

1 **Title:** “Contextual priors do not modulate action prediction in children with autism”

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27 **Abstract**

28 Bayesian accounts of autism suggest that this disorder may be rooted in an impaired ability to
29 estimate the probability of future events, possibly due to reduced priors. Here, we tested this
30 hypothesis within the action domain in children with and without autism using a behavioural
31 paradigm comprising a familiarization and a testing phase. During familiarization, children
32 observed videos depicting a child model performing actions in diverse contexts. Crucially,
33 within this phase, we implicitly biased action-context associations in terms of their
34 probability of co-occurrence. During testing, children observed the same videos but
35 drastically shortened (i.e., reduced amount of kinematics information) and were asked to infer
36 action unfolding. Since during the testing phase movement kinematics became ambiguous,
37 we expected children's responses to be biased to contextual priors, thus compensating for
38 perceptual uncertainty. While this probabilistic effect was present in controls no such
39 modulation was observed in autistic children, suggesting an impairment in using prior
40 information when anticipating others' actions under situations of perceptual uncertainty.

41

42 **Introduction**

43 Understanding what others are doing and what they are going to do next constitutes a major
44 hallmark of social cognition achievement (Sebanz and Knoblich, 2009).
45 Current prediction theories in the action domain suggest that the motor system plays a key
46 role in the anticipation of others' actions (Aglioti et al., 2008; Jeannerod, 2001; Rizzolatti and
47 Craighero, 2004; Wolpert and Flanagan, 2001). Central to these theories is the concept of
48 motor simulation, which assumes that anticipatory mechanisms rely on the covert re-
49 enactment of the motor programs used to perform the observed movements. On this view,
50 prediction would be accomplished by using our motor system as an internal forward model.
51 However, there is evidence challenging this approach by showing that motor experience is

52 not necessarily required to anticipate action unfolding (Vannuscorps and Caramazza, 2015,
53 2016) and that similar predictive performance can be achieved possibly by relying on
54 previous observational experience with others' actions (de Klerk et al., 2016). Nevertheless,
55 despite differences on the role of motor experience in action prediction, most of these views
56 collectively assume that the motor system or, at least, some related structures (i.e., the ventral
57 premotor cortex) are critically involved in predictive processing (Schubotz, 2007).

58 A critical challenge for action prediction accounts in general is explaining how the motor
59 system allows action prediction under situations of perceptual uncertainty (Brass et al., 2007;
60 Kilner et al., 2007). Indeed, similar movement kinematic patterns can be associated with
61 different actions, making it difficult to move backwards from observed data to hidden motor
62 representations about possible underlying causes.

63 Current predictive coding models based on Bayesian inference provide a solution to this
64 “inverse problem” by suggesting that, particularly when visual evidence is ambiguous, the
65 motor system benefits from top-down expectations about others' likely behaviours given
66 previous similar experiences (Friston et al., 2011; Kilner et al., 2007).

67 Top-down predictions, however, are not only based on past visual or motor experience with
68 others movements (Aglioti et al., 2008; Amoruso et al., 2014; Calvo-Merino et al., 2005) but
69 also on prior knowledge about the context in which actions are typically observed (Amoruso
70 et al., 2016; Wurm and Schubotz, 2012). Furthermore, it has been shown (Wurm and
71 Schubotz, 2016) that, when perceptual information is impoverished, context plays a key role
72 in aiding action recognition. Indeed, specific environmental scenarios are often indicative of
73 which actions are likely to occur in them (e.g., we typically crack eggs in the kitchen), thus
74 constraining predictive processing. In Bayesian terms:

75 $P(\text{Action}|\text{Kinematics}, \text{Context}) \propto P(\text{Kinematics}|\text{Action}, \text{Context}) P(\text{Action}|\text{Context})$

76 where $P(\text{Action}|\text{Kinematics}, \text{Context})$ is the posterior probability of an action conditioned on
77 observed kinematics and context, $P(\text{Kinematics}|\text{Action}, \text{Context})$ is the likelihood of the
78 action given observed kinematics and context, and $P(\text{Action}|\text{Context})$ is the prior probability
79 of the action given the context.

80 In other words, the prior defines a hypothesis space of what actions are feasible in that
81 context (Baker et al., 2009). In particular, it has been suggested that priors can be classified in
82 structural and contextual ones (Series and Seitz, 2013). Structural priors reflect default
83 expectations, which may be innate or acquired through long-term learning and, thus, not
84 easily malleable. A common example is given by the expectation that light comes from
85 above. Contextual priors, conversely, are expectations acquired through short-term learning
86 and bound to a specific spatio-temporal framework, cue, or task instruction.

87 Socio-motor impairments, including difficulties in action comprehension, are considered to
88 be among the core deficits associated with Autism Spectrum Disorders (ASD, American
89 Psychiatric Association, 2013). It has been recently proposed that this deficit might be related
90 to atypical predictive processing (Lawson et al., 2014; Pellicano and Burr, 2012; Sinha et al.,
91 2014). Interestingly, these accounts root their explanations in Bayesian inference and suggest
92 that the integration between prior knowledge and sensory evidence may be affected in ASD.
93 Central to some of these accounts is the hypothesis that ASD individuals are less influenced
94 by prior knowledge, leading to a greater reliance on bottom-up sensory signals.

95 Despite the explanatory power of Bayesian approaches on ASD and the great attention they
96 have received in the last years, only a few empirical studies (Chambon et al., 2017) have
97 attempted to test its core assumptions within the action domain (i.e. individuals with ASD do
98 not rely on contextual priors to explain away movement kinematics). Thus, experimental
99 testing is necessary to advance comprehension in the field (Palmer et al., 2017).

