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2 Research Article

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4 **Social tolerance and success-biased social learning underlies the cultural**
5 **transmission of a novel extractive foraging tradition in a wild tool-using**
6 **primate**

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26 **Abstract**

27 The last two decades have seen great advances in the study of social learning (learning
28 from others), in part due to efforts to identify it in the wild as the basis of behavioural
29 traditions.. Theoretical frameworks suggest that both the dynamics of social tolerance and
30 transmission biases (or social learning strategies) influence the pathways of information
31 diffusion in social groups. Bearded capuchins (*Sapajus libidinosus*) inhabiting the semi-arid
32 seasonal caatinga biome of the Serra da Capivara National Park (SCNP) form highly tolerant
33 societies that possess the largest "tool-kit" described for monkeys, a feat likely facilitated by
34 social learning. Here, we used social network analysis and an open diffusion experiment using
35 an extractive foraging task to identify the occurrence of social learning and describe the
36 pathways of social transmission of information in two wild primate populations. The
37 dynamics of social tolerance outside of task introductions predicted opportunities for social
38 learning, but it was tolerance during task introductions that predicted the actual pathways of
39 social information diffusion. Our results also indicated that the capuchins mainly learned from
40 others via direct observation and naïve individuals exhibited an observation bias towards
41 successful males. This study supports the claims of cultural transmission in robust capuchins
42 and empirically supports the role of social tolerance and social learning strategies in human
43 and non-human primate cultural evolution.

44

45 **Significance**

46 The influence of social tolerance in animal social learning has been scarcely
47 investigated empirically. Social tolerance determines who is allowed in proximity to whom
48 and granted access to resources such as food or social information. Therefore, tolerance
49 towards others in proximity is necessary for the spread of social information, linking theories
50 of cultural transmission and animal traditions (or culture). Here, we find evidence that naïve
51 individuals attend to, and potentially learn from, successful conspecifics. Further we find that
52 social tolerance influences pathways of information transmission. Understanding the role of
53 observation biases and social tolerance dynamics in the spread of novel foraging behaviour in
54 a tool-using primate may shed light on the evolutionary forces involved in primate cultural
55 abilities.

56

57 **Introduction**

58 The study of cultural evolution spans a broad range of areas, such as biology,
59 psychology and anthropology, which through combining empirical studies with theoretical

60 and mathematical modelling of cultural change, concerns changes in socially transmitted
61 beliefs, knowledge, customs, skills, attitudes, languages or other behaviours (1-3). Such
62 modelling of the spread of cultural practices in human populations, shows that cultural
63 variation is not a random process, but governed by generalizable rules, and acquired by social
64 learning. The same is true for nonhuman animals, for which a diversity of behavioural
65 traditions have been reported, especially in cetaceans, primates and birds (4). As in humans,
66 animal culture is fuelled by social learning, defined as ‘learning influenced by the observation
67 of, or interaction with, a conspecific, or its products’ (5), and its study is relevant to
68 understanding the evolution of human culture (6). Social learning allows naïve individuals to
69 acquire information about different patterns of behaviour and adopt those that are beneficial.
70 Thus, patterns of behaviour have different transmissibilities. Therefore, in cultural
71 inheritance, transmission biases or social learning strategies, influencing when, what and from
72 whom individuals learn, seem to be the rule rather than the exception (reviewed in 7).
73 However, it is hypothesized that the social tolerance of humans (towards others in proximity)
74 may be one of the underlying factors that facilitates our social learning and innovation,
75 enabling the unique extent of cumulative culture in our species (8).

76 Building on Cavalli-Sforza and Feldman’s (9,10) pioneering importation of theoretical
77 population genetics methods into the study of cultural evolution, Richerson & Boyd (2) used
78 mathematical modelling to describe forces that specify when an individual is more likely to
79 learn from a conspecific, and which conspecifics they may learn from, subdividing the types
80 of transmission biases into *content-based (or direct) biases*, such as ‘copy greater pay-off
81 behaviour’ and *context-based (or indirect) biases*, such as ‘copy the traits exhibited by
82 dominant individuals’. Transmission biases are also termed *social learning strategies* that
83 influence “when to copy”, “what to copy” and “whom to copy” (11,12). Multiple such
84 strategies may be deployed simultaneously, individuals can switch between strategies flexibly,
85 and there is no one-to-one correspondence between psychological heuristics deployed and
86 resulting population-level patterns observed (7). Note, we use the term ‘copy’ synonymously
87 with social learning and not form-copying/imitation.

88 The pathways of information diffusion are also influenced by demographic factors (e.g.,
89 sex, age, social rank, kinship) that shape the occurrence and strength of social relations or social
90 dynamics (13). Coussi-Korbel & Fragaszy (14) outlined how social tolerance (tolerance
91 towards others in proximity) influences opportunities to observe and hence learn from
92 conspecifics. According to this framework, different levels of social tolerance may lead to
93 different patterns of social diffusion of information. *Non-specific social learning*, where

94 information spreads evenly across groups, is predicted to occur in egalitarian species. In
95 contrast, *directed social learning*, which is influenced by the demonstrator's identity resulting
96 in an uneven spread of information across groups (e.g. only within cliques/sub-groups), will
97 occur in more despotic societies due to the asymmetry of social relationships (14). Importantly,
98 care must be taken not to infer social learning strategies (e.g. copy same sex conspecifics)
99 before considering whether the pattern of information diffusion is due to directed social
100 learning (e.g. individuals may only have the opportunity to observe those of the same sex as
101 tolerance of proximity between sexes is lacking). Accordingly, to fully evidence a model-based
102 social learning strategy, an assessment of whether preferential observation of specific
103 individuals is influential in determining the behaviour of observers, is required (15).

