ManyBabies1 Test-Retest Supplementary Materials

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S1. Notes on and deviations from the preregistration

S1.1. Deviations from the preregistration

Below, we have compiled a list of deviations from the preregistered methods and analyses available at https://osf.io/v5f8t.

- All infants with usable data for both test and retest session were included in the
 analyses, regardless of the number of total infants a lab was able to contribute after
 exclusion. This decision is consistent with past decisions in ManyBabies projects to
 be as inclusive about data inclusion as possible (ManyBabies Consortium, 2020).
- A small number of infants whose time between sessions exceeded 31 days were still 57 included in the analyses (n = 3). We included these participants for two reasons. 58 First, the general philosophy in ManyBabies studies has been to err on the side of 59 being inclusive, as long as the data from a given participant adds valid information to 60 the study in question. Secondly, time between test session varied continuously across 61 participants and we planned to assess the impact of time between test on reliability. 62 We expected that including these participants should (if anything) provide additional 63 information (and statistical power) by extending the range of a continuous predictor variable (time between test sessions) in our moderator analyses. 65
 - Consistent with analytic decisions in ManyBabies 1 (ManyBabies Consortium, 2020), total looking times were truncated at 18 seconds (the maximum trial time) in the small number of cases where recorded looking times were slightly greater than 18s (presumably due to small measurement error in recording infant looking times).
 - In assessing differences in IDS preference between test and retest sessions, we preregistered an additional linear mixed-effects model including a by-lab random slope for session. This model yielded qualitatively equivalent results (see R markdown of the main manuscript). However, the model resulted in a singular fit, suggesting that the model specification may be overly complex and that its estimates

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- should be interpreted with caution. We therefore focused only on the first

 preregistered model (including only by-lab and by-participant random intercepts) in

 reporting the analyses in the main manuscript.
- In assessing the reliability of IDS using a linear mixed-effects model predicting IDS

 preference in Session 2 from IDS preference in Session 1, we also assessed the

 robustness of the results by fitting a second preregistered model with more complex

 random effects structure, including a by-lab random slope for IDS preference in

 session 1. This model is included in the main R markdown script and yields

 qualitatively equivalent results to the model reported in the manuscript that includes

 a by-lab random intercept only.
 - We report a series of secondary planned analyses in the Supplementary Materials exploring potential moderating variables of time between test sessions (S2.1.), participant age (S2.2.), method (S2.3.), and the language background of the participants (S2.4.).
- While we fit all models described in the secondary analyses of the preregistration, 89 including models investigating interactions between moderators, we interpret the 90 more complex, three-way interaction models with caution. Our final sample size was 91 smaller than we anticipated, which made our sample less well-powered to investigate 92 more complex relationships between moderators. Moreover, the baseline model for 93 these secondary interaction models was incorrectly specified in the preregistration 94 (lower-order terms for the moderator were incorrectly removed in the planned baseline 95 model), and we opt instead to report estimates using the more conventional method 96 of comparing parameters of interest to models including all predictors except the 97 main predictor of interest (e.g., estimating significance of three-way interaction terms 98 by comparing the model fit to a model including only all lower-order predictors). gg
 - In the by-lab meta-analysis of test-retest reliability (S3), we also separated the data by method, such that the data from one lab was split into data from the head-turn

preference procedure and the central fixation method. We separated the data by both lab and method because differences in IDS preference across methods were observed in MB1 and because this approach was more consistent with the analytic approach in the rest of the manuscript. This decision does not qualitatively affect the conclusions of the meta-analysis, namely that there was no consistent evidence of test-retest reliability.

108 S1.2. Additional notes

While the original idea was to retest infants that contributed data to the original study, some labs had already finished data collection for MB1 but nevertheless agreed to collect a new set of data for the MB1 test-retest spin off project. In addition, one lab already started data collection prior to the preregistration, however, this data had not been inspected or analyz^ed prior to the preregistration. We here present a detailed list of our collected data in relation to the original MB1 study and the preregistration (see Table 1).

Table 1

Additional notes on data collection status of each lab in relation to preregistration and MB1.

Lab	Method	Collection prior to preregistration	MB1 as Session 1
babylab-potsdam	НРР	No	No
babyling-oslo	eye-tracking	No	No
brookes-babylab	central fixation	No	No
InfantCog-UBC	central fixation	No	Yes
infantll-madison	HPP	No	No
lancslab	eye-tracking	No	No
wsi-goettingen	central fixation	Yes (n=14)	Yes
wsi-goettingen	HPP	No	No

S2. Secondary analyses investigating possible moderating variables

116 S2.1. Descriptives and power

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- S2.1.1. Additional descriptive information. To highlight the distributions of the key moderators of interest, we include an additional plot representing the distribution of infant age among the 7 participating labs, split by method and language background (Figure 1).
- S2.1.2. A note on (post-hoc) power. Our final sample size (N = 158) although quite large for typical infant looking time studies had limited power to detect moderation effects. As a heuristic for approximate post-hoc power, we can consider the power to detect differences between correlations for our final sample. For the moderator of language background, we had n = 37 participants with a North American English language background and n = 121 participants with non-North American English backgrounds.

