

1 **Supplements for the manuscript ‘Variation in gaze following across the life span:**  
2 **A process-level perspective’**

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20 **Supplements for the manuscript ‘Variation in gaze following across the life span:**  
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22 **Study 1: Lifespan**

23 **Participants**

24 For further information on the participants in Study 1, please see Supplementary  
25 Table S1.

26 **Analysis**

27 *Model structures*

28 In the paper, we reported the following model structures: linear model:

29 `mean_imprecision ~ age_centered`; quadratic model in R: `mean_imprecision ~ 1 +`

30 `age_centered + I(age_centered^2)`; cubic model: `mean_imprecision ~ 1 +`

31 `age_centered + I(age_centered^2) + I(age_centered^3)`; Gaussian process model:

32 `mean_imprecision ~ gp(age_centered, k=50, c=5/4, scale=TRUE)`. Note the

33 additional parameters in the Gaussian process model. With the default settings, the

34 underlying Gaussian process maths would get solved exactly. By providing the arguments `k`

35 and `c`, we use an approximation process. The higher the value of `k`, the better the

36 approximation: we have used `k=50` for faster processing speed and better diagnostics. `brms`

37 suggests `5/4` as a value for `c`. Adding `scale=TRUE` is supposed to improve sampling speed

38 and convergence.

39 Originally, we fitted the models on a trial-by-trial basis with the following structure

40 in R: `performance ~ age + symmetricPosition + trialNr + (1 +`

41 `symmetricPosition + trialNr | subjID)`. However, the Gaussian Process model was

42 computationally heavy. Therefore, we simplified the model structure, aggregated data on a

43 subject level, and included only age as an effect. We then visually compared the model

44 predictions of the original and the simplified models with each other. As you can see in

45 Figure S1, results of the two models did not differ notably.

## 46 *Changepoint analysis*

47 In our Bayesian changepoint analysis, we restricted the model to a constant mean  
48 (i.e., a flat line with zero degree polynomial) within each segment, and to have minimally 10  
49 data points between two change points (i.e., corresponding to half of the data points we  
50 collected per adult decade) to avoid “overreactions” to individual outlying data points. In a  
51 supplementary analysis, we assessed how different parameter settings effected of our  
52 changepoint analysis. We changed the number of allowed change points, the minimum  
53 number of data points between change points, and the polynomial order. When the model  
54 had more explorative room, for example, by a greater number of change points, smaller  
55 minimum number of data points between change points, higher polynomial order, the model  
56 outputs showed more fine-grained change points. The exact location of the change points  
57 varied slightly. Overall, the interpretation stayed the same as the one we reported in the  
58 paper. While early childhood was characterized by much change, adults showed a relatively  
59 stable level of imprecision. There was a minor change in that elderly adults became slightly  
60 more imprecise again. If you are interested into the details, please have a look at the file  
61 `supplements_changepoint_parameters.html`, which you can find in the GitHub  
62 repository in the `stats` folder.

## 63 **Study 2: Computational cognitive model**

### 64 **Analysis**

#### 65 *Gaze model prediction*

66 Our gaze model predicts that TANGO trials vary in their difficulty, resulting in a  
67 U-shaped pattern: Participants’ imprecision should increase, the further out the target lands  
68 (towards the very left/right sides). Since the task is presented on a screen, there is a natural  
69 border towards one side. Imagine the target lands to the very right side. Participants’  
70 imprecision cannot click further right because the screen ends; all their uncertainty about the  
71 target location faces the inner, left-hand side now. Therefore, the predicted U-shaped  
72 pattern should decrease again towards the screen borders. For previous reliability analyses

73 (Prein et al., 2023), we had increased the trial number for an adult sample ( $N = 70$ ; each 30  
74 trials). Interestingly, here we found the expected shape in the data: the U-pattern decreased  
75 again towards the screen ends (Figure S2).

#### 76 **Inference parameter estimates per individual**

77 As can be seen in Figure S3 and Figure S4, the gaze model estimated the inference  
78 parameter for each individual. Across individuals, the inference parameter varied in the  
79 estimated magnitude and level of uncertainty. In general, estimates for more precise  
80 individuals (i.e., smaller inference parameter value) showed decreased levels of uncertainty.