104 By conducting experiments with wild animal groups, we can establish whether social
105 learning is occurring and then move on to identifying influences on the transmission pathways
106 (here social tolerance dynamics and the types of transmission biases at play), whilst
107 maintaining the ecological validity lacking in laboratory experiments (16,17). Such
108 experimental designs with multiple freely moving demonstrators/models and observers (i.e.
109 open diffusion experiments: 18) were scarce in wild primates until recently, but are important
110 in allowing us to investigate biases in who is attended to and copied (e.g.
111 older/dominant/successful individuals) and the influence of social dynamics (social
112 tolerance), in naturalistic contexts. With the advent of statistical methods such as Network-
113 based Diffusion Analysis (NBDA: 19,20), researchers have demonstrated that social
114 interactions, such as those based on affiliative behaviours (or social tolerance), do indeed
115 represent opportunities for social learning in several wild primates (ring-tailed lemurs: 21;
116 red-fronted lemurs: 22; vervet monkeys: 23). In addition, observation networks, which
117 indicate tolerance of proximity in a competitive context, have evidenced social learning using
118 NBDA in wild (chimpanzees: 24; vervet monkeys: 23) and free-ranging (Barbary macaques:
119 25) primates. Moreover, open-diffusion studies are conducive to registering multiple
120 transmission events, which in turn allows us to investigate transmission biases. Indeed, such
121 studies have found evidence consistent with transmission biases in wild gracile capuchin
122 monkeys (*Cebus sp.*) including *copy most frequent behavioural variant* (26) and *copy highest*
123 *payoff* (27).

124 Studies of wild primates help elucidate factors (cognitive and social) that underlie
125 cultural evolution in human and nonhuman animals (28). The bearded capuchin (*Sapajus*
126 *libidinosus*) is considered an intermediate species in the despotic-egalitarian spectrum within
127 capuchins given their non-linear dominance hierarchies and asymmetric distribution of

128 aggression between dyads. This fosters diversity in social tolerance among individuals (29-
129 31) making them ideal to investigate the role of social tolerance in social learning. In addition,
130 there is no empirical evidence for social learning, nor transmission biases, in wild robust
131 capuchin monkeys (*Sapajus sp.*). Yet, our study population of bearded capuchins (*Sapajus*
132 *libidinosus*) in the Serra da Capivara National Park presents a uniquely large “tool-kit” for
133 capuchins and the largest outside of the great apes. They use stones (for nut/seed-cracking,
134 pulverising stone, digging for tubers, roots and spiders, and throwing in proceptive displays)
135 and sticks (as probes for honey, insects and water) sometimes in combination and serially (32-
136 35). Tool use seems to be acquired by social learning, for which different observation biases
137 have been described in a semi-free ranging group (36,37), making them a good candidate
138 species to investigate social learning in the wild.

139 Here, we report an open-diffusion field experiment, where a novel extractive foraging
140 task – with two possible actions to access rewards – was introduced to two groups of wild
141 bearded capuchins and diffusion of novel solutions tracked. Using NBDA, we investigated
142 whether behaviours spread at a higher rate between individuals who are more strongly
143 connected through more frequent associations and/or interactions, indicating social learning
144 (19). We considered specific social affiliative networks (social proximity, social play,
145 grooming and co-feeding) indicative of social transmissions pathways pertaining to tolerance,
146 and discuss the transmission of behaviours aided by close observation. We also investigate
147 whether naïve individuals display biases in whom they observe interacting with the novel task
148 addressing social learning strategies of “whom to copy”.

149

150 **Results**

151 *Open diffusion experiment*

152 An open diffusion experiment was conducted with two groups of wild bearded
153 capuchins (Jurubeba (JB) and Pedra Furada (PF)) and involved a foraging task (a food-
154 dispensing puzzle-box) which could be solved one of two ways, *lift* or *pull* (Figure 1 & S3.1).
155 This two-action paradigm (38, see 39 for first use in the wild) allows testing for option
156 preferences at the individual or group level. One monkey in each group had been trained as a
157 demonstrator and was included as such in further analyses. The remaining monkeys observed
158 (i.e. head oriented towards the task within a ten-meter radius) a conspecific solving the task
159 at least once, before solving the task themselves. By the end of the open-diffusion experiment
160 a total of 34 individuals across both groups ($N_{JB}=23$ of 40; $N_{PF}=11$ of 30) had solved the task
161 (whether by lifting or pulling) successfully at least once. Due to a bias for ‘lift’ we do not

162 analyse the diffusion of variants but focus on task solution (see SI Appendix, S3). The
163 diffusion pattern of task solution presented a gradual increase in the proportion of informed
164 individuals, stabilising at 57.5% (JB) and 36.7% (PF) of group members (see Figure S3.2).
165 The task was solved a total of 8671 times ($N_{JB} = 4591$, $N_{PF} = 4080$), with 92% of these
166 observed by at least one conspecific. As there was often more than one observer, 33177
167 observation events ($N_{JB} = 15566$, $N_{PF} = 17611$) were recorded.

168

169 *The influence of observation networks during task introductions*

170 To investigate the role of observational learning, we conducted two separate NBDAs
171 for JB and PF using the time of acquisition diffusion analysis (cTADA, see Materials &
172 Methods). For each group, we ran models informed by three observation networks each
173 reflecting different observation distances (indicative of different learning processes) and
174 compared models with social transmission (and asocial transmission, ‘social model’) and
175 without social transmission (‘asocial model’) using likelihood ratio tests (LRT) to quantify
176 the evidence of social transmission and its level of significance (P). Maximum likelihood
177 methods determined which model better explained the observed data (having the lowest AICc
178 and highest Akaike weight).