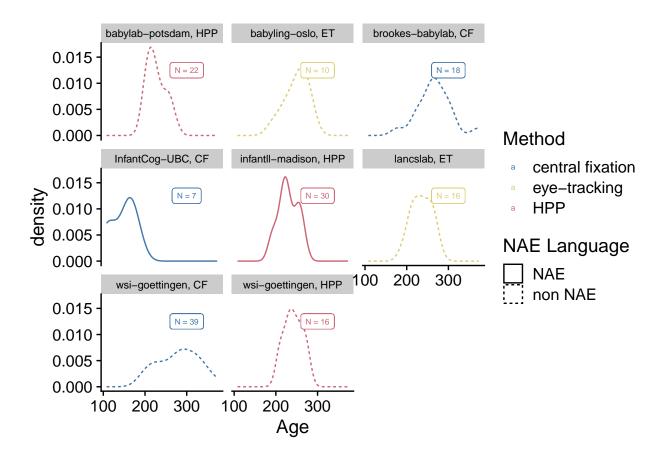


Figure 1. Distribution of participant age for each lab and method. Method is highlighted by color and language status is indicated by line type (solid = North American English; dashed = non-North American English).

Given this sample size, differences between the two samples would have to be substantial in 127 order to have reasonable power to detect a difference: assuming r=0 for one sample, we 128 would only reach 80% power to detect a difference if $r \sim 0.5$ for the other sample. We had 129 slightly more power to detect differences for method, where we had n = 68 HPP 130 observations and n = 90 non-HPP observations. For example, again assuming r = 0 for 131 one sample, we would reach 80% power to detect differences once $r \sim 0.43$ for the second 132 sample. Given the limited power to detect all but large effect sizes in our moderation 133 analyses, we planned to treat any significant results from the moderator analyses with 134 caution. 135

S2.2. Time between test sessions

S2.2.1. Reliability moderated by time between test sessions. The number 137 of days between the first and second testing session varied widely across participants 138 (mean: 10 days; range: 1 - 49 days). We therefore tested for the possibility that the time 139 between sessions might have an impact on test-retest reliability. We fit a linear 140 mixed-effects model predicting IDS preference in Session 2 from IDS preference in Session 1 141 (mean-centered), number of days between testing sessions (mean-centered), and their 142 interaction, including a by-lab random intercept and random slope for IDS preference in 143 Session 1. A more complex random effects structure including additional random slopes for 144 number of days between test sessions and its interaction with IDS preference in Session 1 145 did not converge. We found no evidence that the number of days between test sessions moderated the relationship between IDS preference in Session 1 and 2. Neither the main effect of time between sessions, β =-0.01, SE=0.03, t(148.70)=-0.41, p=.684, nor the interaction term, β =-0.01, SE=0.02, t(149.10)=-0.73, p=.465, showed significant effects.

S2.2.2. Change in preferential looking moderated by time between test 150 sessions. In addition to assessing the influence of moderators on test-retest reliability, we 151 also tested whether the difference in magnitude of the IDS preference between Session 1 and 152 Session 2 depended on moderators of interest. To investigate the influence of time between 153 test sessions, we fit a linear mixed-effects model predicting average IDS preference from Session (centered; Session 1 vs. Session 2), days between test sessions (mean-centered), and their interaction. We included by-lab and by-participant random intercepts (more complex 156 random effects structures did not converge due to singular fits). We found no evidence that 157 the change in preferential looking to IDS between Session 1 and Session 2 was moderated 158 by days between test sessions, $\beta = -0.02$, SE = 0.04, t(156) = -0.48, p = .634. 159

S2.3. Participant age

- S2.3.1. Reliability moderated by participant age. To investigate the possibility that age moderated test-retest reliability, we fit a linear mixed-effects model predicting IDS preference in Session 2 from IDS preference in Session 1 (mean-centered), participant age (mean-centered) and their interaction. The model included a by-lab random intercept and a by-lab random slope for IDS preference in Session 1. We found no evidence that age influenced test-retest reliability as indicated by the interaction between IDS preference in Session 1 and age, β =0.00, SE=0.00, t(76.60)=-0.85, p=.398.
- S2.3.2. Change in preferential looking moderated by participant age. To investigate the potential of moderators to influence the overall magnitude of the IDS effect between Session 1 and 2, we fit a linear mixed-effects model predicting average IDS preference from Session (centered; Session 1 vs. Session 2), participant age (mean-centered), and their interaction. We included by-lab and by-participant random intercepts (more complex random effects structures did not converge due to singular fits). We found no evidence that the change in preferential looking to IDS between Session 1 and Session 2 was moderated by participant age, β =0.00, SE=0.00, t(157.50)=-0.56, p=.577.

176 **S2.4.** Method

S2.4.1. Differences in IDS preference across method. In MB1, infants who participated in the HPP showed a significantly larger magnitude of IDS preference, compared to central fixation and eye-tracking methods. Therefore, in the current study, we also explored whether the magnitude of IDS preference differed as a function of method. We fit a linear mixed-effects model predicting IDS preference from Session and Method (dummy-coded, with central fixation as the reference level), including by-lab and by-participant random intercepts. We found no significant difference in IDS preference across methods, $\chi^2=1.11$, p=.575.

S2.4.2. Reliability moderated by method. We tested whether method 185 (eye-tracking vs. central fixation vs. headturn preference procedure) moderated test-retest 186 reliability by fitting a linear mixed-effects model predicting IDS preference in Session 2 187 from IDS preference in Session 1 (mean-centered), Method (dummy-coded, with central 188 fixation as the reference level) and their interaction. The model included a by-lab random 189 intercept and a by-lab random slope for IDS preference in Session 1 (models with more 190 complex random effects structure including by-lab random effects for Method did not 191 converge). We found no evidence that Method influenced test-retest reliability as indicated 192 by the interaction between IDS preference in Session 1 and age, $\chi^2=3.85$, p=.146. 193