100 Here, we aimed to directly test this hypothesis by comparing the ability to predict others'
101 actions in children with and without autism. To this aim, we developed a behavioural
102 paradigm consisting of a probabilistic learning task (familiarization) followed by an action
103 prediction task (testing). During familiarization, children observed videos depicting a child
104 grasping different objects to either interact or not with another child and were asked to
105 predict action unfolding. Crucially, within this phase, we implicitly biased the association
106 between the action and the context in terms of their probability of co-occurrence to induce
107 contextual priors with different weights. During testing, children observed the same videos
108 but these were shortened so that the amount of visual kinematics information was drastically
109 reduced and were asked again to predict action unfolding.

110 Based on Bayesian models, we hypothesized that during testing, where movement kinematics
111 become ambiguous, typically developing (TD) children would be biased toward contextual
112 priors learned during familiarization, compensating for perceptual uncertainty. Conversely,
113 since individuals with autism may be less influenced by prior knowledge, we expected not
114 such compensation in ASD children.

115

116 **Methods**

117 *Participants*

118 Twenty-four high-functioning ASD children (mean age = 8.66, SD = 1.63 years) and twenty-
119 four TD children (mean age = 9.04, SD = 1.08 years) took part in the current study. We
120 determined the sample size for our mixed within- and between-subject 4×2 design
121 (probability \times group) through the G* power software (Faul, Erdfelder, Buchner, & Lang,
122 2009) by expecting a large effect size ($f = 0.4$) based on a previous study showing a large-
123 association effect ($r = 0.46$) between autistic traits and contextual modulation of action

124 perception (Amoruso et al., 2018), and setting the significance level at 0.05, and the desired
125 power ($1-\beta$) at 0.95.

126 Children with ASD were recruited at the IRCCS Stella Maris Foundation, Pisa, Italy and at
127 the Scientific Institute IRCCS E. Medea, Polo Friuli Venezia Giulia, and had previously
128 received a diagnosis of ASD according to the DSM-5 criteria by independent clinicians and
129 confirmed by a score above the threshold for ASD on the Autism Diagnostic Observation
130 Schedule (ADOS, Lord, Rutter, DiLavore, Risi, Gotham, & Bishop 2012). We controlled also
131 that all patients had no history of neurological or medical problems. TD children were
132 recruited at primary schools in Udine (Italy). The two groups were 1:1 matched for age (ASD
133 = 8.66 ± 1.63 ; TD = 9.04 ± 1.08 ; $t(46) = -0.937$, $p = 0.353$), gender (male: female ratio 22:2),
134 and non-verbal IQ as measured by the Raven's Colored Progressive Matrices (ASD = 107.91
135 ± 10.62 ; TD = 102.5 ± 10.32 ; $t(46) = 1.791$, $p = 0.079$). See Table 1 for participant's
136 characteristics.

137 The study was approved by the local Ethics Committee (Comitato Etico Regionale Unico,
138 Friuli Venezia Giulia, Italy) and was carried out in accordance with the ethical standards of
139 the 1964 Declaration of Helsinki. All participants were naive to the aims and hypothesis of
140 the experiment. Parents/guardians of all participants provided their informed written consent,
141 and children gave their verbal assent to participate.

142 *Stimuli*

143 The videos were recorded in color at 25 frames per second with a Sony Alpha ILCE-7K
144 camera and were further edited with the Adobe Premiere Pro CS3 3.0 Software. Length was
145 matched across videos so that they had an equivalent duration: 25 frames for a total of 1,000
146 ms in the familiarization phase and 15 frames for a total of 600 ms in the testing phase. All
147 videos depicted a male child (10 years-old) performing reach-to-grasp movements on two
148 different objects (i.e., an apple or a glass) with his right hand in front of another child (10

149 years-old). Depending on the kinematics, each object could be grasped to perform either an
150 individual or an interpersonal action (i.e., grasping-to-eat or drink vs. grasping-to-offer).
151 Kinematics differences between individual and interpersonal actions were mainly provided
152 by the type of power grip used: either from the side with a spherical grasp or from the top
153 with a cylindrical grasp, respectively. Crucially, each of these actions was recorded
154 associated with different contextual cues. In the case of the object “apple”, the two possible
155 contextual cues were an orange or a violet dish. In the case of the object “glass”, the two
156 possible cues were a blue or a white tablecloth.

157 *Stimuli validation*

158 A total of 80 videos (Object [5] x Action type [2] x Contextual-cue [2] x Versions [4]) were
159 originally recorded. Please note that by “Versions” we mean different video recordings of the
160 same action. Apart from the apple and the glass videos that were selected for the final
161 experimental design; a cup, a pair of scissors and a spray-cleaner were also used in this initial
162 phase. All the objects could be manipulated by the actor in order to perform either individual
163 (i.e., eating, drinking, cutting or spraying, respectively) or interpersonal actions (i.e., offering
164 it to another child) in the presence of two different contextual cues (i.e., a dish, a tablecloth, a
165 tray, a notebook, or a cloth with different colours, respectively). In order to validate the
166 stimuli, we conducted a preliminary rating study on the 80 videos by asking 10 TD children
167 (4 females; mean age = 9 ± 1.24 years) to watch the first 25 snapshots (i.e., 2 frames before
168 the child made fully contact with the object) and to recognize the action.

169 Overall, the rationale behind this preliminary study was selecting those actions that had
170 comparable discriminability in order to rule out any confounding effect related to intrinsic
171 objects’ properties (e.g., affordance). The Friedman ANOVA yielded an effect of object (X^2
172 = 10.37, $p = 0.03$). Pairwise comparisons with the Wilcoxon matched-pairs test indicated that
173 those objects that elicited the most similar levels of accuracy and were far from being

174 significantly different from each other ($p = 0.918$) were the apple ($M = 86.3$; $SD = 12.7$) and
175 the glass ($M = 86.6$; $SD = 16.3$), while the other objects led to an accuracy level of: cup ($M =$
176 93.8 ; $SD = 12.67$), scissors ($M = 85.9$; $SD = 13.22$) and spray cleaner ($M = 91.6$; $SD = 9.74$).
177 Thus, actions directed to the apple and the glass were selected for the final experiment.

