

1 Unity in diversity: Children from 17 diverse communities process gaze in similar ways

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Abstract

Understanding gaze is critical for human social interaction. In a comprehensive cross-cultural study spanning five continents and 17 distinct cultural contexts, we examined the development of gaze understanding among preschoolers using a single reliable task tailored to each cultural setting. Our data provides evidence that children worldwide process gaze in the same way. Key performance signatures of a computational model that sees gaze following as a form of social vector estimation were found in all communities. Additionally, we found a subtle yet cross-culturally consistent relationship between children's environments and their ability to understand gaze. These results highlight the cross-cultural robustness of core social cognitive skills, as well as similarities in the developmental process, suggesting that fundamental aspects of social cognition and interaction emerge and develop in comparable ways globally and hence are a bedrock of human social cognition.

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Social cognition is a defining aspect of the human species (1–3). It is supposed to enable unique forms of communication and cooperation that underlie cumulative cultural evolution and the formation of complex societies (4–6). The eyes are the proverbial window to the mind and the starting point for a majority of social reasoning processes (7). Gaze is used to infer the focus of visual attention, which is a critical aspect of coordinated activities, including communication and cooperation (8, 9). The ability to follow gaze emerges early in childhood (10–12) and individual differences in children’s gaze following ability predict later life outcomes, most notably, later communicative abilities (13, 14). Underlying this narrative is the wide-held assumption that gaze following is fundamental to human social cognition and, therefore, works and develops in the same way across human societies despite substantial variation in developmental contexts. This claim, however, lacks a solid empirical foundation.

We conducted a large-scale cross-cultural study on the development of gaze-following abilities to study potentially universal processing signatures and their development. Previous developmental studies focused on the onset of gaze following in infancy (15, 16). The 1377 participants in the study lived in 17 different communities spread over five continents (Fig. 1A, Tab. 1). The countries from which data was contributed to the study represented 46 % of the world’s population. Communities covered a broad spectrum of geographical locations, social and political systems, languages, and subsistence styles (see Supplemental Materials).

We used an animated picture book tablet task in which participants were asked to locate a hidden object based on observing an agent’s gaze. Children watched a balloon disappear behind a hedge. An agent followed the trajectory of the balloon with their eyes (Fig. 1B). The key dependent variable was the precision with which children located the agent’s focus of attention, that is, the deviation between where the agent looked (where the object was) and where the child thought the agent looked. We adapted visuals and audio