S2.4.3. Reliability and its interaction with both method and age. 194 more complex linear mixed-effects model (preregistered as part of our planned secondary 195 analyses) including the interaction between IDS preference in Session 1 (mean-centered), 196 Method (dummy-coded, with central fixation as the reference level), participant age 197 (mean-centered), and all lower order interactions, we find evidence for an interaction 198 between method and age in predicting reliability, $\chi^2=6.44$, p=.040. This effect appears to 199 be mainly driven by older infants showing some evidence of test-retest reliability for the 200 HPP, r = 0.45, p = 0.02 (see Figure 3B). However, we believe these tentative findings should be treated with caution, due to the small size of our infant sample once binned by 202 multiple moderating factors. 203

S2.4.4. Change in preferential looking moderated by age and method. We fit a linear mixed-effects model predicting average IDS preference from the three-way interaction of Session (centered; Session 1 vs. Session 2), participant age (mean-centered), Method (dummy-coded, with central fixation as the reference level), and all lower order predictors. We included a by-participant random intercept (more complex random effects structures did not converge due to singular fits). We found no evidence that the change in preferential looking to IDS between Session 1 and Session 2 was moderated by participant age and Method, β =-0.01, SE=0.02, t(155.40)=-0.58, p=.562.

S2.5. Language background

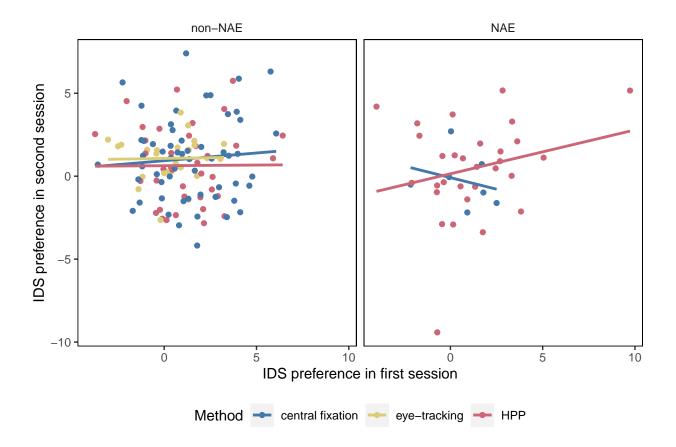


Figure 2. Infants' preference in Session 1 and Session 2 with individual data points and regression lines color-coded by method (CF, ET, or HPP). Results are plotted separately for North American English-learning infants (right panel) and infants learning other languages and dialects (left panel).

S2.5.1. Reliability moderated by language background. NAE-learning infants showed greater IDS preferences than their non-NAE counterparts in MB1. We therefore also assessed whether test-retest reliability interacted with children's language background. A linear mixed-effects model predicting IDS preference in Session 2 based on IDS preference in Session 1 (mean-centered), NAE (centered), and their interaction, including Lab as a random intercept, revealed no interaction, β =0.29, SE=0.18, t(151.30)=1.59, p=.115 (Figure 2).

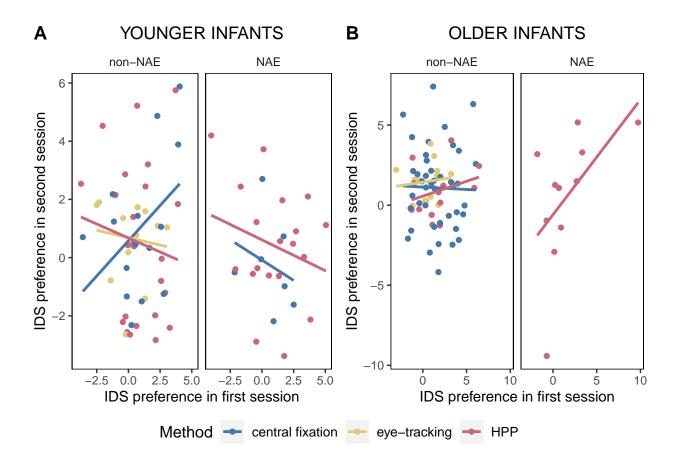


Figure 3. Infants' preference in Session 1 and Session 2 with individual data points and regression lines color-coded by method for (A) younger and (B) older infants (median-split). Results are plotted separately for North American English-learning infants and infants learning other languages and dialects.

S2.5.2. Reliability and its interaction between language background and

²²¹ **age.** We also fit a preregistered linear mixed-effects model predicting IDS preference in Session 2 from the three-way interaction between IDS preference in Session 1 (mean-centered), NAE (centered), participant age (mean-centered), and all lower order interactions. We find evidence for an interaction between language background and age in predicting reliability, β =0.01, SE=0.00, t(63.70)=2.43, p=.018. Figure 3 illustrates that this interaction was driven by a small set of older infants (all from a single lab and participating in the HPP method) showing a somewhat more reliable relationship between

Session 1 and Session 2 looking. Note that the mixed-effects analyses use Age as a continuous predictor — age is median-split in Figure 3 to ease visualization. Given the small number of infants driving the three-way interaction and the confounded nature of this sample (with method and lab), we do not draw strong conclusions from the existence of this three-way interaction, but report it here to spur future investigations into how age and experience interacts with test-retest reliability.

S2.5.3. Change in preferential looking moderated by age and language 234 background. We fit a linear mixed-effects model predicting average IDS preference from 235 the three-way interaction of Session (centered; Session 1 vs. Session 2), participant age 236 (mean-centered), NAE (centered), and all lower order predictors. We included by lab and 237 by-participant random intercepts and by-lab random slope for Session (more complex 238 random effects structures did not converge due to singular fits). We found no evidence that 239 the change in preferential looking to IDS between Session 1 and Session 2 was moderated 240 by participant age and language background, $\beta = 0.01$, SE = 0.02, t(114.60) = 0.95, p = .347. 241

S3. Meta-analysis of test-retest reliability

In addition to the methods for assessing test-retest reliability reported in the main manuscript, we also investigated test-retest reliability across labs using a meta-analytic approach. We used the metafor package (Viechtbauer, 2010) to fit a mixed-effects meta-analytic model on z-transformed correlations for each combination of lab and method using sample size weighting. The model included random intercepts for lab and method. The overall effect size estimate was not significantly different from zero, b = -0.04, 95% CI = [-0.26, 0.19], p = 0.73. A forest plot of the effect sizes for each lab and method is shown in Figure 4.

