- Experience with research paradigms relates to infants' direction of preference.
- Chiara Santolin¹, Gonzalo Garcia-Castro¹, Martin Zettersten², Nuria Sebastian-Galles¹, & Jenny Saffran²
- ¹ Center for Brain and Cognition, Universitat Pompeu Fabra
- ² Waisman Center & Department of Psychology, University of Wisconsin-Madison
- 6 Preprint submitted to peer-review on February 17th, 2020. Updated addressing reviewers'
 7 comments on June 4th, 2020.

8 Author Note

- ⁹ Correspondence concerning this article should be addressed to Chiara Santolin,
- Edifici Merce Rodereda, Calle Ramón Trias Fargas, 25, 08018 Barcelona. E-mail:
- chiara.santolin@upf.edu

2

Abstract

Interpreting and predicting direction of preference in infant research has been a thorny 13 issue for decades. Several factors have been proposed to account for familiarity versus novelty preferences, including age, length of exposure, and task complexity. The current 15 study explores an additional dimension: experience with the experimental paradigm. We 16 re-analyzed the data from 4 experiments on artificial grammar learning in 12-month-old 17 infants run using the Head-turn Preference Procedure (HPP). Participants in these studies 18 varied substantially in their number of laboratory visits. Results show that the number of HPP studies is related to direction of preference: infants with limited experience with the HPP setting were more likely to show familiarity preferences than infants who had amassed more experience with this paradigm. This evidence has important implications for the interpretation of experimental results: experience with a given method or, more broadly, with the lab environment, may affect infants' patterns of preferences.

Keywords: preferential looking, familiarity preference, novelty preference, head-turn preference procedure, linear mixed-effects model

27 Word count: 2765

28

Experience with research paradigms relates to infants' direction of preference.

29 Introduction

The importance of changes in preferential looking has been recognized since at least 30 the 1960s, when Fantz (1964) showed that infants preferentially attend to novel visual 31 stimuli. Subsequent studies extended this evidence to domains including auditory 32 perception and cognition, revealing differences in direction of preference. Rather than 33 representing a binary distinction, direction of preference can be construed as a continuum from more familiar to more novel (e.g., Thiessen et al., 2005). The infant's position along this continuum seems to be determined by a variety of factors related to the task and/or age (e.g., Houston-Price & Nakai, 2004; Aslin, 2007; Hunter & Ames, 1988). However, it is frequently the case that the observed direction of preference does not conform with expectations based on these dimensions; the infancy literature is rife with examples of counterintuitive patterns of preference (e.g., Fiser & Aslin, 2001; Bosch & Sebastián-Gallés, 2001; Dawson & Gerken, 2009; DePaolis, Keren-Portnoy, & Vihman, 2016; Johnson et al., 2009; Jusczyk & Aslin, 1995; Sebastián-Gallés & Bosch, 2009; Thiessen, 2012). One frequently-overlooked factor is that infants do not arrive at the lab as naïve 43 participants. Like adults, they bring significant prior experience that may influence their performance in lab tasks. Researchers attempt to override or sidestep those experiences by 45 using novel stimuli (e.g., unfamiliar languages, shapes or sounds), or by integrating those experiences into their experimental designs (e.g., monolingual vs. bilingual infants; see Sebastian-Galles & Santolin, 2020 for a recent review). But there may also be forms of experience that go unidentified by researchers. One such factor is that many infants participate in multiple (putatively unrelated) experiments over the course of weeks or months. This common practice in infant research reflects the challenges of advancing a field of investigation that is based on a limited and hard-to-recruit population. Researchers are typically very careful to avoid stimulus contagion across unrelated studies, but it is

possible that prior lab experience impacts infants' performance. The purpose of this article is to explore the effect of experience with experimental paradigms on direction of preference in learning tasks.

In an influential model of preferential behavior in infants, Hunter and Ames (1988) 57 hypothesized three central factors to affect the strength and direction of preference: age, 58 familiarization duration, and task complexity. In a given task, younger infants tend to prefer familiar stimuli whereas older infants are more likely to prefer novel stimuli (e.g., Colombo & Bundy, 1983; though see Bergmann & Cristia, 2016, for a meta-analysis 61 suggesting that age does not predict shifts in preference). A shorter exposure to familiar 62 stimuli prior to testing also leads infants to subsequently prefer the familiar items (for reviews, see Rose, Feldman, & Jankowski, 2004). Task complexity refers to the stage of stimulus processing. For example, in a visual recognition task, 4-month-old infants preferred familiar objects before subsequently showing a strong preference for the novel object (Roder, Bushneil, & Sasseville, 2000). Task complexity can also refer to the complexity of the stimuli. For example, sequential stimuli put greater strain on memory resources than materials in which all components are simultaneously available (e.g., Ferguson, Franconeri, & Waxman, 2018). A related dimension is the similarity between stimuli used during familiarization and test: when there is a close perceptual match, infants are more likely to show a novelty preference (e.g., Hunter & Ames, 1988; Thiessen & Saffran, 2003). The combination of these factors informs predictions concerning direction of preference in systematic ways. For example, Thiessen, Hill, and Saffran (2005) manipulated length of exposure and observed a flip from familiarity to novelty preference after doubling the amount of familiarization received by infants. Similarly, Ferguson et al. 76 (2018) manipulated sequential vs. spatial presentation of visual patterns, and observed stronger novelty effects with (a) increasing age and (b) spatial presentation. 78

