

The development of children's ability to track and predict turn structure in conversation

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Abstract

Children begin developing turn-taking skills in infancy but take several years to assimilate their growing knowledge of language into their turn-taking behavior. In two eye-tracking experiments, we measured children's anticipatory gaze to upcoming responders while controlling linguistic cues to upcoming turn structure. In Experiment 1, we showed English and non-English conversations to English-speaking participants, finding minimal differences between the predictive looking behavior of preschoolers and adults. In Experiment 2, we phonetically controlled lexicosyntactic and prosodic cues in English-only speech, finding that children's predictive looking behavior improved from ages one to six, but that even one-year-olds made more anticipatory looks than would be expected by chance. In both experiments, children and adults anticipated more often after hearing questions. Like adults, prosody alone did not improve childrens predictive gaze shifts. But, unlike adults, lexical cues alone were also not sufficient to improve prediction—children's performance was best overall with access to lexicosyntax and prosody together. Our findings support an account in which turn prediction emerges in infancy, but takes several years before becoming fully integrated with linguistic processing.

Keywords: Turn taking, Conversation, Development, Prosody, Lexical, Questions, Eye-tracking, Anticipation

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¹ **1. Introduction**

² Spontaneous conversation is a universal context for using and learning
³ language. Like other types of human interaction, it is organized at its core
⁴ by the roles and goals of its participants. But what sets conversation apart is
⁵ its structure: Sequences of interconnected, communicative actions that take
⁶ place across alternating turns at talk. Sequential, turn-based structures in
⁷ conversation are strikingly uniform across language communities and linguis-
⁸ tic modalities. Turn-taking behaviors are also cross-culturally consistent in
⁹ their basic features and the details of their implementation (De Vos et al.,
¹⁰ 2015; Dingemanse et al., 2013; Stivers et al., 2009). How does this ability
¹¹ develop?

¹² Children participate in sequential coordination with their caregivers start-
¹³ ing at three months of age—before they can rely on any linguistic cues in
¹⁴ taking turns (see, among others, Bateson, 1975; Hilbrink et al., 2015; Jaffe
¹⁵ et al., 2001; Snow, 1977). Of course, infant turn taking is different from
¹⁶ adult turn taking in several ways. Infant turn taking is heavily scaffolded by
¹⁷ caregivers, has distinct timing in comparison to adult turn taking, and lacks
¹⁸ semantic content (Hilbrink et al., 2015; Jaffe et al., 2001). However, children’s
¹⁹ early, turn-structured social interactions are presumably a critical precursor
²⁰ to their conversational turn taking. Along these lines, early non-verbal inter-
²¹ actions establish the protocol by which children come to use language with
²² others. And as they acquire language, they also come to integrate it into
²³ their preverbal turn-taking systems.

²⁴ In this study, we investigate when children begin to make predictions
²⁵ about upcoming turn structure in conversation, and how they integrate lan-
²⁶ guage into their predictions as they grow older. In what follows, we first give
²⁷ a basic review of turn-taking research and the state of current knowledge
²⁸ about adult turn prediction. We then discuss recent work on the develop-
²⁹ ment of turn-taking skills before turning to the details of our own study.

³⁰ *1.1. Turn taking*

³¹ Turn taking itself is not unique to conversation. Many other human activ-
³² ities are organized around sequential turns at action. Traffic intersections and
³³ computer network communication both use turn-taking systems. Children’s
³⁴ early games (e.g., give-and-take, peek-a-boo) have built-in, predictable turn
³⁵ structure (Ratner and Bruner, 1978; Ross and Lollis, 1987). Even monkeys

36 take turns: Non-human primates such as marmosets and Campbell’s monkeys
37 vocalize contingently with each other in both natural and lab-controlled environments (Lemasson et al., 2011; Takahashi et al., 2013). In all these
38 cases, turn taking serves as a protocol for interaction, allowing the participants to coordinate with one another through sequences of contingent action.
39

40 Conversation distinguishes itself from non-conversational turn-taking behaviors by the complexity of the turn sequencing involved. In the examples
41 above (traffic, games, and monkeys) the set of sequence and action types is
42 far more limited and predictable than what we find in everyday talk. For
43 example, conversational turns come grouped into semantically-contingent se-
44 quences of action. The groups can span turn-by-turn exchanges (e.g., sim-
45 ple question-response, “How are you?”—“Fine.”) or sequence-by-sequence
46 exchanges (e.g., reciprocals, “How are you?”—“Fine, and you?”—“Great!”).
47 Sequences of action drive the conversation forward into the next, relevant se-
48 quences of talk (e.g., ”And you?”—“Great!”—“Why’s that?”; Schegloff, 2007).
49 To take a turn, participants need to make predictions about what conversa-
50 tional content will be relevant next. In some cases, relevant next turns are
51 somewhat obvious (e.g., question-response) while, in other cases, there are
52 multiple relevant next actions to choose from or no obvious next action at
53 all (e.g., after a closing).

54 Despite this complexity, conversational turn taking is often precise in its
55 timing. Across a diverse sample of conversations in 10 languages, one study
56 found a consistent average turn transition time of 0–200 msec at points of
57 speaker switch (Stivers et al., 2009). Experimental results and current models
58 of speech production suggest that it takes approximately 600 msec to produce
59 a content word, and even longer to produce a simple utterance (Griffin and
60 Bock, 2000; Levelt, 1989). So in order to achieve 200 msec turn transitions,
61 speakers must begin formulating their response before the prior turn has
62 ended (Levinson, 2013). Moreover, to formulate their response early on,
63 speakers must track and anticipate what types of response might become
64 relevant next. They also need to predict the content and form of upcoming
65 speech so that they can launch their articulation at exactly the right moment.
66 Prediction thus plays a key role in timely turn taking.

67
68

69 1.2. Adults’ turn prediction

70 Adults have a lot of information at their disposal to help make accurate
71 predictions about upcoming turn content. Lexical, syntactic, and prosodic
72 information (e.g., *wh*- words, subject-auxiliary inversion, and list intonation)

73 can all inform addressees about upcoming linguistic structure (De Ruiter
74 et al., 2006; Duncan, 1972; Ford and Thompson, 1996; Torreira et al., 2015).
75 Non-verbal cues (e.g., gaze, posture, and pointing) often appear at turn-
76 boundaries and can sometimes act as late indicators of an upcoming speaker
77 switch (Rossano et al., 2009; Stivers and Rossano, 2010). Additionally, the
78 sequential context of a turn can make it clear what will come next: An-
79 swers after questions, thanks or denial after compliments, et cetera (Schegloff,
80 2007).

81 Prior work suggests that adult listeners primarily use lexicosyntactic in-
82 formation to accurately predict upcoming turn structure (De Ruiter et al.,
83 2006). De Ruiter and colleagues (2006) asked participants to listen to snip-
84 pets of spontaneous conversation and to press a button whenever they antici-
85 pated that the current speaker was about to finish his or her turn. The speech
86 snippets were controlled for the amount of linguistic information present;
87 some were normal, but others had flattened pitch, low-pass filtered speech,
88 or further manipulations. De Ruiter and colleagues found that, with pitch-
89 flattened speech, the timing of participants' button responses was comparable
90 to their timing with the full linguistic signal. But when no lexical information
91 was available, participants' responses were significantly earlier. The authors
92 concluded that lexicosyntactic information¹ was necessary and possibly suf-
93 ficient for turn-end projection, while intonation was neither necessary nor
94 sufficient. Congruent evidence comes from studies varying the predictability
95 of lexicosyntactic and pragmatic content: Adults anticipate turn ends better
96 when they can more accurately predict the exact words that will come next
97 (Magyari and De Ruiter, 2012; see also Magyari et al., 2014). They can also
98 identify speech acts within the first word of an utterance (Gísladóttir et al.,
99 2015), allowing them to start planning their response at the first moment
100 possible (Bögels et al., 2015).

101 The role of prosody for adult turn prediction is still a matter of de-
102 bate. De Ruiter and colleagues' (2006) experiment focused on the role of
103 intonation, which is only a partial index of prosody. Prosodic structure is
104 also tied closely to the syntax of an utterance, and so the two linguistic
105 signals are difficult to control independently (Ford and Thompson, 1996).

¹The “lexicosyntactic” condition only included flattened pitch and so was not exclusively lexicosyntactic—the speech would still have residual prosodic structure, including syllable duration and intensity.

106 Torreira, Bögels and Levinson (2015) used a combination of button-press
107 and verbal responses to investigate the relationship between lexicosyntac-
108 tic and prosodic cues in turn-end prediction. Critically, their stimuli were
109 cross-spliced so that each item had full prosodic cues to accompany the lex-
110 icosyntax. Because of the splicing, they were able to create items that had
111 syntactically-complete units with no intonational phrase boundary at the
112 end. Participants never verbally responded or pressed the “turn-end” but-
113 ton when hearing a syntactically-complete phrase without an intonational
114 phrase boundary. And when intonational phrase boundaries were embedded
115 in multi-utterance turns, participants were tricked into pressing the “turn-
116 end” button 29% of the time. Their results suggest that listeners actually
117 do rely on prosodic cues to execute a response (see also de De Ruiter et al.
118 (2006):525). These experimental findings corroborate other corpus and ex-
119 perimental work promoting a combination of cues (lexicosyntactic, prosodic,
120 and pragmatic) as key for accurate turn-end prediction (Duncan, 1972; Ford
121 and Thompson, 1996; Hirvenkari et al., 2013).

122 In sum, adults accurately and spontaneously make predictions about up-
123 coming turn structure. Their predictions rely on a sophisticated body of
124 knowledge about linguistic structure, non-verbal signals, and social actions.
125 Knowing this, we could expect that children’s acquisition of turn-taking skills
126 is closely tied to their knowledge about language, gaze, gesture, and social
127 cues. But children’s turn taking starts early in infancy, long before their first
128 words or gestures emerge. So a primary role for lexicosyntactic cues doesn’t
129 fit well with children’s pre-verbal turn taking.

130 *1.3. Children’s turn prediction*

131 *1.3.1. Observational studies*

132 The majority of work on children’s early turn taking has focused on ob-
133 servations of spontaneous interaction. Children’s first turn-like structures
134 appear as early as two to three months in proto-conversation with their care-
135 givers (Bruner, 1975, 1985). During proto-conversations, caregivers interact
136 with their infants as if they were capable of making meaningful contributions;
137 they take every look, vocalization, arm flail, and burp as “utterances” in the
138 joint discourse (Bateson, 1975; Jaffe et al., 2001; Snow, 1977). Infants catch
139 onto the structure of proto-conversations quickly. By three to four months
140 they notice disturbances to the contingency of their caregivers’ response and,
141 in reaction, change the rate and quality of their vocalizations (Bloom, 1988;
142 Masataka, 1993). Infants at this age also notice changes to social contingency

143 outside of turn structure. In the Still Face paradigm, caregivers interact with
144 their infants and then suddenly halt, taking on a neutral expression with a
145 sustained gaze. When faced with this sudden disappearance of social contin-
146 gency, infants three months and older try a range of methods to reinitiate the
147 interaction, such as vocalization, reaching, and smiling before looking away
148 or getting upset (Rochat et al., 1998; Toda and Fogel, 1993).

149 The timing of children's responses to their caregivers' speech shows a
150 non-linear pattern of fall-rise-fall from early infancy to middle childhood. A
151 recent study by Hilbrink et al. (2015) finds that infants' turn timing at three
152 months is often too early or too late: They start vocalizing in overlap on
153 40% of their caregivers' turns, and their non-overlapped vocalizations come
154 after an average inter-turn silent gap of 350–900 msec (adult average: 200
155 msec). Between four and nine months, children begin to reduce the number
156 of turns happening in overlap while also improving on their average response
157 latency. But then, later on, children's response latencies slow down again,
158 peaking at average gaps of more than 1000 msec at nine months, with only
159 very gradual improvement after that (Hilbrink et al., 2015). While children's
160 avoidance of overlap is nearly adult-like by nine months, the timing of their
161 non-overlapped responses stays much longer than the 200 msec standard for
162 the next few years (Casillas et al., In press; Garvey, 1984; Ervin-Tripp, 1979).

163 The protracted development of children's timing may be attributable to
164 their linguistic development: Taking turns on time is easier when the response
165 is a simple vocalization rather than a linguistic utterance. Integrating lan-
166 guage into the turn-taking system may be one major factor in children's de-
167 layed responses (Casillas et al., In press). If response planning (i.e., language
168 production) is the primary hurdle in children's spontaneous turn taking, we
169 should find evidence that children understand turn-taking behaviors before
170 they are able to produce the behaviors themselves; this hypothesis has been
171 recently explored in experimental settings.

172 *1.3.2. Experimental studies*

173 Children begin to develop specific expectations about conversational be-
174 havior before they begin to speak. Sometime between four and six months,
175 children begin to attend differently to face-to-face and back-to-back conver-
176 sation; six-month-olds follow conversational speakers more with their gaze
177 when at least one speaker is looking at the other (Augusti et al., 2010). At
178 ten months, infants expect people to look and talk at other people, and not
179 to objects (Beier and Spelke, 2012). At twelve months infants expect to see

180 responses to verbal (but not non-speech) utterances in face-to-face contexts
181 (Thorgrímsson et al., 2015).