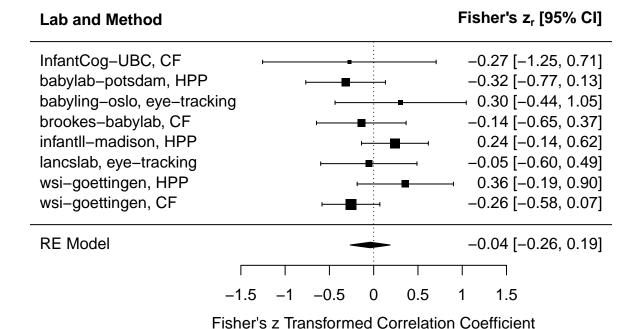


Figure 4. Forest plot of test-retest reliability effect sizes. Each row represents Fisher's z transformed correlation coefficient and 95% CI for a given lab and method (HPP = head-turn preference procedure; ET = eye-tracking; CF = central fixation). The black diamond represents the overall estimated effect size from the mixed-effects meta-analytic model.

S4. Analyses including a more restricted sample

Given that we found that restricting the sample to participants contributing at least 6 ADS and IDS trials in both sessions, we conducted the central analyses with this more restricted infant sample.

₅ S4.1. Descriptives and IDS preference for the restricted sample

The participants in the restricted sample — contributing at least 6 IDS and ADS trials for both sessions — were distributed across the contributing labs, methods, and

Table 2
Statistics of the included labs for the restricted sample (min 6 trials contributed per session). N refers to the number of infants included in the analysis.

Lab	Method	Language	Mean age (days)	N
InfantCog-UBC	central fixation	English	136	5
babylab-potsdam	HPP	German	224	18
babyling-oslo	eye-tracking	Norwegian	250	1
brookes-babylab	central fixation	English	254	15
infantll-madison	HPP	English	233	12
lancslab	eye-tracking	English	235	10
wsi-goettingen	HPP	German	240	13
wsi-goettingen	central fixation	German	281	26

language backgrounds (Table 2). There was no difference in average age between the main sample and the restricted sample, t(204.57) = -0.33, p = .744. There was a robust preference for IDS in both Session 1, t(99) = 6.67, p < .001, and Session 2, t(99) = 4.42, p < .001. We observed no difference in IDS preference between the two sessions, β =-0.34, SE=0.28, p=.225.

Interestingly, while there was a significant simple correlation between IDS preference in session 1 and session 2, r = .22, 95% CI [.02, .40], t(98) = 2.23, p = .028, we found that IDS preference in Session 1 did not significantly predict IDS preference in Session 2 in a linear mixed-effects model including a by-lab random intercept, $\beta=0.12$, SE=0.11, p=.255.

S4.2. Moderator analyses including a more restricted sample

S4.2.1. Time between test sessions. As in the analyses with the full dataset, we found no evidence that the number of days between test sessions moderated the relationship between IDS preference in Session 1 and 2. Neither the main effect of time between sessions, β =-0.03, SE=0.03, t(95.80)=-0.96, p=.342, nor the interaction term, β =-0.01, SE=0.03, t(93.60)=-0.22, p=.828, showed significant effects.

273 S4.2.2. Participant age

To investigate the possibility that age moderated test-retest reliability in the restricted sample, we fit a linear mixed-effects model predicting IDS preference in Session 2 from IDS preference in Session 1 (mean-centered), participant age (mean-centered), and their interaction. The model included a by-lab random intercept and a by-lab random slope for IDS preference in Session 1. We found no evidence that age influenced test-retest reliability as indicated by the interaction between IDS preference in Session 1 and age, β =0.00, SE=0.00, t(43.20)=-0.69, p=.494.

281 S4.2.3. Method

We tested whether method (ET vs. CF vs. HPP) moderated test-retest reliability by fitting a linear mixed-effects model predicting IDS preference in Session 2 from IDS preference in Session 1 (mean-centered), Method (dummy-coded, with CF as the reference level), and their interaction. The model included a by-lab random intercept and a by-lab random slope for IDS preference in Session 1. We found no evidence that Method influenced test-retest reliability as indicated by the interaction between IDS preference in Session 1 and age, $\chi^2=3.85$, p=.146. There was no significant relationship between IDS preference for Session 1 and Session 2 for each method considered separately, CF: $\beta=-0.06$,

²⁹⁰ SE=0.16, p=.704; HPP: β =0.26, SE=0.17, p=.139; eye-tracking: β =-0.04, SE=0.26, p=.866.

292 S4.2.4. Language background

As in the main sample, a linear mixed-effects model predicting IDS preference in Session 2 based on IDS preference in Session 1 (mean-centered), NAE (centered), and their interaction, including Lab as a random intercept, revealed no interaction, β =0.31, SE=0.24, t(95.10)=1.29, p=.199.

Table 3

Correlations between alternative dependent measures

	1	2	М	SD
1. Diff	-		1.21	2.22
2. Prop	.96***	-	0.54	0.07
3. Diff_log_lt	.95***	.96***	0.16	0.30
Note. * p < 0.05; ** p < 0.01; *** p < 0.001				

S5. Alternative dependent variables

To check the robustness of our results, we also investigated whether we obtained similar results with other possible dependent measures: average log-transformed looking times and a proportion-based preference measure. For each alternative dependent variable, we conducted the main analyses of test-retest reliability reported in the manuscript: the overall Pearson correlation, the test-retest linear mixed-effects model, and an inspection of applying stricter inclusion criteria for number of trials contributed.

