

Empathy modulates the temporal structure of social attention

Article

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1	Empathy Modulates the Temporal Structure of Social Attention
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29	Author Contributions.
30	
31	NH carried out the statistical analysis and wrote the initial draft of the paper. BC, EM, and
32	AH conceived of the study design, coordinated data collection and provided modifications to
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35	

36	Abstract
37	Individuals with low empathy often show reduced attention towards social stimuli. A limitation
38	of this literature is the lack of empirical work that has explicitly characterised how this
39	relationship manifests itself over time. We investigate this issue by analysing data from two
40	large eye-tracking datasets (total N = 176). Via growth-curve analysis, we demonstrate that
41	self-reported empathy (as measured by the empathy quotient - EQ) predicts the temporal
42	evolution of gaze behavior under conditions where social and non-social stimuli compete for
43	attention. In both datasets, we found that EQ not only predicted a global increase in social
44	attention, but predicted a different temporal profile of social attention. Specifically, we
45	detected a reliable effect of empathy on gaze towards social images after prolonged viewing.
46	An analysis of switch latencies revealed that low EQ observers switched gaze away from an
47	initially fixated social image more frequently and at earlier latencies than high EQ observers.
48	Our analyses demonstrate that modeling these temporal components of gaze signals may
49	reveal useful behavioral phenotypes. The explanatory power of this approach may provide
50	enhanced biomarkers for conditions marked by deficits in empathy related processes.
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52	Keywords: Eye-tracking; empathy, social attention
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67 To enable successful interactions with the environment, organisms must 68 preferentially attend to socially significant stimuli. Failure to engage with conspecifics can 69 result in exclusion and status loss, which are significant and recurrent fitness threats [1]. 70 Moreover, attending to social stimuli allows the accumulation of strategically beneficial 71 information such as the physical strength of a potential rival, the social standing of a 72 potential ally, or the genetic fitness of a potential mate [2]. In humans, such 'social attention' 73 is also crucial for the development of communicative skills such as language acquisition and 74 emotion recognition [3].

75 Empathy has been defined as the drive to identify with another person's emotions 76 and thoughts, and to respond to these with an appropriate emotion [4]. In order to identify 77 with another's emotions and respond appropriately, it is essential to attend to socially 78 relevant cues such as bodily postures and facial expressions - which provide important 79 information for decoding the emotional states of other people [5,6]. Social attention can 80 therefore be conceptualised as an essential precursor to an empathic response. Support for 81 this view has come primarily from case-control eye-tracking studies, which have 82 demonstrated that individuals with deficits in some empathy related processes also show 83 deficits in social attention. For instance, a recent meta-analysis revealed robust evidence 84 that Autism Spectrum Conditions (ASC) are associated with a reduction in social attention 85 that generalises across a wide range of tasks and stimulus conditions [7]. Influential case-86 control eye-tracking studies have indicated that individuals with ASC exhibit reduced 87 attention to biological relative to non-biological motion patterns [8] and exhibit a preference 88 to direct gaze towards geometric patterns when they compete with videos of social 89 interactions [9]. However, other studies have called into question whether social attention 90 differences are meaningfully related to the aetiology and maintenance of ASC [10,11]. The 91 heterogeneity in reported outcomes is possibly due to the heterogeneous nature ASC and 92 the small sample sizes resulting from the practical issues associated with case-control 93 designs. In this context, it is surprising that there is almost no literature that has attempted to

model *individual*, rather than *group* variation in social attention in the neurotypical
population. One recent study has demonstrated that trait empathy is associated with a gaze
bias towards social rewards in the neurotypical population [12]. Although this observation
indicates that social attention is generally reduced in individuals with low empathy, the *features* of gaze behavior underlying this reduction remain fundamentally unclear.

99 The output of a typical eye-tracking experiment is a continuous stream of spatial 100 coordinates that define the location of an observer's gaze over time. To describe individual/ 101 group differences in social attention, this time series is typically collapsed into the total gaze 102 duration towards areas of interest (AOI's) containing social and non-social stimuli [7]. Whilst 103 total gaze duration is an intuitive and easily interpretable metric, it necessarily involves the 104 removal of informative components of the data contained within the temporal domain. Such 105 an approach may therefore fail in describing more subtle differences between individuals 106 that describe the dynamic nature of social attention. Although some previous studies of 107 social attention have considered the temporal origin of group differences via divergence 108 analyses [13-15] none have provided or tested a quantitative model of the entire time series. 109 To our knowledge, no existing study has provided an explicit model of the temporal structure 110 of social attention and tested predictions about individual-level social gaze behavior over 111 time.