182 There are mixed results regarding when children begin to anticipate turn
183 structure in conversation. One study found that 12-month-olds make more
184 predictive gaze shifts to a responder while watching human verbal conversa-
185 tion compared to conversation-like interactions with objects (Bakker et al.,
186 2011), but another only found a similar effect at 36 months (von Hofsten
187 et al., 2009). However, neither of these two studies had baselines to which the
188 turn-relevant looking behavior could be compared. A baseline measurement
189 is critical because there may be developmental differences in gaze shifting be-
190 tween conversational participants, even if the shifting is not related to turn
191 structure. Such developmental differences could produce artifactual changes
192 in measures of turn-contingent shifting.

193 Keitel and colleagues (2013) addressed the random baseline issue in their
194 study of 6-, 12-, 24-, and 36-month-olds. They asked participants to watch
195 short videos of conversation and tracked their eye movements at points of
196 speaker change. They found that children's anticipatory gaze frequency was
197 only greater than chance for 36-month-olds and adults. Their study was the
198 first to focus on the role of linguistic processing in children's turn predictions.
199 They showed their participants two types of conversation videos: One nor-
200 mal and one with flattened pitch (i.e., with flattened intonation contours),
201 finding that only 36-month-olds were affected by a lack of intonation con-
202 tours. The adult control group made equal numbers of anticipatory looks
203 in the videos, with and without intonation contours, consistent with prior
204 adult findings (De Ruiter et al., 2006). Keitel and colleagues concluded that
205 children's ability to predict upcoming turn structure relies on their ability
206 to comprehend the stimuli (emerging around 36 months), especially with re-
207 spect to semantic access. They also suggest that intonation takes a secondary
208 role in turn prediction, but only *after* children acquire more sophisticated,
209 adult-like language comprehension systems (sometime after 36 months).

210 Although the Keitel et al. (2013) study constitutes a substantial advance
211 over previous work, it has its own limitations. Because these limitations
212 directly inform our own study design, we review them in some detail. First,
213 their estimates of baseline gaze frequency ("random" in their terminology)
214 were not random. Instead, they used gaze switches during ongoing speech
215 as a baseline, during which switching is least likely to occur (Hirvenkari
216 et al., 2013) and thereby maximizing their chances of finding a difference
217 between gaze frequency at turn transitions and their baseline rate. A more

218 conservative baseline would be to compare participants' looking behavior at
219 turn transitions to their looking behavior during randomly selected windows
220 of time throughout the stimulus, including turn transitions. We follow this
221 conservative approach in our work.

222 Second, the conversation stimuli they used were somewhat unusual. The
223 average gap between turns was 900 msec, which is much longer than typical
224 adult timing, where gaps average around 200 msec (Stivers et al., 2009). The
225 speakers in the videos were also asked to minimize their movements while
226 performing a scripted and adult-directed conversation, which would have
227 created a somewhat unnatural stimulus. Additionally, in order to produce
228 more naturalistic conversation, it would have been ideal to localize the sound
229 sources for the two voices in the video (i.e., to have the voices come out of
230 separate left and right speakers). But both voices were recorded and played
231 back on the same audio channel, which may have made it more difficult to
232 distinguish the two talkers. Again, we attempt to address these issues in our
233 current study.

234 Despite these minor methodological issues, the Keitel et al. (2013) study
235 still demonstrates intriguing age-based differences in children's ability to pre-
236 dict upcoming turn structure, and the results suggest that both semantic and
237 intonational development *do* play a role in children's looking patterns. Our
238 current work thus takes this paradigm as a starting point.²

239 1.3.3. Prosodic development

240 The roles of prosody and lexicosyntax in children's turn predictions are
241 currently unknown, but children understand more about prosody than lex-
242 icosyntax early in life. Children begin to acquire prosody in the womb,
243 and can distinguish their native language's rhythm type from others (e.g.,
244 syllable-timed vs. stress-timed) 2–5 days after birth (Mehler et al., 1988;
245 Moon et al., 1993; Nazzi and Ramus, 2003). Beginning between four and five
246 months, infants prefer pauses in speech to be inserted at prosodic bound-
247 aries, and by 6 months they can start using prosodic markers to pick out
248 sub-clausal syntactic units (Jusczyk et al., 1995; Soderstrom et al., 2003).
249 They show preference for the typical stress patterns of their native language
250 over others by 6–9 months (e.g., iambic vs. trochaic), and can use prosodic
251 information to segment the speech stream into smaller chunks from 8 months

²See also Casillas and Frank (2012, 2013).

252 onward (Johnson and Jusczyk, 2001; Jusczyk et al., 1993; Morgan and Saf-
253 fran, 1995). In comparison, children show only a limited lexical inventory at
254 six months, and begin to recognize function words just before their first birth-
255 days, with syntactic categorization beginning around 14 months (Bergelson
256 and Swoley, 2013; Shi and Melancon, 2010). Two-month-olds also notice
257 changes in word order, but this ability appears to rely on prosodic cue-
258 ing (Mandel et al., 1996). Generally speaking then, our current knowledge
259 about children’s linguistic development points to a possible early advantage
260 for prosody in children’s turn-taking predictions.

261 We report here on the role of linguistic processing in children’s predictions
262 about upcoming turn structure, in particular on how children use prosodic
263 and lexicosyntactic information to make their predictions. Prior work has
264 focused mainly on lexicosyntax and intonation, and not on prosody proper
265 (De Ruiter et al., 2006; Keitel et al., 2013, but see Torreira et al., 2015), even
266 though infants seem to acquire the basic rhythmic properties of the prosodic
267 signal first (Mehler et al., 1988; Moon et al., 1993; Nazzi and Ramus, 2003).

268 2. Experiment 1

269 We recorded participants’ eye movements as they watched six short videos
270 of two-person (dyadic) conversation interspersed with attention-getting filler
271 videos. Each conversation video featured an improvised discourse in one of
272 five languages (English, German, Hebrew, Japanese, and Korean); partici-
273 pants saw two videos in English and one in every other language. The partici-
274 pants, all native English speakers, were only expected to understand the two
275 videos in English. We showed participants non-English videos to limit their
276 access to lexical information while maintaining their access to other cues to
277 turn boundaries (e.g., (non-native) prosody, gaze, breath, phrase final length-
278 ening). Using this method, we compared children and adult’s anticipatory
279 looks from the current speaker to the upcoming speaker at points of turn
280 transition in English and non-English videos.

281 2.1. Methods

282 2.1.1. Participants

283 We recruited 74 children between ages 3;0–5;11 and 11 undergraduate
284 adults to participate in the experiment. Our child sample included 19 three-
285 year-olds, 32 four-year-olds, and 23 five-year-olds, all enrolled in a local nurs-
286 ery school. All participants were native English speakers. Approximately



Figure 1: Example frame from a conversation video used in Experiment 1.

287 one-third (N=25) of the children’s parents and teachers reported that their
288 child regularly heard a second (and sometimes third or further) language, but
289 only one child frequently heard a language that was used in our non-English
290 video stimuli, and we excluded his data from analyses. None of the adult
291 participants reported fluency in a second language.

292 *2.1.2. Materials*

293 *Video recordings.* We recorded pairs of talkers while they conversed in
294 a sound-attenuated booth (see sample frame in Figure 1). Each talker was
295 a native speaker of the language being recorded, and each talker pair was
296 male-female. Using a Marantz PMD 660 solid state field recorder, we cap-
297 tured audio from two lapel microphones, one attached to each participant,
298 while simultaneously recording video from the built-in camera of a MacBook
299 laptop computer. The talkers were volunteers and were acquainted with their
300 recording partner ahead of time.

301 Each recording session began with a 20-minute warm-up period of spon-
302 taneous conversation during which the pair talked for five minutes on four
303 topics (favorite foods, entertainment, hometown layout, and pets). Then we
304 asked talkers to choose a new topic—one relevant to young children (e.g.,
305 riding a bike, eating breakfast)—and to improvise a dialogue on that topic.
306 We asked them to speak as if they were on a children’s television show in
307 order to elicit child-directed speech toward each other. We recorded until the
308 talkers achieved at least 30 seconds of uninterrupted discourse with enthу-
309 siastic, child-directed speech. Most talker pairs took less than five minutes
310 to complete the task, usually by agreeing on a rough script at the start. We

311 encouraged talkers to ask at least a few questions to each other during the
312 improvisation. The resulting conversations were therefore not entirely spontane-
313 anous, but were as close as possible while still remaining child-oriented in
314 topic, prosodic pattern, and lexicosyntactic construction.³

315 After recording, we combined the audio and video files by hand, and
316 cropped each recording to the 30-second interval with the most turn activity.
317 Because we recorded the conversations in stereo, the male and female voices
318 came out of separate speakers during video playback. This gave each voice in
319 the videos a localized source (from the left or right loudspeaker). We coded
320 each turn transition in the videos for language condition (English vs. non-
321 English), inter-turn gap duration (in milliseconds), and speech act (question
322 vs. non-question). The non-English stimuli were coded for speech act from
323 a monolingual English-speaker’s perspective, i.e., which turns “sound like”
324 questions, and which don’t: we asked five native American English speakers
325 to listen to the audio signal for each turn and judge whether it sounded
326 like a question. We then coded turns with at least 80% “yes” responses as
327 questions.

328 Because the conversational stimuli were recorded semi-spontaneously, the
329 duration of turn transitions and the number of speaker transitions in each
330 video was variable. We measured the duration of each turn transition from
331 the audio recording associated with each video. We excluded turn transi-
332 tions longer than 550 msec and shorter than 90 msec, including over-
333 lapped transitions, from analysis.⁴ This left approximately equal numbers
334 of turn transitions available for analysis in the English (N=20) and non-
335 English (N=16) videos. On average, the inter-turn gaps for English videos
336 (mean=318, median=302, stdev=112 msec) were slightly longer than for non-
337 English videos (mean=286, median=251, stdev=122 msec). The longer gaps
338 in the English videos could give them a slight advantage: Our definition of

³All of the non-English talkers were fluent in English as a second language, and some fluently spoke three or more languages. We chose male-female pairs as a natural way of creating contrast between the two talker voices.

⁴Overlap occurs when a responder begins a new turn before the current turn is finished. When overlap occurs, observers cannot switch their gaze in anticipation of the response because the response began earlier than expected; participants expect conversations to proceed with “one speaker at a time” (Sacks et al., 1974). As such, they would still be fixated on the prior speaker when the overlap started, and then would have to switch their gaze *reactively* to the responder.

339 an “anticipatory gaze shift” includes shifts that are initiated during the gap
340 between turns (Figure 2), so participants had slightly more time to make
341 anticipatory shifts in the English videos.

342 Questions made up exactly half of the turn transitions in the English
343 ($N=10$) and non-English ($N=8$) videos. In the English videos, inter-turn
344 gaps were slightly shorter for questions (mean=310, median=293, stdev=112
345 msec) than non-questions (mean=325, median=315, stdev=118 msec). Non-
346 English videos did not show a large difference in transition time for questions
347 (mean=270, median=257, stdev=116 msec) and non-questions (mean=302,
348 median=252, stdev=134 msec).

349 *2.1.3. Procedure*

350 Participants sat in front of an SMI 120Hz corneal reflection eye-tracker
351 mounted beneath a large flatscreen display. The display and eye-tracker were
352 secured to a table with an ergonomic arm that allowed the experimenter to
353 position the whole apparatus at a comfortable height, approximately 60 cm
354 from the viewer. We placed stereo speakers on the table, to the left and right
355 of the display.

356 Before the experiment started, we warned adult participants that they
357 would see videos in several languages and that, though they weren’t expected
358 to understand the content of non-English videos, we *would* ask them to an-
359 swer general, non-language-based questions about the conversations. Then
360 after each video we asked participants one of the following randomly-assigned
361 questions: “Which speaker talked more?”, “Which speaker asked the most
362 questions?”, “Which speaker seemed more friendly?”, and “Did the speak-
363 ers’ level of enthusiasm shift during the conversation?” We also asked if the
364 participants could understand any of what was said after each video. The
365 participants responded verbally while an experimenter noted their responses.

366 Children were less inclined to simply sit and watch videos of conversation
367 in languages they didn’t speak, so we used a different procedure to keep them
368 engaged: The experimenter started each session by asking the child about
369 what languages he or she could speak, and about what other languages he
370 or she had heard of. Then the experimenter expressed her own enthusiasm
371 for learning about new languages, and invited the child to watch a video
372 about “new and different languages” together. If the child agreed to watch,
373 the experimenter and the child sat together in front of the display, with
374 the child centered in front of the tracker and the experimenter off to the
375 side. Each conversation video was preceded and followed by a 15–30 second

376 attention-getting filler video (e.g., running puppies, singing muppets, flying
377 bugs). If the child began to look bored, the experimenter would talk during
378 the fillers, either commenting on the previous conversation (“That was a neat
379 language!”) or giving the language name for the next conversation (“This
380 next one is called Hebrew. Let’s see what it’s like.”) The experimenter’s
381 comments reinforced the video-watching as a joint task.

382 All participants (child and adult) completed a five-point calibration rou-
383 tine before the first video started. We used a dancing Elmo for the children’s
384 calibration image. During the experiment, participants watched all six 30-
385 second conversation videos. The first and last conversations were in American
386 English and the intervening conversations were Hebrew, Japanese, German,
387 and Korean. The presentation order of the non-English videos was shuffled
388 into four lists, which participants were assigned to randomly. The entire
389 experiment, including instructions, took 10–15 minutes.