₀₄ S5.1. Correlations between alternative dependent variables

First, we consider the correlations between the three dependent measures we considered for IDS preference: (a) a simple difference score between average IDS and ADS looking times (main manuscript), (b) a difference score between average log-transformed looking times, and (c) the proportion-based preference measure. As expected, the correlations between the alternative dependent measures were very high, all r-values >= 0.95 (Table 3).

Table 4

Coefficient estimates from a linear mixed-effects model predicting

Log LT IDS preference in Session 2.

	Estimate	SE	t	р
Intercept	0.14	0.07	2.05	0.09
Log LT IDS Preference Session 1	-0.06	0.09	-0.68	0.50

311 S5.2. Log-transformed looking times

In these analyses, we calculated IDS preference by first log-transforming looking 312 times for each trial, computing the average log-transformed looking time for IDS and ADS 313 for each participant, and calculating the difference between average IDS and ADS 314 log-transformed looking times. We fit a linear mixed-effects model predicting IDS 315 preference in Session 2 from IDS preference in Session 1, including a by-lab random 316 intercept. As in the analyses using average raw looking times, the results revealed no 317 significant relationship between IDS preference in Session 1 and 2 (Table 4). The Pearson 318 correlation coefficient was also not statistically significant, r = .03, 95% CI [-.12, .19], 319 t(156) = 0.43, p = .670. Applying successively stricter inclusion criteria — by requiring a 320 higher number of valid trials per condition in each session — showed a similar pattern to 321 the main manuscript, such that correlations increased somewhat with stricter inclusion 322 criteria, but substantially reduced the sample size at the same time (Figure 5). 323

S_{24} S5.3. Proportion looking to IDS

Next, we calculated a proportion-based IDS preference measure by computing the average proportion (raw) looking time to IDS relative to total (raw) looking time to IDS and ADS for each subject (i.e., IDS looking time / (ADS looking time + IDS looking

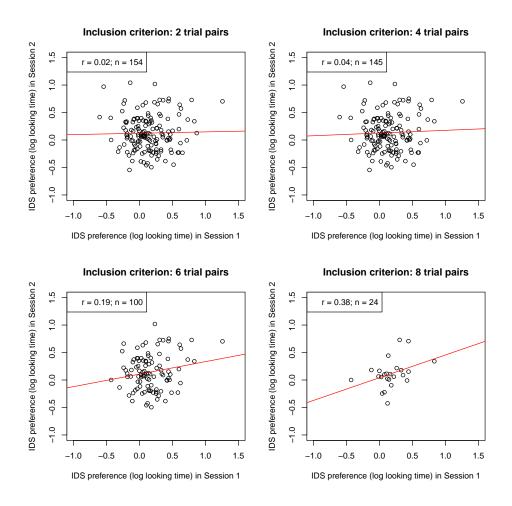


Figure 5. IDS preferences (based on average log-looking times) of both sessions plotted against each other for each inclusion criterion. n indicates the number of included infants, r is the Pearson correlation coefficient as the indicator for reliability.

time)). We fit a linear mixed-effects model predicting proportion-based IDS preference in Session 2 from proportion-based IDS preference in Session 1, including a by-lab random intercept. As in the analyses using other measures of IDS preference, the results revealed no significant relationship between IDS preference in Session 1 and 2 (Table 5). The Pearson correlation coefficient based on proportional IDS looking was also not statistically significant, r = .01, 95% CI [-.15, .16], t(156) = 0.09, p = .927. Stricter inclusion criteria increased the correlation somewhat, as in previous analyses (Figure 6).

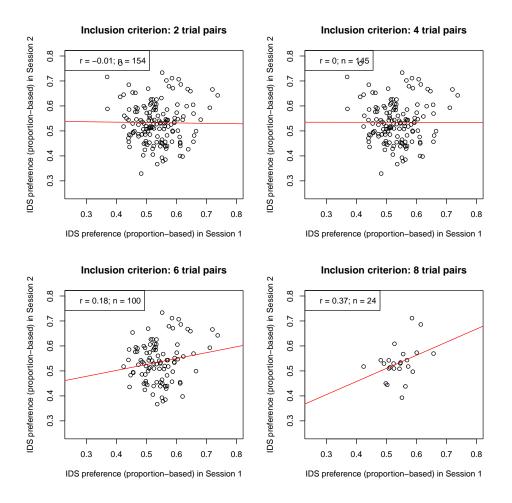


Figure 6. IDS preferences (based on proportion IDS looking) of both sessions plotted against each other for each inclusion criterion. n indicates the number of included infants, r is the Pearson correlation coefficient as the indicator for reliability.

Table 5

Coefficient estimates from a linear mixed-effects model predicting IDS preference
(based on proportion IDS looking) in Session 2.

	Estimate	SE	t	p
Intercept	0.59	0.05	10.70	0.00
IDS Preference (proportion measure) Session 1	-0.10	0.10	-1.01	0.31

S6. Sensitivity of test-retest reliability to trial number inclusion criteria

To conduct a more fine-grained analysis of how stricter trial inclusion criteria affect 336 test-retest reliability, we computed correlations while gradually increasing the number of 337 total valid trials required for inclusion. For this analysis, we required a minimum of one 338 IDS and one ADS trial and gradually increased the number of total valid trials required in 339 both sessions (irrespective of IDS and ADS condition) from 2 to 16 (the maximum number 340 of total trials). Figure 7 depicts the Pearson correlation coefficients for increasingly stricter 341 requirements for the overall trial numbers of a given participant in both sessions. 342 Correlations only increase and reach conventional levels of significance once the number of total required trials for both sessions is greater than 12.



