

Running Head: PATTERN EXTRACTION BY BABOONS

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Extraction of Structural Regularities by Baboons (*Papio papio*):

Adjacent and Non-adjacent Repetition Patterns Differ in Learnability

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1 **Abstract**

2 Several animal species can discriminate between different sequential patterns based on
3 repetitions of items (e.g., ABB *versus* ABA), and generalize their performance to sequences
4 made of novel elements but following the same underlying structure. This achievement
5 suggests that these species possess the ability to abstract the sequences structures beyond
6 exemplars. Developmental studies in humans suggest however that adjacent repetition
7 patterns (e.g. ABB) are easier to extract than patterns based on a non-adjacent (e.g., ABA)
8 repetitions. This study addressed this effect of adjacency from a comparative perspective.
9 Seventeen Guinea baboons (*Papio papio*) were tested with an adaptation of the serial
10 response time task. In Experiment 1, baboons exposed to adjacent or non-adjacent repetition
11 patterns showed shorter response times in responses to novel adjacent and non-adjacent
12 repetition sequences compared to random sequences, suggesting successful generalization of
13 both patterns. In Experiment 2, sequences of the form ABB, ABA and ABC were presented at
14 equal frequencies to test whether baboons would be sensitive to repetitions under these
15 conditions. Response times were faster for adjacent repetitions (ABB) than in the two other
16 conditions. Overall, these results suggest that baboons are sensitive to both adjacent and non-
17 adjacent repetition patterns, but also confirm in baboons that adjacent repetitions are more
18 salient than non-adjacent repetitions.

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20 Key words: rule learning, artificial grammar learning, pattern learning, monkey

21

Introduction

22 When reading the following three-syllable sequences, « ga ti ga » and « li na li », we
23 can memorize each of them individually, but can as well notice that they share a common
24 structure: their first and last syllables are identical while their second syllable differs. The
25 extraction of such an « ABA » pattern requires the processing of the structural properties of
26 the sequences beyond the exemplars, allowing generalization to novel, isomorphic sequences.
27 In humans, sensitivity to the structural organization of words within sentences has been
28 proposed as one of the core mechanisms underlying the formation of grammatical knowledge
29 (e.g., Gomez & Gerken, 1999; Marcus, Vijayan, Rao, & Vishton, 1999). The comparative
30 study of this ability therefore offers an entry into the assessment of non-human animals'
31 syntactic abilities.

32 The extraction of sequential patterns has been extensively studied in laboratories using
33 AAB, ABB and ABA patterns based on repetitions. In the AAB/ABB patterns, there is an
34 immediate repetition of an item within the triplet (« adjacent repetition patterns »). In the
35 ABA pattern, the first and last elements of the triplets are identical, while the element in the
36 middle differs (« non-adjacent repetition pattern »). In a seminal experiment conducted by
37 Marcus, Vijayan, Rao, and Vishton (Experiments 1 and 2, 1999), seven-month-old infants
38 were exposed to triplets of syllables following either an ABB or ABA pattern (e.g., « ga ti ti
39 », « ga ti ga », respectively). They were then tested with pattern-following sequences, and
40 pattern-violating sequences made of novel syllables of the form ABA for the group exposed
41 to ABB sequences (e.g., « wo fe wo »), and *vice-versa*. Infants from both groups showed
42 preference for pattern-violating sequences compared to pattern-following ones, suggesting
43 that they were able to learn and generalize the underlying structures.

44 An extensive number of studies was conducted to assess pattern learning abilities in
45 nonhuman species, following Marcus et al. (1999). Most of these studies used operant
46 conditioning procedures. With these procedures, the subjects are typically first trained to
47 discriminate two different repetition patterns in a Go/No-go (e.g., Murphy, Mondragón, &
48 Murphy, 2008) or a two-alternative forced choice task (e.g., Santolin, Rosa-Salva, Regolin, &
49 Vallortigara, 2016). Individuals are then tested with generalization trials involving sequences
50 made of novel stimuli.

51 Convincing generalization performance was obtained in several species with that
52 procedure, participants showing correct discrimination in generalization trials. In rodents, rats
53 could discriminate an AAB pattern from random ABC sequences, with the A and B items
54 being either vowels (Experiment 1) or consonants (Experiment 2) presented in auditory
55 sequences, and transferred this knowledge to novel phonemes (*Rattus norvegicus*, de la Mora
56 & Toro, 2013; see also Murphy et al., 2008). In birds, Spierings & ten Cate (2016) have
57 shown that budgerigars (*Melopsittacus undulatus*) can discriminate sequences of song
58 elements of the form AAB and ABA and generalize their performance to sequences
59 constructed with a novel set of songs. Chicks also successfully discriminated ABA and ABB
60 patterns, as well as ABB and AAB patterns, in a task relying on a non-sequential presentation
61 of strings of visual shapes (*Gallus gallus*, Santolin, Rosa-Salva, Regolin, & Vallortigara,
62 2016; see also Versace, Spierings, Caffini, ten Cate, & Vallortigara, 2017). In an experiment
63 using a similar procedure, two chimpanzees correctly generalized an ABⁿA pattern to novel
64 colors and shapes (with one to five intervening B elements; *Pan troglodytes*, Sonnweber,
65 Ravignani, & Fitch, 2015; see also Ravignani & Sonnweber, 2017). Finally, cotton-top
66 tamarins respond differentially to novel pattern-following and pattern-violating sequences
67 (AAB vs. ABB) made of speech elements or tones (*Saguinus oedipus*, Neiwirth et al., 2017).
68 All these data suggest that the ability to discriminate between different patterns, based on

69 initial adjacent repetition (AAB), final adjacent repetition (ABB), and non-adjacent repetition
70 (ABA) is widespread among animal species.

71 These studies, however, convey at least one limitation, as none of them has assessed
72 whether those different repetition patterns differ in learnability. Notably, one possibility
73 would be that patterns based on an adjacent repetition are easier to process than a pattern
74 based on a non-adjacent repetition. Discrimination tasks contrasting one repetition pattern
75 against another (e.g. AAB *vs.* ABA) are not suitable to assess this question, as participants
76 can perform the task by processing only one pattern (e.g., by detecting the presence/absence
77 of an immediate repetition; ten Cate & Okanoya, 2012; Corballis, 2009), without processing
78 the full structure underlying the other pattern(s). To solve this problem, researchers working
79 with humans have for instance added random ABC sequences in their designs, and therefore
80 contrasted ABA and ABB patterns against ABC sequences. This procedure revealed that the
81 ability to process adjacent and non-adjacent repetition patterns follow different developmental
82 trajectories, the former appearing earlier than the latter (Gervain, Macagno, Coggi, Pena, &
83 Mehler, 2008; Gervain & Werker, 2013; see also Johnson et al., 2009; Kovács & Mehler,
84 2009). One study suggests that this difference in learnability can also be found in birds (van
85 Heijningen et al., 2013, Experiment 1). These authors trained zebra finches to respond to
86 ABA (S+), but not to AAB/ABB (S-) sequences in a Go/No-go task. Probe tests involving
87 sequences such as ABBA have suggested in this study that the finches responded by attending
88 to the presence/absence of adjacent repetitions only. Therefore, it seems that the extraction of
89 adjacent repetition patterns is easier than the extraction of non-adjacent repetition ones, but
90 evidence from comparative studies remain scarce.