112 The motivation for investigating individual differences in the temporal structure of 113 social attention is not purely data driven. At the theoretical level, prioritised perception of 114 socially relevant signals is one of the most important functions of the visual system. As such, 115 there is a major explanatory burden associated with identifying the features of gaze behavior 116 underlying individual variation in this phenomenon. Neurocognitive theories propose that 117 social attention is mediated by neural circuits that transduce sensory information about conspecifics and translate that information into value signals that bias the spatial allocation 118 119 of gaze over time [16]. In order to more fully appreciate what drives humans to attend to 120 social aspects of the world, one must investigate the individual characteristics that influence 121 this inherently dynamic process. By extension, this research effort may have the corollary of

122 informing explanatory models of disordered social attention. Moreover, influential models 123 propose that attention involves at least two distinct components of initial 'orienting' to and 124 subsequent 'maintaining' of engagement with stimuli [17]. In global eye-tracking metrics, 125 these two processes are conflated - total gaze duration towards social stimuli could reflect 126 some combination of both the orienting and maintaining mechanisms. Delineating these 127 mechanisms requires explicitly modeling the temporal components of the gaze signal. In 128 general, we may expect empathy to primarily influence gaze behavior some time after 129 stimulus presentation because arriving at an empathic response may require sampling many 130 relevant cues from a scene. We may need to attend to multiple subjects in the scene. 131 determine their event roles, recognise their facial expressions/ bodily postures and integrate 132 this information over time before an empathic response is triggered. This idea is consistent 133 with the recent observation that although empathy is predictive of gaze bias towards social 134 images after prolonged viewing, it does not predict the initial saccadic deviation towards 135 social images in a 'global effect' paradigm [12].

136 In the context of the preceding discussion, there is a clear lack of empirical work that 137 has attempted to model the temporal structure of social attention and its relationship with 138 individual social trait characteristics such as empathy. In this study, our goals were to i) 139 characterise the extent of gaze bias towards social stimuli in a large sample of observers ii) 140 model the time course of this social bias iii) determine how empathy modulates the time 141 course of the social bias. We report data from two large eye tracking datasets, with a 142 combined total of 176 observers.

Dataset 1

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144 Method

145 **Participants**

146 Ninety nine participants (58 females, *M* age= 23, *SD* age = 5) were recruited from in 147 and around the University of Reading. Ethical approval for the study was obtained from the 148 Research Ethics Committee of the University of Reading (Ethics ID: 2012/070/BC) and all 149 participants provided informed consent. All participants had normal or corrected to normal

vision. All participants except one female completed the Empathy Quotient (EQ) [16] a reliable, behaviorally validated measure of trait empathy. The mean EQ score was 44.21 (SD = 11.27), and the scores ranged from 25-73. This distribution of scores closely resembles that previously observed in large-scale surveys of the neurotypical population (e.g. [19]: N = 190, M = 44.5, SD = 10.7).

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156 Stimuli

157 Forty pairs of social and nonsocial reward images were taken from the International 158 Affective Picture System (18 pairs [20]) and downloaded from publicly available creative 159 common licensed images databases such as Flickr (22 pairs). All images were the same as 160 used in [10], in which social reward images included one or more humans (e.g. happy 161 individuals) while nonsocial reward images included rewarding nonsocial content (e.g. food. 162 scenery and money - see Supplementary Material S1). All stimuli in the experiment 163 subtended 15.4 x 9.15 degrees of visual angle (DVA), and pairs were separated by 5.29 164 DVA (Fig 1 b).

165 To reduce the influence of extraneous sensory and affective differences between 166 image pairs, all stimulus pairs were matched as closely as possible in terms of low level 167 properties (e.g. luminance, contrast, saliency) as well as perceived valence and arousal -168 see Supplementary Material S1. In addition, to further characterize the influence of low-level 169 confounds, we presented two stimulus types. All image pairs were manipulated via randomly 170 rearranging 10 x 10 pixel grids to create a set of 'scrambled' images in addition to the intact 171 images. The logic of this manipulation is that if simple low-level variability between image 172 pairs drives a gaze bias towards social images, we would expect to find a social bias of 173 similar magnitude for both the intact and scrambled stimulus types. By contrast, if social bias 174 is genuinely driven by the semantic content of the images, we would expect social bias to be 175 substantially reduced for scrambled stimuli.

176 **Procedure**

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Observers were seated 50 cm in front of a Tobii T60 eye-tracker with an inbuilt 1280

178 x 1024 pixel resolution monitor (60hz refresh rate) and sampling rate of 60Hz (Figure 1a). 179 Stimuli were presented using E-Prime 2.0 (Psychology Software Tools, PA, USA [21]) 180 Following a 5-point calibration, participants completed the freeviewing task: Observers were 181 informed that they would be presented with pairs of images side by side for 3 seconds, and 182 that they were free to look wherever they liked during this period. Figure 1b depicts the trial 183 sequence: observers were presented with a fixation cross for 500 ms, followed by a pair of 184 the social and nonsocial stimuli for 3000 ms. To maintain engagement with the task, the 185 color of the fixation cross changed from black to blue on 10% of trials. The participant was 186 asked to report these changes via button press as rapidly as possible. Observers completed 187 80 trials in total (40 image pairs, 2 stimulus types).