Figure 7. Pearson correlation coefficient with increasingly strict trial-level inclusion criteria. The x-axis depicts the required number of overall valid trials in both Session 1 and Session 2. Dots represent corresponding correlation coefficients, with 95 percent CIs. The sample size is shown above each dot.

S7. Patterns of preference across sessions

We also conducted analyses to explore whether there were any patterns of preference reversal across test sessions. While there was no strong correlation in the magnitude of IDS preference between Session 1 and Session 2, here we asked whether infants consistently expressed the same preference across test sessions. Overall, 58.20% of the infants had a consistent preference from test to retest session. Of the 158 total infants, 44.90% of infants showed a consistent IDS preference and 13.30% showed a consistent ADS preference.

Session 2 and 18.40% switched from an ADS preference to an IDS preference.

Next, we explored whether we could detect any systematic clustering of infants with 354 distinct patterns of preference across the test and retest session. We took a bottom-up 355 approach and conducted a k-means clustering of the test-retest difference data (here using 356 log-transformed looking time data). We found little evidence of distinct clusters emerging 357 from these groupings: the clusterings ranging from k=2 (2 clusters) to k=4 (4 clusters) 358 appear to mainly track whether participants are approximately above or below the mean 359 looking time difference for Session 1 and Session 2 (Figure 8A). The diagnostic elbow plot shows little evidence of a qualitative improvement as the number of clusters is increased, which suggests little evidence for a distinctive set of clusters of participants who showed similar patterns of looking across the test and retest sessions (Figure 8B).

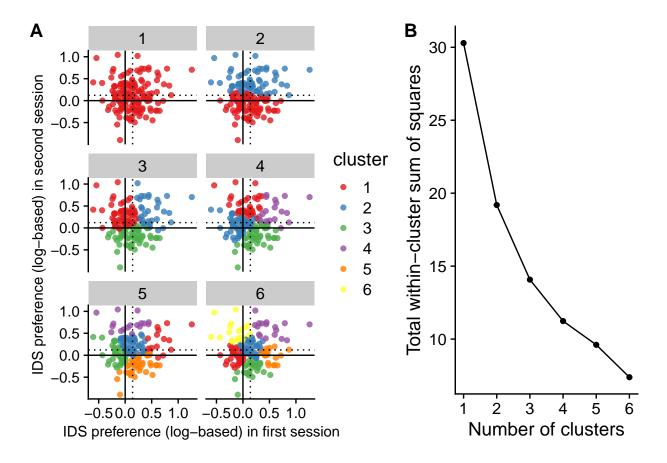


Figure 8. (A) Results from the k-means clustering analysis of IDS preference (based on average log looking times) in Session 1 and 2 for different numbers of k, and (B) the corresponding elbow plot of the total within-cluster sum of squares. In (A), points represent indvidual participants' magnitude of looking time difference at Sessions 1 (x-axis) and 2 (y-axis). The solid line indicates no preference for IDS vs. ADS, the dotted lines indicate mean IDS preference at Session 1 and 2, respectively. Colors indicate clusters from the k-means clustering for different values of k.

S8. Relation between number of contributed trials in each session

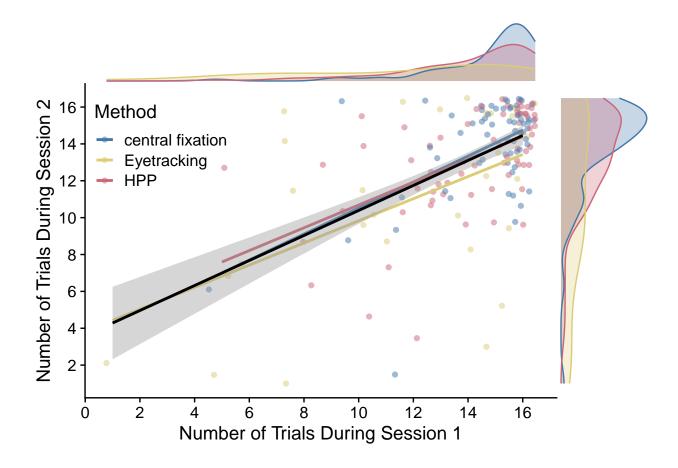


Figure 9. Correlation between the number of trials contributed in Session 1 and Session 2. Each data point represents one infant. Colored lines represent linear fits for each method.

Are there stable individual differences in how likely an infant is to contribute a high number of trials? To answer this question, we conducted an exploratory analysis investigating whether there is a relationship between the number of trials an infant contributed in Session 1 and Session 2. Do infants who contribute a higher number of trials during their first testing session also tend to contribute more trials during their second testing session? A positive correlation between trial numbers during the first and second session would indicate that there is some stability in a given infants' likelihood of remaining attentive throughout the experiment. On the other hand, the absence of a correlation would indicate that the number of trials a given infant contributes is not

predictive of how many trials they might contribute during their next session.

We found a strong positive correlation between number of trials contributed during 375 the first and the second session, r = .58, 95% CI [.47, .67], t(160) = 9.00, p < .001 (Figure 376 9). This result suggests that if infants contribute a higher number of trials in one session, 377 compared to other infants, they are likely to contribute a higher number of trials in their 378 next session. This finding is consistent with the hypothesis that how attentive infants are 379 throughout an experiment (and hence how many trials they contribute) is a stable 380 individual difference, at least for some infant looking time tasks. Researchers should 381 therefore be mindful of the fact that decisions about including or excluding infants based on 382 trials contributed may selectively sample a specific sub-set of the infant population they are 383 studying (Byers-Heinlein, Bergmann, & Savalei, 2021; DeBolt, Rhemtulla, & Oakes, 2020).