91 The aim of the current experiments was therefore to further assess to what extent
92 adjacent repetition patterns may differ in learnability from a non-adjacent repetition pattern.
93 For this purpose, we assessed the sensitivity of non-human primate species, the Guinea

94 baboon (*Papio papio*), to each kind of repetition pattern. Experiment 1 assessed baboons'
95 ability to learn and generalize those patterns. Our group of participants was of baboons split in
96 two halves. One group was exposed to sequences following an adjacent repetition pattern, and
97 the other group was exposed to sequences following a non-adjacent repetition pattern. We
98 then tested their ability to discriminate each repetition pattern against random sequences (for a
99 similar approach with humans, see Gervain & Werker, 2013, Experiment 1). After completing
100 those tests, the groups were tested on the other pattern. This procedure allowed us to assess
101 baboons' sensitivity to each pattern, successively and in the same task and in the same
102 individuals. In Experiment 2, adjacent repetition (ABB), non-adjacent repetition (ABA) and
103 random (ABC) patterns were presented at equal frequencies, to test whether baboons would
104 be sensitive to adjacent and non-adjacent repetitions under these conditions, only to the
105 former, or to none of them.

106 **Experiment 1**

107 Experiment 1 aimed at testing baboons' ability to learn and generalize an adjacent
108 (AA) and a non-adjacent repetition pattern (ABA) by contrasting these patterns with random
109 sequences (AB and ABC, respectively). The baboons were first exposed to sequences of
110 visual shapes instantiating each kind of repetition pattern, and then tested on sequences made
111 of novel shapes either following or violating (random sequences) the repetition pattern.

112

113 **Methods**

114 **Subjects and apparatus**

115 Participants were 17 Guinea baboons (*Papio papio*, 11 females, age range 4-21 years).
116 Six other baboons were tested but not included because they did not complete the experiment
117 within the time allowed to this experiment. The baboons live within a larger group of

118 individuals, within a 700 m² outdoor enclosure and had permanent access to ten Automated
119 Learning Devices for Monkeys (ALDM, for a detailed description, see Fagot & Bonté, 2010;
120 Fagot & Paleressompouille, 2009), each equipped with a 19-inch touch screen and a food
121 dispenser. The main feature of ALDM equipment is that a radio frequency identification
122 reader (RFID) identifies each baboon via microchips implanted in their arms. The baboons
123 can therefore participate to research at will, as the test programs recognize the individuals
124 automatically. The experiment was controlled by the EPrime software (Version 2.0,
125 Psychology Software Tools, Pittsburgh).

126 Prior to the experiment, the baboons have never been tested on sequential pattern
127 learning, but they had participated in numerous computerized experiments using the ALDM
128 test systems, including some experiments on statistical learning, involving chunking of
129 triplets of stimuli (Minier, Fagot, & Rey, 2016), and learning of non-adjacent dependencies
130 (Malassis, Rey, & Fagot, 2018).

131 **Stimuli**

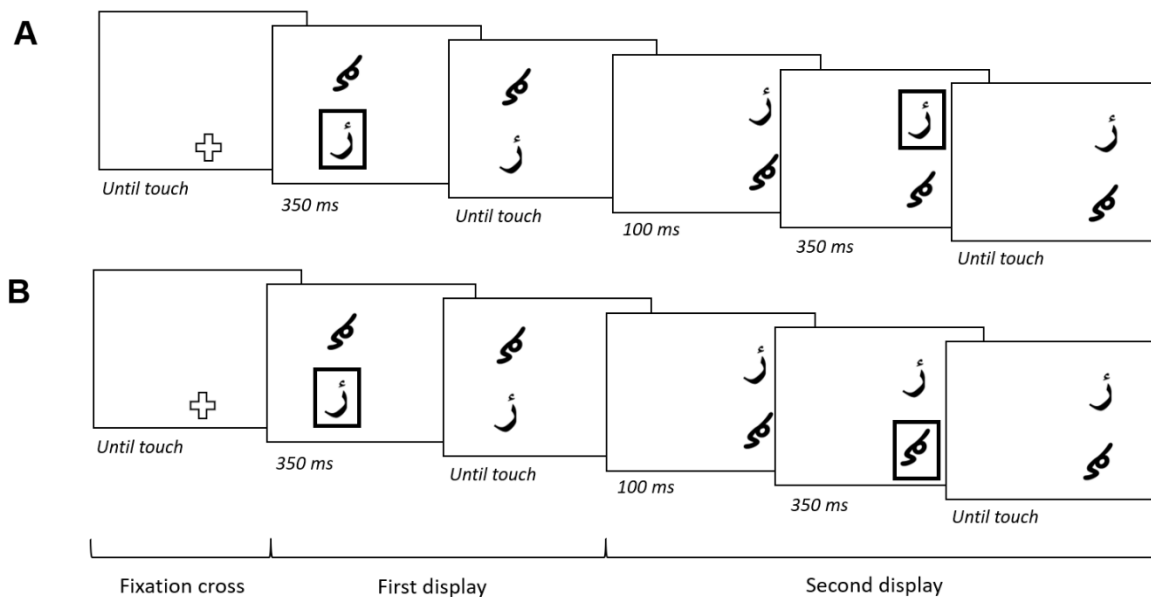
132 They were 192 different yellow visual shapes (available in Supplementary
133 Information) selected from the Unicode 9.0 repertoire considering their apparent
134 discriminability. Each shape measured 180x180 pixels maximum.

135 **Task and test procedure**

136 We used an adaptation of the serial response time task (SRT, Nissen & Bullemer,
137 1987), inspired from Misyak, Christiansen, and Tomblin (2010). This task involves self-paced
138 presentation of sequences of cued stimuli in which response times (RT) provide a continuous
139 measure of the participants' expectations about the incoming stimulus. In this task, RTs are
140 expected to become increasingly fast throughout exposure, if participants are sensitive to the
141 sequences structure. By contrast, they should slow down if a sequence violating this structure
142 is encountered.