390 *2.1.4. Data preparation and coding*

391 To determine whether participants predicted upcoming turn transitions,
392 we needed to define a set of criteria for what counted as an anticipatory gaze
393 shift. Prior work using similar experimental procedures has found that adults
394 and children make anticipatory gaze shifts to upcoming talkers within a wide
395 time frame; the earliest shifts occur before the end of the prior turn, and the
396 latest occur after the onset of the response turn, with most shifts occurring
397 in the inter-turn gap (Keitel et al., 2013; Hirvenkari, 2013; Tice and Henetz,
398 2011). Following prior work, we measured how often our participants shifted
399 their gaze from the prior to the upcoming speaker *before* the shift in gaze
400 could have been initiated in reaction to the onset of the speaker’s response.
401 In doing so, we assumed that it takes participants 200 msec to plan an eye
402 movement, following standards from adult anticipatory processing studies
403 (e.g., Kamide et al., 2003).

404 We checked each participant’s gaze at each turn transition for three char-
405 acteristics (Figure 2): (1) That the participant fixated on the prior speaker
406 for at least 100 msec at the end of the prior turn, (2) that sometime thereafter
407 the participant switched to fixate on the upcoming speaker for at least 100
408 ms, and (3) that the switch in gaze was initiated within the first 200 msec of
409 the response turn, or earlier. These criteria guarantee that we only counted
410 gaze shifts when: (1) Participants were tracking the previous speaker, (2)
411 switched their gaze to track the upcoming speaker, and (3) did so before
412 they could have simply reacted to the onset of speech in the response. Under

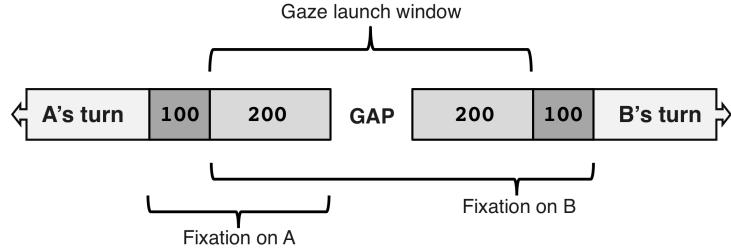


Figure 2: Schematic summary of criteria for anticipatory gaze shifts from speaker A to speaker B during a turn transition.

413 this assumption, a gaze shift that was initiated within the first 200 msec of
 414 the response (or earlier) was planned *before* the child could react to the onset
 415 of speech itself.

416 As mentioned, most anticipatory switches happen in the inter-turn gap,
 417 but we also allowed anticipatory gaze switches that occurred in the final
 418 syllables of the prior turn. Early switches are consistent with the distribution
 419 of responses in explicit turn-boundary prediction tasks. For example, in
 420 a button press task, adult participants anticipate turn ends approximately
 421 200 msec in advance of the turn’s end, and anticipatory responses to pitch-
 422 flattened stimuli come even earlier (De Ruiter et al., 2006). We therefore
 423 allowed switches to occur as early as 200 msec before the end of the prior turn.
 424 For very early and very late switches, our requirement for 100 msec of fixation
 425 on each speaker would sometimes extend outside of the transition window
 426 boundaries (200 msec before and after the inter-turn gap). The maximally
 427 available fixation window was 100 msec before and after the earliest and
 428 latest possible switch point (300 msec before and after the inter-turn gap).
 429 We did not count switches made during the fixation window as anticipatory.
 430 We *did* count switches made during the inter-turn gap. The period of time
 431 from the beginning of the possible fixation window on the prior speaker to the
 432 end of the possible fixation window on the responder was our total analysis
 433 window (300 msec + the inter-turn gap + 300 msec).

434 *Predictions.* We expected participants to show greater anticipation in the
 435 English videos than in the non-English videos because of their increased
 436 access to linguistic information in English. We also predicted that anticipa-
 437 tion would be greater following questions compared to non-questions; ques-
 438 tions have early cues to upcoming turn transition (e.g., *wh*- words, subject-

Age group	Condition	Speaker	Addressee	Other onscreen	Offscreen
3	English	0.61	0.16	0.14	0.08
4	English	0.60	0.15	0.11	0.13
5	English	0.57	0.15	0.16	0.12
Adult	English	0.63	0.16	0.16	0.05
3	Non-English	0.38	0.17	0.20	0.25
4	Non-English	0.43	0.19	0.21	0.18
5	Non-English	0.40	0.16	0.26	0.18
Adult	Non-English	0.58	0.20	0.16	0.07

Table 1: Average proportion of gaze to the current speaker and addressee during periods of talk.

auxiliary inversion), and also make a next response immediately relevant. Our third prediction was that anticipatory looks would increase with development, along with children’s increased linguistic competence.

2.2. Results

Participants looked at the screen most of the time during video playback (81% and 91% on average for children and adults, respectively). They primarily kept their eyes on the person who was currently speaking in both English and non-English videos: They gazed at the current speaker between 38% and 63% of the time, looking back at the addressee between 15% and 20% of the time (Table 1). Even three-year-olds looked more at the current speaker than anything else, whether the videos were in a language they could understand or not. Children looked at the current speaker less than adults did during the non-English videos. Despite this, their looks to the addressee did not increase substantially in the non-English videos, indicating that their looks away were probably related to boredom rather than confusion about ongoing turn structure. Overall, participants’ pattern of gaze to current speakers demonstrated that they performed basic turn tracking during the videos, regardless of language.

2.2.1. Statistical models

We identified anticipatory gaze switches for all 36 usable turn transitions, based on the criteria outlined in Section 2.1.4, and analyzed them for effects

<i>Children</i>	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-0.96146	0.84901	-1.132	0.257446
Age	-0.18268	0.17507	-1.043	0.296725
LgCond= <i>non-English</i>	-3.29347	0.96045	-3.429	0.000606 ***
Type= <i>non-Question</i>	-1.10129	0.86494	-1.273	0.202925
Duration	3.40169	1.22826	2.770	0.005614 **
Age*LgCond= <i>non-English</i>	0.52065	0.21190	2.457	0.014008 **
Age*TypeS= <i>non-Question</i>	-0.01628	0.19437	-0.084	0.933232
LgCond= <i>non-English</i> *	2.68166	1.35016	1.986	0.047013 *
Type= <i>non-Question</i>				
Age*LgCond= <i>non-English</i> *	-0.45632	0.30163	-1.513	0.130315
Type= <i>non-Question</i>				

<i>Adults</i>	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-0.1966	0.6942	-0.283	0.776988
LgCond= <i>non-English</i>	-0.8812	0.9602	-0.918	0.358754
Type= <i>non-Question</i>	-4.4953	1.3139	-3.421	0.000623 ***
Duration	-1.1227	1.9880	-0.565	0.572238
LgCond= <i>non-English</i> *	3.2972	1.6101	2.048	0.040581 *
Type= <i>non-Question</i>				
LgCond= <i>non-English</i> *	1.3626	3.0077	0.453	0.650527
Duration				
Type= <i>non-Question</i> *	10.5107	3.3459	3.141	0.001682 **
Duration				
LgCond= <i>non-English</i> *	-6.3156	4.4926	-1.406	0.159790
Type= <i>non-Question</i> *				
Duration				

Table 2: Model output for children and adults’ anticipatory gaze switches.

of language, transition type, and age with two mixed-effects logistic regressions (Bates et al., 2014; R Core Team, 2014). We built one model each for children and adults. We modeled children and adults separately because effects of age are only pertinent to the children’s data. The child model included condition (English vs. non-English)⁵, transition type (question vs. non-question), age (3, 4, 5), and duration of the inter-turn gap (seconds, e.g., 0.441) as predictors, with full interactions between condition, transition

⁵Because each non-English language was represented by a single stimulus, we cannot treat individual languages as factors. Gaze behavior might be best for non-native languages that have the most structural overlap with participants’ native language: English speakers can make predictions about the strength of upcoming Swedish prosodic boundaries nearly as well as Swedish speakers do, but Chinese speakers are at a disadvantage in the same task (Carlson et al., 2005). We would need multiple items from each of the languages to check for similarity effects of specific linguistic features.

467 type, and age. We included the duration of the inter-turn gap as a predictor
468 since longer gaps also provide more opportunities to make anticipatory
469 switches (Figure 2). We additionally included random effects of item (turn
470 transition) and participant, with random slopes of condition, transition type,
471 and their interaction for participants (Barr et al., 2013).⁶ The adult model
472 included condition, transition type, duration, and their interactions as pre-
473 dictors with participant and item included as random effects and random
474 slopes of condition, transition type, and their interaction for participant.

475 Children’s anticipatory gaze switches showed effects of language condition
476 ($\beta=-3.29$, $SE=0.961$, $t=-3.43$, $p<.001$) and gap duration ($\beta=3.4$, $SE=1.229$,
477 $t=2.77$, $p<.01$) with additional effects of an age-by-language condition in-
478 teraction ($\beta=0.52$, $SE=0.212$, $t=2.46$, $p<.05$) and a language condition-by-
479 transition type interaction ($\beta=2.68$, $SE=1.35$, $t=1.99$, $p<.05$). There were
480 no significant effects of age or transition type alone ($\beta=-0.18$, $SE=0.175$,
481 $t=-1.04$, $p=.3$ and $\beta=-1.10$, $SE=0.865$, $t=-1.27$, $p=.2$, respectively).

482 Adults’ anticipatory gaze switches shows an effect of transition type ($\beta=$
483 4.5 , $SE=1.314$, $t=-3.42$, $p<.001$) and significant interactions between lan-
484 guage condition and transition type ($\beta=3.3$, $SE=1.61$, $t=2.05$, $p<.05$) and
485 transition type and gap duration ($\beta=10.51$, $SE=3.346$, $t=3.141$, $p<.01$).

486 2.2.2. Random baseline comparison

487 We estimated the probability that these patterns were the result of ran-
488 dom looking by running the same regression models on participants’ real
489 eye-tracking data, only this time calculating their anticipatory gaze switches
490 with respect to randomly permuted turn transition windows. This process
491 involved: (1) randomizing the order and temporal placement of the anal-
492 ysis windows within each stimulus (Figure 3; “analysis window” is defined
493 in Figure 2), thereby randomly redistributing the analysis windows across
494 the eye-tracking signal, (2) re-running each participant’s eye tracking data
495 through switch identification (described in 2.1.4), this time using the ran-
496 domly permuted analysis windows, and (3) modeling the anticipatory gazes
497 from the randomly permuted data with the same statistical models we used
498 for the original data (Section 2.2.1; Table 2). Importantly, although the onset
499 time of each transition was shuffled within the eye-tracking signal, the other

⁶The models we report are all qualitatively unchanged by the exclusion of their random slopes. We have left the random slopes in because of minor participant-level variation in the predictors modeled.

500 intrinsic properties of each turn transition (e.g., prior speaker identity, transition type, gap duration, language condition, etc.) stayed constant across
 501 each random permutation.
 502

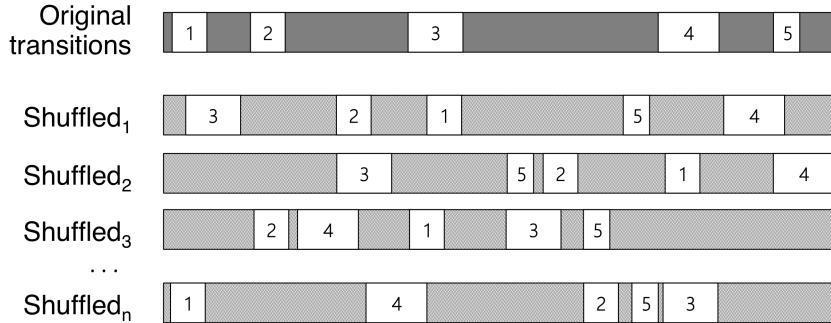


Figure 3: Example of shuffling for five turn transition analysis windows. The windows were ± 300 msec around the inter-turn gap.

503 This procedure effectively de-links participants' gaze data from the turn
 504 structure in the original stimulus, thereby allowing us to compare turn-
 505 related (original) and non-turn-related (randomly permuted) looking behav-
 506 ior using the same eye movements. The resulting anticipatory gazes from the
 507 randomly permuted analysis windows represent an average anticipatory gaze
 508 rate over all possible starting points: a random baseline. By running the real
 509 and randomly permuted data sets through identical statistical models, we
 510 can also estimate how likely it is that predictor effects in the original data
 511 (e.g., the effect of language condition; Table 2) arose from random looking.

512 We completed this baseline procedure on 5,000 random permutations of
 513 the original turn transition analysis windows and compared the t -values from
 514 each predictor in the original models (Table 2) to the distribution of t -values
 515 for each predictor in the 5,000 models of the randomly permuted datasets.⁷
 516 We could then test whether significant effects from the original statistical
 517 models differed from the random baseline by calculating the proportion of
 518 random data t -values exceeded by the original t -value for each predictor,
 519 using the absolute value of all t -values for a two-tailed test. For example,

⁷We report t -values rather than beta estimates because the standard errors in the randomly permuted data models were much higher than for the original data. For those interested, plots of the beta and standard error distributions are available in the Appendix.