S9. Correlations in average looking times between sessions

To what extent are participants looking times between the two sessions related? To 386 test this question, we first investigated whether participants' overall looking times — 387 irrespective of condition — were correlated between the first and second session. There was 388 a robust correlation between average looking time in Session 1 and Session 2: infants with 389 longer looking times during their first session also tended to look longer during their second session, r = .45, 95% CI [.31, .57], t(156) = 6.28, p < .001. This relationship held even after 391 controlling for number of trials in the first and second session, suggesting that the relation between average looking in Session 1 and 2 could not be entirely explained by the 393 correlation in the number of trials contributed between the two sessions (S7), b = 0.42, 95%CI [0.27, 0.58], t(154) = 5.52, p < .001 (Figure 10A). The result is also similar when controlling for participants' average age across the two test sessions, b = 0.44, 95% CI 396 [0.30, 0.59], t(155) = 6.16, p < .001.397

Next, we explored the extent to which average looking times for IDS and ADS stimuli 398 were related. First, we found similar correlations in average looking time to IDS stimuli in 399 Session 1 and 2, r = .38, 95% CI [.24, .51], t(156) = 5.19, p < .001, and ADS stimuli in 400 Session 1 and 2, r = .40, 95% CI [.26, .53], t(156) = 5.49, p < .001 (Figure 10B). To test 401 whether these correlations were specific to looking times for IDS or ADS stimuli alone, we 402 fit linear regression models predicting average looking to IDS (or ADS) stimuli in Session 2 403 from average looking to IDS and ADS stimuli in Session 1. We found that average looking 404 to IDS stimuli in Session 2 could be predicted from average looking to IDS stimuli in Session 1, even after controlling for average looking to ADS stimuli in Session 1, b = 0.21, 95% CI [0.01, 0.41], t(155) = 2.11, p = .037. Conversely, average looking to ADS stimuli in Session 2 could be predicted from average looking to ADS stimuli in Session 1, even after 408 controlling for average looking to IDS stimuli in Session 1, b = 0.36, 95% CI [0.14, 0.58], 409 t(155) = 3.20, p = .002. These results suggest that the condition-specific correlations in 410

average looking time cannot be fully explained by the fact that infants' overall looking times between sessions are correlated.

Finally, we inspected item-level correlations between the two test sessions. 413 Specifically, we investigated the relation between items composed of the same recording 414 clips in Session 1 and Session 2 (but with a reversed order of clips between the two 415 sessions). We fit a linear mixed-effects model predicting item-level looking time in Session 416 2 from item-level looking time in Session 1, including random intercepts for participant, 417 item, and lab, as well as a random slope for item-level looking time in Session 1 for 418 participant and lab. Item-level looking in Session 2 was related to item-level looking in 419 Session 1, $\hat{\beta} = 0.17$, 95% CI [0.07, 0.27], t(5.52) = 3.38, p = .017 (Figure 10C). Similar 420 results hold if looking times are log-transformed. 421

In MB1, the ordering of stimuli was counterbalanced, but some stimuli still appeared 422 earlier in the experiment than others. For example, the IDS1 and ADS1 speech stimuli 423 appeared on trials 1,2,5, or 6, while the IDS8 and ADS8 speech stimuli always occurred on 424 the final two trials (trial number 15 or 16). This means that the interpretation of the 425 correlations between individual speech stimuli must also take into account these stimuli 426 tend to be occurring in earlier or later portions of the experiment (when infants are more 427 or less attentive and show longer looking times in general). To further investigate the 428 impact of trial number on by-item correlations in looking time, we fit an interaction model 429 testing whether the magnitude of the item-level correlation depended on the trial number 430 for a given session. We fit a linear mixed-effects model predicting item-level looking time in 431 Session 2 from the interaction between item-level looking time in Session 1 and trial number in Session 1 (trial numbers across sessions are almost always identical). The model 433 included random intercepts for participant, item, and lab, as well as random slopes for item-level looking time and trial number in Session 1 for participant and lab. We indeed 435 found that the magnitude of the item-level correlations in looking time between sessions 436 depended on trial number, $\hat{\beta} = -0.01$, 95% CI [-0.02, 0.00], t(1, 200.31) = -2.53, p = .012,

- with the strength of the relation between sessions declining as trial number increased.
- While trial number was a strong predictor of Session 2 looking time, $\hat{\beta} = -0.28, 95\%$ CI
- [-0.36, -0.20], t(8.67) = -6.85, p < .001, item-level looking in Session 1 only marginally
- predicted Session 2 looking when controlling for trial number, $\hat{\beta} = 0.10, 95\%$ CI [0.01, 0.20],
- t(6.47) = 2.12, p = .075. Variation in item-level correlations is therefore at least partially
- due to the ordering of the stimuli in the experiment, rather than a sole function of
- differences between the stimuli per se.