The idea behind the current paper emerged from a puzzling pattern of results in a replication of a published study focused on non-linguistic artificial grammar learning in

12-month-olds (Santolin & Saffran, 2019). We observed a flip in preference from novelty to familiarity between the original study and its replication (Santolin & Saffran, 2019), despite the use of identical stimuli and procedures. While there were some differences 83 between the studies (most notably, in the location in which the studies were run), one main factor stood out to us: many of the infants in the study that elicited a novelty preference had participated in prior studies using the Head-turn Preference Procedure (HPP), whereas most of the infants in the study that elicited a familiarity preference were first-time HPP participants. We reasoned that the more familiarity infants had with the lab apparatus and task demands, the more likely they would be to learn rapidly, leading to a novelty preference. To explore this question, we combined the data from these two experiments with the data from two other published artificial grammar learning tasks with 91 similar designs that included 12-month-olds who ranged in the number of lab visits (Saffran et al., 2008, Exp. 1 Language P, and @saffran2003, Exp. 2). Our hypothesis was that the amount of infants' experience with HPP would affect direction of preference.

95 Methods

A brief description of the four experiments included in this analysis, and our rationale for selecting them, is provided in the Supplementary Information (SI), Section 1 (see Fig. 1 for a summary of the results). Infants were aged between 11-13 months in all studies. A fully reproducible repository hosting data and analyses is available at https://osf.io/g95ub/.

We modeled results of all infants (N = 102) who completed the four studies. Number of HPP visits varied from one to six (including the current visit). We fit a linear mixed-effects model including *Looking Time* as the response variable, and *Test Item* (Familiar vs. Novel), HPP (number of experiments completed by infants) and their interaction as fixed effects. We also included by-participant and by-study random intercepts (4 levels: Santolin & Saffran, 2019, 2019; Saffran et al., 2008; Saffran & Wilson,

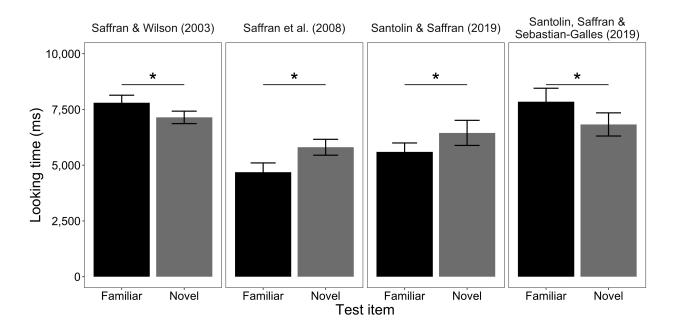


Figure 1. Looking time for familiar and novel test stimuli of the original studies. Stimuli vary based on the experiment. Error bars indicate the standard error of the mean.

107 2003). The *HPP* predictor was coded as a continuous variable indicating each infant's total number of HPP experiments. *Test Item* was centered on familiar test items (Familiar = 0; 109 Novel = 1). Since the experiments differ at distinct levels (e.g., different stimuli, lab 110 location), the model accounted for cross-participant and cross-study differences in looking 111 time. Degrees of freedom were approximated using the Kenward-Rogers approach (e.g., 112 Judd, Westfall, & Kenny, 2012), which can result in non-integer values. See SI, Section 3, 113 for additional details.

We predicted a *Test Item* (familiar vs. novel) by number of *HPP* studies interaction, indicating that the duration of infants' looking towards familiar versus novel items would depend on infants' HPP experience. An interaction could result from at least three different patterns of results: an increase in looking time for novel items, a decrease in looking time for familiar items, or both, as a result of additional HPP experience. 119 Results

The interaction was statistically significant, F(1,100.00) = 11.99, p = .001, suggesting that the effect of Test Items on looking time differences was affected by the number of HPP experiments infants had participated in (Table 1, Fig. 2). In line with our predictions, the size of the difference between looking times on familiar and novel test items changed as a function of number of HPP visits.

The main effect of the HPP predictor was also significant, F(1,133.12) = 4.80, p = .030, indicating that the Test Item by HPP interaction is mainly driven by a significant decrease in looking time to familiar items as the number of HPP visits increases. There was no evidence that a greater number of HPP visits was accompanied by longer looking to novel items, F(1,133.12) = 0.27, p = .606.