188 **Results**

189 Aggregated Social Bias

190 Data reduction was performed via the 'eyetrackingR' package, implemented in the R 191 programming language [22] The display coordinates occupied by the social and nonsocial 192 images on each trial were defined as areas of interest (AOIs). We first analysed the data by 193 aggregating across the time dimension. To this end, we reduced the raw gaze data for each 194 participant into the proportion trial time that gaze was directed into the social AOI and 195 nonsocial AOI. This data was submitted to a general linear model with AOI (social, 196 nonsocial) and stimulus type (intact, scrambled) as fixed effects. Reported significance tests 197 of model coefficients were conducted via likelihood ratio tests of nested models containing 198 the coefficients versus those without them. There was a main effect of AOI, indicating gaze 199 bias towards social images $\chi^2(1) = 104.02$, p < .001. Moreover, the predicted interaction 200 between AOI and stimulus type was detected χ^2 (1) = 18.92, p < .001 (Figure 1c). The bias 201 for social images was larger in the intact condition ($\beta = 0.12$) than scrambled condition ($\beta =$ 202 0.05). Adding EQ to the model revealed a 3 way interaction between AOI, stimulus type and EQ χ^2 (1) = 5.90, *p* = .020. Higher EQ was associated with a larger social bias for intact 203 204 stimuli than scrambled stimuli (Figure 1d).

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INSERT FIGURE 1 HERE

Time-course of Social Bias.

209 Having analysed the aggregated data expressed as total gaze duration, we next 210 aimed to estimate a parsimonious model that described the time course of social bias across 211 participants. For each observer, we first removed trials for which gaze failed to record for 212 more than 60% of a trial (16% of the data). Next, we reduced each observer's gaze data into 213 the proportion of gaze within the social and non-social AOI in each 100ms time bin from the 214 start to end of the trial. We then removed data from the first 100 ms time bin, since it 215 contained 3 SDs less than the mean number valid samples captured within all time bins. No 216 association was detected between EQ and the number of remaining data points when this 217 cleaning strategy was applied r(96) = -.019, p = .851.

218 Figure 2a depicts the time course of gaze proportion into the social AOI for intact 219 stimuli. This gaze bias towards social images is not time invariant (Figue 2a), nor is its time 220 course well described by a linear function (Figure 2b). The global pattern is an initial bias 221 towards the social AOI that peaks within the first 500 ms, followed by a nonlinear decline 222 and a partial recovery towards the end of the trial. To model these nonlinear components of 223 the time course, we proceeded via forward selection and tested the performance of models 224 that included higher-order time regressors [23]. To protect against overfitting, we tested the 225 generalisation performance of each model, using standard leave one out (LOO) cross-226 validation procedures (see Supplementary Material S2, S3). Once linear and guadratic time 227 regressors were added, the addition of higher order terms failed to reduce residuals or 228 improve LOO performance, suggesting that more complex models were prone to overfitting. 229 Therefore, a model with AOI and linear and guadratic time regressors as fixed effects (AIC = 230 -6365.5) was retained as our global model of the time-course of the social bias (Figure 2c). 231 Effect of Empathy on Time Course of Social Bias

232 Having modelled the time course of the social bias pooled across participants, we 233 next attempted to model variation at the individual level. We first tested whether empathy 234 modulates the time-course of the social bias by defining EQ as a predictor of proportion of 235 gaze in the social AOI within each 100 ms time bin. An effect of EQ as a predictor of gaze 236 into the social AOI was detected within 3 'clusters' of contiguous time bins (Figure 2d, see 237 Supplementary Material S4 for a rationale for defining clusters). These were located i) at 238 100-900 ms ii) at 1500-1600 ms iii) at 1800- 2900 ms. Given the multiple tests associated 239 with this analysis, our type 1 error rate may have reached unacceptable levels. Therefore, to 240 protect against false positives, we performed a bootstrapped cluster-based permutation 241 analysis (Supplementary Material S4) akin to that typically applied to electroencephalogram 242 data [24]. After this correction was applied, there was no detectable effect in the second 243 cluster (p = .316), whereas the chances of obtaining the summed statistics observed in the 244 first and last cluster under the null hypothesis were estimated to be at p = .003 and p = .002245 respectively.