520 children's original "language condition" t -value was $|3.429|$, which is greater
 521 than 99.9% of all $|t\text{-value}|$ estimates from the randomly-permuted data mod-
 522 els (i.e., $p = .001$). This leads us to conclude that the effect of language
 523 condition in the original model was highly unlikely to be the result of random
 524 gaze shifting.

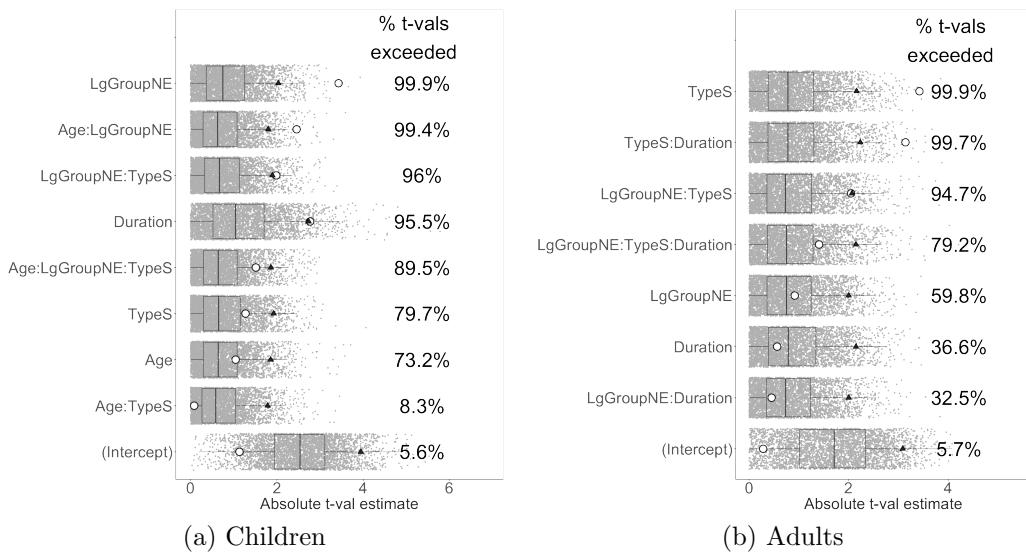


Figure 4: Random-permutation and original $|t$ -values for predictors of children and adults' anticipatory gaze rates. Gray dots = random model estimates, White dots = original model estimates, Triangles = 95th percentile for each t -value distribution.

525 Our baseline analyses revealed that none of the significant predictors from
 526 models of the original, turn-related data can be explained by random looking.
 527 The children's data showed strong evidence of differentiation from the ran-
 528 domly permuted data for all four significant effects in the original model (Ta-
 529 ble 2: Children): the original t -values for language condition, gap duration,
 530 the age-language condition interaction, and the language condition-transition
 531 type interaction were all greater than 95% of t -values for the randomly
 532 permuted data (99.9%, 95.5%, 99.4%, and 96%, respectively; Figure 4a).
 533 Similarly, the adults' data showed significant differentiation from the ran-
 534 domly permuted data for two of the three originally significant predictors—
 535 transition type and the transition type-gap duration interaction (greater than

536 99.9% and 99.7% of random t -values, respectively)—with marginal differen-
 537 tiation for the interaction of language condition and transition type (greater
 538 than 94.7% of random t -values; Figure 4b). The effects of language condi-
 539 tion and transition type for the real and randomly permuted data can also
 540 be observed in Figure 5. We excluded the output of random-permutation
 541 models that did not ultimately converge to remove unreliable model results
 542 from our percentile calculations below (78% and 76% of models for children
 543 and adults, respectively).

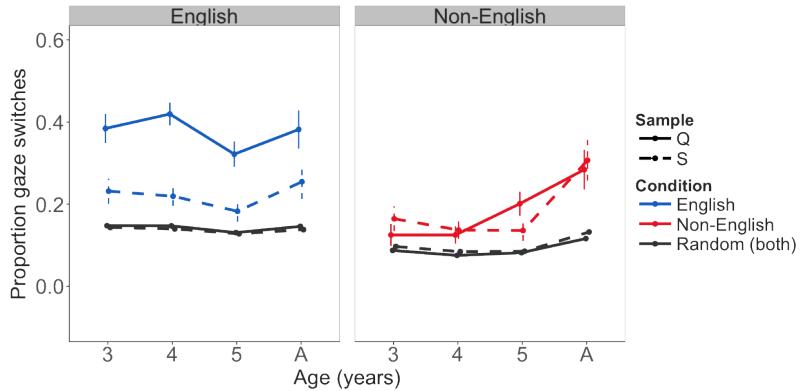


Figure 5: Anticipatory gaze rates across language condition and transition type for the real (red and blue) and randomly permuted (gray) data. Vertical bars represent the standard error.

544 *Developmental effects.* The model of the children’s data revealed a significant
 545 interaction of age and language condition (Table 2) that was highly unlikely
 546 to have derived from random looking (Figure 5). To further explore this
 547 effect, we compared the average effect of language condition across all ages:
 548 we extracted the average difference score for the two language conditions
 549 (English minus non-English) for each subject, computing an overall average
 550 for each random permutation of the data. For each random permutation,
 551 we then made pairwise comparisons of the average difference scores across
 552 ages. Figure 6 plots the real-data difference scores against the random-data
 553 difference score distribution for each pairwise age comparison, showing that
 554 3- and 4-year olds were affected equally by language condition, but that 5-
 555 year-olds affected less than both 3- and 4-year-olds (with 99.52% and 99.96%

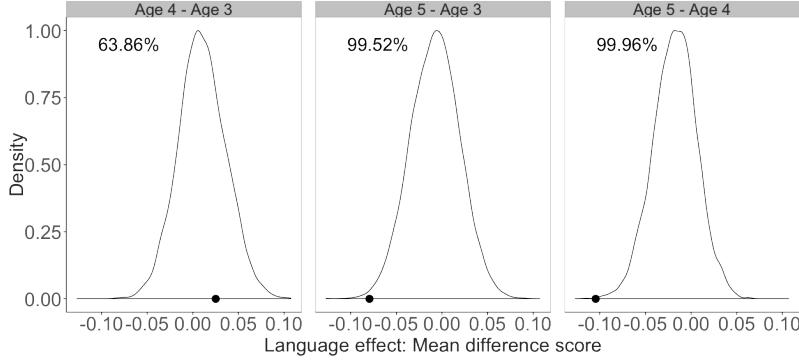


Figure 6: Pairwise comparisons of the language condition effect across ages for the original data (black dots) and the 5,000 randomly permuted datasets (distribution).

of difference scores greater than the randomly permuted data, respectively, i.e., differences of $p < .01$ and $p < .001$.

2.3. Discussion

Children and adults spontaneously tracked the turn structure of the conversations, making anticipatory gaze switches at an above-chance rate across all ages and conditions (Table 1; Figure 5). Children’s anticipatory gaze rates were affected by language condition, transition type, age, and gap duration (Table 2), none of which could be explained by a baseline of random gaze switching (Figure 4a).

Language condition (English vs. non-English) affected children’s anticipations in two ways (Table 2; Figure 5). First, children made more anticipatory switches overall in English videos, compared to non-English videos. This effect suggests that lexical access is important for children’s ability to anticipate upcoming turn structure; children had no lexical access to the speech in the non-English videos, though they did have access to (non-native) prosodic cues and non-verbal behavior, consistent with prior work on turn-end prediction in adults (De Ruiter et al., 2006; Magyari and De Ruiter, 2012) and children (Keitel et al., 2013). Second, children systematically made more anticipatory switches after hearing a question compared to a non-question, but only in the English condition, suggesting that, when children have access to lexical cues, they are more likely to make an anticipatory gaze switch if they

577 can expect an immediate response from the addressee. If so, then children's
578 attention to lexical cues for turn taking may primarily be in monitoring the
579 signal for cues to questionhood (e.g., subject-auxiliary inversion, *wh*-words,
580 etc.).

581 Children's anticipatory gaze switches were also affected by their age, but
582 only in the non-English videos: 3- and 4-year-olds made many more anticipa-
583 tory switches when watching videos in English compared to non-English, but
584 this effect of language condition had attenuated significantly by age 5 (Table
585 2; Figure 5; Figure 6). This interaction suggests that the 5-year-olds were
586 able to leverage anticipatory cues in the non-English videos in a way that
587 3- and 4-year-olds could not, possibly by shifting more attention to the non-
588 native prosodic or non-verbal cues. Prior work on children's turn-structure
589 anticipation proposed that children's turn-end predictions rely primarily on
590 lexicosyntactic structure (and not, e.g., prosody) as they get older (Keitel
591 et al., 2013). The current results suggest more flexibility in children's pre-
592 dictions; when they do not have access to lexical information, older children
593 are likely to find alternative cues to turn taking behavior.

594 Finally, children showed an effect of gap duration (Table 2). This effect
595 is straightforward: longer gaps resulted in longer analysis windows, yielding
596 more time for children to make an anticipatory gaze.

597 Adults' anticipatory gaze rates were also affected by transition type, lan-
598 guage condition, and gap duration (Table 2), none of which could be easily
599 explained by a baseline of random gaze switching (Figure 4b). Like children,
600 adults made more anticipatory switches after hearing questions compared
601 to non-questions, suggesting that anticipation mattered more to them when
602 an immediate response was expected. Also like children, the advantage for
603 questions was driven by lexical access such that adults must have relied on
604 lexicosyntactic cues to questionhood in picking out turns that potentially
605 require an immediate response, though this effect was only marginally di-
606 vergent from the distribution of randomly permuted data ($p = .053$; Figure
607 4b). Finally, adults' anticipation rates were also affected by gap duration,
608 but more so for questions than non-questions (Table 2). This interaction
609 suggests that adults were less likely overall to make anticipatory switches at
610 non-questions (as is evident for adults and children in Figure 5), and so did
611 not benefit from extra time to do so compared to long gaps for questions.

612 *2.3.1. Summary*

613 Children and adults' predictions alike were benefited by access to lexical
614 information (English) and speech act status (questionhood), suggesting that
615 linguistic cues, particularly lexical ones, facilitate their spontaneous predic-
616 tions about upcoming turn structure through the identification of turns with
617 immediate responses. Children's anticipatory gaze rates for questions and
618 non-questions in English was stable across ages and comparable to adult be-
619 havior (Figure 5), suggesting that they can identify questions in native stimuli
620 with adult-like competence by age three. Although participants' ability to
621 recognize questions was facilitated by lexical access (i.e., English vs. non-
622 English), the prosody in the non-English videos was non-native, and so the
623 experimental design can not conclusively show which linguistic cues children
624 relied on in the English videos to identify question turns. Relatedly, though
625 lexical access clearly facilitated participants' anticipatory gaze rate, it was
626 not necessary for participants—especially adults—in order to exceed chance
627 switching rates (Figure 5), suggesting that participants use non-lexical cues
628 (e.g., prosody, non-verbal behavior) to make anticipatory eye movements at
629 least some of the time.

630 Interestingly, adults and children both were strongly affected by transition
631 type, in that they made more anticipatory switches after hearing questions,
632 compared to non-questions. Even in the English videos, when participants
633 had full access to linguistic cues, their rates of anticipation were relatively
634 low—in fact, comparable to the non-English videos—unless the turn was a
635 question (Figure 5). Prior work using online, metalinguistic tasks has shows
636 that participants can use linguistic cues to accurately predict upcoming turn
637 ends. The current results suggest that, in their spontaneous predictions
638 about third-party conversation, both children and adults monitor the lin-
639 guistic structure of unfolding turns for cues to upcoming responses.

640 Children and adults generally behaved relatively similarly in this first ex-
641 periment and our language manipulation (English vs. non-English) was too
642 coarse to comment on when children begin to use different types of native
643 linguistic cues (e.g., prosody vs. lexicosyntax); we would instead need to di-
644 rectly compare lexicosyntactic and prosodic cues in the participants' native
645 language, controlling for the presence of non-verbal cues. To see the emer-
646 gence of anticipatory gaze switching we would also need to include younger
647 children, since participants already reliably made anticipatory gaze switches
648 at age three. Experiment 1 thus lays the analytic groundwork for a method

649 that allows for greater experimental control, which we introduce in Experi-
650 ment 2.

651 **3. Experiment 2**

652 We improved our design by using native-language stimuli, controlling for
653 lexical and prosodic information, eliminating non-verbal cues, and testing
654 children from a wider age range. All of the videos in Experiment 2 were in
655 the participants' native language (American English). To tease apart the
656 role of lexical and prosodic information, we phonetically manipulated the
657 speech signal for pitch, syllable duration, and lexical access. By testing one-
658 to six-year-olds we hoped to find the developmental onset of turn-predictive
659 gaze. We also hoped to measure changes in the relative roles of prosody and
660 lexicosyntax across development.

661 Non-verbal cues in Experiment 1 (e.g., gaze and gesture) could have
662 helped participants make predictions about upcoming turn structure (Rossano
663 et al., 2009; Stivers and Rossano, 2010). Since our focus is on linguistic cues,
664 we eliminated all gaze and gestural signals in Experiment 2 by replacing
665 the videos of human actors with videos of puppets. Puppets are less realistic
666 and expressive than human actors, but they create a natural context
667 for having somewhat motionless talkers in the videos (thereby allowing us
668 to eliminate gestural and gaze cues). Additionally, the prosody-controlled
669 condition included small but global changes to syllable duration that would
670 have required complex video manipulation or precise re-enactment with hu-
671 man talkers, neither of which was feasible. For these reasons, we decided to
672 substitute puppet videos for human videos in the final stimuli.