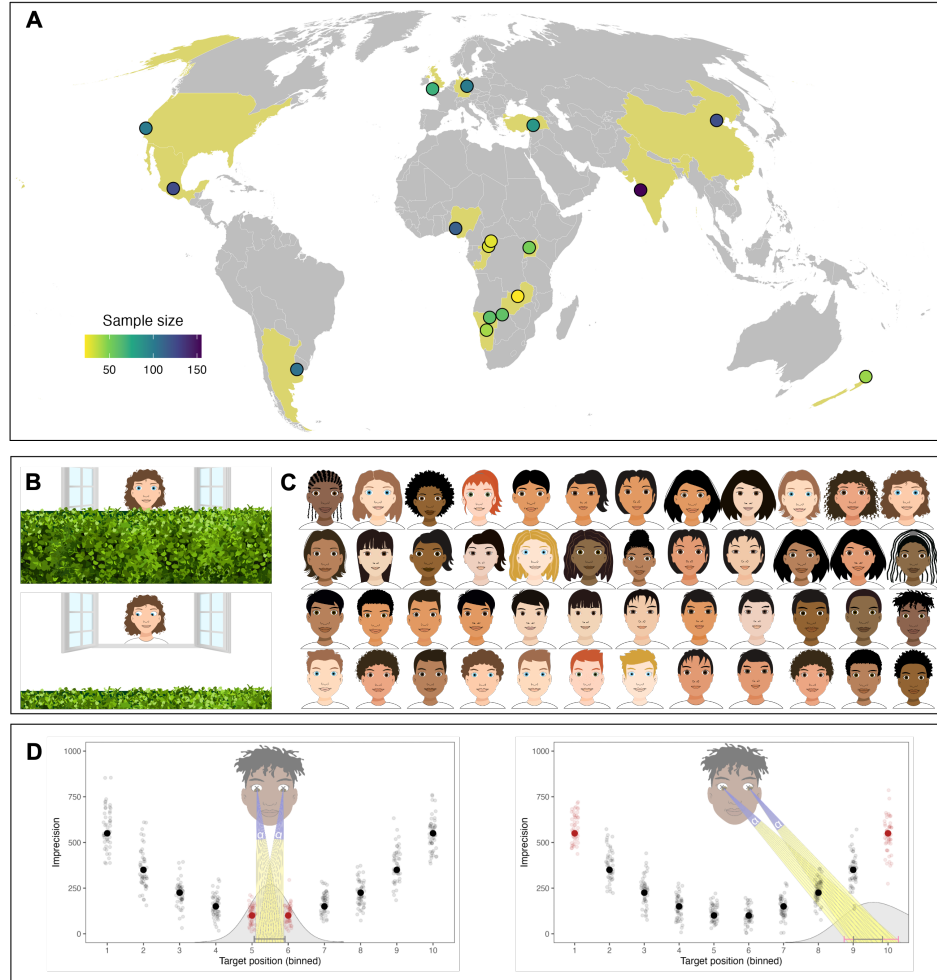


Figure 1. (A) Data collection sites. Points show approximate geographical location of the data collection sites, coloring shows the sample sizes. (B) Screenshot from the task. The scene depicts the choice phase in a test trial. Participants had to use the gaze of the agent to locate the balloon and click on the hedge where they thought the balloon was. Agents, audio recordings and backgrounds were adapted to each cultural setting. (C) Drawings used as agents across cultural settings. (D) Graphical illustration of the cognitive model. Solid and transparent dots show simulated means and individual data points. Individuals infer the target of an agent’s attention by estimating a vector based on the position of the pupils within the eyes. This process is noisy, illustrated by the different vectors. Individuals differ in their level of precision (indicated by α). For a given level of precision, the further out the target lands, the less precise the model predicts individuals to be.

Table 1

Participant demographics.

Continent	Country	Community	N(male)	Age (range)	Language	Market integration	Touchscreen
America	Argentina	Buenos Aires	105 (53)	4.72 (3.00 - 6.96)	Spanish (Rioplatense)	high	0.90
	Mexico	Ocuilan	127 (63)	4.96 (2.57 - 6.95)	Spanish (Mexican)	medium	0.77
	USA	Stanford	98 (54)	4.99 (2.52 - 7.90)	English (American)	high	0.98
Africa	Namibia	Hai om	60 (38)	5.85 (2.74 - 8.34)	Hai om	low	0.05
		Khwe	59 (24)	5.84 (3.38 - 8.63)	Khwedam	low	0.19
		Windhoek	39 (17)	5.69 (2.66 - 8.66)	English (Nigerian)	high	0.95
	Nigeria	Akure	114 (54)	5.07 (2.57 - 7.33)	English (Nigerian)	high	0.91
	Rep. Congo	BaYaka	29 (13)	7.80 (3.94 - 10.56)	BaYaka	low	0.00
		Bandongo	30 (11)	7.45 (3.50 - 10.95)	Lingala	low	0.00
	Sambia	Chimfunshi	22 (5)	5.98 (2.88 - 8.00)	Bemba	medium	0.14
	Uganda	Nyabyeya	125 (62)	5.94 (2.67 - 8.92)	Swahili	medium	0.34
Europe	Germany	Leipzig	100 (48)	4.88 (2.53 - 6.95)	German	high	0.89
	UK	Plymouth	70 (30)	6.02 (2.38 - 8.94)	English (British)	high	0.99
Asia	China	Beijing	123 (62)	5.47 (2.69 - 8.48)	Mandarin	high	0.95
	India	Pune	148 (73)	6.14 (3.06 - 8.83)	English (Indian) / Marathi	high	0.93
	Turkey	Malatya	85 (40)	5.02 (2.75 - 7.12)	Turkish	high	1.00
Oceania	New Zealand	Auckland	43 (19)	5.14 (2.81 - 8.75)	English (New Zealand)	high	0.95

instructions specifically to each of the 17 communities. Previous work demonstrated excellent individual-level measurement properties for this task (17). Thus, in addition to group-level trends, we were able to investigate individual-level variation.