246 With this temporal influence of empathy established, we next proceeded to test 247 models that added EQ as a fixed effect to our initial global model of the time-course 248 (Supplementary Material S5). We first specified a reduced interactive model, which 249 constrained EQ to interact only with AOI but not the time regressors. This led to improved 250 model fit χ^2 (2) = 337.47, *p* <.001, consistent with the previously observed generalised 251 increase in social bias associated with high EQ. Next we specified a fully interactive model, 252 which removed this constraint and allowed EQ to additionally interact with the time regressors. This further improved on the reduced interactive model χ^2 (4) = 72.70, *p* <.001. 253 254 To aid interpretation of this model, its predictions are plotted with the empirical data for 5 255 observers (Figure 2e), whose EQ is ordered from left to right (low to high). The model 256 predicts that EQ is associated with a generalised increase in gaze bias towards the social 257 AOI (i.e. the vertical offset between the blue and green lines), but that this effect is 258 particularly pronounced at the start and end of the trial. Given the complexity of this fully

259 interactive model, we again protected against overfitting via another LOO analysis, which 260 confirmed that this model had the superior performance (Supplementary Material S5). 261 In good agreement with the results of our cluster-based analysis, this confirms that 262 EQ is not only associated with a generalised increase in social bias, but also with a different 263 temporal profile of social bias. Inspection of figure 2e reveals that EQ predicts an initial 264 increase in social attention, but also a more sustained component that maintains social 265 attention at the later portions of the trial. 266 267 **INSERT FIGURE 2 HERE** 268 269 One plausible mechanism for this sustained component is that, after being initially 270 fixated, social images hold attention for longer durations in high empathy individuals than 271 low empathy individuals. To test this possibility, we split trials according to the AOI that was 272 initially fixated and analysed the latency at which observers switched their gaze to the 273 alternate AOI. We reasoned that if empathy was associated with sustained attention on 274 social images, this would be manifested in an interactive effect of EQ and initial AOI on gaze 275 switch latency. Figure 3a depicts the proportion of observers who switched AOI as a 276 function of the initial AOI, EQ (median split for visualisation) and time. Inspection of this 277 figure reveals that low EQ individuals switched from the social AOI more frequently and at 278 earlier latencies than high EQ individuals. The predicted interaction between EQ and initial 279 AOI on switch latency was detected χ^2 (1) = 4.56, p = .030. Higher EQ was associated with 280 later switching from the social AOI relative to the nonsocial AOI (Figure 3b). 281 282 **INSERT FIGURE 3 HERE** 283 284 Dataset 2

285 Our analyses of the first dataset indicate a robust effect of empathy on the time 286 course of social attention. To further validate our initial findings, we next tested their 287 generalisation performance via a re-analysis of an existing, independent dataset [12]. 288 289 Method 290 **Participants** 291 77 participants (42 females; M = 21 years, SD = 3 years) drawn from in and around 292 the University of Reading campus completed the FV task. All participants had normal or 293 corrected to normal vision. 68 (38 female) participants completed the online EQ 294 questionnaire. The study was approved by the University of Reading Research Ethics 295 Committee (Ethics ID: 2010/86/BC). 296 Stimuli 297 The images and image pairings were the same as those described for Dataset 1. 298 Procedure 299 The only procedural differences from those described in Dataset 1 were as follows. 300 Participants were seated at 100 cm from a 1600 x 1200 pixel resolution colour monitor (75hz 301 refresh rate). Eye movements were recorded via a video based eye-tracker with a sampling 302 rate of 500hz (Eyelink 2, SR research). Stimuli were presented via Experiment Builder 303 software [25]. The presentation duration of stimuli in this task was 5000 ms and stimuli 304 subtended 5.59 x 4.19 DVA. 305 Results 306 **Aggregated Social Bias** 307 Inspection of Figure 4 reveals a pattern of results that very closely mirror those obtained from Dataset 1. There was again the same main effect of AOI χ^2 (1) = 91.40, p 308 <.001 and interaction between AOI and stimulus type $\chi^2(1) = 28.61$, p <.001 (Figure 4a). 309 310 The bias for social images was similarly larger in the intact condition ($\beta = 0.13$) than

311 scrambled condition (β = 0.04). Adding EQ to the model revealed the same 3 way

interaction between AOI, stimulus type and EQ χ^2 (1) = 18.21, *p* <.001. Higher EQ was

associated with a larger social bias for intact stimuli, but not scrambled stimuli (Figure 4b).

314

Time-course of Social Bias

We used the same data reduction strategy as reported for Dataset 1. We removed 2.85% trials due to trackloss and again removed data from the first 100 ms timebin. No association was detected between EQ and the number of remaining data points when this cleaning strategy was applied r (67) = -.003, p = .981. The forward selection strategy revealed that a model involving AOI and a linear and quadratic time regressors as fixed effects (Figure 4c) again provided the best fit to the data (AIC -9639.3) and had the best generalisation performance (see supplementary material S6).