Results also hold when reducing the data to infants with less than six HPP visits, indicating that the interaction effect was not driven exclusively by participants with an unusually high number of visits [HPP 1-5: F(1,98.00) = 10.42, p = .002; HPP 1-4: F(1,98.00) = 10.42, p = .002; HPP 1-3: F(1,92.00) = 4.56, p = .035]. Notably, the interaction is significant even with the subset of infants who participated in 1-2 HPP experiments only, HPP 1-2: F(1,78.00) = 4.05, p = .048.

Table 1
Summary of the results of the linear mixed-effects model.

	Coefficient	SEM	95% CI	$oldsymbol{F}$	Den. df	p
Intercept	7,679.0	673.287	[6389.7, 9294.6]	124.638	9.060	< .001
Test Item	-1,398.8	411.324	[-2204.9, -589.1]	11.565	100.000	.001
HPP	-539.7	238.688	[-999.9, -74.7]	4.800	133.117	.030
Test Item \times HPP	667.1	192.645	[247.2, 1028.5]	11.992	100.000	.001

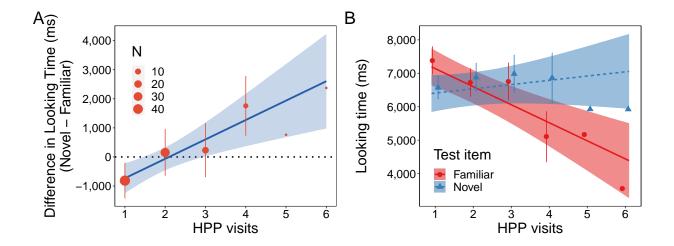


Figure 2. A: Difference in looking time between novel and familiar trials, as a function of HPP visits. shaded bands indicate 95% CIs. Points represente group means, with error bars representing 95% CIs. B: Predicted looking time (in ms) for familiar and novel test items plotted against number of HPP visits. Shaded bands represent +1-1 SEs. Points represent group means with +1-1 SEs as error bars.

Discussion

Experience with the Head-turn Preference Procedure affects direction of preference, at least for the subset of studies examined in this article. The model combined four experiments with 12-month-old infants performing artificial grammar learning tasks.

Infants who had not previously experienced the HPP setting were more likely to show familiarity preferences than infants who had prior experience. One possible explanation for this finding relates to the structure of the HPP task. There are at least two types of information that must be simultaneously encoded during an infant's first HPP experiment:

1) visual-auditory contingency (i.e., sounds appear contingently on the infant looking at the screen), and 2) the experiment stimuli (e.g., word sequences, sound streams). When experiencing HPP for the first time, infants must both learn the structure of the HPP method and solve the learning problem itself (e.g., grammatical pattern learning). Such double-processing of information likely increases the task complexity, biasing results

towards familiarity preferences. Infants who return to the lab for subsequent HPP experiments may be more able to focus on the learning problem, resulting in better learning as evidenced by novelty preferences.

It is important to notice that this effect may not just be limited to experiencing the 152 HPP setting per se, but may also be influenced by the laboratory visit itself. When infants 153 visit the lab for the first time, they face an unusual situation: a new environment with 154 unfamiliar people, testing rooms with a peculiar design (e.g., monochrome walls with big 155 screens), and novel sounds and images (e.g., blinking lights). This is a significant amount 156 of information for a young infant to process at once. In contrast, as infants come back to 157 the lab for subsequent studies, the location, testing room and research staff may become 158 more familiar, reducing the information load (see Rovee-Collier, 1997, for effects of 159 consistent training and testing contexts on reminding infants of details of prior 160 experiences). In the current study, the number of laboratory visits was significantly 161 correlated with the number of HPP visits, r(100 = .92, p < .001, 95% CI = [.88, .94]), 162 therefore the current analyses cannot discern which type of previous experience (HPP 163 procedure and/or lab setting) is responsible for the observed results. 164

Our findings have important implications for the interpretation of directions of 165 preference in future studies. Prior experience with a lab or research paradigm could 166 account for distinct, and sometimes counterintuitive, patterns of preference. We encourage 167 researchers to track number of visits as part of their lab's workflow, and to consider this 168 form of prior experience when preregistering analytic plans and interpreting results. Doing 169 so may be particularly informative when unpredicted directions of preferences emerge, as in the replication that spawned the current set of analyses. It is also possible that apparent 171 null effects may be driven by variability in the number of lab visits; infants with more lab experience may show novelty preferences while infants with less lab experience may exhibit 173 familiarity preferences, leading to an overall lack of preference across the sample. Effects of 174 prior research experience are less likely to be evident in studies with large effect sizes, 175

where there is less intra-infant variability. In addition, apparent age differences may

conceivably be the result not of age per se, but of the number of prior studies, since older

infants are likely to have participated in more experiments than younger infants, on

average. By tracking infants' study participation, it becomes possible to examine these

potential effects, which may be especially apparent in tasks that yield relatively small

effects (as most infant studies do).