## 81 **Simulations**

82 In the manuscript, we have described the gaze model and two alternatives: random  
83 guessing and a center bias. Here, we consider two more alternatives. Let us consider a model  
84 that assumes participants can infer the agent's focus without any noise (for example, by  
85 tracing the line of sight). A model like this would assume that participants follow gaze  
86 without any uncertainty. Therefore, no U curve would be predicted. Furthermore, let us  
87 assume another model in that participants still show no inferential noise, but they vary in  
88 their amount of motor noise, so how accurately participants then click at the corresponding  
89 location. A model like this would assume equal variance across the target locations, so we  
90 would not expect a U-shape here.

91 Please note that our random guessing model acts like 100% noise: the predictions of a  
92 participant's click on a trial level range uniformly from 0 to 1920 (the whole screen range).  
93 The mean comes down to the center, namely 960. However, as you can see in Figure S5A  
94 below, the U shape is weaker compared to the other models. Most importantly, note that  
95 this is only the case when you average across all the trials. When you look at the individual  
96 trial-by-trial level, the models are defined by different data-generating processes, and  
97 therefore, their predictions differ, too (Figure S5C). In our correlational plots (Figure SS5B),  
98 we see that the gaze model is clearly favored. This highlights the benefit that we gain  
99 through the modeling approach: even though a certain (U) pattern in the data could be  
100 elicited from several different models, we can disentangle which process is most likely causing  
101 this pattern.

## 102 **Study 3: Components of gaze understanding**

### 103 **Procedure Theory of mind battery**

104 For the Theory of Mind battery, we followed the standardized procedure as described  
105 below. The battery was administered in German and presented in the order as stated below.

106 **(1) Diverse Beliefs (Wellman & Liu, 2004)**

107 Material: Girl figure, leaf with garage & bush. Experimenter (E): “Look, this is  
108 Linda. Linda wants to find her cat. The cat is either hiding behind the bush or in the garage.  
109 Where do you think the cat is hiding? Behind the bush or in the garage?” (own-belief  
110 question). Child: “Garage”. E: “That’s a good idea! But Linda thinks her cat is behind the  
111 bush. She thinks her cat is behind the bush. So, where will Linda look for her cat? Behind  
112 the bush or in the garage?” (target question)

113 **(2) Knowledge Access (Wellman & Liu, 2004)**

114 Material: Yellow box, pig figure, female figure. E: “Look, here’s a box. What do you  
115 think is in the box?” [Child answers] [Box opens] E: “Let’s see... oh, there’s actually a pig  
116 inside!” [Box is closed] E: “What’s in the box?” [E pulls out figure] E: “That’s Polly. Polly  
117 has never looked in the box. So, does Polly know what’s in the box? (target question) Has  
118 Polly ever looked in the box? (control question)”

119 **(3) Contents False Belief (Wellman & Liu, 2004)**

120 Material: Smarties box, figure of a dog, figure of a boy. E: “Here is a Smartie box.  
121 What do you think is inside?” [Smartie box is opened] E: “Let’s see... there’s actually a dog  
122 inside!” [Smartie box is closed] E: “Okay, what’s in the Smartie box?” [Man’s figure is taken  
123 out] E: “This is Peter. Peter has never looked inside the Smartie box. What does Peter  
124 think is in the box? Smarties or a dog? (target question) Has Peter ever looked in the box?  
125 (memory question)”

126 **(4) Explicit False Belief (Wellman & Liu, 2004)**

127 Material: Figure of a boy, sheet with cupboard & rucksack on it. E: “This is Scott.  
128 Scott wants to find his gloves. The gloves could be in his backpack or they could be in the  
129 closet. In reality, they are in the backpack. But Scott thinks his gloves are in the closet. So,  
130 where will Scott look for his gloves? In the backpack or in the closet? (target question)  
131 Where are his gloves really? In the backpack or in the closet? (reality question)”

132 **(5) Perspective-Taking Level 2 version I (Flavell, Flavell, et al., 1981; Flavell,**  
 133 **Everett, et al., 1981)**

134 Material: Picture of turtle. E: “Look, here’s a picture of a turtle. I’ll put it between  
 135 us. What does the turtle look like to you, is it standing on its feet or lying on its back? (own  
 136 perspective question) And what does it look like to me? Does it stand on its feet for me or  
 137 does it lie on its back? (other perspective question).”