673 As in the first experiment, we recorded participants' eye movements as
674 they watched six short videos of dyadic conversation, and then analyzed
675 their anticipatory glances from the current speaker to the upcoming speaker
676 at points of turn transition.

677 *3.1. Methods*

678 *3.1.1. Participants*

679 We recruited 27 undergraduate adults and 129 children between ages 1;0–
680 6;11 to participate in our experiment. We recruited our child participants
681 from the Children's Discovery Museum in San Jose, California, targeting ap-
682 proximately 20 children for each of the six 1-year age groups (range=20–23).
683 All participants were native English speakers, though some parents (N=27)

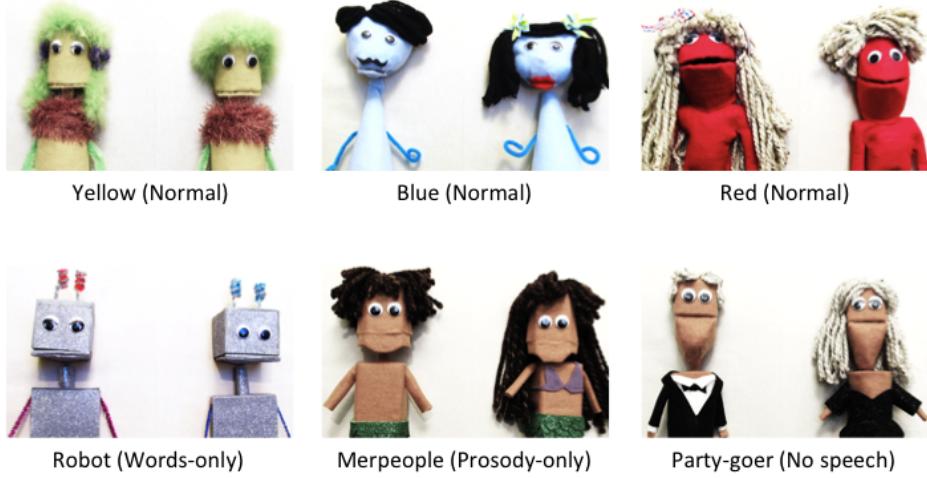


Figure 7: The six puppet pairs (and associated audio conditions). Each pair was linked to three distinct conversations from the same condition across the three experiment versions.

684 reported that their child heard a second (and sometimes third) language at
 685 home. None of the adult participants reported fluency in a second language.
 686 We ran Experiment 2 at a local children’s museum because it gave us access
 687 to children with a more diverse range of ages.

688 *3.1.2. Materials*

689 We created 18 short videos of improvised, child-friendly conversation (Fig-
 690 ure 7). To eliminate non-verbal cues to turn transition and to control the
 691 types of linguistic information available in the stimuli we first audio-recorded
 692 improvised conversations, then phonetically manipulated those recordings to
 693 limit the availability of prosodic and lexical information, and finally recorded
 694 video to accompany the manipulated audio, featuring puppets as talkers.

695 *Audio recordings.* The recording session was set up in the same way as
 696 the first experiment, but with a shorter warm up period (5–10 minutes) and
 697 a pre-determined topic for the child-friendly improvisation ('riding bikes',
 698 'pets', 'breakfast', 'birthday cake', 'rainy days', or 'the library'). All of the
 699 talkers were native English speakers, and were recorded in male-female pairs.
 700 As before, we asked talkers to speak "as if they were on a children's television
 701 show" and to ask at least a few questions during the improvisation. We cut
 702 each audio recording down to the 20-second interval with the most turn

703 activity. The 20-second clips were then phonetically manipulated and used
704 in the final video stimuli.

705 *Audio Manipulation.* We created four versions of each audio clip: *Normal*,
706 *words only*, *prosody only*, and *no speech*. That is, one version with a full
707 linguistic signal (*normal*), and three with incomplete linguistic information
708 (hereafter “limited cue” conditions). The *normal* clips were the unmanipu-
709 lated, original audio clips.

710 The *words only* clips were manipulated to have robot-like speech: We
711 flattened the intonation contours to each talker’s average pitch (F0) and
712 we reset the duration of every nucleus and coda to each talker’s average
713 nucleus and coda duration.⁸ We made duration and pitch manipulations
714 using PSOLA resynthesis in Praat (Boersma and Weenink, 2012). Thus,
715 the *words only* versions of the audio clips had no pitch or durational cues
716 to upcoming turn boundaries, but did have intact lexicosyntactic cues (and
717 residual phonetic correlates of prosody, e.g., intensity).

718 We created the *prosody only* clips by low-pass filtering the original record-
719 ing at 500 Hz with a 50 Hz Hanning window (following de Ruiter et al., 2006).
720 This manipulation creates a “muffled speech” effect because low-pass filter-
721 ing removes most of the phonetic information used to distinguish between
722 phonemes. The *prosody only* versions of the audio clips lacked lexical infor-
723 mation, but retained their intonational and rhythmic cues to upcoming turn
724 boundaries.

725 The *no speech* condition served as a non-linguistic baseline. For this
726 condition, we replaced the original clip with multi-talker babble: We overlaid
727 different child-oriented conversations (not including the original one), and
728 then cropped the result to the duration of the original video. Thus, the
729 *no speech* audio clips lacked any linguistic information to upcoming turn
730 boundaries—the only cue to turn taking was the opening and closing of the
731 puppets’ mouths.

732 Finally, because low-pass filtering removes significant acoustic energy, the
733 *prosody only* clips were much quieter than the other three conditions. Our
734 last step was to downscale the intensity of the audio tracks in the three other
735 conditions to match the volume of the *prosody only* clips. We referred to the
736 conditions as “normal”, “robot”, “mermaid”, and “birthday party” speech

⁸We excluded hyper-lengthened words like [wau] ‘woooow!’. These were rare in the clips.

737 when interacting with participants.

738 *Video recordings.* We created puppet video recordings to match the ma-
739 nicipated 20-second audio clips. The puppets were minimally expressive;
740 the experimenter could only control the opening and closing of their mouths;
741 their head, eyes, arms, and body stayed still. Puppets were positioned look-
742 ing forward to eliminate shared gaze as a cue to turn structure (Thorgrímsson
743 et al., 2015). We took care to match the puppets' mouth movements to the
744 syllable onsets as closely as possible, specifically avoiding any mouth move-
745 ment before the onset of a turn. We then added the manipulated audio clips
746 to the puppet video recordings by hand.

747 We used three pairs of puppets used for the *normal* condition—‘red’,
748 ‘blue’ and ‘yellow’—and one pair of puppets for each limited cue condition:
749 “robots”, “merpeople”, and “party-goers” (Figure 8). We randomly assigned
750 half of the conversation topics (‘birthday cake’, ‘pets’, and ‘breakfast’) to the
751 *normal* condition, and half to the limited cue conditions (‘riding bikes’, ‘rainy
752 days’, and ‘the library’). We then created three versions of the experiment,
753 so that each of the six puppet pairs was associated with three different con-
754 versation topics across the different versions of the experiment (18 videos
755 in total). We ensured that the position of the talkers (left and right) was
756 counterbalanced in each version by flipping the video and audio channels as
757 needed.

758 The duration of turn transitions and the number of speaker changes
759 across videos was variable because the conversations were recorded semi-
760 spontaneously. We measured turn transitions from the audio recording of
761 the *normal*, *words only*, and *prosody only* conditions. There was no audio
762 from the original conversation in the *no speech* condition videos, so we mea-
763 sured turn transitions from the video recording, using ELAN video editing
764 software (Wittenburg et al., 2006).

765 There were 85 turn transitions for analysis after excluding transitions
766 longer than 550 msec and shorter than 90 msec. The remaining turn transi-
767 tions had slightly more questions than non-question ($N=50$ and $N=35$, re-
768 spectively), with transitions distributed somewhat evenly across conditions
769 (keeping in mind that there were three *normal* videos and only one lim-
770 ited cue video for each experiment version): *Normal* ($N=36$), *words only*
771 ($N=13$), *prosody only* ($N=17$), and *no speech* ($N=19$). Inter-turn gaps for
772 questions (mean=365, median=427) were longer than those for non-questions
773 (mean=302, median=323) on average, but gap duration was overall com-
774 parable across conditions: *Normal* (mean=334, median=321), *words only*

775 (mean=347, median=369), *prosody only* (mean=365, median=369), and *no
776 words* (mean=319, median=329). The longer gaps for question transitions
777 could give them an advantage because our anticipatory measure includes
778 shifts initiated during the gap between turns (Figure 2).

779 *3.2. Procedure*

780 We used the same experimental apparatus and procedure as in the first
781 experiment. Each participant watched six puppet videos in random order,
782 with five 15–30 second filler videos placed in-between (e.g., running pup-
783 pies, moving balls, flying bugs). Three of the puppet videos had *normal*
784 audio while the other three had *words only*, *prosody only*, and *no speech* au-
785 dio. This experiment required no special instructions so the experimenter
786 immediately began each session with calibration (same as before) and then
787 stimulus presentation. The entire experiment took less than five minutes.

788 *3.2.1. Data preparation and coding*

789 We coded each turn transition for its linguistic condition (*normal*, *words
790 only*, *prosody only*, and *no speech*) and transition type (question/non-question)⁹
791 and identified anticipatory gaze switches to the upcoming speaker using the
792 methods from Experiment 1.

793 *3.3. Results*

794 Participants' pattern of gaze indicated that they performed basic turn
795 tracking across all ages and in all conditions. Participants looked at the
796 screen most of the time during video playback (82% and 86% average for
797 children and adults, respectively). Children and adults primarily kept their
798 eyes on the person who was currently speaking: They gazed at the current
799 speaker between 44% and 69% of the time, looking back at the addressee
800 between 11% and 14% of the time (Table 2). They tracked the current
801 speaker in every condition—even one-year-olds looked more at the current
802 speaker than at anything else in the three limited cue conditions (40% for
803 *words only*, 43% for *prosody only*, and 39% for *no speech*). There was a steady
804 overall increase in looks to the current speaker with age and added linguistic
805 information (Tables 3 and 4). Looks to the addressee also decreased with
806 age, but the change was minimal.

⁹We coded *wh*-questions as “non-questions” for the *prosody only* videos. Polar questions had a final rising prosodic contour, but *wh*-questions did not (Hedberg et al., 2010).

Age group	Speaker	Addressee	Other onscreen	Offscreen
1	0.44	0.14	0.23	0.19
2	0.50	0.13	0.24	0.14
3	0.47	0.12	0.25	0.16
4	0.48	0.11	0.29	0.12
5	0.54	0.11	0.20	0.14
6	0.60	0.12	0.18	0.10
Adult	0.69	0.12	0.09	0.10

Table 3: Average proportion of gaze to the current speaker and addressee during periods of talk across ages.

Condition	Speaker	Addressee	Other onscreen	Offscreen
Normal	0.58	0.12	0.17	0.13
Words only	0.54	0.11	0.24	0.10
Prosody only	0.48	0.12	0.26	0.15
No speech	0.44	0.13	0.26	0.18

Table 4: Average proportion of gaze to the current speaker and addressee during periods of talk across conditions.

807 3.3.1. Statistical models

808 We identified anticipatory gaze switches for all 85 usable turn transitions,
 809 and analyzed them for effects of language condition, transition type, and age
 810 with two mixed-effects logistic regressions (Bates et al., 2014; R Core Team,
 811 2014). We again built separate models for children and adults because effects
 812 of age were only pertinent to the children’s data. The child model included
 813 condition (normal/prosody only/words only/no speech; with no speech as
 814 the reference level), transition type (question vs. non-question), age (1, 2, 3,
 815 4, 5, 6), and duration of the inter-turn gap (in seconds) as predictors, with
 816 full interactions between language condition, transition type, and age. We
 817 again included the duration of the inter-turn gap as a control predictor and
 818 added random effects of item (turn transition) and participant, with random
 819 slopes of transition type for participants (Barr et al., 2013). The adult model
 820 included condition, transition type, their interactions, and duration as a
 821 control predictor, with participant and item included as random effects and

822 random slopes of condition and transition type.