A related hypothesis suggests that less-common directions of preference for studies 182 addressing a given topic (e.g., rule learning) likely represent sign errors (a sampling error in 183 which the estimated effect has the wrong sign, e.g. a novelty preference is incorrectly 184 estimated to be a familiarity preference; see also Gelman & Carlin, 2014) as opposed to 185 true infant preferences (Bergmann, Rabagliati, & Tsuji, 2019; Rabagliati, Ferguson, & 186 Lew-Williams, 2019). While this may be the case, it is also possible that some 187 discrepancies in preferential looking are related to factors like those investigated in the 188 current study: prior experience with the testing environment. For this reason, unexpected 189 directions of preference may actually be meaningful and informative about the state of 190 infant learners in specific studies. 191

These results also suggest extensions of models of the factors inducing different 192 patterns of preference (e.g., Hunter & Ames, 1988). The current results suggest that the 193 dimension of task complexity could be expanded beyond the specific task content (e.g., how 194 complex are the stimuli presented) to include infants' familiarity with the paradigm. Our 195 findings, in fact, suggest that the learning outcome of a given task is constrained by how 196 much task experience infants have accumulated through prior lab visits. Therefore, the amount of novel information infants must process in parallel during a study increases the 198 task demands, and the likelihood of showing a familiarity preference. This may well include 199 the novelty of the experimental paradigm. Ongoing efforts in the infant research 200 community to facilitate large-scale replications of studies (e.g., Consortium, 2020) provide 201 a unique opportunity to determine whether experience with different paradigms influences 202

preferential behavior. Expanding our findings to other paradigms (e.g., infant-controlled preferential looking procedures, visual-world paradigms) would continue to advance our understanding of how task/laboratory experience modulates infants' performance. These efforts, in turn, will bring us closer to connecting our research paradigms with the pressing questions about infant behavior that we hope to answer.

208 References

```
Aslin, R. N. (2007). What's in a look? Developmental Science, 10(1), 48–53.
200
          https://doi.org/10.1111/j.1467-7687.2007.00563.x
210
   Bergmann, C., & Cristia, A. (2016). Development of infants' segmentation of words from
211
           native speech: A meta-analytic approach. Developmental Science, 19(6), 901–917.
212
          https://doi.org/10.1111/desc.12341
   Bergmann, C., Rabagliati, H., & Tsuji, S. (2019). What's in a looking time preference?
214
          https://doi.org/10.31234/osf.io/6u453
215
   Bosch, L., & Sebastián-Gallés, N. (2001). Evidence of Early Language Discrimination
216
           Abilities in Infants From Bilingual Environments. Infancy, 2(1), 29–49.
217
           https://doi.org/10.1207/S15327078IN0201 3
218
   Colombo, J., & Bundy, R. S. (1983). Infant response to auditory familiarity and novelty.
219
           Infant Behavior & Development, 6(3), 305-311.
220
           https://doi.org/10.1016/S0163-6383(83)80039-3
221
   Consortium, T. M. (2020). Quantifying sources of variability in infancy research using the
222
          infant-directed-speech preference. Advances in Methods and Practices in
223
           Psychological Science, 3(1), 24–52. https://doi.org/10.1177/2515245919900809
224
   Dawson, C., & Gerken, L. (2009). From Domain-Generality to Domain-Sensitivity:
225
           4-Month-Olds Learn an Abstract Repetition Rule in Music That 7-Month-Olds Do
226
           Not. Cognition, 111(3), 378–382. https://doi.org/10.1016/j.cognition.2009.02.010
227
   DePaolis, R. A., Keren-Portnoy, T., & Vihman, M. (2016). Making sense of infant
228
           familiarity and novelty responses to words at lexical onset. Frontiers in Psychology,
229
           7, 715. https://doi.org/10.3389/fpsyg.2016.00715
230
```