179 We found evidence ($\Delta AIC > 2$ between social and asocial models, as per 19), for social
180 learning of task solution for all observation networks (at different distances) in both groups
181 (Table 1). These results were confirmed by the percentage of events that occurred by social
182 transmission (%ST) and a likelihood ratio test comparing the asocial to the social model
183 (LRT), with $P < 0.05$ indicating evidence of an effect consistent with social transmission. The
184 95% confidence intervals (CI) were consistent with evidence of social learning ($s' = 0$ not
185 included in the interval), but the effect of social transmission was small for JB and large
186 CI95% ranges indicate uncertainty in the strength of the effect for PF (Table 1). In JB, NBDA
187 found stronger support (ΔAIC) for the social model over the asocial model when individuals
188 observed task manipulations within 1m than when they were beyond 1m. However, we found
189 contrasting results in PF, with stronger evidence of social learning when individuals observed
190 task manipulations from beyond 5m than within 5m or 1m. Differences in social structure
191 and dynamics between both groups may explain these contrasting results (see SI Appendix
192 S4 & Discussion).

193 Individual-level variables (ILVs), of sex, age, rank, neophobia, and task
194 monopolisation, were included in the models to assess their potential influence on the social or

195 asocial learning rates to avoid erroneous identification of social learning (Table 2). ILVs may
196 influence only the asocial learning rates (additive model), or both learning rates equally
197 (multiplicative model) or independently (unconstrained model). In both groups, inclusion of a
198 variable describing the latency to touch novel objects (of those that entered 5m of the object)
199 improved the model fit for observation networks within 1m and 5m of the task in both groups,
200 indicating that increasing ‘neophobia(latency to touch)’ slowed asocial and social learning
201 rates. In JB, sex influenced social diffusion in observation networks beyond 5m, indicating that
202 males had higher learning rates than females. Finally, in PF, the best model for the observation
203 network beyond 5m was influenced by a variable describing the avoidance of novel objects
204 (including those that never entered 5m of the object) with decreasing ‘neophobia(avoidance)’
205 accelerating asocial and social learning rates (see SI Appendix S8). The ILV monopolisation
206 of resources did not improve model fit in any case (SI Appendix S5).

207 Additive models were a better fit than multiplicative models when testing social
208 transmission in all observation networks except for that beyond 5m in JB, for which the
209 multiplicative model was a better fit. Thus, excepting one case, the social transmission of task
210 solving was a direct consequence of observing those manipulating the task (or observational
211 learning) rather than indirect social learning processes (40).

212

213 *The social context fostering cultural transmission*

214 Social networks were created and SNA metrics were calculated for each group. PF was a more
215 cohesive group (network density: PF range 0.11–0.74, mean 0.34 ± 0.28 ; JB range 0.11–0.56,
216 mean 0.28 ± 0.18) than JB, the larger group, that had a more cliqued social structure (clustering
217 coefficient: PF range 0.29–0.79, mean 0.60 ± 0.18 , JB range 0.16–0.67, mean 0.47 ± 0.17 ; see SI
218 Appendix S4).

219 In a second set of NBDA models we investigated whether social structure and dynamics
220 outside the experimental context of task introduction predicted the information transmission
221 observed (41). In both groups, the $\Delta AIC < 2$ indicated insufficient support of social
222 transmission when the NBDA was informed with the socio-positive networks (social
223 proximity, co-feeding and grooming). However, results of the LRT indicated significant
224 evidence ($P < 0.05$) consistent with social transmission when models were informed by co-
225 feeding in JB (social model $\exp(0.5 * \Delta AIC) = 1.89x$ more support than asocial model) and
226 grooming in PF (social model 1.91x more support than asocial model). Accordingly, in both
227 cases, less than 60% of learning events (57% when using the co-feeding network in JB; 26%
228 when using the grooming network in PF) occurred by social transmission and a large CI95%

229 range of the s' parameter indicated uncertain strength of the social transmission effect (see
230 Table 1). Sex influenced learning rates in both cases, with males learning faster than females
231 (PF, grooming: 154x faster; JB, co-feeding: 17x faster), although results must be taken with
232 caution due to the wide CI95% (Table 2). Consistent with the weak social transmission effects,
233 the best models were obtained using multiplicative approaches, indicating that co-feeding and
234 grooming relations provided opportunities for indirect social learning processes such as
235 local/stimulus enhancement, or social/response facilitation.

236 To further explore how the patterns of affiliative social relations established outside
237 task introductions may predict observation opportunities in a social learning context, we
238 conducted permutation-based linear mixed model regressions (Table 3). In JB, observation
239 networks within 1m and 5m of the task were predicted by co-feeding, grooming and proximity
240 within 1m outside of task introductions. In most of those cases, the socio-positive-observation
241 relationship was significantly influenced by social rank similarity (Table 3) such that frequent
242 affiliative partners that are more similar in rank, more frequently observed each other during
243 task introductions than those more dissimilar in rank. Observations beyond 5m of the task were
244 only predicted by grooming and proximity within 1m and, in both cases, relationships were
245 influenced by sex similarity. In PF, observation networks within 1m and 5m of the task were
246 predicted by co-feeding, social play and proximity within 1m. Only grooming significantly
247 predicted observations within 1m of the task and no socio-positive network significantly
248 predicted observations beyond 5m (Table 3).

249

250 *Transmission biases*

251 To investigate model-based biases, a GLM analysis with a gamma log link function
252 was run to determine whether any model characteristics (sex, age, dominance, task success)
253 predicted whom was most frequently observed by naïve conspecifics when interacting with the
254 task. When the overall fitted model was compared against the intercept-only model, a
255 significant fit was found (likelihood ratio chi-square = 80.127, d.f. = 15, $p < 0.001$), indicating
256 that at least one of the factors/covariates characterising models was predictive of observation
257 frequency. Both the ratio of successful to unsuccessful task manipulations (success ratio) as a
258 main factor and the interaction between sex and success ratio were predictive of the frequency
259 with which a model was observed by naïve conspecifics (Table 4). Sociograms (Figure 2)
260 indicate that, for both groups, successful males were attended to by naïve individuals most
261 often. Unfortunately, there was insufficient variation in the task option used (lift/pull) in each

262 group to enable analysis of whether the observation bias translated into learning of the
263 behaviour exhibited by ‘successful males’ (see Fig S6.1).