143 Figure 1 illustrates the procedure used in each trial. A trial (see Figure 1) began by the
144 display of a fixation cross (120x120 pixels) presented at the bottom of the screen. Touching
145 this stimulus triggered the display of two shapes, presented one above the other (upper/lower
146 location balanced) on the left side of the screen. One of the two shape was surrounded by a
147 square (empty fill, white border; hereafter “cue”) which served to designate the positive
148 stimulus (S+) that the baboon had to touch. The square appeared around S+, and then
149 disappeared from the screen after 350 ms, in contrast to the two stimuli which remained on
150 the screen until the subject touched one of them. A correct touch of S+ immediately triggered
151 the display of a second pair of shapes. Presentation of this second pair of shapes used the
152 same procedure (for shape and cue presentation) as for the first pair, but the two stimuli were
153 now presented on the right side of the screen, and the cue appeared around S+ after a delay of
154 100 ms. A correct selection of S+ in this second display delivered grains of dry wheat. An
155 incorrect selection of S- on either the first or the second display immediately stopped the trial
156 and triggered a 3-sec timeout without food reward. The baboons could respond to each pair of
157 shapes during or after the appearance of the cue. Trials in which the baboon failed to select a
158 stimulus within five seconds (from trial onset) were aborted and presented again immediately.
159 The procedure outlined above was followed for the presentation of the adjacent repetition
160 pattern. It was also used for presenting the non-adjacent repetition pattern, with the exception
161 that these trials involved three instead of two displays, which were presented sequentially in
162 left, center and right part of the screen.

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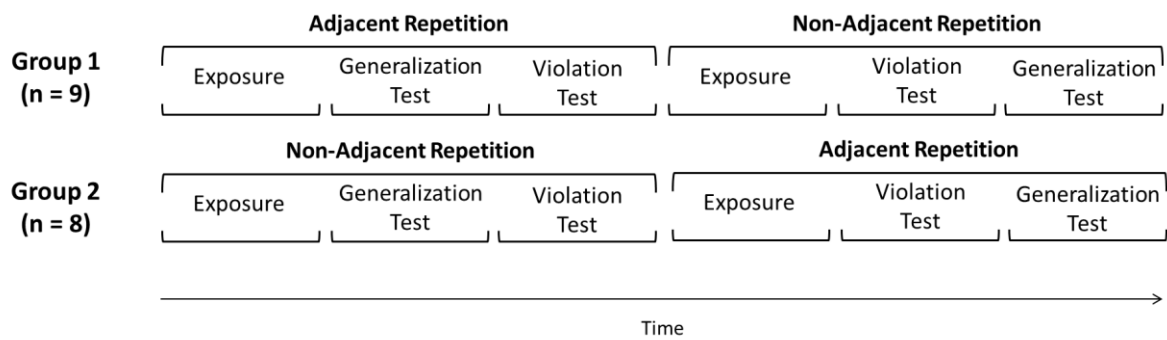


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Figure 1. Timeline of a trial in Experiment 1. A. Trial presenting an adjacent repetition pattern-following sequence (of the form AA). B. Pattern-violating sequence (of the form AB). The task was similar for non-adjacent repetition trials, except that three instead of two displays were presented.

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Our procedure aimed at comparing baboons' responses to repetition patterns to their responses to random sequences. Figure 2 provides an overview of the experiment timeline. The baboons were tested with the two repetition patterns, one after the other, in a within-subjects design (order counterbalanced across participants; Group 1: $n = 9$, Group 2: $n = 8$). For each pattern, the baboons performed three phases: an exposure phase, a generalization test, and a violation test. These phases were presented sequentially, and the order of the two test phases was counterbalanced across baboons. Our collection of 192 shapes was divided in six sets of 32 shapes each (see Supplementary Information). Each set served to generate the sequences for a given phase and pattern. The sets involved in generalization and violation phases were counterbalanced across baboons. Table 1 provide a summarized description of the phases and conditions involved in Experiment 1.



184
 185 *Figure 2.* Timeline of Experiment 1. The two pattern experiments (adjacent repetition, non-
 186 adjacent repetition) were conducted one after the other, their order being counterbalanced
 187 across participants. The order of the test phases was also counterbalanced across participants
 188 within each pattern experiment.
 189

190 Two types of sequences were presented to the baboons during the exposure phase:
 191 baseline (95% of the trials) and violation (5%) sequences. For the adjacent repetition pattern,
 192 baseline sequences were of the form AA, and violation sequences were of the form AB, with
 193 the A and B stimuli being drawn from the same set of shapes. For the non-adjacent pattern,
 194 these two types of sequences were of the form ABA and ABC, respectively, A, B and C
 195 stimuli being all drawn from the same set of shapes. In violation trials, the cue appeared
 196 around the distractor (i.e., around “B” instead of “A” for the adjacent repetition pattern, and
 197 around “C” instead of “A” for the non-adjacent repetition pattern). Baboons were expected to
 198 select the cued stimuli. Because 95% of the sequences followed an adjacent or a non-adjacent
 199 repetition pattern, it was possible to predict the stimulus to be selected in the last display.
 200 Patterns learning therefore predicts a decrease in RTs on the last stimulus of baseline
 201 sequences throughout exposure. The introduction of a small amount of violation sequences
 202 aimed at disentangling learning of repetition patterns from a mere effect of task practice. We
 203 assumed that repetition pattern learning will be indicated by slower RTs on violation trials
 204 compared to baseline trials.

205 During the test phases (i.e., the generalization and violation tests), sequences of novel
 206 shapes were presented in 20% of the trials. In the generalization test, these sequences

207 followed the repetition pattern. In the violation test, these were random sequences (i.e., of the
208 form AB for adjacent pattern, and ABC for non-adjacent repetition pattern). The remaining
209 80% of the trials were baseline sequences, randomly intermixed with the novel sequences.
210 Reinforcement schedules were similar to those of the exposure phase: baboons were expected
211 to select the cued stimuli and were reinforced to do so. Note that novel pattern-following and
212 pattern-violating sequences were presented at similar frequencies in generalization and
213 violation tests, respectively. Rote learning of the novel sequences therefore predicts no
214 difference in RTs between tests, while generalization of the learned patterns predicts slower
215 RTs in the violation compared to the generalization test.

216 For each pattern, the exposure phase involved 25 sessions of 160 trials (five repetitions
217 of each shape as S+ within a block), and each test phase involved five sessions of 160 trials
218 (four repetitions of each exposure shape and one presentation of each novel shape as S+
219 within a block). To complete the experiment, the baboons required 18 days on average (range
220 10-26). Custom-written Python codes (www.python.org) were used to generate the sequences.

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Table 1 Description of the phases and conditions involved in Experiment 1

Pattern	Phase	Condition	Sequences Structure	Percentage of Trials	Stimulus Set (n°)
Adjacent Repetition	Exposure	Baseline	AA	95%	1
		Violation	AB	5%	
	Generalization Test	Baseline	AA	80%	1
		Generalization	AA	20%	2 or 3
Violation Test	Baseline	AA	80%	1	
	Violation	AB	20%	2 or 3	
Non-Adjacent Repetition	Exposure	Baseline	ABA	95%	4
		Violation	ABC	5%	
	Generalization Test	Baseline	ABA	80%	4
		Generalization	ABA	20%	5 or 6
	Violation Test	Baseline	ABA	80%	4
		Violation	ABC	20%	5 or 6

Note that the stimulus sets used in the test phases were counterbalanced across participants. Stimuli used in the different sets are available in *Supplementary Information*.