178 *Task and procedure*

179 To test whether children with ASD benefit of prior probabilistic information to predict action
180 unfolding, we developed a friendly paradigm consisting of a familiarization and a testing
181 phase (see Fig. 1 and supplementary Videos 1–8 for examples).
182 During familiarization, videos depicted a child grasping either an apple from a plate or a glass
183 from a tablecloth to perform individual or interpersonal actions (i.e., eating/drinking and
184 giving the object to the other child, respectively). Videos were stopped 2 frames before the
185 child model made full contact with the object. Thus, even though participants observed the
186 pre-shaping of the hand during the reaching phase of the movement and not the grasping
187 movement itself, the amount of visual information given was high. In a two-alternative forced
188 choice (2AFC) task, participants were asked to watch the videos and predict action unfolding.
189 Importantly, during the familiarization, we implicitly manipulated prior expectations by
190 setting the probability of co-occurrence between actions and contextual-cues (i.e., the colour
191 of the plate and the tablecloth) to 10%, 40%, 60% or 90% (see Fig. 1A-B).
192 During the testing phase (Fig. 1C), children observed the same videos and were asked again to
193 predict action unfolding. However, in this case, the amount of sensory evidence was
194 drastically reduced by shortening the videos so to occlude the last 400ms (i.e., last 10
195 frames). We reasoned that during this phase, where movement kinematics became
196 ambiguous, children's responses would be biased toward contextual priors acquired during
197 familiarization, thus compensating for perceptual uncertainty. Of note, even though
198 participants observed only the initial part of the videos (i.e., either the initial 25 or 15 frames

199 depending on the phase), when these were originally recorded, the child actor was asked to
200 perform the full action to provide veridical kinematics information.

201 Each participant performed the full experimental session in the same day. Before the
202 experimental session, children were initially introduced to the objects and received
203 demonstrations about the different possible ways of manipulating them. More specifically,
204 we gave children specific examples by showing the overall action with the original objects
205 used in the videos and explicitly named the associated labels (e.g., “This is how we grasp an
206 apple when we want to eat it”). Furthermore, children were asked to grasp the objects and
207 perform the complete action themselves to promote simulation.

208 The experimental session consisted of two blocks and lasted ~40min. Each of these blocks
209 comprised a familiarization phase (80 trials) immediately followed by a testing phase (40
210 trials). Thus, the overall experiment consisted of 240 trials, 160 of familiarization and 80 of
211 testing. Short breaks were allowed between blocks and phases. Neither explicit information
212 about the associations between contextual-cues and actions nor trial-by-trial feedback was
213 provided. Thus, participants were completely naïve to the existence of underlying statistical
214 regularities. This was further confirmed after the experiment during the debriefing session, in
215 which participants were explicitly informed about the existence of action-color associations.

216 Interestingly, some children reported to have noticed a relationship between them but,
217 critically, only two TD children reported the *exact* content of the abstract rule (e.g., “every
218 time an orange plate appeared it was more likely to see the child grasping the apple for
219 eating”). This ensures that, overall, no explicit knowledge about the underlying associations
220 was evident for either TD or ASD children and that observed effects are unlikely to stem
221 from attentional aspects regarding the ability or inability to attend to a contextual cue.

222 It is worth noting that half of the overall trials involved the object “apple” and the other half
223 the object “glass”. While for one object we biased 10-90% the action-cue association (i.e., in

224 90% of the trials the action “eat” was combined with the presence of the orange plate, and in
225 the other 10% of the trials it was combined with a violate plate), for the other object we
226 biased 60-40% (i.e., in 60% of the trials the action “drink” was combined with the presence
227 of the blue tablecloth, and in the other 40% it was combined with a white tablecloth).
228 Importantly, this probabilistic structure was kept identical within participants in the two
229 consecutive experimental blocks but the actions associated to high or low probabilities were
230 counterbalanced between participants. Importantly, during the testing phase, all possible
231 action-contextual cues associations were equally presented (10 trials for each action); for a
232 total of 20 trials for each of the 4 probabilities of context-action associations established in
233 the familiarization phase.

234 *Trial structure*

235 The trial structure was the same in the familiarization and testing phase, with the exception of
236 video duration. Trials started with a fixation cross lasting 3,000 ms and it was followed by the
237 video-clip presentation lasting 1,000 ms in the familiarization and 600 ms in the testing
238 blocks. After the video-clip, a frame was presented with the verbal descriptors of the two
239 possible actions (e.g., “mangiare” or “bere” and “offrire”, in English “to eat” or “to drink”
240 and “to give”, respectively; one located on the right and the other on the left) written in white
241 on a black background. This frame remained on the screen until a response was recorded.
242 Participants were requested to give their responses by pressing with the index finger the
243 computer keys “z” (for left choices) or “m” (for right choices). The keys were covered with
244 white stickers in order to facilitate localizing their position on a QWERTY keyboard. The
245 location of the two descriptors was counterbalanced, ensuring that in half of the trials a
246 descriptor was presented on the left and, in the other half, on the right. This procedure
247 enabled us to prevent participants from planning their response in advance on the basis of the
248 spatial location of the descriptors.