264

265 **Discussion**

266 *Detecting social learning*

267 The results herein provide the first empirical evidence for social learning in wild robust
268 capuchin monkeys (*Sapajus sp*), supporting the claims of culture in this clade (42), and as
269 suggested by a similar study of wild gracile capuchins (*Cebus capucinus*: 27). Hoppitt (43)
270 empirically demonstrated that observation networks are a direct and powerful way to detect
271 social transmission, even when there is no social structure information or when other networks
272 (e.g. affiliative) cannot provide evidence of social learning. Observation networks predicted
273 the diffusion of social information regarding task solutions in all cases, with the additive
274 model being the best model in all but one case. This indicates that social learning of the
275 foraging tasks was a direct consequence of observation (e.g. observational learning: 3). These
276 findings for this exceptional tool-using species echo the belief that observational learning
277 supports the maintenance of complex cultural behaviours such as tool use (44, but see 45). In
278 contrast, the multiplicative model selection when NBDA was applied using socio-positive
279 networks outside of task introductions (grooming, social play, social proximity and co-
280 feeding) to inform the social model reiterates that these relationships reflect opportunities to
281 learn by indirect learning processes such as social facilitation, or local enhancement (see
282 Materials & Methods).

283

284 *The role of social dynamics and social tolerance*

285 The NBDA findings, the regressions of observation networks with socio-positive
286 networks, and the fact that the individual level variable (ILV) of monopolization did not
287 improve the model fit in any case, points to the importance of social tolerance in the
288 dissemination of information through these groups, as seen in several species (chimpanzees:
289 46,47; ravens: 48; squirrel monkeys: 49; Barbary macaques: 25).

290 The observational learning, identified above, requires behavioural coordination in space
291 and time, for which individuals must tolerate others in proximity (14). Moreover, socio-
292 positive networks outside of task introductions that were consistent with social transmission in
293 the NBDA analysis were co-feeding (for JB) and grooming (for PF). These represent
294 interactions that are more indicative of tolerance than proximity associations in cohesive
295 groups (e.g. captive starlings: 50) and may be more important than in more fluid fission-fusion

296 groups where proximity associations can predict social learning (e.g. great tits: 51; whales: 52).
297 In a highly competitive context as in our experiment, where the task is a monopolizable
298 resource, social affiliation is a determining factor to be tolerated near, or granted access to, the
299 task (53-55). As visual attention is required for behavioural coordination and any bias in
300 individuals' attention towards closely-bonded conspecifics increases the likelihood of
301 acquiring information from those particular individuals via social learning (56) we conducted
302 network regressions. Here, social tolerance represented by grooming, social play, social
303 proximity and co-feeding (outside of task introductions) were variously predictive of who
304 observed whom (at either 1m or 5m) during task presentations, confirming the importance, as
305 predicted by Coussi-Korbel & Fragazy (14), of social tolerance and diverse relationships in
306 the pathways of information transmission.

307 Together, results of NBDA and network regressions provide further support that
308 tolerance in the bearded capuchin groups was key in enabling social learning. This corresponds
309 with the assertion of Pasqueretta et al. (57) that, across 78 primate groups, more tolerant groups
310 (with little clustering or variance in individuals' centrality) have more efficient networks in
311 terms of information flow. The sociograms and SNA metrics (see SI Appendix S4), indicate
312 this may be true for PF as a more cohesive group and within each clique/sub-group in the
313 larger JB group, each of which was generally composed of a large male and encircling females.
314 Indeed, the influence of ILVs in the NBDA analysis and network regressions attests to the
315 influence of social structure on the diffusion of information with greater evidence of potential
316 directed social learning (14) in the cliquy JB than PF group. In observation networks beyond
317 5m, sex influenced the diffusion of information in JB, with males having a higher learning rate
318 than females, whereas in the cohesive PF it was neophobia(avoidance). Thus, during task
319 introductions in JB, each male had preferential access to the task in front of an audience mostly
320 composed of females from his clique. In fact, in these wild groups, female capuchins are less
321 frequent tool users (34, 58), meaning that natural social diffusion of behavioural traits such as
322 tool-use skills and access to these resources is favoured amongst adult males. Finally, in
323 network regressions rank similarity (close-range observations) and sex similarity
324 (observations>5m) was a determining factor in JB, but no ILVs were relevant in the more
325 cohesive PF. We note however, that contrary to expectations of directed social learning, the
326 spread of novel task solutions was faster and more encompassing in the 'cliquy' JB than the
327 more cohesive PF. This may be explained by the lower connectivity in observation networks
328 (see Table S4.1) of PF than JB, due to the greater number of central individuals (large males)
329 in JB that naïve individuals (encircling females) may observe. However, the identity of the

330 ‘innovator’ or trained individual may also have been influential, being a mid-ranking female
331 in JB, yet a subordinate, hence less observed (15), male in PF.

332

333 *Model-based observation biases*

334 Selective attention by naïve individuals towards proficient individuals (those with high
335 success ratio) was found. Such may translate into a ‘copy successful’ social learning strategy
336 and resonates with what has been described for tool-aided nut cracking in semi-free (*Sapajus*
337 *spp.*: 36, 37) and wild (*Sapajus libidinosus*: 59) capuchins. As in many natural foraging
338 situations, the naïve individual could be attending to these models for scrounging opportunities.
339 However, the motivation to do so, coupled with social tolerance, allows learning opportunities.
340 Consequently, if more proficient individuals are attended to, there are better chances of a
341 beneficial novel behaviour spreading through the social group and fostering cultural evolution.