222

223 **Data analyses**

224 In the current design, the short delay (100 ms) before the appearance of the cue is too
 225 short to allow the participants to respond before the cue. Therefore, we do not expect the
 226 baboons to make more errors in violation trials compared to baseline or generalization trials.
 227 This was confirmed by our data: the proportion of errors remained low on average (5.76%)
 228 and did not distinguish violation from generalization trials. Statistical analyses were therefore
 229 restricted to RTs. With this test design, the analyses focused on the responses times obtained

230 for touching the stimulus in the last display (i.e. the second display for adjacent repetition
231 pattern, and the third display for non-adjacent repetition pattern). RT corresponded to the time
232 elapsed between the appearance of the stimuli from the last display and the participant's
233 selection of one of the two shapes. Incorrect trials were removed from the data set as well as
234 trials with any abnormally long RT (i.e., with a RT > 3 s on at least one display; 0.22% of the
235 trials). Sessions were grouped in five (Exposure) and two (Test) blocks of 800 trials each for
236 statistical analyses. Statistical analyses were done with R (V3.1.1, www.R-project.org).

237 Learning during the exposure phase was first assessed by comparing baboons' RTs on
238 baseline and violation trials, for each pattern. This was evaluated with a repeated-measures
239 ANOVA involving the Condition (Baseline, Violation), Pattern (Adjacent, Non-adjacent) and
240 Block (1-5) as within-participant factors. Post-hoc comparisons were performed using
241 Tukey's HSD test ($p < .05$).

242 Two statistical analyses were conducted on the test phases. The first one aimed at
243 verifying that baboons discriminated the novel shapes from the shapes used in baseline
244 sequences, which predicts an increase in RTs. One-sided paired t-tests assessed whether the
245 RTs were greater for the novel compared to baseline sequences. The second analysis assessed
246 whether the increase in RTs (if any) was greater for violation sequences, compared to the
247 generalization sequences. For each participant, mean difference in RTs between the novel and
248 the baseline sequences was computed for each test. These data were then analyzed using one-
249 sided paired t-tests.

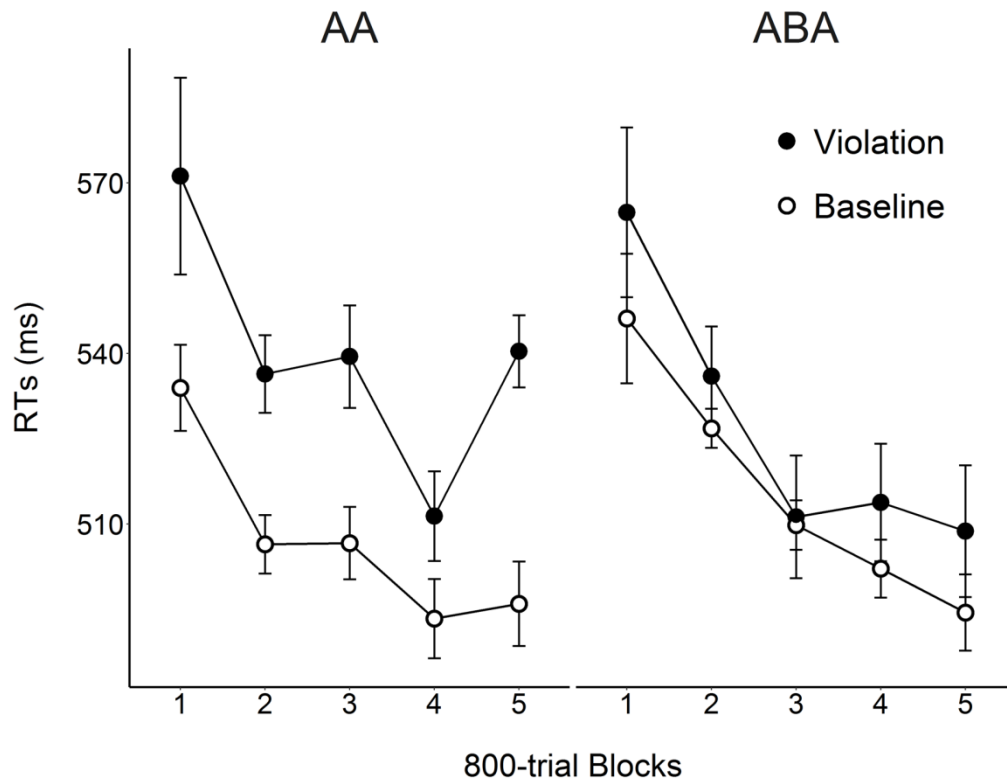
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251 **Results and Discussion**

252 **Exposure phase**

253 Mean RTs are reported in Figure 3 for each pattern, condition and block. The main
254 effect of condition was significant, $F(1, 16) = 15.95, p = .001, \text{partial } \eta_p^2 = .50$, corresponding

255 to overall faster RTs in the baseline (Mean \pm SD = 512 \pm 50 ms) than in the violation
256 condition (533 \pm 69 ms). Also significant was the main effect of Block, $F(1, 16) = 17, p <$
257 $.001$, partial $\eta_p^2 = .52$, showing that RTs decreased with practice. No main effect of Pattern
258 was found, $F(1, 16) = 0.03, p = .87$ (adjacent repetition pattern: 524 \pm 58 ms; non-adjacent
259 repetition pattern: 521 \pm 64 ms). A significant Condition \times Pattern interaction was however
260 found, $F(1, 16) = 7.31, p = .02$, partial $\eta_p^2 = .31$. Post-hoc tests revealed that mean RTs for
261 the different types of sequences differed significantly from each other, in the following order:
262 AA (507 \pm 43 ms) $<$ ABA (516 \pm 57 ms) $<$ ABC (527 \pm 71 ms) $<$ AB (540 \pm 67 ms). No other
263 interaction was found (all $ps > .44$). This significant increase in response time observed on the
264 violation sequences for the adjacent and non-adjacent repetition patterns suggests that
265 baboons successfully learned both repetition patterns. However, baboons showed differential
266 responses to the two patterns: effect size of the difference between baseline and violation
267 conditions was moderate to high for the adjacent repetition pattern (Cohen's $d = 0.58$) and
268 small for the non-adjacent repetition pattern ($d = 0.17$). Further, baboons were faster to
269 respond to baseline sequences of the adjacent repetition pattern than of the non-adjacent
270 pattern, and showed their slowest RTs in reaction to violations of the former.
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273 *Figure 3.* Response times (RTs) obtained in the exposure phase of Experiment 1. Left:
 274 adjacent repetition pattern; right: non-adjacent repetition pattern. Bars represent standard
 275 errors after subtraction of the individuals' mean RTs.