322 Effect of Empathy

An effect of EQ as a predictor of social bias was detected within a cluster from 2800 -5000ms (*corrected* p =.009 - Figure 4d). We again tested models that added EQ as a fixed effect to our initial model of the global data. The reduced interactive model again improved model fit χ^2 (2) = 335.98, p <.001. Moreover, a fully interactive model further improved on the reduced interactive model χ^2 (4) = 85.14, p <.001. EQ was primarily predictive of social bias towards the end of the trial (Figure 4d).

An analysis of switch latencies did not detect an interaction between initial AOI and EQ χ^2 (1) =3.52, p =.060, but the effect was similar in magnitude and direction to that observed in Dataset 1. Higher EQ was again associated with later switching from the social AOI relative to from the nonsocial AOI (Figure 4e, Figure 4f).

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Discussion

INSERT FIGURE 4 HERE

In this study our major novel contributions were as follows: We i) provide an explicitmodel of the time course of social attention, ii) determine how the parameters of this model

339 are modulated by social trait characteristics of the observer iii) test this model by making 340 quantitative predictions about the allocation of an individual's gaze over time. Across two 341 large datasets, we found a number of similar findings. i) Observers exhibit a robust gaze 342 bias towards social images ii) EQ is reliably associated with an increase in this bias iii) This 343 effect of EQ is not time invariant - a model that allowed empathy to interact with the temporal 344 components of the gaze bias provided a superior fit to a model that assumed a time-345 invariant effect of empathy. Specifically, empathy was found to reliably maintain gaze bias 346 towards social images after prolonged viewing. iv) Higher EQ was associated with less 347 frequent, and later switching from an initially fixated social image.

348 At the most fundamental level, our finding that gaze behavior is predicted by the 349 social trait characteristics of the observer emphasises that the mechanisms underlying social 350 attention are deeply enmeshed with other aspects of social cognition. The dynamic influence 351 of empathy on gaze behavior suggests that empathy is not a passive affective resonance 352 with the emotions of others and that wider contextual influences play feed-forward roles in 353 how emotions are perceived and experienced. This fits with neurocognitive theories of 354 empathy, which propose that empathy is implemented by a network of recursively connected 355 cortical and subcortical sites [26]. It also fits well with multi-stage models of empathy, which 356 propose that prolonged attention to social stimuli reflects a form of evidence gathering so 357 that appropriate empathic responses can be generated [27,28].

358 Our findings appear consistent with recent pharmacological work, which indicates 359 that administration of oxytocin (associated with the experience of empathy in humans and 360 mesolimbic dopaminergic activity involved in responding to rewards) predicts maintained 361 periods of eye-contact in Macaque monkeys [14]. We speculate the similarity of these 362 findings with our own reflect some common mechanism that promotes prolonged perceptual 363 selection of socially relevant inputs. Computational models of alternative forced choice 364 behavior have been proposed that explicitly relate gaze behavior to value coding. The 'gaze 365 cascade model' proposes that gaze and value coding mutually interact, resulting in an 366 increased gaze towards preferred stimuli over time [29]. A consistent observation from both

367 of our datasets is that trait empathy is better able to predict gaze toward social rewards 368 towards the end of the trial. One potential interpretation of this observation is that trait 369 empathy is related to enhanced motivational salience of social stimuli. By extension, we 370 speculate that the individual differences in the temporal evolution of eye-movement behavior 371 observed in our study reflects some online behavioral correlate of the value-coding process. 372 This inference relies on electrophysiological studies that show value-coding is a dynamic 373 process, and requires accumulation of evidence over time [30]. This interpretation of 374 empathy being related to the value coding of social rewards is also consistent with the 375 observation that higher empathy is associated with greater reward-related striatal activation 376 in response to socially stimuli [30]. Our free-viewing task, of course, did not require 377 observers to make an explicit choice between two stimuli. Recent computational modeling of 378 binary choice behavior indicates that impressive predictions of choice behavior can be 379 generated by models that incorporate gaze behavior and the reward value of competing 380 stimuli [32]. In this context, an interesting question concerns whether empathy similarly 381 predicts different trajectories of social attention and different gaze cascade effects in choice-382 based paradigms.

383 In interpreting our findings, it is important to acknowledge that gaze behavior in 384 response to complex rewarding scenes is likely to reflect the output of many dissociable and 385 fundamental processes. As such, the pattern of results we found could also be driven by 386 some combination of component processes found to vary as a function of empathy. This 387 may include individual differences in gaze perception [33] expression recognition [34] 388 temporal integration [15] and a precedence of local over global processing [35]. Our data 389 cannot clarify the relative contribution of these factors. Moreover, gaze behavior is strongly 390 determined by low-level properties, such as luminance contrast and spatial frequency profile. 391 Although we attempted to protect against these issues with our matching procedures and 392 use of scrambled control stimuli, our stimuli are still not immune to these issues. However, 393 no study involving complex, naturalistic visual stimuli is completely resistant to these potential confounds. 394

395 In the absence of longitudinal data, a claim about the directionality of the causal 396 relationship between empathy and social attention observed here is clearly over-reaching. 397 Based on the available developmental literature, however, there are sensible grounds for 398 proposing that some aspects of social attention precede empathy. Newborns exhibit robust 399 orienting responses to conspecific stimuli (particularly faces) [36], whereas the cognitive 400 components of empathy (such as theory of mind) emerge several years in development [37]. 401 In this context, our study could motivate well-controlled developmental studies that track the 402 temporal structure of social attention across development and its shared trajectory with the 403 development of empathic abilities.