Fantz, R. L. (1964). Visual Experience in Infants: Decreased Attention to Familiar
Patterns Relative to Novel Ones. *Science*, 146(3644), 668–670.

```
https://doi.org/10.1126/science.146.3644.668
233
    Ferguson, B., Franconeri, S. L., & Waxman, S. R. (2018). Very young infants learn
234
           abstract rules in the visual modality. PloS One, 13(1), e0190185.
235
          https://doi.org/10.1371/journal.pone.0190185
236
   Fiser, J., & Aslin, R. N. (2001). Unsupervised statistical learning of higher-order spatial
237
          structures from visual scenes. Psychological Science, 12(6), 499–504.
238
          https://doi.org/10.1111/1467-9280.00392
239
    Gelman, A., & Carlin, J. (2014). Beyond power calculations: Assessing type s (sign) and
240
           type m (magnitude) errors. Perspectives on Psychological Science, 9(6), 641–651.
241
          https://doi.org/https://doi.org/10.1177/1745691614551642
242
   Houston-Price, C., & Nakai, S. (2004). Distinguishing novelty and familiarity effects in
243
          infant preference procedures. Infant and Child Development, 13(4), 341–348.
          https://doi.org/10.1002/icd.364
245
   Hunter, M. A., & Ames, E. W. (1988). A multifactor model of infant preferences for novel
246
          and familiar stimuli. In Advances in infancy research, Vol. 5. (pp. 69–95).
           Westport, CT, US: Ablex Publishing.
248
   Johnson, S. P., Fernandes, K. J., Frank, M. C., Kirkham, N., Marcus, G., Rabagliati, H., &
          Slemmer, J. A. (2009). Abstract Rule Learning for Visual Sequences in 8- and
           11-Month-Olds. Infancy: The Official Journal of the International Society on
251
           Infant Studies, 14(1), 2–18. https://doi.org/10.1080/15250000802569611
    Judd, C. M., Westfall, J., & Kenny, D. A. (2012). Treating stimuli as a random factor in
```

- Judd, C. M., Westfall, J., & Kenny, D. A. (2012). Treating stimuli as a random factor in
 social psychology: A new and comprehensive solution to a pervasive but largely
 ignored problem. *Journal of Personality and Social Psychology*, 103(1), 54–69.
 https://doi.org/10.1037/a0028347
- Jusczyk, P. W., & Aslin, R. N. (1995). Infants' detection of the sound patterns of words in

```
fluent speech. Cognitive Psychology, 29(1), 1–23.
258
          https://doi.org/10.1006/cogp.1995.1010
250
   Rabagliati, H., Ferguson, B., & Lew-Williams, C. (2019). The profile of abstract rule
260
          learning in infancy: Meta-analytic and experimental evidence. Developmental
261
           Science, (1), e12704. https://doi.org/10.1111/desc.12704
262
   Roder, B. J., Bushneil, E. W., & Sasseville, A. M. (2000). Infants' Preferences for
263
          Familiarity and Novelty During the Course of Visual Processing. Infancy, 1(4),
264
          491–507. https://doi.org/10.1207/S15327078IN0104 9
265
   Rose, S. A., Feldman, J. F., & Jankowski, J. J. (2004). Infant visual recognition memory.
266
           Developmental Review, 24(1), 74–100. https://doi.org/10.1016/j.dr.2003.09.004
267
   Saffran, J., Hauser, M., Seibel, R., Kapfhamer, J., Tsao, F., & Cushman, F. (2008).
268
           Grammatical pattern learning by human infants and cotton-top tamarin monkeys.
269
           Cognition, 107(2), 479–500. https://doi.org/10.1016/j.cognition.2007.10.010
270
   Saffran, J. R., & Wilson, D. P. (2003). From Syllables to Syntax: Multilevel Statistical
271
          Learning by 12-Month-Old Infants. Infancy, 4(2), 273–284.
          https://doi.org/10.1207/S15327078IN0402_07
273
   Santolin, C., & Saffran, J. R. (2019). Non-Linguistic Grammar Learning by 12-Month-Old
274
          Infants: Evidence for Constraints on Learning. Journal of Cognition and
           Development, 20(3), 433-441. https://doi.org/10.1080/15248372.2019.1604525
276
   Sebastian-Galles, N., & Santolin, C. (2020). Bilingual acquisition: The early steps. Annual
           Review of Developmental Psychology (in Press).
278
   Sebastián-Gallés, N., & Bosch, L. (2009). Developmental shift in the discrimination of
          vowel contrasts in bilingual infants: Is the distributional account all there is to it?
280
           Developmental Science, 12(6), 874-887.
281
```

https://doi.org/10.1111/j.1467-7687.2009.00829.x

282

- ²⁸³ Thiessen, E. D. (2012). Effects of inter- and intra-modal redundancy on infants' rule
- learning. Language Learning and Development, 8(3), 197–214.
- https://doi.org/10.1080/15475441.2011.583610
- Thiessen, E. D., Hill, E. A., & Saffran, J. R. (2005). Infant-Directed Speech Facilitates
- Word Segmentation. Infancy, 7(1), 53–71.
- https://doi.org/10.1207/s15327078in0701_5
- Thiessen, E. D., & Saffran, J. R. (2003). When cues collide: Use of stress and statistical
- cues to word boundaries by 7- to 9-month-old infants. Developmental Psychology,
- 39(4), 706–716. https://doi.org/10.1037/0012-1649.39.4.706

Appendix A

S1: Experiments included in the linear mixed-effects model.