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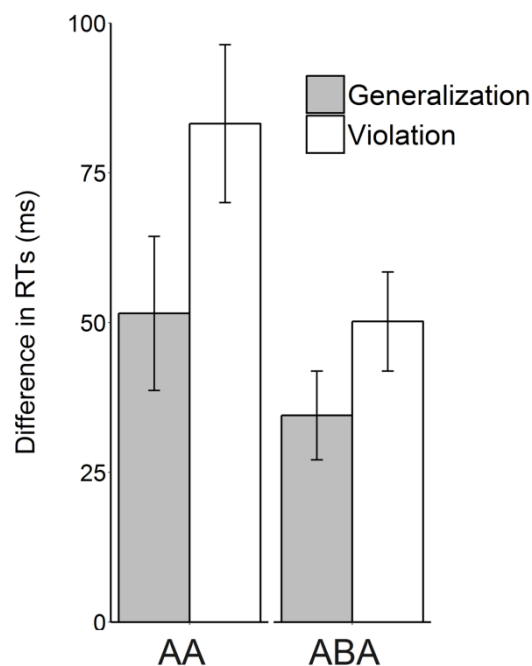
277 **Test phases**

278 Mean differences in RTs between the novel and baseline sequences are reported in
 279 Figure 4 for each pattern and test. A significant positive difference was observed for every
 280 pattern and test type: for the adjacent repetition pattern, RTs were greater in generalization
 281 trials compared to baseline trials of the generalization test ($M \pm SD = 538 \pm 68$ ms, and $487 \pm$
 282 33 ms, respectively, $t(16) = 4.01$, $p < .001$, Cohen's $d = 0.96$), as well as in violation trials
 283 compared to baseline trials of the violation test (569 ± 61 ms, and 486 ± 37 ms, respectively, t
 284 $(16) = 6.31$, $p < .001$, $d = 1.64$). For the non-adjacent repetition pattern, a similar increase in
 285 RTs was observed (generalization *vs* baseline trials of the generalization test: 535 ± 54 ms,
 286 and 500 ± 36 ms, respectively, $t(16) = 4.66$, $p < .001$, $d = 0.76$; violation *vs* baseline trials of
 287 the violation test: 542 ± 51 ms, and 492 ± 33 ms, $t(16) = 6.01$, $p < .001$, $d = 1.18$). These

288 results confirmed that the baboons discriminated the novel shapes used to construct the
289 generalization and violation sequences from the shapes instantiating baseline sequences.

290 We then assessed whether this difference in RTs was greater for the violation
291 compared to the generalization test. A significant difference was found for the adjacent
292 repetition pattern (Mean difference \pm SD = 83 ± 54 ms, and 52 ± 53 ms, respectively, $t(16) =$
293 2.75 , $p < .01$, $d = 0.60$), as well as for the non-adjacent repetition pattern (50 ± 34 ms, and 35
294 ± 31 ms, respectively, $t(16) = 2.12$, $p = .03$, $d = 0.48$).

295 In sum, all the data reported above suggest successful extraction and generalization of
296 both adjacent and nonadjacent repetition patterns, as indicated by baboons' shorter RTs on the
297 last element of generalization sequences compared to the RTs obtained on the same element
298 in violation sequences.



299

300 *Figure 4.* Mean response times (RTs) difference between the novel and baseline conditions
301 of each test phase of Experiment 1. Left: adjacent repetition pattern; right: non-adjacent
302 repetition pattern. Positive values indicate longer RTs in the novel condition compared to the
303 baseline condition. Bars represent standard errors.
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Experiment 2

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Methods

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Subjects, stimuli and task

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Experiment 1 has suggested that the baboons learned and generalized both adjacent and non-adjacent repetition patterns. However, the possibility remains that RT differences observed between the repetition and random sequences derived from the novelty of the last element within this last type of sequences. Moreover, one limitation of the first experiment is that it involved sequences of different lengths for the adjacent repetition and non-adjacent repetition patterns (two and three items, respectively). Those issues were addressed in Experiment 2. This second experiment used three-item sequences following an adjacent repetition pattern, a non-adjacent repetition pattern, or a random pattern (of the form ABB, ABA, and ABC, respectively). Those three types of sequences were presented at equal frequency. This procedure aimed at assessing whether baboons would show sensitivity to repetitions even under those conditions. Six months elapsed between the two experiments, during which the baboons were exposed to different and unrelated tasks.

Data were collected on a total of 13 baboons (six females, age range 4-21 years). Among them, nine had participated in Experiment 1. Eleven other baboons were tested but not included in the analyses because they did not complete the experiment in the allotted time. The apparatus was the same as in the previous experiment.

Experiment 2 used 32 different shapes which have served previously to construct test sequences in Experiment 1 (Set 6). The use of shapes already presented in test sequences was

328 not problematic in itself, because the low frequency of these sequences in Experiment 1 (five
329 repetitions) prevented their learning.

330 In Experiment 2, the cue was no longer a square surrounding the positive stimulus
331 (S+). It was instead a short disappearance of the stimulus giving the impression that it blinks
332 once. This change was introduced because another experiment (Fagot, Malassis, & Medam,
333 2017) suggested that baboons' attention to the task was increased with this procedure. On
334 each display, the two shapes were first presented during 200 ms, after which the S+
335 disappeared for a short period of 80 ms. The two shapes then remained displayed on the
336 screen until the participant touched one of them. Spatial and temporal parameters of the trials
337 were identical to Experiment 1.

338 **Procedure**

339 Three 3-shapes patterns were presented to the baboons at equal frequencies: random
340 pattern (ABC), pattern involving a final adjacent repetition (ABB) and pattern involving a
341 non-adjacent repetition (ABA). A, B and C stimuli were all drawn from the same set of 32
342 shapes (Set 6). Ninety-six sequences were created for this design. Thirty-two A-B pairs of
343 shapes were chosen pseudo-randomly, and used as sequences' beginning for the three
344 patterns. They were followed at equal frequencies by an A, B or C shape, rendering the third
345 shape unpredictable (i.e. with a probability of 0.33 in each case). The frequency of each shape
346 within each position was balanced among the three patterns. Experiment 2 involved a total of
347 ten sessions of 96 randomly intermixed trials. Each session comprised 32 sequences for each
348 of the three patterns described above. The 13 baboons required 3.5 days on average to
349 complete Experiment 2 (range 1-6).

