Saffran, & Robe-Torres, 2009;
76	Howard, Howard, Japikse, & Eden, 2006; Menghini, Hagberg, Caltagirone, Petrosini, &
77	Vicari, 2006; Vicari et al., 2005) and non-linguistic perceptual processing (Ahissar,
78	Protopapas, Reid, & Merzenich, 2000; Christmann, Lachmann, & Steinbrink, 2015;
79	Giraud & Ramus, 2013; McAnally & Stein, 1996; Sperling, Lu, Manis, & Seidenberg,
80	2005). Since phonological deficits can also be observed in people without
81	developmental dyslexia (Huettig et al., 2017), and not all children with developmental
82	dyslexia show phonological processing deficits (Lachmann & van Leeuwen, 2008;
83	Morris et al., 1998), phonological deficits may not automatically lead to developmental

84 dyslexia. Thus, at least in some individuals, developmental dyslexia may also have other85 causes.

87	One possible cause is a more domain-general statistical learning deficit, not
88	specific to phonological processing. Particularly, a great deal of evidence has shown that
89	auditory statistical learning is impaired in dyslexic children, adolescents, and adults
90	(Gabay et al., 2015; Dobo et al., 2021; Kahta et al., 2019; Vandermosten et al., 2019),
91	although some other studies suggest that children with and without developmental
92	dyslexia had no difference in statistical learning performance (Witteloostuijn et al.,
93	2019). Despite much behavioral evidence on statistical learning deficits, the underlying
94	neural mechanisms of these deficits remain unclear. This study aimed to investigate the
95	neural correlates of statistical learning in individuals with developmental dyslexia using
96	electroencephalography (EEG), which exhibits a high sensitivity to capture statistical
97	learning even when behavioural measures may not indicate learning effects (Koelsch,
98	Busch, Jentschke, & Rohrmeier, 2016).
99	
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109 auditory brainstem response (ABR; Skoe, et al., 2015), P50 (Daikoku et al., 2017; 110 Paraskevopoulos et al., 2012), N100 (Sanders et al., 2002), and mismatch negativity 111 (MMN; Koelsch et al., 2016; Moldwin, Schwartz, & Sussman, 2017) as well as later ERP 112 components such as the P200 (Balaguer et al., 2007; Cunillera et al., 2006) and the N400: 113 (François, et al., 2013). Typically, a mismatch negativity (MMN) is elicited in response to 114 physical deviants in oddball sequences. In such experiments, a series of standard stimuli 115 is interspersed with physical deviants (oddballs; e.g. sound differing in pitch or location 116 deviants; Christmann, Lachmann, & Berti, 2014; Garrido et al., 2008; Rinne, Antila, & 117 Winkler, 2001; Sussman, Winkler, & Schröger, 2003; Winkler & Czigler, 2012). For 118 example, if several sounds are presented from the right side, a sound presented on the 119 left side elicits an MMN, which is generated mainly in the auditory cortex (Garrido et al., 120 2008).

121

122 Because developmental dyslexia is related to sensory processing dysfunctions. 123 including those of the auditory cortex (Clark et al., 2014; Goswami, 2014; for a review 124 see Gu & Bi, 2020), the MMN has been used to investigate the neural basis of 125 developmental dyslexia (Kujala et al., 2000). A reduced MMN amplitude in children 126 (Lachmann et al., 2005; for an overview see Bishop, 2007) and young adults (Schulte-127 Körne et al., 2001) with developmental dyslexia reflected impaired performance in 128 syllable and tone discrimination and impaired tuning to native language speech 129 representations (Bruder et al., 2011). The MMN has therefore been suggested as a 130 neurophysiological endophenotype for developmental dyslexia (Neuhoff et al., 2012). 131 However, there is also one study suggesting that only certain aspects of auditory 132 processing may be affected: whereas the pitch MMN was shown to be impaired in that 133 study, was the location MMN enhanced (Kujala et al., 2006). Notably, such MMN effects

134 to physical changes do not require statistical learning because the perceptual 135 regularities underlying the generation of the classical MMN can be extracted on a 136 moment-to-moment basis, such as a series of stimuli coming from the right side 137 interrupted by a stimulus from the left side. However, statistical learning can also be 138 reflected in the MMN (François, Cunillera, Garcia, Laine, & Rodriguez-Fornells, 2017; 139 Koelsch et al., 2016; Moldwin et al., 2017). Tsogli, Jentschke, Daikoku, and Koelsch 140 (2019) presented sequences of tone triplets that could contain location deviants 141 (comprising stimuli from an irregular location) and statistical deviants (comprising 142 triplet endings with a low probability given the two preceding triplet items; this low 143 transition probability could only be represented based on statistical learning, i.e., after 144 extensive exposure to many triplets). Both statistical and location deviants elicited 145 prominent mismatch ERP responses approximately 150–250 ms after stimulus onset. 146 ERP effects elicited by statistical deviants are termed statistical MMN (sMMN; Koelsch 147 et al., 2016) to distinguish them from the MMN to physical deviance, e.g., elicited by 148 location deviants (Paavilainen, Karlsson, Reinikainen, & Näätänen, 1989; Sams, 149 Paavilainen, Alho, & Näätänen, 1985). In contrast to the MMN eliecited by physical 150 deviance, the elicitation of the sMMN does not occur on a moment-to-moment basis. 151 Instead, it requires a more extended learning period to encode the underlying statistical 152 regularities and to store them in long-term memory. Thus, the sMMN is suitable to 153 investigate acquisition of knowledge regarding the statistical regularity of sound 154 sequences.