342 Here, in the early transmission phase of a tradition, more proficient males were the
343 preferred target of naïve individuals while in Coelho et al.’s (37) study of an established nut-
344 cracking tradition in a different population, age and dominance rank of models was influential
345 as these characteristics reliably correlated with proficiency. Moreover, a study of wild vervet
346 monkeys posed a foraging task (60), suggests that dominant females, rather than dominant
347 males, are favoured as a source of social information in species with female philopatry. This
348 stands in opposition to our findings with bearded capuchins, despite them also displaying
349 female philopatry, yet we note this may be explained by the male domination of tool use in this
350 species (34, 58). Further investigations of observation biases for apparent learning purposes (in
351 naïve individuals) are required to elucidate the factors involved in diverse species and contexts
352 to build a better picture of the variety of ways social learning strategies may be combined (7).
353 Such is not confined to model-based biases as indicated by wild vervet monkeys exhibiting a
354 content/direct pay-off bias alongside a bias to copy higher rank individuals (61). Future studies
355 will benefit from ensuring diverse trait variants (e.g., task options) are used by individuals
356 within a group, even whilst perhaps manipulating their relative payoff, to enable investigation
357 of the extent to which observation biases translate into social learning.

358

359 *Final remarks*

360 As shown by our study, social tolerance facilitates the diffusion of social information
361 within primate groups. Through a consequent increase of social learning opportunities such
362 tolerant groups should generate more diverse local traditions (62; evidenced in orangutans:
363 63) and, according to the ‘cultural intelligence hypothesis’ – that states that cultural effects

364 had a role in the evolution of social tolerance and animal intelligence – ultimately enhanced
365 cognitive abilities (64,65) necessary for cultural evolution.

366 Populations of *Sapajus libidinosus* have produced the oldest known nonhuman tools
367 (stone hammers and anvils) outside Africa (3000 years:66) and, through their percussive
368 activities, unintentionally produce stone flakes and cores that closely resemble those
369 previously thought to be intentionally produced by extinct hominins (67). Moreover,
370 capuchins in caatinga biomes live in unusually large group sizes (68), a demographic factor
371 known to favour the emergence and spread of novel behaviour in early humans (69,70).
372 Likewise, increased terrestriality has recently been linked to increased tool-kit size in these
373 populations (71). Although no extant species is a perfect comparator for extinct hominins, our
374 findings, regarding the importance of social tolerance and observation biases (alongside the
375 influence of population size and terrestriality) for the transmission of novel foraging
376 behaviour, contributes to increasing understanding of human technological evolution and
377 cumulative culture in the second half-century of cultural evolution research.

378

379 **Material and Methods**

380 Data on group composition, behavioural data sampling and recording methods, and
381 calculation of social ranks, that informed the analysis, are reported in SI Appendix S1 & S2.

382

383 *Open diffusion experiment*

384 The Lift-Pull task consisted of a matte white acrylic box 20(w) x 30(h) x 20(d) cm. The
385 two functional parts were a blue rectangular plate (10 x 8 cm), that could be lifted perpendicular
386 to the front of the box, and a green knob (7 x 5 cm) at the end of a protruding rod, that could
387 be pulled away from the box. Both actions, when successfully executed, triggered release of
388 the same quantity and quality of food rewards (a mixture of corn, peanuts and raisins) into a
389 tray below (Figure 1 & S3.1). The blue plate or green knob would then automatically return to
390 its initial position.

391 A pilot study with semi-free capuchin monkeys, at Tiete Park, Sao Paulo, indicated it
392 would be unlikely that an individual capuchin would monopolise the task in order to learn it
393 and act as demonstrator. We therefore seeded each wild group with a trained demonstrator by
394 attracting an individual and demonstrating how to solve the task out of sight of other group
395 members. In JB, a mid-ranking adult female (CHI), received a single training session
396 encompassing 4 full demonstrations of the action lift and, when the open diffusion phase
397 started, she was the first to solve the task in her group and did so using lift. In PF, a mid-ranking

398 juvenile male (Lim) received a single demonstration of the pull action, after his group had
399 departed. He then immediately approached the box and solved the task using pull. Note, that
400 half of 14 asocial learning controls failed to solve the task in 4x5minute trials (see SI Appendix,
401 S3). To further the goal of seeding one option in each group (lift for JB; pull for PF), the task
402 was initially presented, to the entire group, with the appropriate option functioning and the
403 alternative locked. This was maintained until approximately 10% of each group (5 monkeys in
404 JB over 1 day; 3 monkeys in PF over 2 days) had solved the task at least once using the seeded
405 action.

406 The open diffusion phase began with the task presented to each group with both options
407 functioning for a total of 14 days for each group between September-October 2012 for JB and
408 between March-April 2013 for PF. The task was offered for as long as the group remained in
409 the experimental area or until they had consumed a maximum of 2kg of food rewards per day.
410 On average, JB spent 90 minutes around the task, before setting off on their daily routes,
411 resulting in 19 hours and 20 minutes of experimentation. PF would spend, on average, 120
412 minutes around the task, resulting in 29 hours and 15 minutes of experimentation. For further
413 details see SI Appendix (S3).

414 To accurately record the activities at the task and the opportunities monkeys had of
415 observing conspecifics solving it, the experiment was filmed with two video cameras: one
416 zoomed in on the task to clearly register task manipulations (Figure S3.1), and one covering a
417 5m radius around the task (Figure 1). CGC also narrated the identity of all monkeys within a
418 10m radius of the box and whether they had their heads oriented towards the task when a
419 demonstration occurred, indicating observational opportunities of task solving. Videos were
420 coded independently by two researchers registering (i) which individuals solved the task, (ii)
421 latency from the beginning of the experiment until each successful task solution (iii) which of
422 the two options (lift or pull) was used, (iv) the identity and (v) distance of monkeys observing
423 the task being solved by a conspecific, regardless of the option used (lift or pull) (for inter-
424 observer reliability, see SI Appendix S9).