<i>Children</i>	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-3.57414	0.48576	-7.358	1.87e-13 ***
Age	0.02543	0.10260	0.248	0.8042
Type= <i>non-Question</i>	-0.81873	0.59985	-1.365	0.1723
Duration	4.17672	0.62446	6.689	2.25e-11 ***
Age*Type= <i>non-Question</i>	0.15116	0.13643	1.108	0.2679
Condition= <i>normal</i>	0.36710	0.43296	0.848	0.3965
Age*Condition= <i>normal</i>	0.12919	0.10227	1.263	0.2065
Condition= <i>normal</i> *	0.91059	0.72095	1.263	0.2066
Type= <i>non-Question</i>				
Age*Condition= <i>normal</i>)*	-0.37542	0.16963	-2.213	0.0269 *
Type= <i>non-Question</i>				
Condition= <i>muffled</i>	-1.63429	0.86390	-1.892	0.0585 .
Age*Condition= <i>muffled</i>	0.39317	0.18907	2.080	0.0376 *
Condition= <i>muffled</i> *	1.77190	1.24864	1.419	0.1559
Type= <i>non-Question</i>				
Age*Condition= <i>muffled</i> *	-0.47057	0.28703	-1.639	0.1011
Type= <i>non-Question</i>				
Condition= <i>robot</i>	-0.26741	0.59071	-0.453	0.6508
Age*Condition= <i>robot</i>	0.13740	0.13568	1.013	0.3112
Condition= <i>robot</i> *	-1.02193	1.01227	-1.010	0.3127
Type= <i>non-Question</i>				
Age*Condition= <i>robot</i> *	0.08946	0.22349	0.400	0.6890
Type= <i>non-Question</i>				
<i>Adults</i>	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-3.4557	0.7199	-4.800	1.58e-06 ***
Type= <i>non-Question</i>	0.4292	0.6089	0.705	0.480916
Duration	4.7500	1.2480	3.806	0.000141 ***
Condition= <i>normal</i>	1.2556	0.5633	2.229	0.025805 *
Condition= <i>normal</i> *	-0.9452	0.7631	-1.239	0.215475
Type= <i>non-Question</i>				
Condition= <i>muffled</i>	0.3349	0.8965	0.374	0.708692
Condition= <i>muffled</i> *	0.6627	1.2138	0.546	0.585108
Type= <i>non-Question</i>				
Condition= <i>robot</i>	1.5938	0.7208	2.211	0.027023 *
Condition= <i>robot</i> *	-1.1265	0.9109	-1.237	0.216201
Type= <i>non-Question</i>				

Table 5: Model output for children and adults' anticipatory gaze switches.

823 Children's anticipatory gaze switches showed an effect of gap duration
 824 ($\beta=4.18$, $SE=0.624$, $t=6.689$, $p<.001$), a two-way interaction of age and lan-
 825 guage condition (for prosody only speech compared to the no speech reference
 826 level; $\beta=0.393$, $SE=0.189$, $t=2.08$, $p<.05$), and a three-way interaction of
 827 age, transition type, and language condition (for normal speech compared to
 828 the no speech reference level; $\beta=-0.375$, $SE=0.17$, $t=-2.213$, $p<.05$). There

were no significant effects of age or transition type alone (Table 3.3.1), with only a marginal effect of language condition (for prosody only compared to the no speech reference level; $\beta=-1.634$, $SE=0.864$, $t=-1.89$, $p=.06$)

Adults' anticipatory gaze switches showed effects of gap duration ($\beta=4.75$, $SE=1.248$, $t=3.806$, $p<.001$) and language condition (for normal speech $\beta=1.256$, $SE=0.563$, $t=2.229$, $p<.05$. and words only speech $\beta=1.594$, $SE=0.721$, $t=2.211$, $p<.05$ compared to the no speech reference level). There were no effects of transition type ($\beta=0.429$, $SE=0.609$, $t=0.705$, $p=.48$).

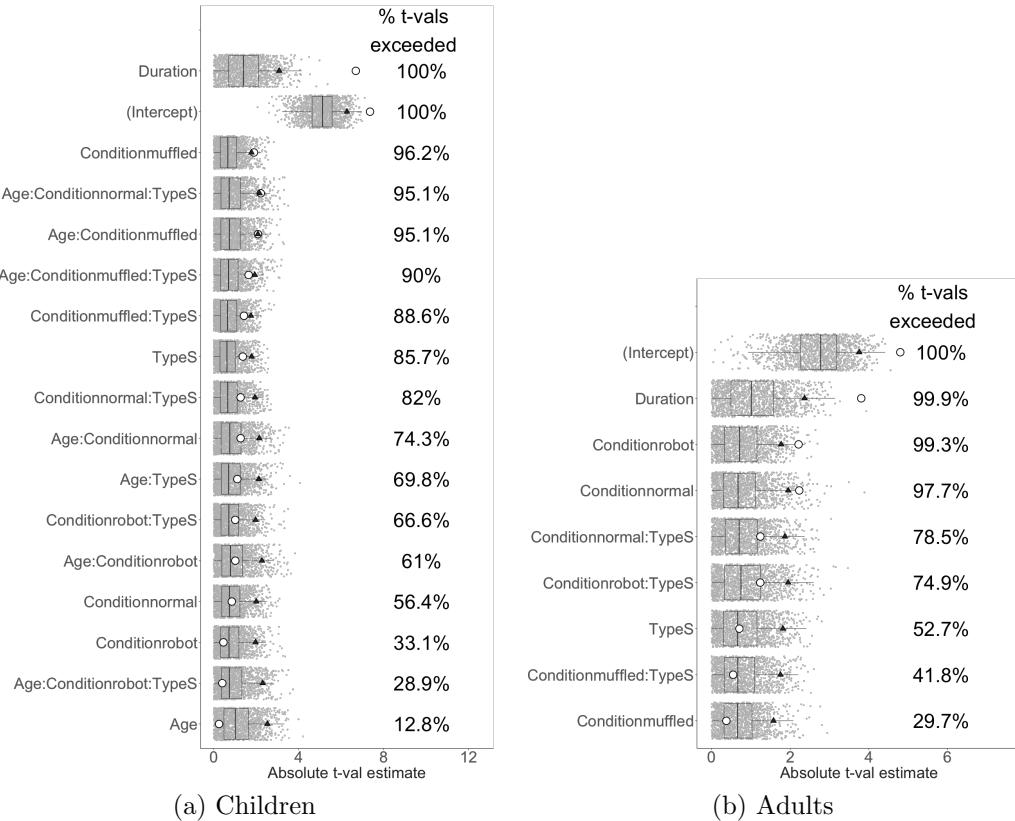


Figure 8: Random-permutation and original $|t\text{-values}|$ for predictors of children and adults' anticipatory gaze rates. Gray dots = random model estimates, White dots = original model estimates, Triangles = 95th percentile for each t -value distribution.

837 *3.3.2. Random baseline comparison*

838 Using the same technique described in experiment 1 (Section 2.2.2), we
 839 created and modeled 5,000 random permutations of participants' anticipa-
 840 tory gaze. Our baseline analyses revealed that none of the significant pre-
 841 dictors from models of the original, turn-related data (Table 5: Children)
 842 can be explained by random looking. In the children's data, the original
 843 model's *t*-values for language condition (prosody only), gap duration, the
 844 two-way interaction of age and language condition (prosody only) and the
 845 three-way interaction of age, transition type, and language condition (normal
 846 speech) were all greater than 95% of the randomly permuted *t*-values (96.2%,
 847 100%, 95.1%, and 95.1%, respectively; Figure 8a). Similarly, the adults' data
 848 showed significant differentiation from the randomly permuted data for all
 849 originally significant predictors: gap duration and language condition for
 850 normal speech and words-only speech (greater than 100%, 96.8%, and 98.7%
 851 of random *t*-values, respectively; Figure 8b). The effects of language condi-
 852 tion and transition type for the real and randomly permuted data can also
 853 be observed in Figure 5. We excluded the output of random-permutation
 854 models that did not ultimately converge to remove unreliable model results
 855 from our percentile calculations below (69% and 70% of models for children
 856 and adults, respectively).

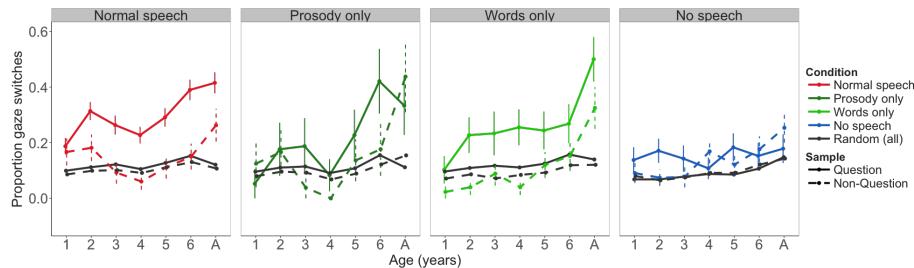


Figure 9: Anticipatory gaze rates across language condition and transition type for the real (blue, dark green, light green, and red) and randomly per-
 muted (gray) data. Vertical bars represent the standard error.

857 *Developmental effects.* The model of the children's data revealed two signif-
 858 icant interactions with age, neither of which derived from random looking
 859 (Table 5; Figure 9). The first was a significant interaction of age and lan-
 860 guage condition (for prosody only compared to the no speech reference level),

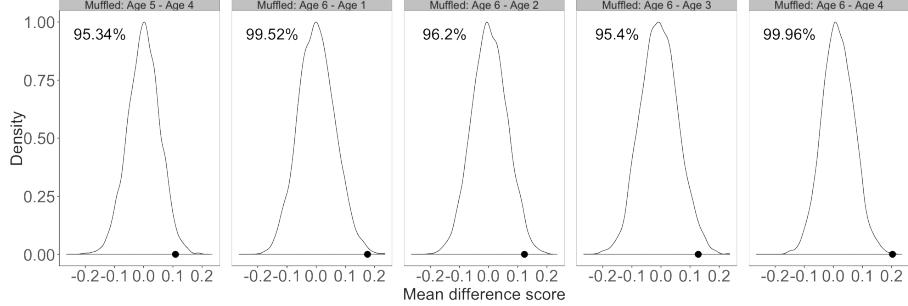


Figure 10: Significant pairwise comparisons of the prosody only-no speech linguistic condition effect, across ages, for the original data (black dots) and the 5,000 randomly permuted datasets (distribution).

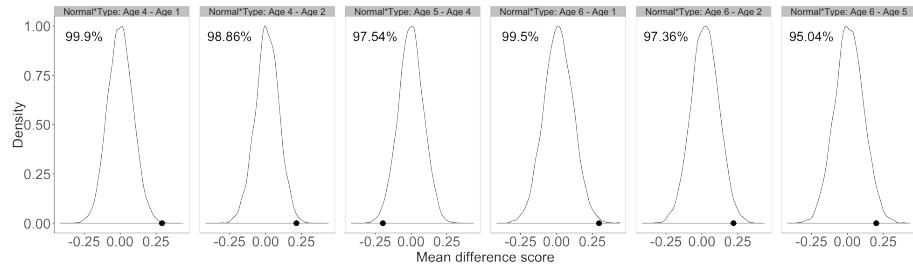


Figure 11: Significant pairwise comparisons of the normal speech-no speech language condition effect for questions status, across ages, for the original data (black dots) and the 5,000 randomly permuted datasets (distribution).

suggesting a different age effect between the two linguistic conditions. As in Experiment 1, we further explored the age effect by extracting the average difference score over subjects for this age-language condition interaction in each random permutation of the data, making pairwise comparisons between the six age groups. Figure 10 shows the comparisons that demonstrate a significant difference in age for the no speech baseline vs. the prosody only condition, revealing that anticipatory gaze in the prosody only condition significantly improves with age, especially at ages 5 and 6 (all with difference scores greater than 95% of the random data scores; $p < .05$).

The second interaction with age was a three-way interaction of age, tran-

871 sition type, and language condition (for normal speech compared to the no
872 speech baseline). We again created pairwise comparisons of the average dif-
873 ference scores for this age-transition type-language condition interaction in
874 each random permutation of the data, with significant differences shown in
875 Figure 11. These pairwise comparisons showed that the effect of transition
876 type in the normal speech condition became larger with age, with significant
877 improvements by age 4 over ages 1 and 2 (99.9% and 98.86%, respectively),
878 by age 5 over age 4 (97.54%), and by age 6 over ages 1, 2, and 5 (99.5%,
879 97.36%, and 95.04%), all significantly different from chance ($p < .05$).

880 *3.4. Discussion*

881 As in Experiment 1, children and adults spontaneously tracked the turn
882 structure of the conversations, making anticipatory gaze switches at an above-
883 chance rate across all ages but not in all conditions (Table ??; Figure 9).
884 In the normal speech condition, however, when children had access to full
885 linguistic information, they made more anticipatory gaze switches than ex-
886 pected by chance even at age one. Children’s anticipatory gaze rates were
887 affected by gap duration, plus interactions of language condition, transition
888 type and age (Table 5), none of which could be explained by a baseline of
889 random gaze switching (Figure 8a).

890 Two of the three linguistic conditions affected children’s anticipatory gaze
891 switches, compared to the no-speech reference level: prosody-only speech and
892 normal speech. Prosody only speech resulted in fewer anticipatory gazes over-
893 all, compared to the no-speech baseline, though the effect was only marginal
894 ($p=.06$). However, prosody-only speech *did* significantly differ from the no
895 speech condition in its interaction with age: whereas age gains were negligible
896 in the no speech condition, 5- and 6-year-olds in the prosody-only condition
897 showed significantly more anticipatory gaze switches than younger children
898 (Figure 10), going from below- and at-chance anticipatory gaze rates to being
899 well above chance (Figure 9).