155

This study investigated how developmental dyslexia affects the classical MMN in
response to location deviants and the sMMN in response to statistical deviants in adults.
Assuming that dyslexia adversely affects the inferences of sensory statistics in the

auditory cortex (Jaffe-Dax, Kimel, & Ahissar, 2018; Lieder et al., 2019), we hypothesised
that statistical and location deviants would elicit weaker MMNs in participants with
developmental dyslexia than the controls. Confirmation of this hypothesis would
provide evidence that developmental dyslexia is associated with both auditory
processing dysfunction and statistical learning of auditory sequences.

165 Materials and Methods

166 **Participants**

Twenty-one adults diagnosed with developmental dyslexia (11 females, mean 167 168 age = 26 years, SD = 5.3) and 20 age- and gender-matched control participants without a 169 diagnosis of dyslexia (14 females, mean age = 26 years, SD = 3.1) were screened for 170 eligibility to participate in this study. We excluded one individual from the 171 developmental dyslexia group and one from the control group because both performed 172 in the nonverbal intelligence test Standard Progressive Matrices (Raven & Court, 1998) 173 with an IQ score < 70 (two SD below the normal range). Three other individuals with 174 developmental dyslexia were excluded because they did not receive a clear diagnosis in 175 childhood. After these exclusions, our study sample included 17 adults with 176 developmental dyslexia and 19 control participants (Table 1), who all met the following 177 inclusion criteria: German as the native language, right-handedness (Edinburgh 178 Inventory; Oldfield, 1971), no history of neurological or audiological disorders, no 179 diagnosis of a general or specific language impairment, no mental retardation, and no 180 formal musical training for more than 5 years (beyond regular school lessons). 181

The study protocol was approved by the Ethics Committee of the Max-Planck-Institute (approval number: 2018/352). All participants were informed of the purpose of the study and the procedures in place to ensure their safety and the confidentiality of their personal data. They all provided written informed consent to participate in this study.

187

188 Evaluation of Spelling and Reading Skills

189 Spelling skills were assessed with the Rechtschreibtest (Ibrahimović & Bulheller, 190 2013). In this test, participants were asked to fill in the missing words of a text, mainly 191 composed of irregular German words, that a skilled and German native speaking 192 experimenter read aloud. Silent text reading speeds and reading comprehension skills 193 were assessed with the Lesegeschwindigkeits-und Verständnistest für die Klassen 5–12 194 (LVGT 5-12; Schneider, Schlagmüller, & Ennemoser, 2017). In this test, participants 195 were asked to read as much of the text as possible within 4 min and to fill in each gap in 196 the text with one of the three possible options.

197

198 Stimuli

199 Sounds

200 We used the same stimuli and sequences as in a previous study (Tsogli et al., 201 2019). Each sound consisted of a Shepard tone (Shepard, 1964), combined with the 202 sound produced by one of six different percussion instruments (i.e., a surdo, a 203 tambourine, agogô bells, a hi-hat, castanets, or a woodblock). We obtained the 204 percussive sounds from the Philharmonia Orchestra website 205 (http://www.philharmonia.co.uk/explore/sound_samples). We used six distinct

206	Shepard tones based on six frequencies (i.e., F_3 [174.61 Hz], G_3 [196.00 Hz], A_3 [220.00
207	Hz], B ₃ [246.94 Hz], C \mathbb{Z}_4 [277.18 Hz], and D \mathbb{Z}_4 [311.13 Hz]), each tone resulting from the
208	superposition of nine sinusoidal components spaced an octave apart. The specific
209	combinations of Shepard tones and percussive sounds were counterbalanced across
210	participants. Examples of sounds are provided in Appendix A.
211	
212	Another set of six sound combinations was created for a practice phase at the
213	start of each experiment. These sounds were similar to those used in the main
214	experiment but differed in terms of the frequencies providing bases for Shepard tones
215	(i.e., E_3 [164.81 Hz], $F\mathbb{Z}_3$ [184.99 Hz], $G\mathbb{Z}_3$ [207.65 Hz], $A\mathbb{Z}_3$ [233.08 Hz], C_4 [261.62 Hz],
216	and D_4 [293.66 Hz]) and the percussive sounds used (i.e., the sounds of a woodblock, a
217	tambourine, agogô bells, castanets, a hi-hat, and a bass drum). An additional target
218	Shepard tone based on C2 $_5$ (554.37 Hz), that did not have an accompanying percussive
219	sound, was used for a cover task and a passive listening component of the experimental
220	procedure (see Experimental Procedure).
221	All sound stimuli had a tone duration of 220 ms, including rising and falling
222	periods of 10 ms and 20 ms, respectively, a constant loudness, and a sampling frequency
223	of 44,100 Hz with 16 bit resolution.
224	
225	Triplet Sequences
226	The stimuli descibed above, hereafter referred to as sounds A to F, were
227	combined into sound triplets. Each 220 ms sound was followed by an 80 ms pause, such
228	that the total duration of each triplet was 900 ms. As shown in Figure 1, sounds A and B
229	and sounds C and D were paired to create two distinct two-sound sequences (i.e., AB
230	and CD) that served as the first two sounds of each triplet, which are hereafter referred