249

250 **Results**

251 *Neuropsychological results*

252 Before the experiment, we assessed social perception abilities in all children using the Italian
253 Version of the NEPSY-II (Urgesi et al., 2011). This battery includes two tests that evaluate
254 theory of mind (ToM, i.e., the capacity to interpret other's intentions, desires and beliefs), and
255 affect recognition (i.e., identification of emotional facial expressions). In addition, the ToM
256 test can be divided in a verbal (ToM A) and a contextual (ToM B) part, with the former one
257 evaluating intention understanding of social situations from verbal or pictorial descriptions;
258 and the latter one, assessing the capacity to understand how certain emotions are linked to
259 specific contexts (see Table 1). As in Narzisi et al. (2013), the scores obtained for each
260 participant at each test were expressed as scaled scores with respect to the normative values
261 for the corresponding chronological ages, thus improving the approximation of the data to the
262 normal distribution.

263 We compared social perceptual abilities between groups by means of independent sample t-
264 tests. A significant difference between groups was observed in ToM A [$t(46) = -4.22, p =$
265 0.0001 ; ASD children, $M = 6.27, SD = 4.66$; TD children, $M = 11.12, SD = 3.13$], with ASD
266 children scoring lower than the TD ones. However, no significant differences between groups
267 were observed for the ToM B [$t(46) = -0.54, p = 0.58$; ASD children, $M = 9.42, SD = 3.08$;
268 TD children, $M = 9.87, SD = 2.63$] or for the affect recognition test [$t(46) = 1.50, p = 0.13$;
269 ASD children, $M = 6.77 \pm 1.95$, TD children, $M = 7.76 \pm 2.57$], even though the ASD group
270 tended to have lower scores than the TD one.

271 *Action prediction results*

272 Behavioral reaction time (RT) performance was acquired during the testing phase. RT were
273 trimmed at 2.5 standard deviations (SD) above their mean. In addition, RTs < 100ms were
274 considered accidental button presses and removed from the analysis. We run a repeated-
275 measures analysis of variance (RM-ANOVA) considering group (ASD, TD) as between-
276 subjects variable and the different probability conditions (10%, 40%, 60%, 90%) and blocks
277 (1, 2) as within-subjects variables. The analysis yielded a main effect of group ($F_{1,46} = 6.98$, p
278 $= 0.011$, $\eta_p^2 = 0.13$), indicating that TD children (mean = 1608.13; SD = 571.05) were faster
279 than the ASD ones (mean = 1969.95; SD = 491.57); and a main effect of block ($F_{1,46} = 11.52$,
280 $p = 0.001$, $\eta_p^2 = 0.20$), showing an overall learning effect with reduced RT in the second block
281 (mean = 1709.17; SD = 540.3) as compared to the first one (mean = 1868.92; SD = 573.4).
282 No other effects were observed (all $ps > 0.32$).

283 Accuracy data from the familiarization and testing phases were converted into d' prime values
284 (d'), a bias-corrected measure of sensitivity in discriminating between 2 categories
285 (Macmillan and Kaplan, 1985), in this case, between individual vs. interpersonal actions. We
286 also calculated a measure of response criterion (c), which reflects the existence of a bias in
287 providing a specific response.

288 We run an exploratory RM-ANOVA in the familiarization phase considering group (ASD,
289 TD) as between-subjects variable and the different probability conditions (10%, 40%, 60%,
290 90%) and blocks (1, 2) as within-subjects variables. This analysis yielded non-significant
291 differences (all $ps > 0.22$). Nevertheless, we decided to collapse the 4 probability conditions
292 during familiarization due to their unequal number of trials resultant from the probabilistic
293 manipulation. An independent t-test comparing the overall d' scores obtained during the
294 familiarization revealed no differences ($t(46) = -1.305$, $p = 0.198$) between the TD (M =
295 1.414, SD = 0.37) and ASD (M = 1.239, SD = 0.53) groups (see Fig. 2). Furthermore, non-
296 significant differences were observed for the c values ($t(46) = -1.683$, $p = 0.099$), suggesting

297 that predictive performance in both groups was comparable when the amount of visual
298 kinematics information was high and that their responses were not biased in terms of
299 identifying individual or interpersonal actions.

300 Then we run the RM-ANOVA on the d' responses obtained during the testing phase. Here,
301 we find a main effect of probability ($F_{3, 138} = 5.212, p = 0.001, \eta_p^2 = 0.10$) and an interaction
302 between probability and group ($F_{3, 138} = 3.271, p = 0.023, \eta_p^2 = 0.06$). No effects or
303 interactions including the factor block were found (all $ps > 0.13$). Post-hoc comparisons
304 performed with the Newman-Keuls test on the interaction (MSE = 0.41039, $df = 119.67$)
305 indicated that, within the ASD group, performance under the different probability conditions
306 did not differ (all $ps > 0.55$). Conversely, TD children were better at predicting others'
307 actions under the highest probability condition as compared to the low and intermediate ones
308 (90% vs. 10%, $p = 0.0001$; 90% vs. 40%, $p = 0.002$; 90% vs. 60%, $p = 0.04$). Moreover, the
309 performance of the TD and ASD groups only differed for the highest probability condition
310 (90%). Specifically, the 90% condition in the TD group significantly differed from the 90%
311 ($p = 0.04$), 60% ($p = 0.01$), 40% ($p = 0.01$), and 10% ($p = 0.02$) conditions of the ASD group,
312 with larger d' values in the TD than in the ASD group (see Fig. 2).

313 Again, no significant main effects or interactions were observed for response criterion c
314 values (all $ps > 0.11$).

315 *Correlations results*

316 Finally, we examined the relationship between the probabilistic effect and measures of social
317 perception, symptomatology and psychopathology. In order to account for the probabilistic
318 effect, we calculated a beta index for each participant based on their performance on the
319 familiarization and the testing phases, separately. Briefly, standardized beta coefficients were
320 estimated across trials at the individual participant level by running a regression analysis with

321 accuracy and probability as the dependent and independent variables, respectively. Thus, this
322 measure provides an estimate of the strength of the probabilistic effect on children's
323 performance. Since responses were coded on the basis of movement kinematics, positive
324 betas indicate that, as the probability of action-context association increases, the prediction of
325 action unfolding increases, in keeping with the use of contextual priors to disambiguate
326 action kinematics. In contrast, negative betas reflect the inverse relationship, that is, as
327 probability increases, action prediction decreases, pointing to a tendency to respond
328 counterintuitively with respect to the context. Betas close to zero reflect that children mostly
329 relied on body kinematics and were less influenced by the context. Thus, a negative
330 relationship between symptom severity and individual beta value would point to weaker use
331 of contextual prior in more impaired children.