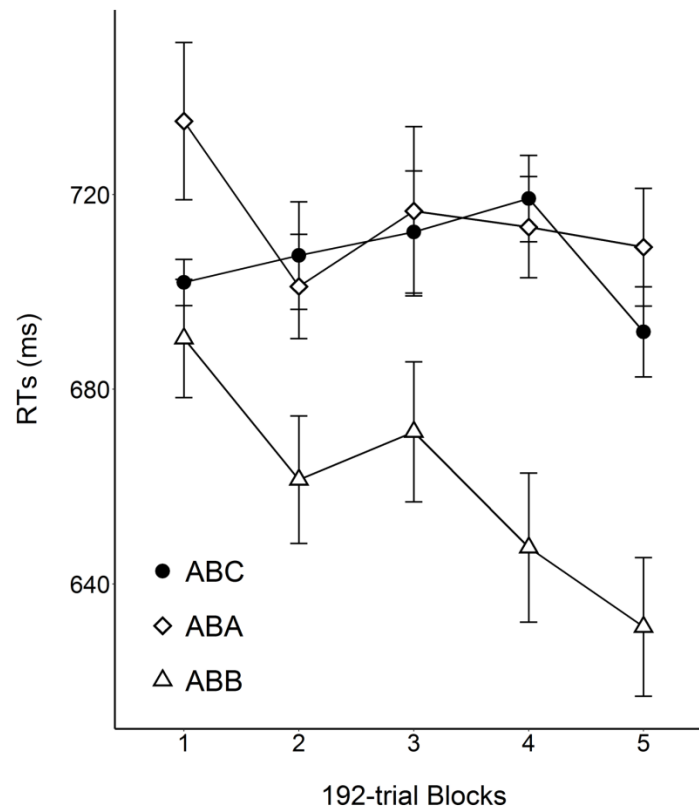
350 **Results and Discussion**

351 Sessions were grouped for the analysis in five blocks of 192 trials each. Incorrect trials
352 were removed from the data set (35 %), as well as trials with any abnormally long RT (i.e.,
353 with a RT > 3 s on either the first, second or third target; 0.5%). Analyses were conducted on
354 baboons' RTs for touching the stimulus in the last display (i.e., the time elapsed between the
355 apparition of the last stimulus and the participant's touch). Mean RTs are reported in Figure 5
356 for each pattern and block. A two-way repeated-measures ANOVA revealed a main effect of
357 Pattern, $F(2, 24) = 11.79, p < .001, \text{partial } \eta_p^2 = .50$. Tukey's HSD test indicated significantly
358 faster RTs for the ABB pattern ($M \pm SD = 660 \pm 66$ ms) compared to the ABA (715 ± 77 ms,
359 Cohen's $d = 0.76$), and ABC patterns (707 ± 62 ms, $d = 0.72$). No significant difference was
360 observed between the ABA and ABC patterns ($p = .41$). The main effect of block was close to
361 significance, $F(1, 12) = 3.90, p = .07, \text{partial } \eta_p^2 = .50$, as was the Pattern x Block interaction,
362 $F(2, 24) = 2.55, p = .10, \text{partial } \eta_p^2 = .50$, suggesting greater acceleration throughout the task
363 for the ABB pattern, compared to the two others.

364 To analyze the response times obtained at the onset of the experiment, we computed
365 two-tailed paired t-tests comparing the mean RTs obtained in the ten first trials for the
366 different patterns. These analyses indicated a tendency for shorter RTs for the ABB pattern
367 (689 ± 90 ms) than for both the ABA pattern (729 ± 123 ms), $t(12) = 1.86, p = .09, d = 0.37$,
368 and the ABC pattern (723 ± 67 ms), $t(12) = 1.89, p = .08, d = 0.43$. Here again, no difference
369 was found between the ABA and ABC patterns, $t(12) = 0.20, p = .84$.

370 These results therefore suggest an initial speed advantage for the adjacent repetition
371 pattern (ABB), which increased throughout the experiment. The acceleration observed for the
372 adjacent repetition pattern, but not for the two others, suggests that repeated exposure to
373 adjacent repetitions may prompt their subsequent recognition, even when their probability is
374 equated with other types of sequences (the last element being a repetition of either the first or

375 second shape, or being new, with a probability of .33). Conversely, no difference in RTs was
 376 found between ABA and ABC patterns, revealing that baboons show sensitivity to adjacent
 377 repetitions only under those conditions.



378
 379 *Figure 5.* Response times (RTs) for each pattern across blocks in Experiment 2. Bars
 380 represent standard errors after subtraction of the individuals' mean RTs.
 381

382 **General Discussion**

383 In Experiment 1, baboons' response times were longer for random sequences
 384 compared to adjacent and non-adjacent pattern-following sequences, whether the sequences
 385 were made of familiar (during exposure phase) or novel shapes (test phases). Earlier studies
 386 supporting this conclusion contrasted ABA against AAB (and/or ABB) repetition patterns
 387 (e.g. Santolin et al., 2016; Spierings & ten Cate, 2016). However this comparison is unable to
 388 rule out the hypothesis that the participants performed the task only by processing the
 389 adjacent repetition pattern, and treating the other by exclusion (e.g. ten Cate & Okanoya,

390 2012; Gervain & Werker, 2013). The main strength of our research is to contrast each
391 repetition pattern against random sequences. The current data therefore confirm that baboons'
392 pattern learning abilities are not restricted to patterns based on adjacent repetition. Baboons
393 are also sensitive to non-adjacent repetition patterns, at least when those patterns are not
394 interleaved with sequences containing adjacent repetitions.

395 In the second experiment, however, baboons show sensitivity to adjacent repetitions
396 only. One critical feature of Experiment 2 is that the three types of patterns were presented at
397 equal frequencies. This result confirms that adjacent repetition are more salient for baboons
398 than non-adjacent repetitions, in line with previous results obtained in human infants (e.g.,
399 Gervain et al., 2008; Gervain & Werker, 2013; Johnson et al., 2009).

400 One possible explanation for baboons' lack of sensitivity to non-adjacent repetitions in
401 Experiment 2 is that they may have focused exclusively on adjacent repetition. Indeed, the
402 presence of adjacent repetitions may impede the exploitation of more complex structural
403 regularities. The current results are reminiscent of two comparative studies aimed at studying
404 birds' ability to discriminate sequences of the form A^nB^n from sequences of the form $(AB)^n$
405 (van Heijningen, Visser, Zuidema, & ten Cate, 2009; Ravnani, Westphal-Fitch, Aust,
406 Schlumpp, & Fitch, 2015). In those studies, a large amount of participants based their
407 decisions on the detection of immediate repetitions within the former type of sequences,
408 rather than on the processing of the full underlying grammars. Taken together, these studies
409 and ours suggest that processing of complex grammars might be demonstrated in future
410 studies with nonhuman species, if the test design avoids the presentation of adjacent
411 repetitions within the sequences, to promote extraction of less salient regularities.

412 One limitation of our study is that the exact structure that the baboons have extracted
413 remains uncertain. For instance in Experiment 1, baboons may have learned that the first and

414 last elements are identical in the ABA patterns, without processing the fact that second
415 element differs. Addressing this issue would also require additional experiments, contrasting
416 for instance ABA against AAA sequences. Another limitation is that baboons were tested on
417 visual pattern learning only. Whether the current findings would hold with auditory sequences
418 remains unknown. Past studies with humans have revealed that sequential regularities
419 extraction is highly constrained by several modality-specific and stimulus-specific features
420 (for a review, Frost, Armstrong, Siegelman, & Christiansen, 2015). Studying those
421 constraints in a comparative perspective will be mandatory to better understand the evolution
422 of pattern learning abilities (Milne, Wilson, & Christiansen, 2018; Neiworth et al., 2017;
423 Santolin & Saffran, 2018).