to as the triplet roots. Sounds E and F were the sounds that could be used as the last
sound of a triplet, which is hereafter referred to as the triplet ending. Combining the
two roots and two triplet endings yielded four possible triplets (i.e., ABE, ABF, CDE, and
CDF). The assignment of different sounds to roles as triplet roots or endings was
counterbalanced across participants as a way of ensuring that any possible acoustical
differences between sounds would be cancelled out through particpants and not bias
the neural responses of interest.

238

Exposition Sequences. The sequence of sounds used during an exposition phase
comprised 400 sound triplets with a total sequence duration of about 6 min. The triplets
were presented in a pseudo-randomised order with no two sequentially adjacent
triplets being identical. Each of the two roots had an equal probability of occurring in a
given triplet regardless of the ending sound of the previous triplet. Each statistically
deviant triplet was followed by at least 3 triplets that were not statistically deviant.

245

246 Each sound stimulus was presented from either a speaker to the participant's 247 right or a speaker to the participant's left. These speakers were positioned at 60° angles 248 in the azimuthal plane. For each participant, one side was pseudo-randomly selected as 249 the standard side for stimuli to be presented from, and the other side was the deviant 250 one. The lateralization of the stimuli was balanced across blocks and counterbalanced 251 between participants, and whether the location was "standard" or "deviant" was 252 considered in the data analyses. For the triplet root sounds (i.e., sounds A, B, C, and D), 253 95% of the stimuli were presented from the standard side and the remaining 5% from 254 the deviant side. For the triplet endings (i.e., sounds E and F), 80% of the stimuli were 255 presented from the standard side and the remaining 20% from the deviant side. The

triplets with endings presented from the deviant side were considered location

257 deviants.

258

259 To generate statistical deviants, we set distinct probabilities for a transition (i.e., 260 transitional probability) from a given root to a given ending within a triplet. The 261 sensitivity to transitional probability is one of the important aspects of statistical 262 learning mechanisms (Perruchet & Pacton, 2006; Saffran, Aslin, & Newport, 1996). The 263 transitional probability for a given triplet ending was either 90% or 10% depending on 264 root identity (Figure 1a). Triplets containing low-probability root-to-ending transitions 265 were considered statistical deviants. Each statistically deviant triplet was followed by 266 \geq 3 triplets that were not statistically deviant. The location and statistical deviance 267 created four triplet categories (Table 2). That is, standard triplets, which accounted for 268 72% of all triplets, were neither location deviant nor statistically deviant. The remaining 269 28% of triplets could be statistically deviant only (8% of all triplets), location deviant 270 only (18% of all triplets), or both statistically deviant and location deviant (2% of all 271 triplets).

272

273 Behavioural Testing Sequences. Each behavioural testing phase of the 274 experiment consisted of twelve trials, with each trial involving a pair of triplets 275 separated by a 335-ms pause. The triplets in each pair had the same root but different 276 endings with low vs. high transition probabilities, respectively (i.e., statistically deviant 277 vs. not deviant). The order in which the statistically deviant and standard triplets were 278 played was counterbalanced across trials. Varying the order in which the same-root 279 triplets within a pair were played created four sequentially distinct pairings (i.e., ABE to 280 ABF, ABF to ABE, CDE to CDF, and CDF to CDE). Each sequentially distinct pairing was

- 281 played three times during each behavioural testing phase, and consecutively played
- 282 pairings had alternating triplet roots (i.e., each AB pairing was followed by a CD pairing
- that was in turn followed by an AB pairing). During behavioural testing phases, all
- triplets were played from both speakers, i.e., the task focussed on statistical regularity
- 285 whereas location was not a factor.



296

297 Figure 1. Triplets and stream in this study. (A) Four types of triplets generated from 298 six different sounds (designated with letters from A to F) produced by pairing a Shepard 299 tone with various percussive sounds. Each triplet consists of a root containing two 300 conserved sounds (AB or CD) and a triplet ending with a high (90%) or a low (10%)301 transitional probability given the triplet's root. Each triplet root has a 50% probability 302 of occurring regardless of the ending sound of the previous triplet. (\mathbf{B}) Example of a 303 possible triplet sequence including standard triplets (triplet endings in black boxes). 304 triplets with a statistically deviant ending (triplet endings in blue boxes) or a location 305 deviant ending (triplet ending in red box), and a triplet with a doubly deviant ending 306 (i.e., statistically deviant and location deviant; triplet ending in purple box). Reprinted, 307 with permission, from Tsogli et al. (2019). Abbreviations: ISI, inter-stimulus interval; p. 308 & prob., probability.