332 The Pearson correlations between the beta index and the neuropsychological measures run
333 separately for each group yielded non-significant effects neither in ASD (ToM A: $r = -0.16$,
334 $p = 0.42$; ToM B: $r = -0.001$, $p = 0.99$; emotion recognition: $r = 0.05$, $p = 0.78$) nor TD
335 children (ToM A: $r = -0.16$, $p = 0.42$; ToM B: $r = -0.35$, $p = 0.08$; emotion recognition: $r =$
336 0.11 , $p = 0.60$).

337 Then, within the ASD group, we correlated the beta index with the ADOS calibrated severity
338 scores (Gotham et al., 2009) to check if autism severity explained the observed effect.

339 However, no association was found between the ADOS CSS and the beta indexes
340 (familiarization phase: $r = -0.01$, $p = 0.94$; testing phase: $r = -0.02$, $p = 0.91$).

341 Finally, we correlated the beta index with the Child Behaviour Checklist (CBCL) subscales
342 (see Table 2). The CBCL (Achenbach and Ruffle, 2000), constitutes a parent report for the
343 screening of emotional and behavioural problems in childhood. Of note, one of the CBCL
344 subscales measures children's anxiety. This is of particular interest since a series of recent
345 studies indicate the existence of a strong negative association between anxiety levels and

346 unpredictability in ASD (Boulter et al., 2014; Chamberlain et al., 2013). In other words,
347 unexpected events constitute a potential stressor that triggers increase levels of anxiety in
348 ASD individuals. Indeed, all the constellation of symptoms characterized as “restricted and
349 repetitive behaviours” (e.g., insistence on sameness, rituals and difficulty in tolerating
350 change) are thought to represent compensatory strategies to mitigate uncertainty and make
351 life as much predictable as possible.

352 The analysis involving the CBCL subscales showed a significant negative correlation
353 between the beta index obtained during the testing phase and anxiety ($r = -0.559$, $p = 0.036$,
354 one-tailed; Bonferroni corrected). See Fig. 3. Importantly, no correlation between anxiety and
355 beta was found for the familiarization ($r = 0.058$, $p = 0.79$), suggesting that the association
356 was specifically related to atypical predictive processing in uncertain environments.

357

358 **Discussion**

359 Bayesian models postulate that human observers optimize sensory processing by combining
360 perceptual information with previous knowledge about the statistical regularities of the world
361 (Knill and Pouget, 2004). An appealing Bayesian approach to ASD suggests that autistic
362 impairments might be related to a reduced influence of prior probabilistic knowledge when
363 interpreting variations in incoming inputs (Pellicano and Burr, 2012). Furthermore, since
364 priors can be operationalized as expectations in predictive coding models (Kording and
365 Wolpert, 2004), this deficit can also be interpreted in terms of poor predictive abilities
366 (Schuwerk et al., 2016; Sinha et al., 2014).

367 Here, we tested this hypothesis by examining how ASD and TD children tracked the
368 probabilities of action-context co-occurrence and used this knowledge to predict others’
369 actions under situations of perceptual uncertainty (i.e., low amount of movement kinematics
370 information). Overall, we found that both groups were able to accurately predict action

371 unfolding when perceptual information about kinematics was high (i.e., during
372 familiarization). However, during testing, where movement kinematics became ambiguous,
373 only TD children were able to capitalize on priors (i.e. probabilistic distribution of the action-
374 context associations) to help disambiguation and accurately predict action unfolding.
375 Conversely, ASD children were able to predict the correct action above chance level, but
376 their responses were not biased by the prior probabilistic distributions. Pearson analysis
377 correlating task performance with social cognition and symptom severity found no evidence
378 that ASD-related differences in these domains could account for the pattern of observed
379 responses. Interestingly, however, the anxiety subscale of the CBCL negatively correlated
380 with ASD performance during testing but not during familiarization, suggesting that higher
381 levels of anxiety were associated with the reduced effect of probabilistic knowledge during
382 action prediction only in uncertain environments.

383 It has been proposed that social impairments in ASD may be linked to difficulties in
384 perceiving and recognizing other people' actions (Fecteau et al., 2006; Iacoboni and
385 Dapretto, 2006). According to this theory, this might be due to deficits in processing
386 biological motion (Blake et al., 2003) or, more broadly, to aberrant activity in the action
387 observation network (AON) that prevents from properly coding others' movement kinematics
388 (Theoret et al., 2005). However, this theory has been recently challenged by evidence
389 showing that individuals with ASD exhibit functionally intact perceptual signals for
390 interpreting others' behaviours (Cusack et al., 2015) as well as comparable activity in the
391 AON (Dinstein et al., 2010). For a systematic review see (Hamilton, 2013). In line with these
392 latter studies, we found no differences between ASD and TD children either during
393 familiarization or during testing in the weakly associated condition (i.e., 10%), suggesting
394 that both groups were able to correctly identify observed actions on the basis of perceptual
395 movement kinematics. This is also in keeping with previous studies investigating action

396 prediction in adults with ASD and showing normal bottom-up sensory processing (Chambon
397 et al., 2017) and with Bayesian accounts, suggesting that it is not sensory processing itself
398 what is compromised in ASD but its interpretation.