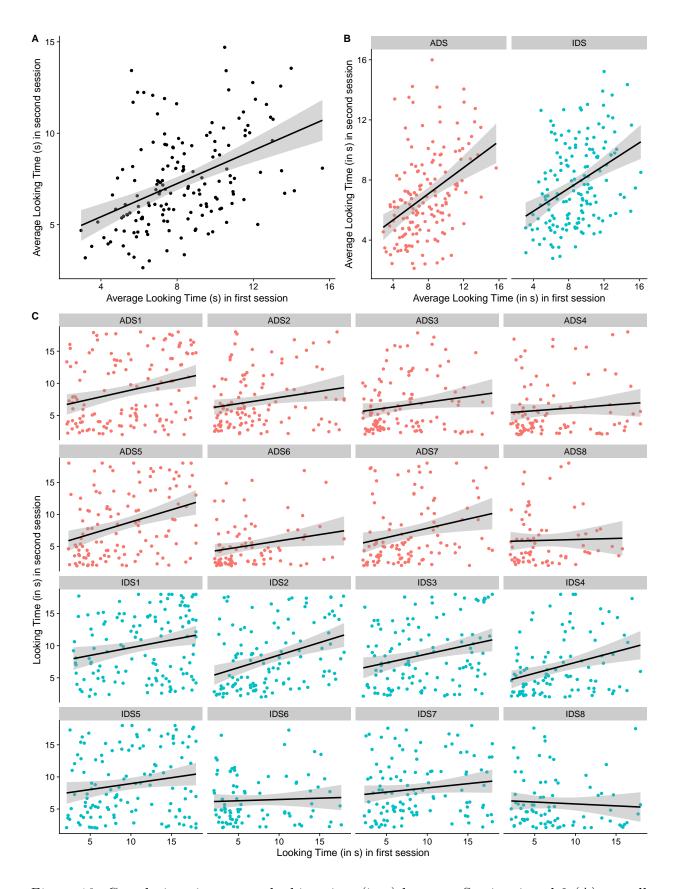


Figure 10. Correlations in average looking time (in s) between Session 1 and 2 (A) overall, (B) by condition, and (C) by item.

Table 6

Linear mixed-effects model results predicting IDS

preference in Session 2 from IDS preference in

Session 1 at the stimulus level.

Term	\hat{eta}	95% CI	t	df	p
Intercept	1.02	[0.14, 1.90]	2.27	6.55	.060
Diff 1	0.07	[-0.01, 0.14]	1.79	718.46	.074

S10. By-item-pair preference scores across sessions

Finally, we inspected on a more fine-grained item level whether IDS preference in 446 Session 1 was related to IDS preference in Session 2. To do so, we exploited the fact the 447 specific IDS and ADS stimuli were paired together in test orders in both sessions, such that one IDS stimulus (e.g., IDS1) always occurred adjacently to a specific ADS stimulus (e.g., 449 ADS1). We therefore computed stimulus-specific IDS preference scores by calculating the 450 difference in raw looking time for each of the eight IDS-ADS stimulus pairs for each 451 participant (whenever both trials in a given pair were available). We then fit a linear mixed-effects model predicting stimulus-specific IDS preference in Session 2 from stimulus-specific IDS preference in Session 1, including by-participant and by-lab random intercepts (models with more complex random effects structure, including by-item random 455 effects, failed to converge). There was a marginal, but non-significant relation in 456 stimulus-specific IDS preference between the two test sessions (Table 6).

S11. Overall looking times and test-retest IDS preference

S11.1. Correlations between overall looking time and IDS preference

We also investigated whether preferential looking behavior varied as a function of infants' tendencies to look for longer or shorter periods of time on average across all stimuli. We found no evidence for a correlation between infants' average looking time and the magnitude of IDS preference (Figure 11), either in Session 1 (r = .05, 95% CI [-.11, .20], t(156) = 0.61, p = .543) or in Session 2 (r = -.06, 95% CI [-.21, .10], t(156) = -0.70, t(156) = -0.70,

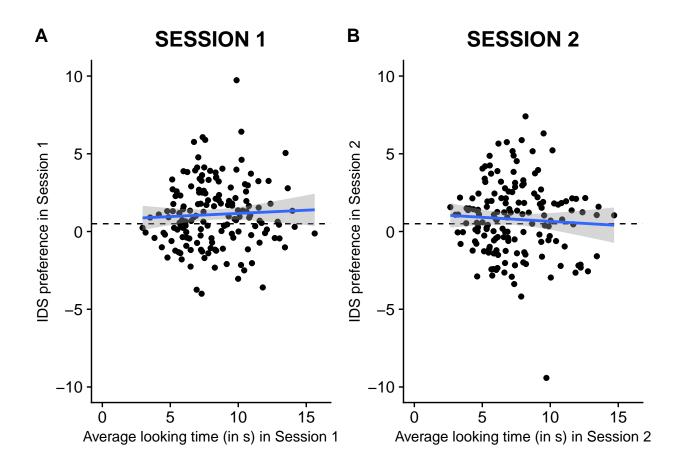


Figure 11. Correlations between average looking time (in s) and IDS preference in (A) Session 1 and (B) Session 2.

66 S11.2. Does average looking time moderate test-retest reliability?

• marginal effect

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• direction of the effect is consistent with a slight increase in test-retest reliability as average looking time increases (Figure 12), but no evidence for robust differences between long and short lookers.

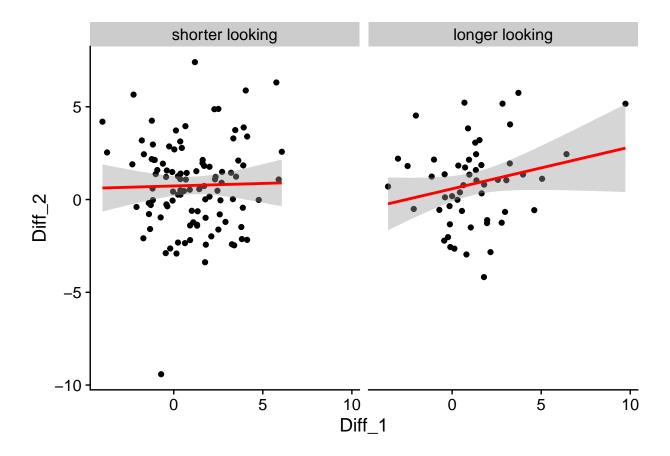


Figure 12. Correlations between average looking time (in s) and IDS preference in (A) Session 1 and (B) Session 2.

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