309 Experimental Design

310 The participants completed a multi-stage experiment with six blocks while 311 undergoing EEG monitoring (see 'Collection and Analysis of EEG Data') inside an 312 electromagnetically shielded chamber. Immediately before the experiment, the 313 participants were provided with instructions concerning the procedures for the 314 experiment's different phases (see below). To ensure that only implicit learning could 315 occur, the instructions did not include any description of the possible location or 316 statistical deviance of the triplets. The participants then completed a 1-min practice 317 session in which they were asked to press a key as soon as possible after hearing a 318 target sound (i.e., C25 [554.37 Hz]). If necessary, each participant repeated the practice 319 session until they had correctly pressed the key after 80% of the target sound 320 presentations. 321

322 Each of the six blocks included an exposition phase comprising the passive 323 listening of the sequence of 400 triplets and a subsequent behavioural testing phase in 324 which the participants performed actions based on a triplet sequence. During the 325 exposition phase, participants were instructed to react to the high-pitched tones (cover 326 tasks, 0.67% of all tones) while they were exposed to the sequence of 400 triplets 327 described above (see Exposition Sequences). At the same time, they were watching a 328 silent movie (nature or wildlife documentaries) played on a monitor in front of them. 329 While the cover task was not particularly demanding (e.g., in terms of attentional 330 resources) it minimized the possibility of the participants intentionally focusing their 331 attention on statistical properties.

332

During a behavioural testing phase, the participants listened to paired
statistically deviant and standard triplets (see Behavioural Testing Sequences). After
each pair of triplets, a participant was asked to choose which triplet in the pair sounded
more familiar and to rate their confidence in a given answer on a scale ranging from 1
(no certainty) to 5 (certainty).

338

339 Acquisition and Analysis of EEG Data

340 We obtained 64-channel EEG data (Brain Amp, Brain Products, Munich, 341 Germany) through cap-mounted electrodes placed over the participants' scalps in 342 accordance with the extended international 10-20 system. The left mastoid electrode 343 was used as reference electrode and the neck electrode was used as ground electrode. 344 The electrodes were clustered into six regions of interest: a frontal left region (F7, F5, 345 F3, FT7, FC5, and FC3), a frontal middle region (F1, FZ, F2, FC1, FCZ, and FC2), a frontal 346 right region (F8, F6, F4, FT8, FC6, and FC4), a central left region (T7, C5, C3, TP7, CP5, and CP3), a central middle region (C1, Cz, C2, and CPZ), and a central right region (T8, 347 348 C6, C4, TP8, CP6, and CP4). Horizontal and vertical electro-oculograms were recorded 349 bipolarly through electrodes placed at the outer canthi of the eyes and above and below 350 the right eye. Electrode impedance was kept < 5 k Ω . Signals were recorded with a 0.25– 351 1,000-Hz bandpass filter and a 500-Hz sampling rate.

352

EEG data were analysed in EEGLAB 13 (Delorme & Makeig, 2004) in MATLAB R2018b (The MathWorks, Natick, Massachusetts). Continuous raw data files were rereferenced to the algebraic mean of the left and right mastoid electrodes and filtered with a 0.5-Hz high-pass filter and a 30-Hz low-pass filter implemented with finite

357 impulse response designs and Blackman windows of 550 points and 2,750 points, 358 respectively. Channels with excessive noise were identified through visual inspection 359 and interpolated when necessary. The mean number of interpolated channels per 360 participant was 0.22. Independent component analysis was used for linear 361 decomposition of continuous data to remove the contributions of artefacts affecting 362 scalp sensors (e.g. slow drifts, eye blinks or movement, and muscle artefacts). Epochs 363 were removed from further analyses if the amplitude changes exceeded $\pm 45 \,\mu$ V in any 364 channels, including the electro-oculograms (less than 10% of the trials). The epochs of 365 target stimuli (cover tasks) were removed in the analysis. In the end, 97.2% (SD±2.6%) 366 and 94.8% (SD±8.0%) was preserved in the dyslexic and control groups, respectively. 367 We also performed student's t-test between groups. There were no significant 368 differences (t(37), 0.69, p=0.50). Selective response averaging was conducted 369 separately for standard triplets, triplets with a location deviant (but not with a 370 statistical deviant), triplets with a statistical deviant (but not a location deviant), and 371 triplets with both a location deviant and a statistical deviant (see 'Exposition 372 Sequences'). 373

Averages were computed using a 100 ms pre-stimulus baseline. We directly addressed the hypothesis of the present study, focusing on the average MMN amplitudes measured within a 150–250 ms time window. In addition, we analysed an obvious positive component approximately 100–140 ms after the onset of the stimuli, henceforth referred to as P120. That is, we also investigated the average amplitudes within a 100–140 ms time window.

381 Statistical Analysis

382 Between-Group Comparisons of Participant Characteristics

Statistical analyses were conducted using jamovi version 1.2 (The jamovi Project,
2020). We used Bonferroni-corrected *t*-tests (dividing 0.05 by the number of tests)
when comparing demographic characteristics and scores on tests of intelligence,
spelling skills, and reading abilities between the dyslexia and the control group.