As the first step, we investigated developmental gains. Across all 17 communities, we found a substantial increase in average levels of precision with age (fixed effect in Bayesian regression model (18): $\beta = -0.30$, 95% HDI (-0.39 - -0.21); range of community-level effects in model: $\beta_{min} = -0.06$, 95% HDI (-0.17 - 0.06) to $\beta_{max} = -0.58$, 95% HDI (-0.69 - -0.46). Nevertheless, there were also marked differences between communities (see Fig. 2A). In a cross-validation procedure, in which we trained a regression model on a subset of the data (training data) to later predict the held out data (testing data) (19), we found that a model that assumed cross-cultural variation to outperform simple models in 100% of cases (see

Supplemental Material). At first glance, it seems that highly market-integrated communities around the globe showed higher levels of precision compared to less market-integrated communities (see Tab. 1). However, believe that these results are better understood in terms of exposure to technology and we discuss them in more detail below. Furthermore, average differences between communities were small compared to differences between individuals: communities did not form homogeneous clusters but largely overlapping distributions in that some individuals from communities with a lower average level of precision performed better compared to some individuals from a setting with a very high average level of precision. Similarly, in all communities, some 4-year-olds outperformed children two years older than them (see Fig. 2A).

Consistent developmental gains alone cannot inform us about the cognitive processes children use when locating the agent’s focus of attention. Recent computational work modeled gaze following as social vector estimation (20). When observing the eyes, onlookers estimate a vector from the center of the eye through the pupil. The focus of attention is the location wherever the estimated vectors from both eyes hit a surface (Fig. 1C). The estimation process is not perfect, but each individual has a systematic level of uncertainty, which is conceptualized as the cause of individual differences. Importantly, this process model predicts a clear performance signature in our cross-cultural gaze following task: Trials in which the agent looks further away from the center (i.e. to the left or right side of the screen) should result in lower levels of precision, compared to trials in which the agent looks closer to the middle. This prediction is best understood by considering a similar phenomenon: pointing a torch light to a flat surface. The width of the light beam represents each individual’s level of uncertainty in vector estimation. When the torch is directed straight down, the light beam is concentrated in a relatively small area. When the torch is rotated to the side, the light from one half of the cone must travel further than the light from the other half to reach the surface. As a consequence, the light is spread over a wider area.