404 Our findings have several important implications for the design of future studies. We 405 observed that empathy can take effect on behavior several seconds after stimulus onset. 406 Spontaneous mimicry, related to certain components of empathy [38] can also take effect 407 several seconds after stimuli onset (e.g. in response to reward [39]). Findings like these may 408 question the sensitivity of methods that rely on much briefer stimulus exposures, such as 409 visual probe paradigms [40-42] in detecting differences between groups that vary in 410 empathic traits. There is widespread enthusiasm for the idea that electrophysiological 411 methods with high temporal resolution may further clarify the temporal brain dynamics of 412 empathy [43,44] and distinguish between competing explanatory models. Based on the 413 findings reported in this paper, we are additionally enthusiastic about the prospect of 414 paradigms that employ concurrent recording of both EEG and gaze data. Capitalising on the 415 high temporal resolution shared by these methods may lead to theoretical advancement by 416 providing insight into the time-course of the neural signatures underlying empathy and their 417 behavioral correlates. Motivated accounts of empathy suggest that observers may 418 dynamically increase or decrease attention to social cues to regulate their emotional 419 responses [28]. Paradigms that concurrently monitor gaze allocation and autonomic arousal 420 over time could explicitly test the predictions of such models.

421 In general, our data demonstrate that considering the temporal structure of gaze 422 signals may provide impetus towards enhanced behavioral phenotypes for conditions 423 marked by deficits in one or more empathy related processes (ASC, Psychopathy, Bipolar 424 Disorder, Schizophrenia [45-47]). More broadly, follow up experimentation of this variety can 425 also help us answer the more fundamental question: What features of gaze behavior 426 differentiate between individuals with and without these conditions? Failing to capitalize on 427 the high-dimensional, time-varying nature of gaze signals necessarily entails restricting the 428 information available for answering this question. 429

431		References
432 433	1.	Boyer P, Bergstrom B. Threat-detection in child development: An evolutionary perspective. Neurosci Biobehav Rev. 2011 Mar;35(4):1034–41.
434 435 436	2.	Hayden BY, Parikh PC, Deaner RO, Platt ML. Economic principles motivating social attention in humans. Proceedings of the Royal Society of London B: Biological Sciences. 2007 Jul 22;274(1619):1751–6.
437 438	3.	Chevallier C, Kohls G, Troiani V, Brodkin ES, Schultz RT. The Social Motivation Theory of Autism. Trends Cogn Sci. 2012-4;16(4):231–9.
439 440	4.	Adams RB Jr, Adams RB, Ambady N, Shimojo S, Nakayama K. The Science of Social Vision: The Science of Social Vision. OUP USA; 2011. 471 p.
441 442 443	5.	Hafri A, Trueswell JC, Strickland B. Encoding of event roles from visual scenes is rapid, spontaneous, and interacts with higher-level visual processing. Cognition. 2018 Jun 1;175:36–52.
444 445	6.	Klein JT, Shepherd SV, Platt ML. Social Attention and the Brain. Curr Biol. 2009 Nov 3;19(20):R958–62.
446 447	7.	Chita-Tegmark M. Social attention in ASD: A review and meta-analysis of eye-tracking studies. Res Dev Disabil. 01/2016;48:79–93.
448 449 450	8.	Klin A, Lin DJ, Gorrindo P, Ramsay G, Jones W. Two-year-olds with autism orient to non-social contingencies rather than biological motion. Nature. 2009 May 14;459(7244):257–61.
451 452 453 454	9.	Pierce K, Marinero S, Hazin R, McKenna B, Barnes CC, Malige A. Eye Tracking Reveals Abnormal Visual Preference for Geometric Images as an Early Biomarker of an Autism Spectrum Disorder Subtype Associated With Increased Symptom Severity. Biol Psychiatry. 2016 Apr 15;79(8):657–66.
455 456	10.	Fischer J, Koldewyn K, Jiang YV, Kanwisher N. Unimpaired Attentional Disengagement and Social Orienting in Children with Autism. Clin Psychol Sci. 2014 Mar;2(2):214–23.
457 458	11.	Shah P, Gaule A, Bird G, Cook R. Robust orienting to protofacial stimuli in autism. Curr Biol. 2013 Dec 16;23(24):R1087–8.
459 460 461	12.	Chakrabarti B, Haffey A, Canzano L, Taylor CP, McSorley E. Individual differences in responsivity to social rewards: Insights from two eye-tracking tasks. PLoS One. 2017 Oct 18;12(10):e0185146.
462 463 464	13.	Sasson N, Tsuchiya N, Hurley R, Couture SM, Penn DL, Adolphs R, et al. Orienting to social stimuli differentiates social cognitive impairment in autism and schizophrenia. Neuropsychologia. 2007 Jun 18;45(11):2580–8.
465 466 467	14.	Dal Monte O, Piva M, Anderson KM, Tringides M, Holmes AJ, Chang SWC. Oxytocin under opioid antagonism leads to supralinear enhancement of social attention. Proc Natl Acad Sci U S A. 2017 May 16;114(20):5247–52.
468 469 470 471	15.	Nakano T, Tanaka K, Endo Y, Yamane Y, Yamamoto T, Nakano Y, et al. Atypical gaze patterns in children and adults with autism spectrum disorders dissociated from developmental changes in gaze behaviour. Proc Biol Sci. 2010 Oct 7;277(1696):2935– 43.