388 Analyses of Exposition Phase Data

389 We used separate ANOVA models to analyse the effects of stimulus deviance on 390 P120 and MMN effect amplitudes during the exposition phases. The sMMN response is 391 had a smaller amplitude size compared with the location MMN. Further, the sMMN is a 392 relatively new ERP component, for which the neural mechanism has not yet been 393 elucidated in detail. To determine those neural generators, studies using fMRI or MEG 394 would be most suited, however, such studies still are lacking despite many studies using 395 EEG methodology. Thus, this study directly compared between standard and deviant 396 but did not use the subtraction waveform that is typically used in MMN studies. One 397 ANOVA model featured the within-participant factor of sound location (i.e., location 398 standard triplet endings or location deviant triplet endings), and the other featured the 399 within-participant factor of transitional probability (i.e., statistically standard triplet 400 endings or statistically deviant triplet endings). Both ANOVA models included the 401 between-participants factor of groups (i.e., the developmental dyslexia vs. the control 402 group) and three within-participant factors: the distinction between ERP responses in 403 the frontal vs. central areas of the brain; the distinction between ERP responses in the 404 left, medial, and right areas of the brain (i.e., ERP response lateralisation); and the 405 different experimental blocks. The posterior region was not included because of low

406	amplitude sizes.	To boost our	signal-to-no	ise ratio, our	·ANOVA m	odels included three
	1		0	,		

- 407 blocks rather than the actual six by merging the data from pairs of blocks.
- 408

409	We selected $p < 0.05$ as our threshold for statistical significance and used a false
410	discovery rate method for the post-hoc testing of significant effects. To determine
411	whether the Rechtschreibtest and LVGT 5–12 scores correlated with the P120 and MMN $$
412	effect amplitudes, we calculated Pearson correlation coefficients.
413	
414	Analyses of Behavioural Testing Phase Data
415	We used a two-tailed <i>t</i> -test to determine whether the frequency of correct answers
416	during behavioural testing exceeded chance levels (i.e., >.5). We also used ANOVA
417	models to compare the dyslexia and control groups in terms of response accuracies and
418	reaction times (RT), and included experimental blocks as a within-subject factor in
419	these analyses. As in our analyses of exposition phase data, we assumed three blocks
420	rather than the actual six in order to boost our signal-to-noise ratios. We selected $p < .05$
421	as our threshold for statistical significance and used Bonferroni correction for the post-
422	hoc testing of significant effects. We conducted Pearson correlation analysis to

 $423 \quad determine \ whether \ response \ accuracy \ rates \ correlated \ with \ confidence \ ratings \ and$

424 dyslexia test scores.

425

426 **Results**

427 Participant Characteristics

428 Relative to the control group, the developmental dyslexia group had lower
429 average scores for spelling abilities, reading comprehension, and reading speed (Table

430 1). However, the two groups were comparable in terms of age and intelligence (age:

431
$$t(34) = 1.41, p = .167; IQ: t(34) = .50, p = .620).$$

432

433 **EEG Results**

434 P120 and MMN Responses to Location Deviance

435 First, we examined whether developmental dyslexia affects ERP responses to location 436 deviants (only standards and deviants presented on high-probability triplet endings 437 were included in this analysis). Location deviants elicited a P120 followed by a location 438 MMN, which was maximal at anterior frontal electrodes (Figure 2a–b). The location 439 MMN appeared in both the dyslexia and control groups, but had a smaller amplitude in 440 the dyslexia group. Compared with the standards, the location deviants elicited larger 441 P120 responses in the control group, whereas the location deviants did not elicit larger 442 P120 responses in the dyslexia group (see Appendix B for mean P120 and location MMN 443 amplitudes in each condition). These observations were reflected in an ANOVA, 444 indicating a significant interaction between sound location (standard, deviant) and group for the P120: F(1, 34) = 5.52, p = .025, $\eta^2 p = .14$; and the location MMN: F(1, 34) =445 5.16, p = .03, $\eta^2 p = .13$; see the Appendix C for complete results. When analysing ERP 446 447 responses separately in each group, location deviants elicited a significant location 448 MMN in both the dyslexia group and the control group (p < .001; Figure 2c). Compared 449 with the standards, location deviants elicited larger P120 responses in the control group 450 (p = .011), but there was no significant difference in the dyslexia group (p = .49). Higher 451 spelling scores on the Rechtschreibtest correlated with larger location deviance-452 induced P120 effects (r = .37, p = .03; see the Appendix C). No other correlations 453 between language aptitude test scores and EEG responses were observed.

454

455 P120 and MMN Responses to Statistical Deviance

456 Next, we examined whether developmental dyslexia affects ERP responses to 457 statistical deviants (only triplet endings without location change were included in this 458 analysis). Statistical deviants elicited a P120, followed by a sMMN, which was maximal 459 over anterior frontal electrodes (Figure 3a-b; see the Appendix B for mean amplitudes 460 of the P120 and sMMN components in each condition). Statistical deviants elicited a 461 P120 in both groups, and an sMMN in the control group but not in the dyslexia group. 462 These observations were reflected in an ANOVA, indicating a significant interaction 463 between transitional probability (high vs. low probability), frontal vs. central, and Group: F(1, 34) = 5.50, p = .025, $\eta^2 p = .14$ (for complete results, see the Appendix D). 464 465 Post-hoc tests revealed that the sMMN response at frontal electrodes was significant in 466 the control group (p = .032) but not in the dyslexia group (p = .74; Figure 3c). At the 467 central electrodes, the sMMN was not significant in both groups.