The selected experiments consist of an artificial grammar learning task with 293 12-month-old infants. These experiments are characterized by variability in the number of 294 infants' prior HPP visits¹, the first author noted that there appeared to be an association 295 between the number of prior studies completed by the infants and the direction of 296 preference. The analysis was included in the original manuscript submission but was 297 removed from later revisions based on reviewer suggestions.]. They include all studies run 298 in the two senior authors' labs that included (a) 11- to 13-month-old participants; (b) 299 HPP; (c) artificial grammar learning (linguistic or non-linguistic); (d) 2 to 5 minutes of 300 exposure; (e) an a priori hypothesis that infants would show learning; (f) visit numbers 301 recorded at the time of testing. The studies are thus as well matched as is possible given 302 the retrospective nature of this analysis. 303

Saffran & Wilson (2003) demonstrated that 12-month-old infants can compute multiple regularities from a finite-state grammar. Infants were able to first segment words from running speech based on transitional probabilities, then detect permissible orderings of the segmented words. Test items consisted of grammatical and ungrammatical sentences that could only be discriminated based on word-level information (transitional probabilities between syllables were not informative about the "grammaticality" of test items). Infants showed a significant familiarity preference: F(1, 38) = 5.37, p < .05.

Saffran, Hauser, Seibel, Kapfhamer, Tsao, & Cushman (2008) demonstrated that
infants could detect simple phrases (i.e., clusters of nonsense words grouped together based
on statistical regularities) from artificial grammars. In Exp. 1, infants in the Predictive
Language condition were familiarized with a grammar including predictive (statistical)

 $^{^{1}}$ At the time of publication of Saffran & Wilson (2003)

dependencies between words. The test items consisted of familiar sentences vs. novel 315 sentences violating the grammar. Infants showed a significant novelty preference: t(11) =316 2.52, p < .05.317 Santolin & Saffran (2019) is a conceptual replication of Saffran et al. (2008) using 318 non-linguistic sounds (e.g., computer alert sounds) to implement the grammars. Infants 319 exposed to the Predictive language showed a significant novelty preference: t(26) = 2.45, p 320 = .021, d = 0.47.321 Santolin & Saffran (2019) is a conceptual replication of Saffran et al. (2008) using 322 non-linguistic sounds (e.g., computer alert sounds) to implement the grammars. Infants 323

santonn & Saffran (2019) is a conceptual replication of Saffran et al. (2008) using non-linguistic sounds (e.g., computer alert sounds) to implement the grammars. Infants exposed to the Predictive language showed a significant novelty preference: t(26)=2.45, p=.021, d=0.47.

We replicated the Predictive Language condition of Santolin & Saffran (2019) at the
University Pompeu Fabra, Barcelona (Santolin, Saffran & Sebastian-Galles, 2019)(2019),
using identical stimuli and procedures. We found significant discrimination of the test
stimuli but observed the opposite direction of preference: infants listened longer to familiar
than novel strings: t(23) = 2.30, p = .030, d = 0.47. All results are shown in Figure 1 of
the main manuscript.

332 S2: Participants information

We retrieved data from 102 infants who had participated in a range of 1-6 HPP visits. Three of the experiments were run in Madison, WI (University of Wisconsin-Madison): Saffran & Wilson, 2003 (Exp. 2; N=40, mean age: 11.5 months); Saffran et al., 2008 (Exp. 1, Condition P-Language: N=12, mean age: 12.8 months); Santolin & Saffran, 2019 (Condition 1; N=26, mean age: 12.9 months). One study was run in Barcelona, Spain (Universitat Pompeu Fabra): Santolin, Saffran & Sebastian-Galles, 2019 (N=24, mean age: 13 months).

S3: Linear mixed-effects model - additional information

We fit a model predicting looking time (LT) including Item (Familiar vs. Novel), 341 number of Head-turn Preference Procedure experiments completed by infants (HPP), and 342 their interaction ($Item \times HPP$) as fixed effects. Participant and study [4 levels: Santolin 343 & Saffran (2019), Santolin, Saffran & Sebastian-Galles (2019), Saffran et al. (2008), Saffran 344 & Wilson (2003)] were included as random effects. Following Barr, Levy, Scheepers, & Tily 345 (2013), we fit a model with the maximal random effects structure including random 346 intercepts by-participant and by-study, and random slopes of HPP by-participant and 347 by-study. However, due to lack of convergence, we pruned the random effects structure until convergence was achieved (e.g., Brauer & Curtin, 2018). The final model included by-participant and by-study random intercepts only. This model accounts for cross-participant variability in overall looking time (as some infants look longer than 351 others), and for cross-study differences in overall looking time. The model was fit using the 352 1me4 R package (Bates, Kliegl, Vasishth, & Baayen, 2015). We used the Anova function 353 from the car R package (Fox & Weisberg, 2019) to perform F-tests on fixed effects using 354 Kenward-Roger's approximation of the degrees of freedom (e.g., Judd, Westfall, & Kenny, 355 2012). 356

S4: Results sub-setting data to participants with less than 6, 5, 4, and 3 HPP studies

Consistent with the results of the entire dataset, we found a statistically significant interaction of *Test Item* with the number of *HPP* visits when reducing the sample to the infants who participated in less than 6, 5, 4, and 3 HPP experiments. Below, a table reporting the output of the linear mixed-effects model fitted on the original and reduced samples (Table S4.1).