- 472 16. Chang SWC, Brent LJN, Adams GK, Klein JT, Pearson JM, Watson KK, et al.
 473 Neuroethology of primate social behavior. Proc Natl Acad Sci U S A. 2013 Jun 18;110
 474 Suppl 2:10387–94.
- 475 17. Posner MI, Snyder CR, Davidson BJ. Attention and the detection of signals. J Exp
 476 Psychol. 1980 Jun;109(2):160–74.
- 477 18. Lawrence EJ, Shaw P, Baker D, Baron-Cohen S, David AS. Measuring empathy:
 478 reliability and validity of the Empathy Quotient. Psychol Med. 2004 Jul;34(5):911–9.
- 479 19. Baron-Cohen S, Wheelwright S. The empathy quotient: an investigation of adults with
 480 Asperger syndrome or high functioning autism, and normal sex differences. J Autism
 481 Dev Disord. 2004 Apr;34(2):163–75.
- 482 20. [PDF]International Affective Picture System (IAPS): Technical ... Unifesp. Available
 483 from: https://www2.unifesp.br/dpsicobio/adap/instructions.pdf
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- 486 22. eyetrackingR [Internet]. [cited 2018 Jun 7]. Available from: http://www.eyetracking 487 r.com/
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- 491 24. Maris E, Oostenveld R. Nonparametric statistical testing of EEG- and MEG-data. J
 492 Neurosci Methods. 2007 Aug 15;164(1):177–90.
- 493 25. SR Research Experiment Builder 1.10.165 [Computer software]. (2011). Mississauga,
 494 Ontario, Canada: SR Research Ltd.
- 495 26. Decety J. The neurodevelopment of empathy in humans. Dev Neurosci. 2010 Aug 31;32(4):257–67.
- 497 27. Bird G, Viding E. The self to other model of empathy: providing a new framework for
 498 understanding empathy impairments in psychopathy, autism, and alexithymia. Neurosci
 499 Biobehav Rev. 2014 Nov;47:520–32.
- 500 28. Zaki J. Empathy: a motivated account. Psychol Bull. 2014 Nov;140(6):1608–47.
- Shimojo S, Simion C, Shimojo E, Scheier C. Gaze bias both reflects and influences
 preference. Nat Neurosci. 2003 Nov 9;6:1317.
- 503 30. Kim H, Sul JH, Huh N, Lee D, Jung MW. Role of Striatum in Updating Values of Chosen
 504 Actions. J Neurosci. 2009 Nov 25;29(47):14701–12.
- 505 31. Chakrabarti B, Bullmore E, Baron-Cohen S. Empathizing with basic emotions: common and discrete neural substrates. Soc Neurosci. 2006;1(3-4):364–84.
- 507 32. Krajbich I, Armel C, Rangel A. Visual fixations and the computation and comparison of value in simple choice. Nat Neurosci. 2010 Oct;13(10):1292–8.
- 509 33. Pantelis PC, Kennedy DP. Deconstructing atypical eye gaze perception in autism
 510 spectrum disorder. Sci Rep. 2017 Nov 8;7(1):14990.