468

The effects of statistical learning on P120 amplitudes were prominent at medial electrodes (Figure 3b–c), as reflected by the significant interaction between transitional probability and lateralisation (F[2, 68] = 4.15, p = .020, $\eta^2 p = .11$). Post-hoc tests revealed that the P120 responses to statistical deviants were larger than those to the standards at medial electrodes (p = .009); however, no such effects were apparent in the left (p = .23) or right region (p = .53).

475

The effects of statistical learning on P120 amplitudes gradually increased as the experiment progressed to later blocks, as reflected by a significant interaction between transitional probability, lateralisation, and experimental blocks (*F*[4, 136] = 2.78, *p* = 479 .029, $\eta^2 p = .08$). Post-hoc tests revealed that the amplitudes of P120 responses to

- 480 statistical deviants were larger in the third experimental block than in the first (p =
- 481 .002) and second blocks (p = .004). Furthermore, in the analysis of ERP responses over
- 482 the central region, the P120 response amplitudes elicited by statistical deviants were
- 483 significantly different from those elicited by standards during the third block (p < .001);
- 484 however, this was neither observed for the first (p = .15) nor the second block (p = .85).



495 Figure 2. Location MMN results. (a) Mean ERP responses to triplet endings as 496 recorded from the FZ electrode. Grey areas indicate the time windows used for 497 quantifying the P120 (\sim 100–140 ms) and location MMN (150–250 ms) components. 498 Averaged ERP responses to location standard (blue) and deviant (red) triplet endings, 499 as well as differences between them (black), are shown separately. (b) Isopotential 500 maps showing the scalp distributions of differences between the ERPs evoked by 501 location deviant triplet endings and those evoked by location standard triplet endings in 502 the control (left) and dyslexia (right) groups. (c) Interactive effects of dyslexia and 503 sound location at an anterior frontal region (average of F7, F5, F3, FT7, FC5, FC3, F1, FZ, 504 F2, FC1, FCZ, FC2, F8, F6, F4, FT8, FC6, and FC4). Location deviant triplet endings 505 elicited larger P120 amplitudes than location standard triplet endings in the control 506 group, but this was not the case in the dyslexia group. Significant location MMN effects 507 were observed in both groups but were larger in the control group than in the dyslexia

508 group. Error bars indicate the standard deviation of the mean. Abbreviations: ERP,



509 event-related potential; location MMN, location mismatch negativity.

519 Figure 3. Statistical MMN results. (a) Mean ERP responses to triplet endings as 520 recorded from the Fz electrode. The grey areas indicate the time windows used for 521 quantifying the P120 (\sim 100–140 ms) and sMMN (150–250 ms) component. Averaged 522 ERP responses to statistically standard (blue) and deviant (red) triplet endings, as well as differences between them (black), are shown separately. (b) Isopotential maps 523 524 showing the scalp distributions of differences between ERPs evoked by statistically 525 deviant triplet endings and those evoked by statistically standard triplet endings. (c) 526 Interactive effects of dyslexia and transitional probability at an anterior frontal region 527 (average of F7, F5, F3, FT7, FC5, FC3, F1, FZ, F2, FC1, FC2, FC2, F8, F6, F4, FT8, FC6, and 528 FC4). In the medial electrodes, statistically deviant triplet endings elicited larger P120 529 amplitudes than the statistically standard triplet endings in both groups. Significant 530 sMMN effects were observed in anterior frontal brain areas in the control group, but no

- 531 sMMN effects were observed in the dyslexia group. Error bars indicate standard
- 532 deviation of the mean. Abbreviations: ERP, event-related potential; sMMN, statistical
- 533 mismatch negativity.

534 Behavioural Results

535 During the exposition phase, the participants discovered on average 94.3% (*SD* = 536 .02) of the acoustical deviants (i.e., they showed a high performance in the cover task 537 where participants had to detect high pitched tones). This indicates that the 538 participants paid attention to the acoustical stimuli, while the task was relatively simple 539 to carry out.

540

541 At the end of the exposition phase in each block, the participants listened to 542 paired statistically deviant and standard triplets. A participant was asked to choose which triplet in the pair sounded more familiar and to rate their confidence in a given 543 544 answer on a scale ranging from 1 (no certainty) to 5 (certainty). To evaluate the 545 performance in the behavioural testing phase, we used a two-tailed *t*-test to determine 546 whether the frequency of correct answers during behavioural testing exceeded chance levels (i.e., p > .05). Further, we used ANOVAs to compare response accuracies, RT, and 547 548 confidence ratings between dyslexics and controls. ANOVAs were computed with Group 549 (dyslexia vs. control) as a between-subjects factor and experiment block as a within-550 subject factor (three blocks; the first, second and third blocks, instead of the actual six 551 blocks of the experiment to obtain a higher signal-to-noise ratio).