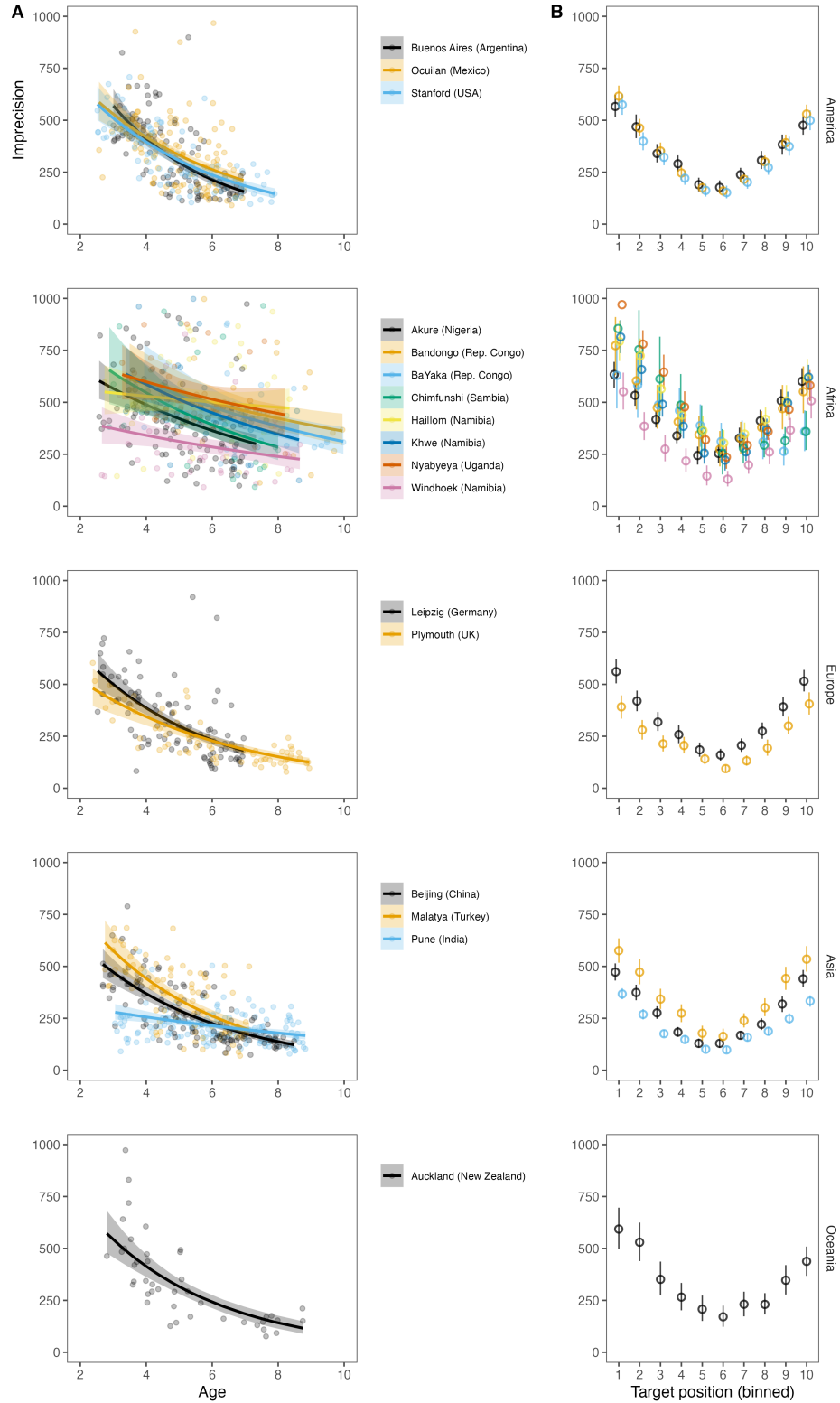


Figure 2. A) Developmental trajectories by cultural settings. Transparent dots show aggregated data for each individual. The color denotes the different cultural settings. The developmental trajectories are predicted based on a model of the data aggregated for each participant. B) Performance by screen section and cultural setting. Each bin covers 1/10th of the screen. Points show means and error bars 95% confidence intervals for the data within

The processing signature was clearly visible across all 17 communities. Precision decreased when the agent looked at locations further away from the center (fixed effect: $\beta = 0.48$, 95% HDI (- 0.54); range of community-level effects: $\beta_{min} = 0.58$, 95% HDI (0.51 - 0.66) to $\beta_{max} = 0.22$, 95% HDI (0.05 - 0.38). Visualization of the data showed the predicted u-shaped pattern in all communities (see Fig. 2B). These results indicate a universal cognitive process used by children in all communities. There is, however, an alternative way in which the u = shaped pattern might arise: it would also arise when participants ignored the agent’s gaze and instead always selected the middle of the screen (center bias). To rule out this alternative explanation, we directly compared three cognitive models that made different assumptions about how participants’ responses were generated: the focal vector-based gaze estimation model described above, a center-bias model where participants always select the center, and a random guessing model where participants select random locations. For every community, we found overwhelming support for the gaze estimation model (min $BF_{10} > 1000$ for comparisons with both alternative models). Taken together, children from all 17 communities processed gaze in similar ways.