399 While no differences between groups were observed during familiarization, the testing phase
400 revealed different patterns of behavioural performance. TD children showed a clear bias
401 towards priors learnt during familiarization that was completely absent in the ASD group.
402 Specifically, when the amount of perceptual evidence was drastically reduced (i.e., shortened
403 videos), TD children were better at predicting others' actions in the high-probability
404 condition as compared to those with lower probability. This is in accordance with Bayesian
405 proposals suggesting that ambiguous visual evidence leads to an increased reliance on
406 contextual priors and greater perceptual bias. It can be argued, however, that children were
407 not interpreting diminished movement kinematics by using learnt priors but rather ignoring
408 them and solely responding based on the contextual cues. In other words, an alternative
409 explanation to our findings could be phrased in terms of general learning of cue-response
410 pairings rather than Bayesian inference. Nevertheless, this seems unlikely. Indeed, both ASD
411 and TD children performed above chance level (~68 % and ~66 %, respectively) in the
412 testing phase for the low-probability condition (i.e., 10%), namely, when contextual cues
413 violated the expectations triggered by the high-probability condition (e.g., observing the
414 action "give" in the context of an orange plate when this cue was strongly associated with the
415 action "eat"). In these "incongruent" testing trials, children still tended to predict the action
416 signalled by movement kinematics rather than that expected from the context (i.e., answer
417 "give" instead of "eat"), thus indicating they were using kinematics to predict action
418 unfolding when contextual priors were not reliable. Accordingly, it has been shown that when
419 contextual cues are not available, information from observed movement kinematics forms the
420 basis for action inference (Soriano et al., 2018). It is worth noting, however, that the current

421 design does not allow to determine whether contextual modulations occurred directly on the
422 kinematics level or on other higher levels of action representation (i.e., intention), which
423 remains an open question for future studies.

424 Interestingly, performance in the testing phase remained stable in the ASD group
425 irrespectively of the probabilistic condition (i.e., ~68 %), suggesting that they mainly based
426 their responses on movement kinematics and did not benefit from using contextual priors. It
427 is unlikely that this pattern of results reflects a greater ability of ASD children to ignore the
428 color cue and to focus on the task at hand (i.e., anticipating action unfolding based on
429 kinematics). Indeed, were this the case, better performance of ASD as compared to TD
430 children would be expected in the 10% probability condition, where context and movement
431 kinematics pointed to incongruent predictions. Conversely, performance of ASD and TD
432 children was comparable when context was unreliable and only differed, with better
433 predictive performance in the TD group, when context provided reliable cues. Accordingly,
434 previous studies have shown that ASD individuals typically exhibit enhanced access to the
435 details of a scene, which reflects a general bias toward processing local features and elements
436 (Dakin and Frith, 2005). On this view, what would be actually expected is a higher ASD
437 attentional bias directed toward the color feature (Wang et al., 2015). Indeed, there is
438 evidence (Maule et al., 2017) showing that ASD individuals are better than TD controls at
439 recognizing whether colors were part of an original set or not. Furthermore, no differences
440 were observed between ASD and TD children in the non-verbal contextual task of the
441 NEPSY-II (see also Narzisi et al., 2013), which evaluates the capacity to interpret how
442 others' intentions can be linked to specific contexts, making it unlikely that the observed
443 results could be explained by a deficit in contextual integration. Thus, the overall pattern of
444 results suggests that, even though ASD children perceive and recognize action kinematics to
445 the same extent as TD children when enough perceptual information is provided (i.e., in the

446 familiarization phase) or when context provides unreliable cues (i.e., 10% condition during
447 testing), they do not benefit of prior probabilistic information to predict action unfolding.
448 A limitation of the current study, however, is that it cannot disentangle whether ASD children
449 were unable to track regularities and learn the action-context associations or learnt them but
450 did not use them to predict others' actions. Previous evidence seems to support the former
451 option by showing that ASD children exhibit a general deficit in using arbitrary cues to make
452 inferences (Ames and Jarrold, 2007) as well as impairments in tracking implicit regularities
453 and form action predictions (Schuwerk et al., 2016).

454 However, it is worth mentioning that other studies have shown that ASD children are able to
455 use statistical information from the environment to a similar extent as TD children. For
456 instance, Manning et al. (2017) tested how ASD and TD children used reward probabilities in
457 a decision-making task under stable or volatile contexts (i.e., fixed vs. fluctuating
458 probabilities). Based on the Bayesian proposal suggesting that autistic observers are less
459 biased to prior information (Pellicano and Burr, 2012), they hypothesized that ASD children
460 would assign more weight to recent trials and would not flexibly update their behaviour in
461 response to uncertainty in volatile contexts. In contrast, they found similar learning rates and
462 updating profiles in both groups, suggesting that, at least under some situations, ASD
463 children can use contextual priors to interpret sensory information.

464 Yet, difficulties may arise when ASD individuals are presented with more complex tasks in
465 which uncertainty is linked to social information. Indeed, in our task, uncertainty was tightly
466 related to the possibility of an agent interacting –or not- with another. Thus, it is likely that
467 the differences between groups detected by our paradigm may have arisen from the social
468 setting of the task.

469 Finally, our finding of a negative correlation between anxiety and task performance in the
470 testing phase is in line with current proposals suggesting that atypical predictive abilities in

471 ASD are associated to increased levels of anxiety (Sinha et al., 2014). In particular, the non-
472 relying on prior knowledge to explain away sensory information makes the world to appear
473 more unstable and unpredictable and this is sufficient to trigger a stress response (Boulter et
474 al., 2014; Chamberlain et al., 2013). This might explain a wide range of ASD symptoms,
475 such as insistence on sameness and repetitive behaviors, which can be conceived as coping
476 strategies to reduce anxiety. Notably, no correlation was observed during familiarization, thus
477 highlighting to a specific link between anxiety traits and poor predictive processing only
478 under situations of perceptual uncertainty.