552

The two-tailed *t*-test revealed the frequency of correct answers was significantly higher than chance levels in the control group (p = .017, *Cohen's* d = .325) but not the dyslexic group (p = .520). ANOVAs revealed no effects on response accuracies and confidential rating in both the dyslexia and control groups (Figure 4, and see Appendix E). As for the reaction time, a significant main effect of the experimental block was noted (F[2, 68] = 7.92, p < .001, $\eta^2 p = .189$). Post-hoc tests revealed that the reaction

559	time in the second and last blocks was significantly faster than that in the first block
560	(2nd: $p = .006$; 3rd: $p < .001$). No other effect was found in the ANOVA of behavioural
561	results.
562	
563	To determine whether response accuracy rates correlated with confidence
564	ratings, we also conducted a Pearson correlation analysis. Response accuracies did not
565	correlate with confidence ratings in both the control ($r = -0.18$, p = .19) and dyslexia (r
566	= -0.05 , $p = .70$) groups. Similarly, response accuracies did not correlate with reading

- 567 and spelling test scores.



578 **Figure 4**. Behavioural testing results across experimental blocks. Error bars

579 indicate standard deviations. The ANOVA revealed no significance difference in

580 correctness ratios and reaction time between dyslexia and control groups.

581 **Discussion**

582 In this study, we assessed the relationship between developmental dyslexia and 583 ERP responses to a continuous stream of sound triplets in which some triplet endings 584 were either location deviants or statistical deviants. We found that statistical deviants 585 elicited small sMMN and that location deviants elicited a large location MMN. Compared 586 with controls, the participants with developmental dyslexia exhibited a smaller location 587 MMN and no sMMN. Whereas there was a tendency to a higher proportion of of correct 588 responses and decreased reaction times in the control compared to the dyslexia 589 group, there was no significant between-group difference for those two variables, 590 reflecting the ability to learn the statistical characteristics of stimuli. Our findings 591 indicate that the neural functions underlying the sMMN and the location MMN are 592 impaired in individuals with developmental dyslexia. This is in agreement with previous 593 evidence of a smaller MMN in developmental dyslexics (Gu & Bi, 2020; Lachmann et al., 594 2005; Neuhoff et al., 2012). Our findings also suggest that at the group level, the MMN 595 may be more sensitive to mild difficulties than behavioural measures. 596 597 The location MMN is a response reflecting auditory perceptual memory 598 operations that are instantly updated after new information is obtained (Bendixen, 599 Prinz, Horváth, Trujillo-Barreto, & Schröger, 2008; Sussman & Winkler, 2001). The 600 sMMN, on the other hand, relies on memory representations that are formed from the 601 implicit knowledge of sequential statistical structure (i.e., knowledge of stimulus 602 transitional probabilities; cf., e.g., Koelsch et al., 2016; Tsogli et al., 2019). Notably, the

603 sMMN effects elicited in our experiment were maximal at anterior frontal electrodes

604 (Figure 3b); however, the location MMN effect was more broadly distributed from

605 central to frontal areas (Figure 2b). This suggests the possibility that neural sources of 606 the sMMN are different from those of the location MMN. However, our findings showed 607 that people with dyslexia also exhibited smaller location MMN. 608 A past study found a diminished pitch MMN but an enhanced location MMN in 609 dyslexia (Kujala et al., 2006). We assumed that the combination of location and 610 statistical MMN in our paradigm may lead to different results from Kujala's study. 611 Another possibility is that the acoustic perceptual dysfunction of dyslexic 612 individuals also influenced neural processing of a sMMN. For example, past studies have 613 showed faster decay of dyslexics' perceptual memory trace behaviorally (Jaffe-Dax et al., 614 2017; Lieder et al., 2019) and in brain activity (Jaffe-Dax 2015, 2017), specifically in the 615 auditory cortex (Gertsovski & Ahissar, 2022 Perrachione et al., 2016). They proposed 616 that it reduces learning of complex rules (e.g. Virtala et al., 2021) and particularly 617 hampers acquisition of categories (Gertsovaki & Ahissar, 2022). 618 619 Both location MMN and sMMN were preceded by P120 responses. Compared 620 with the controls, the participants with developmental dyslexia exhibited no difference

621 in P120 responses between location deviants and standards. Previous studies have

622 reported the atypicalities of the P1 component in sound processing and dyslexia

623 (Stefanics et al., 2011) and implicit and statistical learning (Jaffe-Dax et al., 2017). The

624 P120 found in our study may be a P1-like component in terms of the positive

625 component temporally adjacent to the N1 or MMN effects. Further, given the

626 fundamental differences in the memory system between ERPs elicited by statistically

627 deviant stimuli and location deviants as stated above, the P120 components elicited by

628 statistically deviant stimuli may also be distinguishable from those of the P120

629 components elicited by location deviants. However, considering the difference in the

630 time window and that this effect was not hypothesised, it will not be discussed further

631 until future studies replicate this effect. Further studies are therefore necessary to

elucidate the neural basis underlying P120 components.