425

426 *Network-based diffusion analysis*

427 NBDA was applied to test for social learning (19). We informed the model with (1) the
428 diffusion of the novel trait as the time or order in which each group member first solved the
429 task and (2) a social network detailing the strength of connection between group members. We
430 used both continuous time of acquisition diffusion analysis (cTADA) and order of acquisition
431 diffusion analysis (OADA) following the guidelines and R codes provided by Hasenjager et al.

432 (72). Results of both analyses were similar hence we report results for the more powerful
433 cTADA (40).

434 We built networks of varying conspecific observation distances during task
435 manipulations and socio-positive networks, outside of task introductions (grooming, social
436 proximity, co-feeding and social play). Observation networks during task introductions were
437 used to identify social learning, whereas the socio-positive networks outside task introductions
438 were used to identify potential pathways of social transmission of information. All networks
439 were weighted with the strength of connections between individuals calculated using relative
440 measures (see SI Appendix S4 and Figure S4.1&2 for sociograms). Finally, since NBDA is
441 susceptible to Type I error (20), six non-colinear individual-level variables (ILVs: sex, age,
442 dominance rank, two measures of neophobia and one measure of monopolisation), were
443 included in the NBDA (see SI Appendix S5).

444 NBDA compared a purely asocial learning model with a social learning model to test
445 whether the order or time of diffusion (task solving) followed the pattern of relations of the
446 social networks. Multi-model inference was used to determine the best models (lowest AICc)
447 in terms of ILV selection and influence on social and asocial learning rates (unconstrained,
448 additive and multiplicative models). In an unconstrained model, the effect that each ILV has
449 on asocial and social learning rates is estimated independently. The additive model assumes
450 that ILVs only influence asocial learning so that social transmission occurs as an independent
451 process from asocial learning, meaning the total rate of trait acquisition is the sum of the rates
452 of asocial learning and social transmission (i.e. social influence adds to the chances of
453 individual learning). The additive model is, therefore, likely to be appropriate if individuals
454 can acquire the trait as a direct consequence of observation (44) such as observational learning,
455 including imitation. Conversely, the multiplicative model assumes that ILVs equally influence
456 both learning rates and, therefore, the behaviour of the demonstrator influences the naïve
457 individual's behaviour in a manner that leads indirectly to learning (i.e., whereby the social
458 influence of the demonstrator multiplies the chances of individual learning, such as
459 local/stimulus enhancement (see SI Appendix S6).

460

461 *Correlations between socio-positive networks and observation networks*

462 To test whether affiliative relations outside of task introductions correlate with
463 observation opportunities during task introductions we used permutation-based mixed models
464 where the socio-positive network was entered as the independent matrix and the observation
465 network as the dependent matrix. Each model was informed with other independent variables

466 or fixed effects (sex, age and dominance ranks) and random effects (individual identity was
467 used to control for the number of observations, a confounding factor that may bias effect sizes
468 and hinder the interpretation of results: 73). Different models (combinations of fixed and
469 random effects) were tested for the same pair-wise matrix comparison and the Akaike
470 Information Criterion (AIC) used to determine which model better explained the data and
471 Variance Inflation Factor (VIF) to determine collinearity of variables (with none found) before
472 each regression. The best model was used for regressions with the observed networks and each
473 permutation of the dependent matrix. A total of 10,000 permutations were run for each pair-
474 wise network comparison and p-values calculated based on the distribution of the regression
475 coefficients of each permutation-based regression (see SI Appendix S7).

476

477 *Model based observation biases*

478 Four variables were considered as possible model-based factors/covariates in the GLM
479 analysis: sex, age, dominance, and success at solving the task. Sex and age-group was known
480 for all experimental subjects and dominance rank was determined for the months before,
481 during, and after task introductions in each group. Individual's task-solving success was
482 calculated as a success ratio by dividing the frequency of successful manipulations by the total
483 frequency of (successful + unsuccessful) manipulations of the functional blue plate or green
484 knob. The observation records used, portrayed the frequency with which naïve observers (those
485 yet to solve the task) saw the task being solved (and by whom), from close range (within 5m).
486 To control for individual differences in the frequency of solving the task (and consequent
487 observation opportunities), observation matrices were normalised, along each column based
488 on the maximum value of that column, using UCINET 6.0's "Normalization" procedure. To
489 represent the relative frequency a given individual was observed by naïve conspecifics, while
490 it solved the task, in-degree (which indicates how many edges arrive at the node) was
491 calculated. A generalized linear regression analysis was then conducted to test whether the
492 model-based factors (sex, age, dominance rank) or covariate (success ratio) were predictive of
493 the response variable, the frequency an individual was observed by naïve individuals (in-degree
494 for the demonstrator normalised for relative frequency of successful manipulations).

495

496 **Data, Materials and Software Availability**

497 Data available at https://osf.io/s4fct/?view_only=43c4790a2025472b840c4c87c35ef1ec;
498 UCINET, SOCPprog, Gephi, R code adapted from Hasenjager et al. (72).

499

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506

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Figure 1. The two-action extractive foraging task. Food rewards slide down a shaft into a tray at the bottom of the task. Here, an adult male solves the task using *lift* (the blue flap) instead of *pull* (the green nob), with three monkeys observing within 5-metres.