Subset	Term	Coefficient	SEM	95% CI	${f F}$	Den. df	p
Original	Intercept	7,679.0	673.3	[6389.7, 9294.6]	124.6	9.1	< .001
	Test Item	-1,398.8	411.3	[-2204.9, -589.1]	11.6	100.0	.001
	HPP	-539.7	238.7	[-999.9, -74.7]	4.8	133.1	.030
	$\textit{Test Item} \times \textit{HPP}$	667.1	192.6	[247.2, 1028.5]	12.0	100.0	.001
HPP 1-5	Intercept	7,611.1	719.3	[6188.6, 9275.3]	107.4	10.5	< .001
	Test Item	-1,491.1	452.1	[-2348.4, -578.5]	10.9	98.0	.001
	HPP	-500.8	278.7	[-1070.2, 98.2]	3.0	131.9	.083
	$Test\ Item\ \times\ HPP$	726.2	224.9	[294.2, 1145]	10.4	98.0	.002
HPP 1-4	Intercept	7,611.1	719.3	[6188.6, 9275.3]	107.4	10.5	< .001
	$Test\ Item$	-1,491.1	452.1	[-2348.4, -578.5]	10.9	98.0	.001
	HPP	-500.8	278.7	[-1070.2, 98.2]	3.0	131.9	.083
	$Test\ Item\ \times\ HPP$	726.2	224.9	[294.2, 1145]	10.4	98.0	.002
HPP1-3	Intercept	7,470.0	794.6	[6007.1, 9172.6]	83.8	14.3	< .001
	Test Item	-1,349.9	532.3	[-2426.9, -267.6]	6.4	92.0	.013
	HPP	-395.8	366.6	[-1182.6, 316.1]	1.1	122.3	.299
	Test Item \times HPP	627.9	294.1	[3, 1252.6]	4.6	92.0	.035
HPP 1-2	Intercept	7,301.9	1009.9	[5253.3, 9362.1]	48.9	23.9	< .001
	$Test\ Item$	-1,783.7	726.7	[-3199.6, -360.7]	6.0	78.0	.016
	HPP	-261.3	586.8	[-1449.8, 991.7]	0.2	107.0	.667
	Test Item \times HPP	969.5	481.8	[38.3, 2010.4]	4.0	78.0	.048

In all models, the $Item \times HPP$ interaction term was statistically significant [HPP 1-5: F(1, 98) = 10.42, p = .002; HPP 1-4: F(1, 98) = 10.42, p = .002; HPP 1-3: F(1, 92) = 4.56, p = .035; HPP 1-2: F(1, 78) = 4.05, p = .048]. These results provide evidence that the effect of HPP on the preference pattern we observed in our main analysis is not entirely dependent on any subsample of the HPP variable.

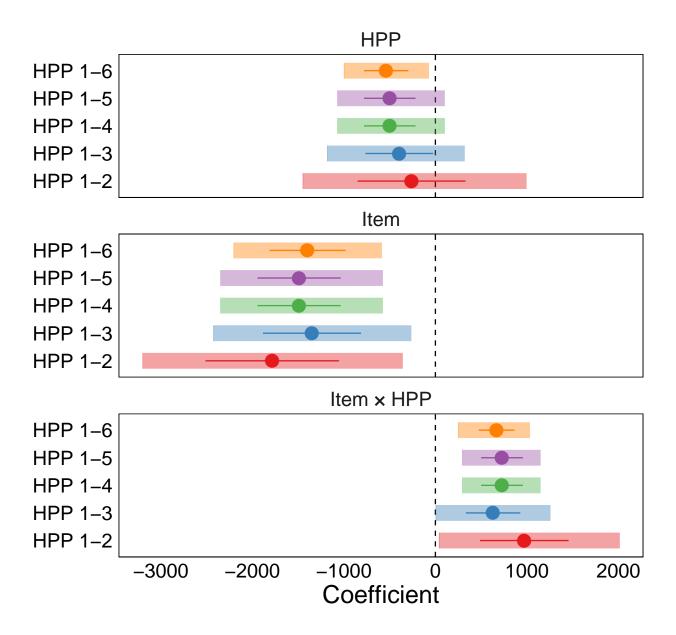


Figure A1. Graphical representation of the coefficients of the Test Item by HPP visits interaction term of the model fitted on the complete data-set (reported in the main manuscript), and of the models fitted on the reduced data-sets. Black dots represent the point estimate of the coefficient, black whiskers represent the standard error of the mean, and grey boxes represent the bootstrapped 95% confidence interval around the point estimate.