424 Baboons' results in the current study are in line with previous comparative researches
425 suggesting that the ability to extract structural properties beyond exemplars is widespread
426 among animal species. However, future studies may reveal differences between humans and
427 other species, notably regarding the type of elements over which structural regularities can be
428 extracted. In the current study, patterns are based on identity relationships at the token level
429 (i.e., "A" being an arbitrary element which is repeated). However, in human language,
430 structural regularities can also be found over *functional* categories (ten Cate & Okanoya,
431 2012; Corballis, 2009) and it remains unknown if nonhuman species have the capacity of
432 extracting such language-like patterns. In addition, results from a recent neuroimaging study
433 suggest that brain areas involved in sequential patterns encoding partly differ between humans
434 and monkeys, possibly reflecting interspecific differences in the way the different features of
435 a pattern are encoded (Wang, Uhrig, Jarraya, & Dehaene, 2015).

436 The nature of the mechanisms underlying the processing of the structural properties of
437 sequences is a matter of a longstanding and still ongoing debate. Marcus et al. (1999; see also
438 Marcus, 2001) argued that generalization to novel isomorphic sequences involves rule-

439 governed mechanisms and symbolic representations of the patterns. However, numerous
440 alternative accounts were proposed for this behavior (e.g., Altmann, 2002; Altmann &
441 Dienes, 1999; Brooks & Vokey, 1991; Christiansen & Curtin, 1999; McClelland & Plaut,
442 1999; Negishi, 1999; Pothos, 2005; Seidenberg & Elman, 1999). Most of these accounts
443 argue that statistical learning mechanisms can explain generalization in these tasks, and do not
444 postulate the existence of rule-based mechanisms. In a complementary perspective, Brooks
445 and Vokey (1991) proposed that the subjects may compare the novel exemplars to a set of
446 memorized exemplars, instead of comparing them to an abstract, symbolic, representation of
447 these sequences' structure. Although such alternative accounts are still debated (for a
448 discussion, see e.g., Endress et al., 2007), they may ultimately offer a more parsimonious
449 explanation of generalization in pattern learning studies, than the hypothesis of rule learning.
450 As a third alternative, Endress, Nespors, and Mehler (2009) proposed that repetitions detection
451 is supported by a specialized mechanism, belonging to a set of « perceptive and memory
452 primitives ». These mechanisms are hypothesized to be automatic, would require little or no
453 exposure to the stimuli to be recruited, and would take place in early stages of processing
454 (Endress, Dehaene-Lambertz, & Mehler, 2007; Gervain et al., 2008). Endress, Carden,
455 Versace, and Hauser (2010) proposed that some of these perceptual and memory primitives
456 may be neither language- nor human-specific. Results from our second experiment are
457 consistent with this last hypothesis, as baboons demonstrated a strong sensitivity to adjacent
458 repetitions. However, this effect was not observed for non-adjacent repetitions, which
459 suggests that if such mechanism exists, it is highly limited by memory constraints.

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462

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References

Altmann, G. T. M. (2002). Learning and development in neural networks-the importance of prior experience. *Cognition*, 85(2), B43-50.

Altmann, G. T. M., & Dienes, Z. (1999). Rule Learning by Seven-Month-Old Infants and Neural Networks. *Science*, 284(5416), 875–875.
<https://doi.org/10.1126/science.284.5416.875a>

Brooks, L. R., & Vokey, J. R. (1991). Abstract analogies and abstracted grammars: Comments on Reber (1989) and Mathews et al. (1989). *Journal of Experimental Psychology. General*, 120(3), 316–323. <http://dx.doi.org/10.1037/0096-3445.120.3.316>

Cate, C. ten, & Okanoya, K. (2012). Revisiting the syntactic abilities of non-human animals: natural vocalizations and artificial grammar learning. *Philosophical Transactions of the Royal Society of London B: Biological Sciences*, 367(1598), 1984–1994.
<https://doi.org/10.1098/rstb.2012.0055>

Christiansen, M. H., & Curtin, S. (1999). Transfer of learning: rule acquisition or statistical learning? *Trends in Cognitive Sciences*, 3(8), 289–290.

Corballis, M. C. (2009). Do rats learn rules? *Animal Behaviour*, 78(4), e1–e2.
<https://doi.org/10.1016/j.anbehav.2009.05.001>

de la Mora, D. M., & Toro, J. M. (2013). Rule learning over consonants and vowels in a non-human animal. *Cognition*, 126(2), 307–312.
<https://doi.org/10.1016/j.cognition.2012.09.015>

Endress, A. D., Carden, S., Versace, E., & Hauser, M. D. (2010). The apes’ edge: positional learning in chimpanzees and humans. *Animal Cognition*, 13(3), 483–495.
<https://doi.org/10.1007/s10071-009-0299-8>

- 494 Endress, A. D., Dehaene-Lambertz, G., & Mehler, J. (2007). Perceptual constraints and the
495 learnability of simple grammars. *Cognition*, *105*(3), 577–614.
496 <https://doi.org/10.1016/j.cognition.2006.12.014>
- 497 Endress, A. D., Nespors, M., & Mehler, J. (2009). Perceptual and memory constraints on
498 language acquisition. *Trends in Cognitive Sciences*, *13*(8), 348–353.
499 <https://doi.org/10.1016/j.tics.2009.05.005>
- 500 Fagot, J., & Bonté, E. (2010). Automated testing of cognitive performance in monkeys: use of
501 a battery of computerized test systems by a troop of semi-free-ranging baboons (*Papio*
502 *papio*). *Behavior Research Methods*, *42*(2), 507–516.
503 <https://doi.org/10.3758/BRM.42.2.507>
- 504 Fagot, J., Malassis, R., & Medam, T. (2017). The processing of positional information in a
505 two-item sequence limits the emergence of symmetry in baboons (*Papio papio*), but
506 not in humans (*Homo sapiens*). *Learning & Behavior*. [https://doi.org/10.3758/s13420-](https://doi.org/10.3758/s13420-017-0290-1)
507 [017-0290-1](https://doi.org/10.3758/s13420-017-0290-1)
- 508 Fagot, J., & Paleressompouille, D. (2009). Automatic testing of cognitive performance in
509 baboons maintained in social groups. *Behavior Research Methods*, *41*(2), 396–404.
510 <https://doi.org/10.3758/BRM.41.2.396>
- 511 Frost, R., Armstrong, B. C., Siegelman, N., & Christiansen, M. H. (2015). Domain generality
512 versus modality specificity: the paradox of statistical learning. *Trends in Cognitive*
513 *Sciences*, *19*(3), 117–125. <https://doi.org/10.1016/j.tics.2014.12.010>
- 514 Gervain, J., Macagno, F., Cogoi, S., Pena, M., & Mehler, J. (2008). The neonate brain detects
515 speech structure. *Proceedings of the National Academy of Sciences*, *105*(37), 14222–
516 14227. <https://doi.org/10.1073/pnas.0806530105>
- 517 Gervain, Judit, & Werker, J. F. (2013). Learning non-adjacent regularities at age 0 ; 7. *Journal*
518 *of Child Language*, *40*(4), 860–872. <https://doi.org/10.1017/S0305000912000256>