900 The normal speech condition showed a significantly more gains in antici-
901 patory gaze for questions with age compared to the no-speech condition. As
902 in Experiment 1, children consistently made more anticipatory gazes after
903 hearing questions when they had access to lexical material. But now, with
904 a larger age span in Experiment 2, we can begin to see a developmental
905 path for the question effect. At least for normal speech, greater anticipatory
906 looking for questions is not present from the start: 4-year-olds are the first
907 to make significant gains on 1- and 2-year-olds, but then there are further

908 significant gains at age 5, and again age 6 (Figure 11). While children at ages
909 five and six showed adult-like differentiation of questions and non-questions
910 in the normal speech condition, 1-year-olds had nearly identical switch rates
911 for the two transition types. This suggests that the participants' tendency
912 to make more anticipatory switches for questions emerges after age one and
913 continues developing, with rapid gains in ages three through age six. Fi-
914 nally, children showed a straightforward effect of gap duration (Table 5), as
915 in Experiment 1.

916 Adults' anticipatory gaze rates were affected by gap duration and two
917 of the language conditions (Table 5), none of which could be explained by
918 a baseline of random gaze switching (Figure 8b). Adults made more an-
919 ticipatory switches overall for the normal speech and words-only conditions
920 compared to the no-speech condition, falling in-line with past work showing
921 that adults primarily use lexical information in making predictions about
922 upcoming turn structure (De Ruiter et al., 2006). Though adults did make
923 more anticipatory switches for questions than non-questions on average in
924 the two lexical conditions (Figure 9), the effect of transition type was not
925 significant in either (Table 5), unlike Experiment 1. Like children, adults
926 also showed a straightforward effect of gap duration.

927 3.4.1. Summary

928 Children and adults both showed more anticipatory gaze switches with
929 increased linguistic information, but only for a subset of the linguistic con-
930 ditions and transition types. We had expected to see the most anticipatory
931 switches in the *normal* condition and the least anticipatory switches in the
932 *no speech* condition because they contained the most and least linguistic in-
933 formation, respectively. We had also expected to replicate our finding from
934 Experiment 1 that questions result in more anticipatory switches than non-
935 questions, with the added hypothesis that the question effect is driven by
936 lexicosyntactic cues. We additionally anticipated an overall increase in an-
937 ticipatory switches with age. Finally, since the development of prosodic skills
938 partially precedes the development of lexicosyntax, we expected to see more
939 switches in the *prosody only* condition compared to the *words only* condition
940 in the youngest age groups.

941 In fact, children and adults did show more anticipatory gaze switching
942 in the normal condition compared to the no-speech condition, but for chil-
943 dren this effect only emerged in the form of anticipations following question
944 turns, which increased with age faster than it did with no linguistic cues at

945 all (i.e., in the no-speech condition). Adults also showed significantly higher
946 anticipatory switch rates in the words only condition but no effects of transi-
947 tion type, alone or within linguistic conditions. Taken together, these results
948 partly replicate the findings from Experiment 1: participants make more an-
949 ticipatory switches when they have access to lexical information and, when
950 they do, tend to make more anticipatory switches for questions compared to
951 non-questions.

952 We had anticipated significant gains in anticipatory switching with age,
953 but children only showed significant developmental increases in the prosody
954 only condition and the normal condition (for question transitions). Rather
955 than showing an early advantage for prosody over lexical information, chil-
956 dren did not show significant improvement (or even above-chance perfor-
957 mance) until age five. In contrast, their anticipatory gaze switches were, on
958 average, already above chance at age one in the normal condition (with both
959 lexical and prosodic information) and at age three in the words only condi-
960 tion (with lexical information). These findings do not support an early role
961 for prosody in children’s spontaneous predictions about upcoming turn struc-
962 ture. On the contrary, their predictions were best when lexical information
963 was present, especially when following a question.

964 However, these results do not support the idea that lexical information is
965 sufficient for children (as is proposed for adults; De Ruiter et al., 2006). On
966 the contrary, children showed significant gains in the normal speech condition,
967 but *not* in the words only condition. Notably, the normal speech condition,
968 in addition to having both lexical and prosodic information, is also the one
969 most likely to occur in our participants’ daily lives (compared to muffled or
970 robotic speech) and therefore may have an extra advantage over the other
971 conditions.

972 Finally, on average, participants at all ages made anticipatory gaze switches
973 more often than would be expected by chance in the *no speech* condition. One
974 interpretation of this finding is that participants were not using linguistic in-
975 formation to predict upcoming turns at all—the only cue to turn taking in
976 the *no speech* condition was the alternating mouth movements of the con-
977 versing puppets. But this interpretation is not compatible with the effects
978 of linguistic information that do emerge in the adult and child data: An
979 advantage for lexical cues shaped by transition type and age.

980 The core aims of Experiment 2 were to gain better traction on the indi-
981 vidual roles of prosody and lexicosyntax in children’s turn predictions, and to
982 expand our age range to capture more developmental change. We found that

983 effects of linguistic processing and age *were* present in the dataset, but were
984 primarily related to the question effect, and primarily occur in the normal
985 speech condition. The relation between prosody and lexicosyntax in young
986 children’s anticipations is still not clear: The results of the prosody only ma-
987 nipulation suggest that children do not reliably use low-pass filtered speech
988 to anticipate upcoming turn structure until age five, but on the other hand,
989 the words only condition did not show significant overall differences from the
990 no-speech baseline, including its interactions with age or transition type. If
991 anything, the results do not support the idea that there is a strong early
992 advantage for prosody in turn predictions.

993 4. General Discussion

994 Children begin to develop conversational turn-taking skills long before
995 their first words (Bateson, 1975; Hilbrink et al., 2015; Jaffe et al., 2001;
996 Snow, 1977). As they acquire language, they also acquire the information
997 needed to make accurate predictions about upcoming turn structure. Until
998 recently, we have had very little data on how children weave language into
999 their already-existing turn-taking behaviors.

1000 In two experiments investigating children’s anticipatory gaze to upcom-
1001 ing speakers, we found evidence that turn prediction develops early in child-
1002 hood and that spontaneous predictions are primarily driven by participants’
1003 expectation of an immediate response in the next turn. In making predic-
1004 tions about upcoming turn structure, children used a combination lexical
1005 and prosodic cues; neither prosodic nor lexical cues alone were sufficient to
1006 support increased anticipatory gaze. We also found no early advantage for
1007 prosody over lexicosyntax, and instead found that children were unable to
1008 make above-chance anticipatory gazes in the prosody only condition until age
1009 five. We discuss these findings with respect to the role of linguistic cues in
1010 predictions about upcoming turn structure, the importance of questions in
1011 spontaneous predictions about conversation, and children’s developing com-
1012 petence as conversationalists.

1013 4.1. Predicting turn structure with linguistic cues

1014 Prior work with adults has found a consistent and critical role for lex-
1015 icosyntax in predicting upcoming turn structure (De Ruiter et al., 2006;
1016 Magyari and De Ruiter, 2012), with the role of prosody still under debate
1017 (Duncan, 1972; Ford and Thompson, 1996; Torreira et al., 2015). Knowing

1018 that children comprehend more about prosody than lexicosyntax early on
1019 (see introduction; also see Speer and Ito, 2009 for a review), we thought it
1020 possible that young children would instead show an advantage for prosody
1021 over lexicosyntax. Our results suggest that, on the contrary, when presented
1022 with *only* prosodic information, children's spontaneous predictions about up-
1023 coming turn structure are limited until age five. Do children instead rely on
1024 lexicosyntax, as adults are proposed to do?

1025 We found no evidence that, for children, lexicosyntax alone is "suffi-
1026 cient" (equal to full linguistic information) for spontaneous turn prediction
1027 (De Ruiter et al., 2006, pg. 531): In both experiments, children's perfor-
1028 mance was best in conditions when they had access to the full linguistic
1029 signal. Adults on the other hand, showed significant gains in anticipatory
1030 gaze switching in both conditions with lexical cues, compared to a no speech
1031 baseline.

1032 Participants appeared to use linguistic cues to make more anticipatory
1033 gaze switches in both experiments. Questions are often marked both prosod-
1034 ically and lexicosyntactically for their speech act status. The close link be-
1035 tween prosodic and syntactic structure makes it difficult to tease apart how
1036 predictive processing in one domain is distinct from the other in turn pre-
1037 diction. That being said, compared to prosodic contours (e.g., final rising
1038 intonation), lexicosyntactic cues like *wh*-words, *do*-insertion, and subject-
1039 auxiliary inversion are frequent, categorical, and early-occurring in the utter-
1040 ance. Children may therefore have an easier time picking out and interpreting
1041 lexical cues to questionhood on the fly. Question turns started yielding sig-
1042 nificantly more anticipatory gazes by age 3–4 in the normal speech condition
1043 of Experiment 2, by which time children frequently hear and use a variety of
1044 polar *wh*-questions (Clark, 2009). Furthermore, while lexicosyntactic ques-
1045 tion cues were available on every instance of *wh*- and *yes/no* questions in our
1046 stimuli, prosodic question cues were only salient on *yes/no* questions and,
1047 even then, the mapping of prosodic contour to speech act (e.g., high final
1048 rises for polar questions) is far from one-to-one.

1049 4.1.1. The question effect

1050 In both experiments, anticipatory looking was primarily driven by ques-
1051 tion transitions, a pattern that had not been previously reported in other
1052 observer gaze studies, on children or adults (Keitel et al., 2013; Hirvenkari,
1053 2013; Tice and Henetz, 2011). Questions make a speaker switch immediately
1054 relevant, helping the listener to predict with high certainty what will happen

1055 next (i.e., an answer from the addressee). As mentioned, linguistic cues to
1056 questionhood also often occur early in the utterance, giving observers time
1057 to plan a gaze switch. Because the form of a question can constrain the
1058 type of response that will come next (e.g., a location after a *where* question),
1059 questions can even help listeners predict specific upcoming content in the
1060 next turn.

1061 Our results suggest that question turns start significantly affecting chil-
1062 dren’s predictions between ages three and four, but further testing with finer-
1063 grained age samples with stimuli focused on specific communicative acts and
1064 linguistic cues is needed to better sketch out the developmental trajectory of
1065 this effect. For example, the effect size and age of emergence might differ by
1066 question type (e.g., *wh-* vs. *yes-no*) or the location of question-identifying
1067 cues within the unfolding utterance (early vs. late), but we would not be
1068 able to see it given the current design.

1069 Prior work on children’s acquisition of questions indicates that they may
1070 already have some understanding about question-answer sequences by the
1071 time they begin to speak: Questions make up approximately one third of
1072 the utterances children hear, before and after the onset of speech, and even
1073 into their preschool years, even though the types and complexity of questions
1074 change throughout development (Casillas et al., In press; Fitneva, 2012; Hen-
1075 ning et al., 2005; Shatz, 1979).¹⁰ For the first few years, many of the questions
1076 directed to children are “test” questions—questions that the caregiver already
1077 has the answer to (e.g., “What does a cat say?”), but this changes as children
1078 get older. Questions help caregivers to get their young children’s attention
1079 and to ensure that information is in common ground, even if the responses are
1080 non-verbal or infelicitous (Bruner, 1985; Fitneva, 2012; Snow, 1977). So, in
1081 addition to having a special interactive status (for adults and children alike),
1082 questions are a core characteristic of many caregiver-child interactions, mo-
1083 tivating a general benefit for questions in turn structure anticipation.

1084 All that being said, our current data do not tell us what it is about ques-
1085 tions that makes children and adults more likely to anticipatorily switch their
1086 gaze to addressees. Other request formats, such as imperatives, compliments,
1087 and complaints make a response from the addressee highly likely in the next
1088 turn (Schegloff, 2007). Rhetorical and tag questions, on the other hand, take

¹⁰There is substantial variation question frequency by individual and socioeconomic class (Hart and Risley, 1992).

1089 a similar form to prototypical polar questions, but often do not require an
1090 answer. So, though it is clear that participants anticipated responses more
1091 often for questions than non-questions, we do not yet know whether their
1092 predictive action is limited to turns formatted as questions or is generally
1093 applicable to turn structures that project an immediate response from the
1094 addressee.

1095 Much recent work on prediction during turn taking has focused on partic-
1096 ipants' use of linguistic cues to predict the end of the current turn (Torreira
1097 et al., 2015; Magyari and De Ruiter, 2012; De Ruiter et al., 2006; Ford and
1098 Thompson, 1996; Duncan, 1972), in some cases finding that lexical informa-
1099 tion was sufficient for prediction. Our current results suggest that, sponta-
1100 neous predictions are instead driven by predictions about what is *beyond* the
1101 end of the current turn—that questions are sufficient for prediction.

1102 To integrate these results and understand how listeners make predictions
1103 for turn taking, it is crucial to account for the participants' role in the in-
1104 teraction. The results we present here are based on predictions about third-
1105 party conversation, which enables participants to follow interactions with no
1106 chance of actually participating. Although recent work has shown that sim-
1107 ilar anticipatory eye gazes do occur in spontaneous conversation (Holler and
1108 Kendrick, 2015), we do not yet know if the same question advantage occurs,
1109 or which linguistic cues seem to drive it. It is possible that participants track
1110 conversation for cues to upcoming opportunities/obligations to speak and,
1111 when one is found, focus more attention on predicting the precise timing of
1112 the current turn's end. To answer these questions we will need to innovate
1113 first-person prediction measures that can be used in real-time interaction.

1114 4.1.2. Early competence for turn taking?