What is left to explain are the marked community- and individual-level differences. In addition to the gaze-following task, caregivers filled out a small questionnaire about children’s access to screen-based technology and the composition of their households. On an individual level, we found that across communities, children with access to touchscreen devices had higher levels of precision ($\beta = -0.17$, SE = 0.04, 95% CrI = -0.25 - -0.09). On a community level, we also saw that average performance was lowest in the communities in which touchscreen devices were the least frequent (Fig. 3). Thus, methodological factors likely explained the marked differences between communities.

On an individual level, other factors likely generated individual differences because there was substantial variation even in communities where almost all children had access to touchscreens. Previous work suggested that social cognition develops in social interaction

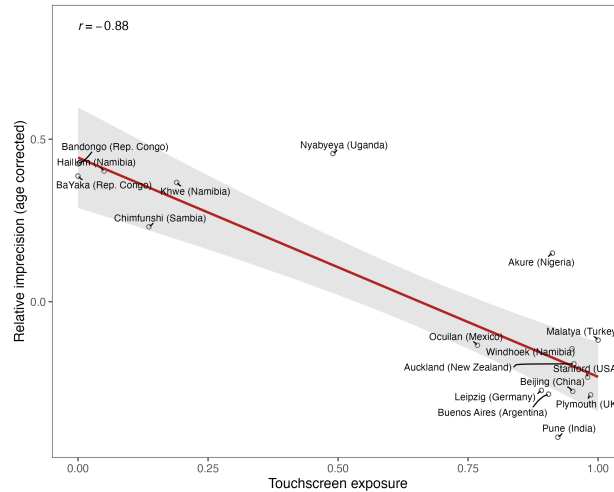


Figure 3. Correlation between performance in the gaze-following task and the exposure to touchscreens on a cultural setting-level. Relative imprecision accounts for differences in age between settings in that it corresponds to the random intercept estimate for each setting in a model predicting performance by age.

(21–23), in particular, with older siblings (24, 25). We found tentative support for this idea in the form of a small but consistent effect in the direction that participants living together with more children older than themselves (relative to other children from the same setting) had slightly higher levels of precision ($\beta = -0.02$, $SE = 0.01$, $95\% \text{ CrI} = -0.04 - 0.01$).

Following and understanding gaze is a foundational building block of human social cognition (8, 9). A substantial body of work has explored the developmental onset of gaze following in a few selected cultural settings (10–12, 15, 16, 26). The data reported here provides evidence that children all over the world process gaze in the same way. Key performance signatures of a model that sees gaze following as a form of social vector estimation were found in the data of all 17 distinct cultural communities. The cognitive processes underlying gaze following are thus likely to be rooted in humans’ evolved cognitive architecture, which is – presumably – later refined during social interaction (27–29). The phylogenetic roots of these processes might possibly lie much deeper as primates from a wide range of species follow gaze (30–33). Yet, if they also show the same processing signatures

has yet to be explored.

There are important limitations to this study. The methodological factors that influenced performance might have overshadowed individual and community-level differences that originate from other sources. Importantly, this does not affect our interpretation of the data as evidence for shared cognitive processing because the key processing signatures were present in all communities. The role that social interaction plays as a potential driver of development and source of individual differences is vastly under-explored in the current study. The small effect of the exposure to older children does not indicate which aspects of social interaction enhance gaze processing. Naturalistic data directly recording social interactions across communities would offer crucial information to close this explanatory gap. Recent work in the field of language acquisition has shown how technological innovations can be used to close this explanatory gap (34, 35).

The evidence presented here holds far-reaching implications. Our work pioneers a methodological approach that introduces solid individual-level measurement to the cross-cultural study of cognitive development. As such, it serves as a blueprint for future research on a broad spectrum of cognitive abilities. Most importantly, this study offers a much-needed empirical foundation for theories on human nature. The finding that children from diverse cultures deploy similar cognitive processes in interpreting gaze points to an evolutionary basis of basic social cognition, which is refined during development, presumably through social interaction.

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