- 511 34. Harms MB, Martin A, Wallace GL. Facial emotion recognition in autism spectrum
 512 disorders: a review of behavioral and neuroimaging studies. Neuropsychol Rev. 2010
 513 Sep;20(3):290–322.
- 514 35. Dakin S, Frith U. Vagaries of visual perception in autism. Neuron. 2005 Nov 3;48(3):497–507.
- 516 36. Johnson MH, Dziurawiec S, Ellis H, Morton J. Newborns' preferential tracking of face-517 like stimuli and its subsequent decline. Cognition. 1991 Aug 1;40(1):1–19.
- 518 37. Leslie AM, Friedman O, German TP. Core mechanisms in "theory of mind." Trends
 519 Cogn Sci. 2004 Dec 1;8(12):528–33.
- 38. Pfeifer JH, Iacoboni M, Mazziotta JC, Dapretto M. Mirroring others' emotions relates to
 empathy and interpersonal competence in children. Neuroimage. 2008 Feb
 15;39(4):2076–85.
- 523 39. Sims TB, Van Reekum CM, Johnstone T, Chakrabarti B. How reward modulates
 524 mimicry: EMG evidence of greater facial mimicry of more rewarding happy faces.
 525 Psychophysiology. 2012 Jul;49(7):998–1004.
- 526 40. Zhao X, Zhang P, Fu L, Maes JHR. Attentional biases to faces expressing disgust in 527 children with autism spectrum disorders: an exploratory study. Sci Rep. 2016 Jan 528 13;6:19381.
- 41. Moore DJ, Heavey L, Reidy J. Attentional Processing of Faces in ASD: A Dot-Probe
 Study. J Autism Dev Disord. 2012 Oct 1;42(10):2038–45.
- 42. Quintana DS, Westlye LT, Hope S, Nærland T, Elvsåshagen T, Dørum E, et al. Dosedependent social-cognitive effects of intranasal oxytocin delivered with novel Breath
 Powered device in adults with autism spectrum disorder: a randomized placebocontrolled double-blind crossover trial. Transl Psychiatry. 2017 May 23;7(5):e1136.
- 535 43. Neumann DL, Westbury HR. The psychophysiological measurement of empathy.
 536 Psychology of empathy. 2011;119–42.
- 44. Suzuki Y, Galli L, Ikeda A, Itakura S, Kitazaki M. Measuring empathy for human and robot hand pain using electroencephalography. Sci Rep. 2015 Nov 3;5:15924.
- 539 45. Derntl B, Finkelmeyer A, Toygar TK, Hülsmann A, Schneider F, Falkenberg DI, et al.
 540 Generalized deficit in all core components of empathy in schizophrenia. Schizophr Res.
 541 2009 Mar;108(1-3):197–206.
- 542 46. Jones AP, Happé FGE, Gilbert F, Burnett S, Viding E. Feeling, caring, knowing:
 543 different types of empathy deficit in boys with psychopathic tendencies and autism
 544 spectrum disorder: Comparing empathy deficits in boys with psychopathic tendencies
 545 and ASD. J Child Psychol Psychiatry. 2010 Nov 13;51(11):1188–97.
- 546
 547. Shamay-Tsoory S, Harari H, Szepsenwol O, Levkovitz Y. Neuropsychological evidence
 of impaired cognitive empathy in euthymic bipolar disorder. J Neuropsychiatry Clin
 548 Neurosci. 2009 Winter;21(1):59–67.
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Figures



Figure 1. a) Schematic of experimental setup and *b)* trial sequence. c) Gaze proportion as a
function of AOI and stimulus type. Red points indicate individual data. *d)* Gaze proportion as
a function of AOI, stimulus type and EQ. Error bars are +/- 1 SEM.



Figure 2. a) Shows the time series fit to the gaze proportion into the social AOI with only AOI 559 560 as a fixed effect (no effect of time). b) Shows a fit to the same data with AOI and a linear 561 time regressor as fixed effects. c) Shows the the data fit with AOI and linear and quadratic 562 time regressors. d) Shows t statistics for the test that EQ is a linear predictor of gaze 563 proportion into the social AOI within each 100ms time bin. Shaded areas demarcate the time 564 bins wherein the statistic reaches the (uncorrected) threshold for rejecting the null 565 hypothesis. e) Shows predictions of the fully interactive model for 5 observers. The panel 566 headers indicate the observer's EQ score. Solid lines are model predictions, points are the 567 empirical data. 568 569





573 *Figure 3.* a) Shows proportion of observers who switched to the alternate AOI as a function

of initial AOI, EQ (median split) and time b) Shows switch latency as a function of initial AOI

575 and EQ. Error bars are +/- 1 SEM.



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Figure 4. a) Gaze proportion as a function of AOI and stimulus type. Red points indicate individual data. b) Gaze proportion as a function of AOI, stimulus type and EQ. Error bars are +/- 1 SEM. c) Shows the fit to gaze proportion time series with AOI and a linear and quadratic time regressors as fixed effects. Data is shown for the social AOI. d) Shows *t* statistics for the test that EQ is a linear predictor of gaze proportion into the social AOI within each 100ms time bin. Shaded areas demarcate the time bins wherein the statistic reaches

- 585 the (uncorrected) threshold for rejecting the null hypothesis. e) Shows proportion of
- 586 observers who switched to the alternate AOI as a function of initial AOI and EQ (median
- 587 split) f) Shows switch latency as a function of initial AOI and EQ. Error bars are +/- 1 SEM.