1115 One of the core aims of our study was to test whether children show an
1116 early competence for turn taking, as is proposed by studies of spontaneous
1117 mother-infant interaction and theories about the mechanisms underlying hu-
1118 man interaction (Hilbrink et al., 2015; Levinson, 2006). Although children
1119 and adults' gaze patterns were quite similar in Experiment 1, children in Ex-
1120 periment two showed change with age in their anticipatory looking, even in
1121 the most prototypical speech condition ("normal" speech), and 6-year-olds
1122 still did not achieve adult-like levels of anticipatory looking on average. This
1123 may indicate that children rely more in non-verbal cues in anticipating turn
1124 transitions or, alternatively, that adults are better at flexibly adapting to
1125 the turn-relevant cues present at any moment. When they *did* have full lin-

1126 guistic information, children on average still made above-chance anticipatory
1127 gazes to responders, suggesting that at least some turn-taking competence
1128 for third-party conversation is present in infancy, though it is a weak effect.

1129 Taken together, the data suggest that turn-taking skills do begin to
1130 emerge in infancy, but that adult-like competence is not achieved until much
1131 later, mirroring results from children's spontaneous interactions with their
1132 caregivers (infants: Hilbrink et al., 2015; Jaffé et al., 2001; Snow, 1977; older
1133 children: Casillas et al., In press; Garvey, 1984; Ervin-Tripp, 1979). It is
1134 possible, however, that first-person measures of children's predictions would
1135 show more frequent turn structure anticipations at younger ages.

1136 4.2. Limitations and future work

1137 Although Experiments 1 and 2 have offered new findings regarding the
1138 relation to speech act in online turn predictions and the emergence of those
1139 predictions in the first six years, there remain at least two major limitations
1140 to our work: speech naturalness and participant role.

1141 Following prior work (De Ruiter et al., 2006; Keitel et al., 2013), we used
1142 phonetically manipulated speech in Experiment 2, resulting in speech sounds
1143 that children don't usually hear in their natural environment. Many prior
1144 studies have used phonetically-altered speech with infants and young children
1145 (cf. Jusczyk, 2000), but almost none of them have done so in a conversational
1146 context. Children could have had trouble processing the *words only* and
1147 *prosody only* conditions because they were unfamiliar, and not just because
1148 they had less linguistic information available. Future work could instead
1149 carefully script or cross-splice parts of turns to control for the presence or
1150 absence of linguistic cues for turn transition.

1151 The prediction measure we present is based on an observer's view of
1152 third-party conversation but, because participants' role in the interaction
1153 could affect their online predictions about turn taking, an ideal experimen-
1154 tal measure would capture first-person behavior. First-person measures of
1155 spontaneous turn prediction will be key to revealing how participants dis-
1156 tribute their attention over linguistic and non-verbal cues while taking part
1157 in everyday interaction, the implications of which relate to theories of online
1158 language processing for both language learning and everyday talk.

1159 4.3. Conclusions

1160 Conversation plays a central role in children's language learning. It is
1161 the driving force behind what children say and what they hear. Adults use

1162 linguistic information to accurately predict turn structure in conversation,
1163 which facilitates their online comprehension and allows them to respond rel-
1164 evantly and on time. In the current study we have investigated how children's
1165 predictions about turn structure changes as their linguistic skills develop in
1166 the first six years. We found that, although some basic knowledge about turn
1167 taking exists at age one, the integration of linguistic cues into children's pre-
1168 dictions about turn structure takes time and is, like adults, primarily driven
1169 by sequences of action—in our case, by questions and their answers.

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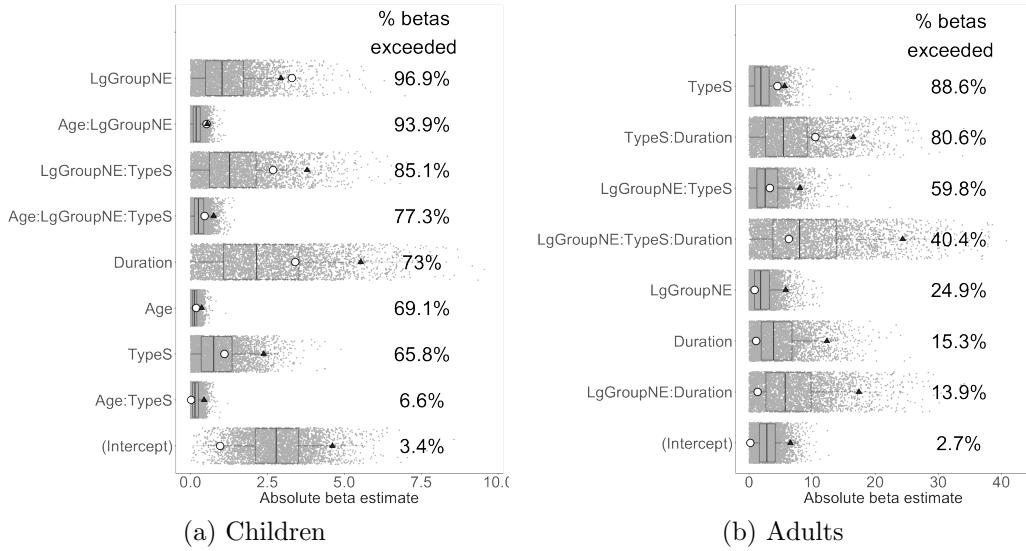


Figure .1: Random-permutation and original $|\beta\text{-values}|$ for predictors of children and adults' anticipatory gaze rates in Experiment 1.

1373 In all of the following plots, the gray dots represent the randomly per-
 1374 mitted data's model estimates for the value listed (beta or standard error),
 1375 the white dots represent the model estimates from the original data, and the
 1376 triangles represent the 95th percentile for each distribution being shown.

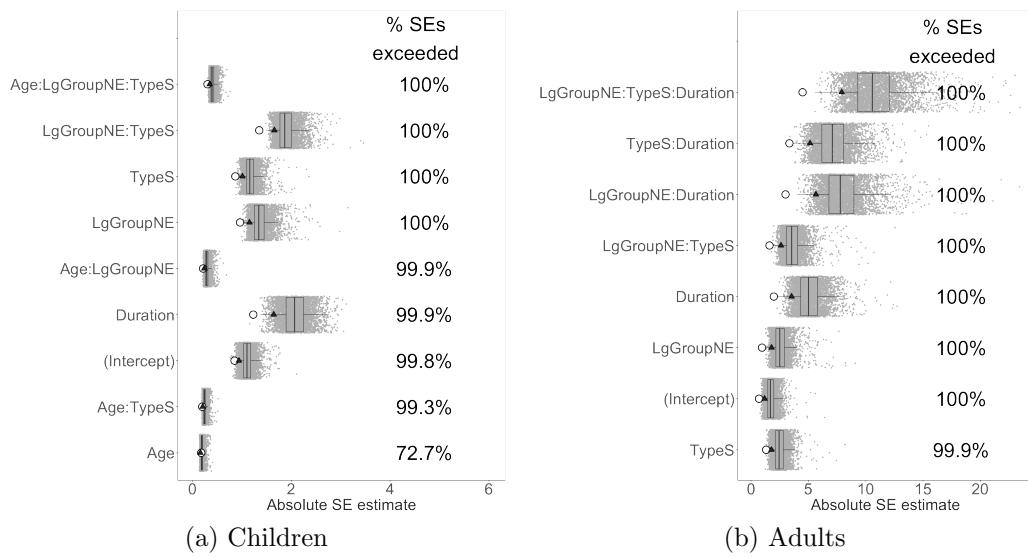


Figure .2: Random-permutation and original $|SE\text{-values}|$ for predictors of children and adults' anticipatory gaze rates in Experiment 1.

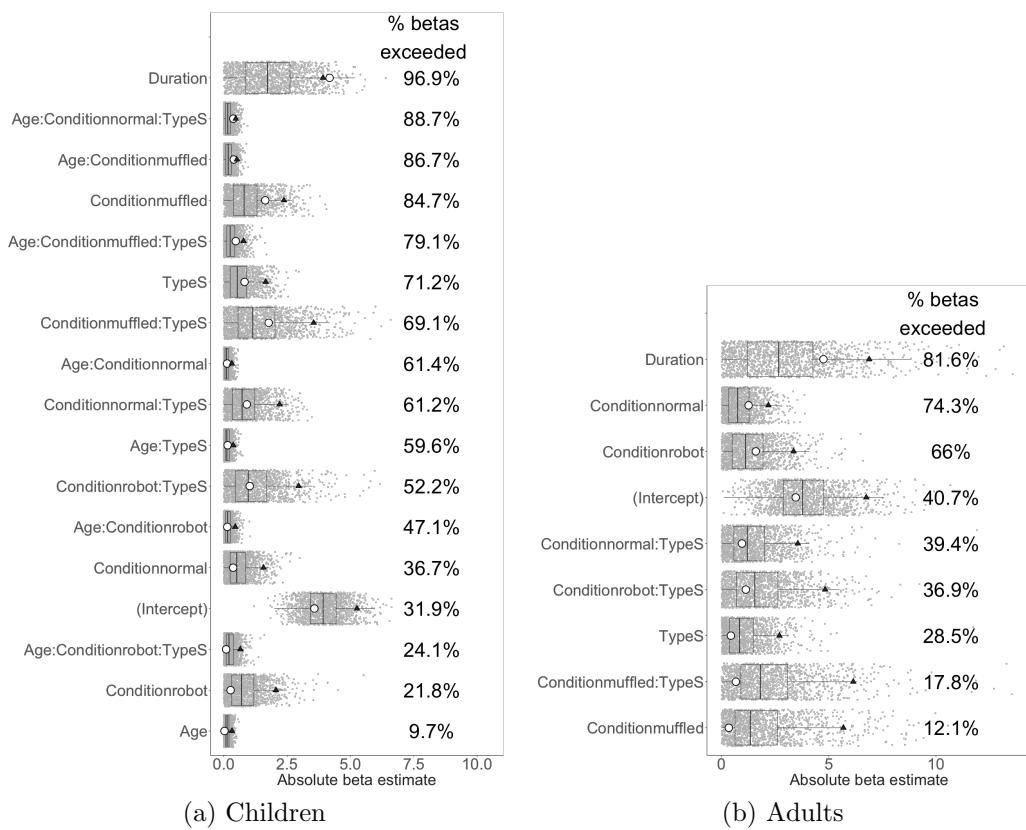


Figure .3: Random-permutation and original $|\beta\text{-values}|$ for predictors of children and adults' anticipatory gaze rates in Experiment 2.

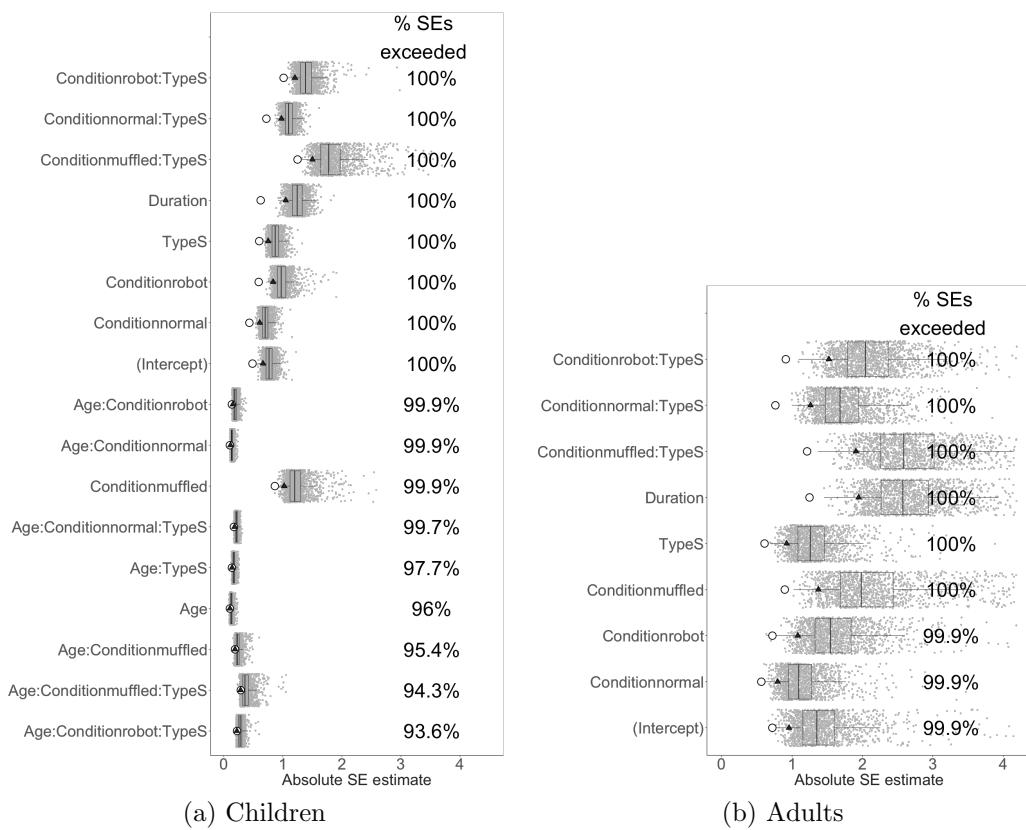


Figure .4: Random-permutation and original $|SE\text{-values}|$ for predictors of children and adults' anticipatory gaze rates in Experiment 2..