479 Interestingly, Manning et al. (2017) also explored the possible link between children's anxiety
480 and task performance. However, they did not find correlations between these aspects.
481 Nevertheless, the task used by Manning et al. did not comprise social situations and this may
482 explain their negative result. Thus, our findings point to the fact that the link between anxiety
483 and prediction abilities in ASD might be particularly related to handling uncertainty in social
484 environments. While here we provide preliminary support for this suggestion, we
485 acknowledge that our current design does not allow to establish, whether ASD predictive
486 deficits are specific for the social domain or can also be extended to the non-social one. Of
487 note, recent evidence (Ego et al., 2016; Tewolde et al., 2018) seems to point to the fact that
488 ASD children can actually predict non-social/physical events such as the trajectory of a flying
489 bird or the movement of a car. Nevertheless, future studies directly comparing the differential
490 role of social vs. non-social cues within the same sample of participants are necessary to shed
491 light on this aspect.

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501

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613 hard to recognize. *Psychonomic bulletin & review*.

614

615 **Figure Legends**

616 **Fig.1. Experimental design.** (A) Full schema of probabilities allocation during a
617 familiarization block (80 trials) in a hypothetical participant. In the example, low and high
618 probabilities are assigned to actions performed with the object “apple”, while intermediate
619 probabilities are assigned to the object “glass”. This distribution was counterbalanced across
620 participants. The experiment included 2 blocks of familiarization (80 trials each) and 2 of
621 testing (40 trials each). During the familiarization (B) children observed videos depicting a
622 child performing individual or interpersonal actions and predicted action unfolding. During
623 this phase, we implicitly manipulated action-context associations in terms of their probability
624 of co-occurrence to 10%, 40%, 60% and 90%. During testing (C), participants performed the

625 same task but in this case the amount of perceptual information was drastically reduced (i.e.,
626 videos lasted 600ms).

627 **Fig. 2. Action prediction results.** Participants' performance (d') in predicting the course of
628 the observed actions during the familiarization and the testing phases for action-context
629 “weak association” (10%), “intermediate associations” (40/60%) and “strong association”
630 (90%). Asterisks indicate significant comparison ($p < 0.05$). Error bars represent SEM.

631

632 **Fig. 3. Pearson correlation.** Relationship between beta coefficients for ASD participants
633 based on their performance on the prediction task and the parent-report anxiety scores. CBCL
634 questionnaires were returned by 23 out of the 24 ASD children's parents. In addition, one
635 child was removed from the analysis since the anxiety score had a Cook distance of 2.5.
636 Thus, the analysis was performed on a reduced sample ($n = 22$).

637

638 **Table 1**

	ASD	TD
<i>N</i> (male:female ratio)	22:2	22:2
Age mean (SD)	8.66 (1.63)	9.04 (1.04)
Age range	7-12	7-11
IQ mean (SD)	107.91 (10.62)	102.08 (10.2)
IQ range	80-130	80-120
ToM A	6.27 (4.66)	11.12 (3.13)
ToM B	9.42 (3.08)	9.87 (2.63)
Emotion Recognition	6.77 (1.95)	7.76 (2.57)
ADOS CSS mean (SD)	5.75 (1.29)	
ADOS CSS range	4-9	

639 Table 2.

Syndrome	Mean t scores (SD)	range
Anxious/depressed	56.31 (6.17)	50-69
Withdrawn/depressed	61.5 (7.07)	50-79
Somatic complaints	53.27 (4.38)	50-64
Social problems	62.4 (7.04)	50-78
Thought problems	61.27 (7.76)	50-74
Attention problems	61.22 (9.12)	52-92
Rule breaking behaviour	56.13 (5.68)	50-72
Aggressive behaviour	56.09 (4.83)	50-67