Table 1. Results for continuous time of acquisition diffusion analysis (cTADA)

Network informing the social model	Jurubeba					Pedra Furada				
	ΔAIC	Akaike weights (ω)	LRT pvalue	%ST	CI95%	ΔAIC	Akaike weights (ω)	LRT (p)	%ST	CI95%
Observation<1m	25.87*	1	28.57 <0.001*	71.96	s'=6.20 L=1.42 U=40.29	2.11*	0.74	6.39 0.011*	83.79	s'=314.94 L=45.44 U=2539.47
Observation<5m	13.45*	1	16.16 <0.001*	75.46	s'=1.45 L=0.31 U=18.95	3.82*	0.87	8.11 0.004*	85.68	s'=317.50 L=47.47 U=2113.76
Observation>5m	13.34*	1	16.04 <0.001*	68.86	s'=22.67 L=5.39 U=105.67	8.89*	0.99	13.18 <0.001*	78.24	s'=199.79 L=46.61 U=1348.82
Grooming	-0.16	0.48	2.54 0.111	33.11	s'=1987.93 L=0.00 U=8988.62	1.29	0.66	5.58 0.018*	26.32	s'=12312.89 L=672.77 U=56683.98
Social proximity	-0.65	0.42	2.05 0.152	44.59	s'=5.01 L=2.19 U=85.51	-3.00	0.18	0.21 0.645	0.00	s'=0.00 L=0.00 U=6.92
Co-feeding	1.28	0.65	3.98 0.046*	56.78	s'=8.53 L=0.06 U= ∞	-4.29	0.11	0.00 1.00	0.00	s'=0.00 L=0.00 U=7.89
Social play	-2.70	0.21	0.00 1.00	0.00	s'=0.00 L=0.00 U=59.85	-4.29	0.11	0.00 1.00	0.00	s'=0.00 L=0.00 U=17.04

An ΔAIC of at least 2 points indicates a better fit of one model over the other: $\Delta AIC > 2$ = social model selected (*) and $\Delta AIC < 2$ = asocial model selected. Akaike weights (ω) represent the weight or likelihood of a model relative to other candidate models (74). Observation networks were collected during task introductions whereas the remaining networks were collected outside of task introductions. $\Delta AIC = AIC_{\text{asocial model}} - AIC_{\text{social model}}$. %ST: the percentage of events that occurred by social transmission. LRT: a likelihood ratio test comparing both agent-based models. CI95%: 95% confidence intervals for the social parameter s' which determines the strength of social transmission relative to asocial learning. L: Lower value of the CI95%. U: Upper value of the CI95%. When models provided the same results using different approaches and rates, LRT and CI95% were calculated for those with better estimates of the s' parameter. * indicates models that provide evidence of social transmission. For interpretation of CI95% for the s' parameter, refer to Table 1 and SI in Hasenjager et al. (72).

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797 **Table 2.** Contribution of the ILVs in social learning for models providing evidence of social
 798 transmission.

Group	Network	Approach	ILV	Effect (95%CI)
Jurubeba	Observation within 1 m relative frequency	cTADA Additive	Neophobia (latency touch)	0.38 (0.19, 0.73)
	Observation within 5 m relative frequency	cTADA Additive	Neophobia (latency touch)	0.32 (0.12, 0.64)
	Observation beyond 5 m relative frequency	cTADA Multiplicative	Sex	6.19 (1.95, 27.72)
	Co-feeding	cTADA Multiplicative	Sex	16.75 (5.54, 72.34)
Pedra Furada	Observation within 1 m relative frequency	cTADA Additive	Neophobia (latency touch)	0.68 (0.15, 39.81)
	Observation within 5 m relative frequency	cTADA Additive	Neophobia (latency touch)	0.76 (0.15, 70.32)
	Observation beyond 5 m relative frequency	cTADA Additive	Neophobia (avoidance)	1.57 (0.17, 14.41)
	Grooming	cTADA Multiplicative	Sex	154.42 (14.38, 3,352.05)

799 **Approach:** cTADA= Continuous time of acquisition NBDA; Additive, indicative of direct social
 800 learning/observational; Multiplicative, indicative of indirect social learning such as stimulus enhancement. **ILV:**
 801 Individual level variable. **Effect:** The degree to which social and asocial learning increase as measures of the
 802 parameters (ILVs) increase, calculated as $\exp(\text{MLE})$. For example, in the first line social and asocial learning rates
 803 decrease by a factor of 0.38x per 1 value increase of neophobia(latency to touch the novel object).

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809 **Table 3.** Results of the permutation-based linear mixed model regressions.