633

634 It has recently been considered that a domain-general statistical learning

635 impairment, rather than a specific impairment in phonological processing (Ramus et al.,

636 2003; Vellutino et al., 2004), may underlie developmental dyslexia. For example,

637 individuals with developmental dyslexia show weaker domain-general statistical

638 learning across sensory domains (Hung, Frost, & Pugh, 2018), such as auditory (Arciuli

639 & Conway, 2018; Dobó et al., 2021; Gabay et al., 2015; Kahta & Schiff, 2019;

640 Vandermosten et al., 2019) and visual stimuli (Sigurdardottir et al., 2017). Furthermore,

641 statistical learning impairment in developmental dyslexia is not limited to speech

642 stimuli but also occurs in non-speech stimuli (Plakas, van Zuijen, van Leeuwen,

643 Thomson, & van der Leij, 2013). Our results support these past findings; our study

644 explored effects of deviation from predictions based on physical and statistical

645 properties using non-speech auditory stimuli, and our findings showed that statistical

646 learning processes and basic auditory processes are affected in individuals with

647 developmental dyslexia.

648

649 Some new theoretical frameworks have been proposed to explain statistical

650 learning impairment in dyslexia. For example, a multicomponent memory network,

referred to as the Statistical Learning and Reading (SLR) model (Lee et al., 2022),

652 represents a domain-general system that consists of a short-term memory subsystem,

- an explicit declarative long-term memory subsystem, and an implicit procedural long-
- 654 term memory subsystem. The two long-term memory subsystems are distinguishable in

terms of the attentional demands required for encoding and storage of information,
with more controlled attention (i.e., a top-down selective attention to learning stimuli)
than automatic attention (i.e., a bottom-up involuntary attention to salient stimuli)
needed in the explicit declarative subsystem, and, conversely, more automatic attention
than controlled attention needed in the implicit procedural subsystem. The model
output is a statistically optimal representation as manifested by the neural and
behavioural response of statistical learning and reading activities.

662

663 In this study, participants conducted the behavioural testing at the end of each 664 block. In this experimental paradigm, it would be possible that they tended to pay more 665 attention to the tone sequence in the later blocks since they anticipated they may 666 experience a test later on. However, it is also important to see the time course of 667 statistical learning effects. Indeed, this study detected that participants did not show 668 statistical learning effects in the first block and a difference between groups. However, 669 the difference gradually became larger particularly in the second block whereafter it 670 stayed at about the same level (Figure 4, top). Further, the control group but not the 671 dyslexic group showed an above-chance level in behavioral performance of statistical 672 learning. However, no significant group difference may suggest the possibility that the 673 two groups learned the sequence equally. Further research is necessary to reveal why 674 and how the MMN is more sensitive to group difference than behavioural measures.

675

In conclusion, our findings lay forth evidence that location deviants elicit a
distinct classical MMN, which was larger in controls than in individuals with
developmental dyslexia. Statistical deviants elicited a sMMN in controls, whereas in
individuals with developmental dyslexia, sMMN was not recognizable. Our findings add

- to and are consistent with the so far scarce evidence showing that statistical learning
- and the underlying neural correlates may be impaired in individuals with
- 682 developmental dyslexia. Thus, exploring those neural correlates may contribute to a
- 683 better understanding of the cognitive processes underlying the acquisition of the rules
- and regularities that guide the arrangement of elements in ordered sequences such as
- 685 language and music.

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1005 Tables

1006 Table 1. Demographic and diagnostic data (raw scores) in the control and

1007 dyslexia groups.

	Control	Dyslexia	t(34)	р	Cohen's
					d
N (females)	19 (13)	17 (12)			
Age	25.6 (±3.3)	23.8 (±4.2)	1.41	.167	.472
Intelligence	28.0 (±3.7)	27.0 (±3.9)	0.79	.436	.263
Spelling	70.8 (±4.1)	46.5 (±9.6)	9.71	<.001	3.373
Text reading					
Reading	48.5 (±10.6)	32.5 (±8.9)	4.91	<.001	1.637
Comprehension					
Reading Speed	1190.7 (±232.8)	803.94 (±202.4)	5.29	<.001	1.766
(number					
of read words)					

1008 Note. Averages are reported as mean ± standard deviation. The intelligence scores and
 1009 spelling and reading test scores are raw scores.

1010

1012 Table 2. The 2x2 types of triplet endings based on location and statistical

1013 constraints.

		Transition probability				
		High (90%)	Low (10%)			
Sound	Standard (80%)	Standards (72%)	Statistical Deviant (8%)			
Location	Deviant (20%)	Location Deviant (18%)	Double Deviant (2%)			

1014

1016 Extended data

1017

1018 Appendix B. Mean ERP amplitudes under different conditions. ERP, event-related

- 1019 potential.
- 1020
- 1021 Appendix C. ANOVA results for the effects of location deviance. (a) Correlation
- analysis and ANOVA results for P120 component amplitudes and (b) ANOVA results for
- 1023 location MMN effects. Location MMN, location mismatch negativity.
- 1024
- 1025 Appendix D. ANOVA results for the effects of statistical deviance. (a) ANOVA results

1026 for P120 component amplitudes and (b) sMMN effects. sMMN, statistical mismatch

- 1027 negativity.
- 1028
- 1029 Appendix E. ANOVA results for data from a behavioural test assessing the 1030 statistical learning of stimulus characteristics.