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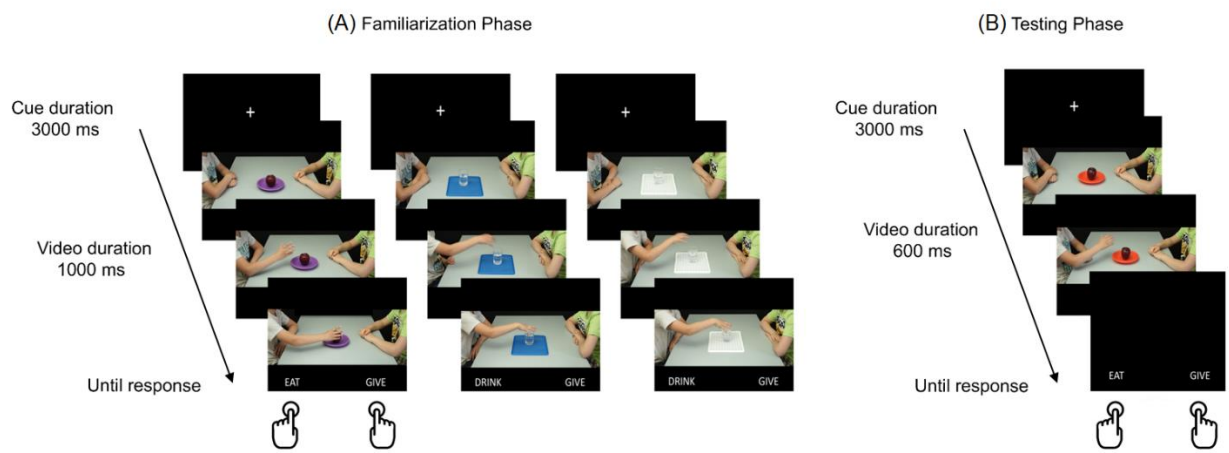
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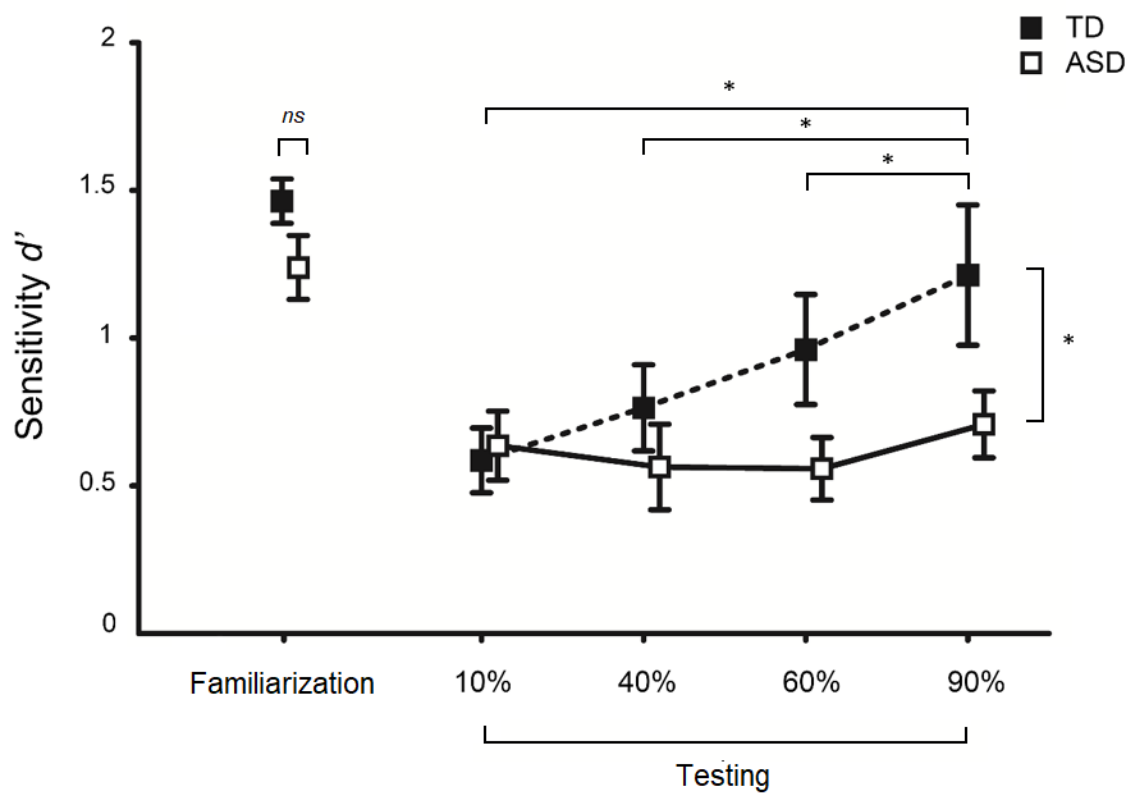
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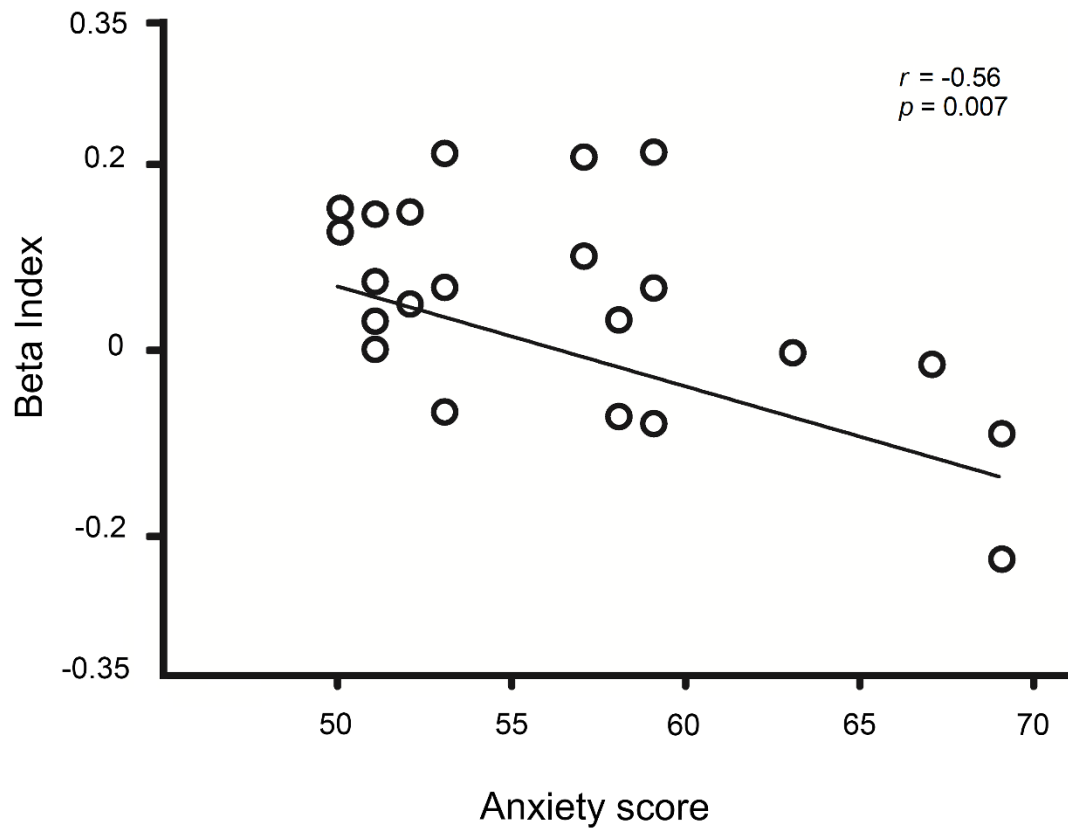
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