- 519 Gomez, R. L., & Gerken, L. (1999). Artificial grammar learning by 1-year-olds leads to
520 specific and abstract knowledge. *Cognition*, *70*(2), 109–135.
- 521 Heijningen, C. A. A. van, Visser, J. de, Zuidema, W., & Cate, C. ten. (2009). Simple rules can
522 explain discrimination of putative recursive syntactic structures by a songbird species.
523 *Proceedings of the National Academy of Sciences*, pnas.0908113106.
524 <https://doi.org/10.1073/pnas.0908113106>
- 525 Johnson, S. P., Fernandes, K. J., Frank, M. C., Kirkham, N., Marcus, G., Rabagliati, H., &
526 Slemmer, J. A. (2009). Abstract Rule Learning for Visual Sequences in 8- and 11-
527 Month-Olds. *Infancy: The Official Journal of the International Society on Infant*
528 *Studies*, *14*(1), 2–18. <https://doi.org/10.1080/15250000802569611>
- 529 Kovács, Á. M., & Mehler, J. (2009). Flexible Learning of Multiple Speech Structures in
530 Bilingual Infants. *Science*, *325*(5940), 611–612.
531 <https://doi.org/10.1126/science.1173947>
- 532 Malassis, R., Rey, A., & Fagot, J. (2018). Non-adjacent Dependencies Processing in Human
533 and Non-human Primates. *Cognitive Science*. <https://doi.org/10.1111/cogs.12617>
- 534 Marcus, G. F. (2001). *The Algebraic Mind*. MIT Press.
- 535 Marcus, G. F., Vijayan, S., Rao, S. B., & Vishton, P. M. (1999). Rule Learning by Seven-
536 Month-Old Infants. *Science*, *283*(5398), 77–80.
537 <https://doi.org/10.1126/science.283.5398.77>
- 538 McClelland, J. L., & Plaut, D. C. (1999). Does generalization in infant learning implicate
539 abstract algebra-like rules? *Trends in Cognitive Sciences*, *3*(5), 166–168.
540 [https://doi.org/10.1016/S1364-6613\(99\)01320-0](https://doi.org/10.1016/S1364-6613(99)01320-0)
- 541 Milne, A., Wilson, B., & Christiansen, M. (2018). Structured sequence learning across
542 sensory modalities in humans and nonhuman primates. *Current Opinion in Behavioral*
543 *Sciences*, *21*, 39–48. <https://doi.org/10.1016/j.cobeha.2017.11.016>

- 544 Minier, L., Fagot, J., & Rey, A. (2016). The Temporal Dynamics of Regularity Extraction in
545 Non-Human Primates. *Cognitive Science*, 40(4), 1019–1030.
546 <https://doi.org/10.1111/cogs.12279>
- 547 Misyak, J. B., Christiansen, M. H., & Bruce Tomblin, J. (2010). Sequential Expectations: The
548 Role of Prediction-Based Learning in Language. *Topics in Cognitive Science*, 2(1),
549 138–153. <https://doi.org/10.1111/j.1756-8765.2009.01072.x>
- 550 Murphy, R. A., Mondragón, E., & Murphy, V. A. (2008). Rule Learning by Rats. *Science*,
551 319(5871), 1849–1851. <https://doi.org/10.1126/science.1151564>
- 552 Negishi, M. (1999). Do infants learn grammar with algebra or statistics? *Science (New York,*
553 *N.Y.)*, 284(5413), 435; author reply 436-437.
- 554 Neiworth, J. J., London, J. M., Flynn, M. J., Rupert, D. D., Alldritt, O., & Hyde, C. (2017).
555 Artificial Grammar Learning in Tamarins (*Saguinus oedipus*) In Varying Stimulus
556 Contexts. *Journal of Comparative Psychology (Washington, D.C. : 1983)*, 131(2),
557 128–138. <https://doi.org/10.1037/com0000066>
- 558 Nissen, M. J., & Bullemer, P. (1987). Attentional requirements of learning: Evidence from
559 performance measures. *Cognitive Psychology*, 19(1), 1–32.
560 [https://doi.org/10.1016/0010-0285\(87\)90002-8](https://doi.org/10.1016/0010-0285(87)90002-8)
- 561 Pothos, E. M. (2005). The rules versus similarity distinction. *The Behavioral and Brain*
562 *Sciences*, 28(1), 1–14; discussion 14-49.
- 563 Ravignani, A., & Sonnweber, R. (2017). Chimpanzees process structural isomorphisms across
564 sensory modalities. *Cognition*, 161, 74–79.
565 <https://doi.org/10.1016/j.cognition.2017.01.005>
- 566 Ravignani, A., Westphal-Fitch, G., Aust, U., Schlumpp, M. M., & Fitch, W. T. (2015). More
567 than one way to see it: Individual heuristics in avian visual computation. *Cognition*,
568 143, 13–24. <https://doi.org/10.1016/j.cognition.2015.05.021>

- 569 Santolin, C., Rosa-Salva, O., Regolin, L., & Vallortigara, G. (2016). Generalization of visual
570 regularities in newly hatched chicks (*Gallus gallus*). *Animal Cognition*, *19*(5), 1007–
571 1017. <https://doi.org/10.1007/s10071-016-1005-2>
- 572 Santolin, C., & Saffran, J. R. (2018). Constraints on Statistical Learning Across Species.
573 *Trends in Cognitive Sciences*, *22*(1), 52–63. <https://doi.org/10.1016/j.tics.2017.10.003>
- 574 Seidenberg, M. S., & Elman, J. L. (1999). Do Infants Learn Grammar with Algebra or
575 Statistics? *Science*, *284*(5413), 433–433.
576 <https://doi.org/10.1126/science.284.5413.433f>
- 577 Sonnweber, R., Ravignani, A., & Fitch, W. T. (2015). Non-adjacent visual dependency
578 learning in chimpanzees. *Animal Cognition*, *18*(3), 733–745.
579 <https://doi.org/10.1007/s10071-015-0840-x>
- 580 Spierings, M. J., & ten Cate, C. (2016). Budgerigars and zebra finches differ in how they
581 generalize in an artificial grammar learning experiment. *Proceedings of the National*
582 *Academy of Sciences of the United States of America*, *113*(27), E3977-3984.
583 <https://doi.org/10.1073/pnas.1600483113>
- 584 van Heijningen, C. A. A., Chen, J., van Laatum, I., van der Hulst, B., & ten Cate, C. (2013).
585 Rule learning by zebra finches in an artificial grammar learning task: which rule?
586 *Animal Cognition*, *16*(2), 165–175. <https://doi.org/10.1007/s10071-012-0559-x>
- 587 Versace, E., Spierings, M. J., Caffini, M., Ten Cate, C., & Vallortigara, G. (2017).
588 Spontaneous generalization of abstract multimodal patterns in young domestic chicks.
589 *Animal Cognition*, *20*(3), 521–529. <https://doi.org/10.1007/s10071-017-1079-5>
- 590 Wang, L., Uhrig, L., Jarraya, B., & Dehaene, S. (2015). Representation of Numerical and
591 Sequential Patterns in Macaque and Human Brains. *Current Biology*, *25*(15), 1966–
592 1974. <https://doi.org/10.1016/j.cub.2015.06.035>
- 593