Appendix B

Session info

```
R version 3.6.3 (2020-02-29) Platform: x86 64-apple-darwin15.6.0 (64-bit) Running under:
   macOS Catalina 10.15.4
         Matrix products: default BLAS:
371
    Library/Frameworks/R.framework/Versions/3.6/Resources/lib/libRblas.0.dylib
372
   LAPACK:
373
    /Library/Frameworks/R.framework/Versions/3.6/Resources/lib/libRlapack.dylib
374
         locale: [1]
   en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
         attached base packages: [1] stats graphics grDevices utils datasets methods base
377
         other attached packages: [1] purrr_0.3.4 kableExtra_1.1.0 ggplot2_3.3.1 here_0.1
378
         [5] tibble 3.0.1 dplyr 1.0.0 magrittr 1.5 knitr 1.28
379
         [9] papaja_0.1.0.9942
         loaded via a namespace (and not attached): [1] Rcpp_1.0.4.6 pillar_1.4.4
381
   compiler 3.6.3 highr 0.8
382
         [5] base64enc 0.1-3 tools 3.6.3 digest 0.6.25 viridisLite 0.3.0 [9] evaluate 0.14
383
   lifecycle_0.2.0 gtable_0.3.0 pkgconfig_2.0.3
384
         [13] rlang_0.4.6 rstudioapi_0.11 yaml_2.2.1 xfun_0.14
385
         [17] xml2_1.3.1 httr_1.4.1 withr_2.2.0 stringr_1.4.0
386
         [21] hms 0.5.3 generics 0.0.2 vctrs 0.3.0 webshot 0.5.2
         [25] rprojroot 1.3-2 grid 3.6.3 tidyselect 1.1.0 glue 1.4.1
388
         [29] R6_2.4.1 rmarkdown_2.1 bookdown_0.18 readr_1.3.1
389
         [33] backports 1.1.7 scales 1.1.1 ellipsis 0.3.1 htmltools 0.4.0
390
         [37] rvest 0.3.5 colorspace 1.4-1 stringi 1.4.6 munsell 0.5.0
391
         [41] crayon_1.3.4
392
```

93 References

414

415

```
Barr, D. J., Levy, R., Scheepers, C., & Tily, H. J. (2013). Random effects structure for
394
           confirmatory hypothesis testing: Keep it maximal. Journal of Memory and
395
           Language, 68(3), 255–278. https://doi.org/10.1016/j.jml.2012.11.001
396
   Bates, D., Kliegl, R., Vasishth, S., & Baayen, H. (2015). Parsimonious Mixed Models.
           arXiv:1506.04967 [Stat]. Retrieved from http://arxiv.org/abs/1506.04967
   Brauer, M., & Curtin, J. J. (2018). Linear mixed-effects models and the analysis of
390
          nonindependent data: A unified framework to analyze categorical and continuous
400
          independent variables that vary within-subjects and/or within-items. Psychological
401
           Methods, 23(3), 389-411. https://doi.org/10.1037/met0000159
402
   Fox, J., & Weisberg, S. (2019). An R companion to applied regression (Third). Thousand
403
           Oaks CA: Sage. Retrieved from
404
          https://socialsciences.mcmaster.ca/jfox/Books/Companion/
405
   Judd, C. M., Westfall, J., & Kenny, D. A. (2012). Treating stimuli as a random factor in
          social psychology: A new and comprehensive solution to a pervasive but largely
407
          ignored problem. Journal of Personality and Social Psychology, 103(1), 54–69.
408
          https://doi.org/10.1037/a0028347
409
   Saffran, J., Hauser, M., Seibel, R., Kapfhamer, J., Tsao, F., & Cushman, F. (2008).
410
           Grammatical pattern learning by human infants and cotton-top tamarin monkeys.
411
           Cognition, 107(2), 479–500. https://doi.org/10.1016/j.cognition.2007.10.010
412
   Saffran, J. R., & Wilson, D. P. (2003). From Syllables to Syntax: Multilevel Statistical
413
```

Santolin, C., & Saffran, J. R. (2019). Non-Linguistic Grammar Learning by 12-Month-Old
Infants: Evidence for Constraints on Learning. *Journal of Cognition and*

Learning by 12-Month-Old Infants. Infancy, 4(2), 273–284.

https://doi.org/10.1207/S15327078IN0402_07

Development, 20(3), 433–441. https://doi.org/10.1080/15248372.2019.1604525

Santolin, C., Saffran, J. R., & Sebastian-Galles, N. (2019). Non-linguistic artificial

grammar learning in 12-month-old infants: A cross-lab replication study. In.

Potsdam, Germany.