Group	Socio-positive network	Observation <1m	Observation <5m	Observation >5m	
JURUBEBA	Co-feeding	$r_{\text{COF}} = 2.492$ ($p < 0.001$)*	$r_{\text{COF}} = 3.889$ ($p < 0.001$)*	$r_{\text{COF}} = 1.035$ ($p = 0.429$)	
		$\text{SR} = -0.072$ ($p < 0.001$)*	$\text{SR} = -0.082$ ($p < 0.001$)*	$\text{SS} = -6.04\text{E-}04$ ($p = 0.023$)*	
	Grooming	$r_{\text{GRO}} = 1.575$ ($p < 0.001$)*	$r_{\text{GRO}} = 2.939$ ($p < 0.001$)*	$r_{\text{GRO}} = 1.463$ ($p < 0.001$)*	
		$\text{SR} = -0.084$ ($p < 0.001$)*	$\text{SR} = -0.099$ ($p < 0.001$)*	$\text{SS} = -4.45\text{E-}04$ ($p < 0.001$)*	
	Social play	$r_{\text{PLA}} = 0.559$ ($p = 0.086$)	$r_{\text{PLA}} = 0.689$ ($p = 0.134$)	$r_{\text{PLA}} = -0.139$ ($p = 0.594$)	
		$\text{SR} = -0.084$ ($p < 0.001$)*	$\text{SR} = -9.89\text{E-}05$ ($p = 0.079$)	$\text{SS} = -6.22\text{E-}04$ ($p < 0.001$)*	
	Proximity 1 m	$r_{\text{PRI}} = 0.772$ ($p < 0.001$)*	$r_{\text{PRI}} = 1.298$ ($p < 0.001$)*	$r_{\text{PRI}} = 0.556$ ($p = 0.012$)*	
		$\text{SR} = -7.89\text{E-}05$ ($p = 0.018$)*	$\text{SR} = -0.091$ ($p < 0.001$)*	$\text{SS} = -5.16\text{E-}04$ ($p < 0.001$)*	
	PEDRA FURADA	Co-feeding	$r_{\text{COF}} = 3.066$ ($p = 0.002$)*	$r_{\text{COF}} = 3.335$ ($p < 0.001$)*	$r_{\text{COF}} = 0.064$ ($p = 0.358$)
			$\text{SS} = 0.001$ ($p = 0.717$)	$\text{SS} = 5.76\text{E-}04$ ($p = 0.737$)	$\text{AS} = 5.71\text{E-}04$ ($p = 0.345$)
		Grooming	$r_{\text{GRO}} = 1.364$ ($p = 0.045$)*	$r_{\text{GRO}} = 1.621$ ($p = 0.064$)	$r_{\text{GRO}} = 0.693$ ($p = 0.363$)
			$\text{SS} = 0.001$ ($p = 0.565$)	$\text{SS} = 0.001$ ($p = 0.634$)	$\text{AS} = 5.35\text{E-}04$ ($p = 0.349$)
Social play		$r_{\text{PLA}} = 2.260$ ($p = 0.042$)*	$r_{\text{PLA}} = 2.927$ ($p = 0.043$)*	$r_{\text{PLA}} = 1.460$ ($p = 0.858$)	
		$\text{SS} = 0.001$ ($p = 0.776$)	$\text{SS} = 0.001$ ($p = 0.779$)	$\text{AS} = -5.64\text{E-}04$ ($p = 1.000$)	
Proximity 1 m		$r_{\text{PRI}} = 1.395$ ($p < 0.001$)*	$r_{\text{PRI}} = 1.568$ ($p < 0.001$)*	$r_{\text{PRI}} = 0.313$ ($p = 0.513$)	
		$\text{SS} = 0.001$ ($p = 0.627$)	$\text{SS} = 0.001$ ($p = 0.646$)	$\text{AS} = -5.02\text{E-}04$ ($p < 0.001$)*	

810 r: regression coefficient of the independent network. GRO: Grooming. COF: Co-feeding. PLA: Social play. PRI:
811 Proximity 1 m. SS: regression coefficient for sex similarity. AS: regression coefficient for age similarity. SR:
812 regression coefficient for social rank similarity. p: p-values. Only regression coefficients and p-values of
813 significant variables in the fitted model are reported. *: $p < 0.05$

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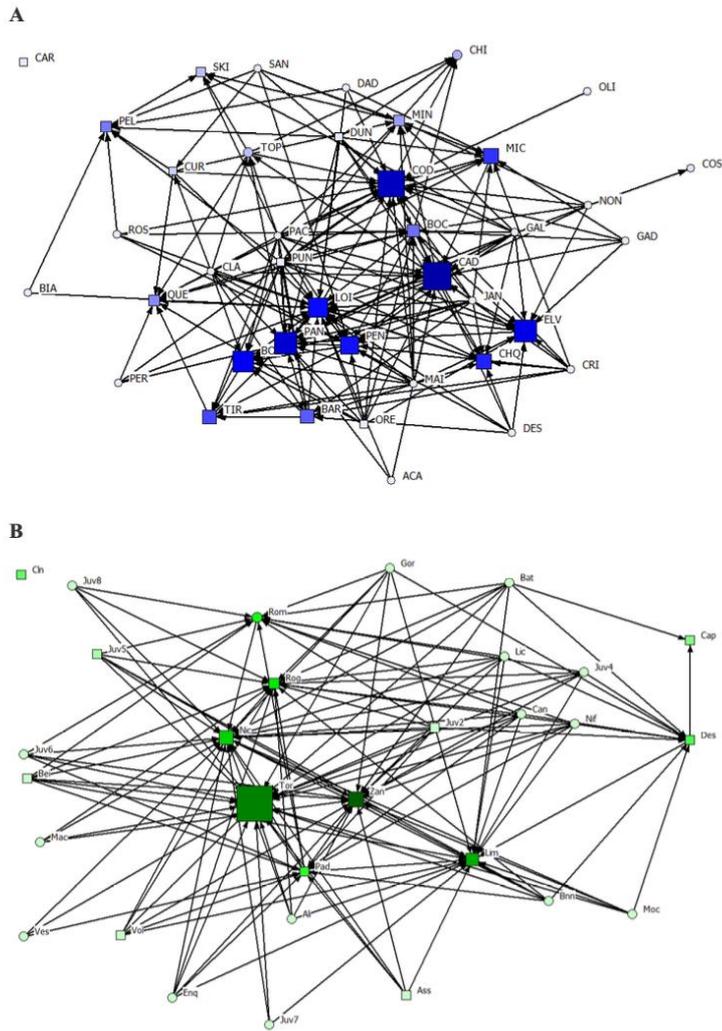
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825 **Table 4.** Results for the generalized linear model analysis, testing for model-based biases
 826 in the choice of observational targets during the lift-pull task experiment.

Model fit	Likelihood ratio Chi- square	d.f.	Sig.
Group	2.740	1	0.098
Sex	0.049	1	0.825
Age	0.526	1	0.468
Dominance	0.638	2	0.727
Success ratio	3.959	1	0.047**
Sex*Age	0.924	1	0.336
Sex*Dominance	1.978	2	0.372
Age*Dominance	2.056	2	0.358
Sex*Success ratio	7.349	1	0.007**
Age*Success ratio	3.168	1	0.075
Dominance*Success ratio	1.093	2	0.579

827 ** p<0.05

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830 **Figure 2.** Observation within 5-meter radius networks for (A) Jurubeba and (B) Pedra Furada.
 831 Larger nodes indicate higher in-degree centrality (frequency that a monkey was observed by a
 832 naïve conspecific while task solving); Darker colours indicate higher success ratios in task
 833 solving. Circle=female and Square=male.

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