138 **(6) Perspective-Taking Level 2 version II (Flavell, Flavell, et al., 1981;**  
 139 **Flavell, Everett, et al., 1981)**

140 Material: Picture of worm between two pillows. E: “Now we have a worm lying  
 141 between two pillows. I’ll put it between us again. How does it look to you, is the worm lying  
 142 on the red or the blue cushion? (own perspective question) And what does it look like to me  
 143 - is the worm lying on the red or the blue cushion? (other perspective question)”

## 144 Analysis

### 145 *Animal vs. human faces*

146 In Study 1 and Study 2, we presented the TANGO (Prein et al., 2023) with animal  
 147 characters. For Study 3, we exchanged the animals with human faces, modelled in  
 148 appearance after the local population. We decided to do so in order to enhance the social  
 149 context of this task and to make it more comparable to the Theory of Mind task battery  
 150 (where there is live interaction with the experimenter). To ensure the change from animal to  
 151 human faces did not notably change children’s responses, we conducted an exploratory  
 152 analysis. We conducted a GLMM analysis with the following model structure in R: `click ~`  
 153 `age_scaled + stimuli + symmetric_position + trial_nr + (1 +`  
 154 `symmetric_position + trial_nr | subj_id)`; where `stimuli` denoted either human or  
 155 animal faces. The estimate for the fixed effect of `stimuli` was small and the 95% CrI  
 156 included zero:  $\beta = 0.16$ ; 95% CrI [-0.06; 0.37]). Therefore, we concluded that the animal  
 157 vs. human version of the TANGO did not differ substantially.

158 *Model comparisons*

159 To identify which (social-)cognitive components were needed to best explain the  
160 TANGO score, we compared GLMMs that predicted the mean imprecision in gaze  
161 understanding by age + the respective task score: imprecision in non-social vector  
162 estimation, the ToM aggregate score, and/ or the aggregate of the two perspective-taking  
163 tasks (subset of ToM battery). For example, the model notation in R: `tango_mean ~`  
164 `age_centered + magnet_scaled + perspective_scaled`). The model including the  
165 non-social vector estimation task (magnet) and the two perspective-taking tasks won, as  
166 indicated by the model comparison results shown in Supplementary Table S2.



**References**

167

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Supplementary Tables

**Table S1**

*Participants in Study 1. Age is given in years.*

Age group	n	Age mean	Age range	Age SD
3	19 (7 female)	3.62	3.04 - 3.99	0.31
4	17 (9 female)	4.45	4.05 - 4.91	0.30
5	22 (13 female)	5.56	5.08 - 5.99	0.31
6	24 (16 female)	6.50	6.1 - 6.99	0.28
7	39 (20 female)	7.48	7.04 - 7.95	0.25
8	41 (20 female)	8.46	8.03 - 8.98	0.27
9	56 (29 female)	9.46	9.01 - 9.96	0.28
10	35 (22 female)	10.49	10.01 - 11	0.28
11	54 (26 female)	11.43	11.01 - 11.96	0.28
12	43 (19 female)	12.41	12.01 - 12.99	0.30
13	42 (19 female)	13.50	13.09 - 13.99	0.27
14	20 (14 female)	14.37	14.05 - 14.98	0.23
15	21 (11 female)	15.56	15.05 - 15.98	0.30
16	19 (10 female)	16.51	16.17 - 16.97	0.24
17	19 (10 female)	17.53	17.01 - 17.95	0.28
18	2 (0 female)	18.00	18 - 18	0.00
19	5 (4 female)	19.00	19 - 19	0.00
20	40 (25 female)	23.02	20 - 29	2.77
30	40 (21 female)	34.42	30 - 39	3.00
40	40 (24 female)	44.17	40 - 49	2.92
50	40 (21 female)	54.38	50 - 59	3.04
60	40 (21 female)	63.73	60 - 69	2.56
70	40 (20 female)	72.75	70 - 79	2.44

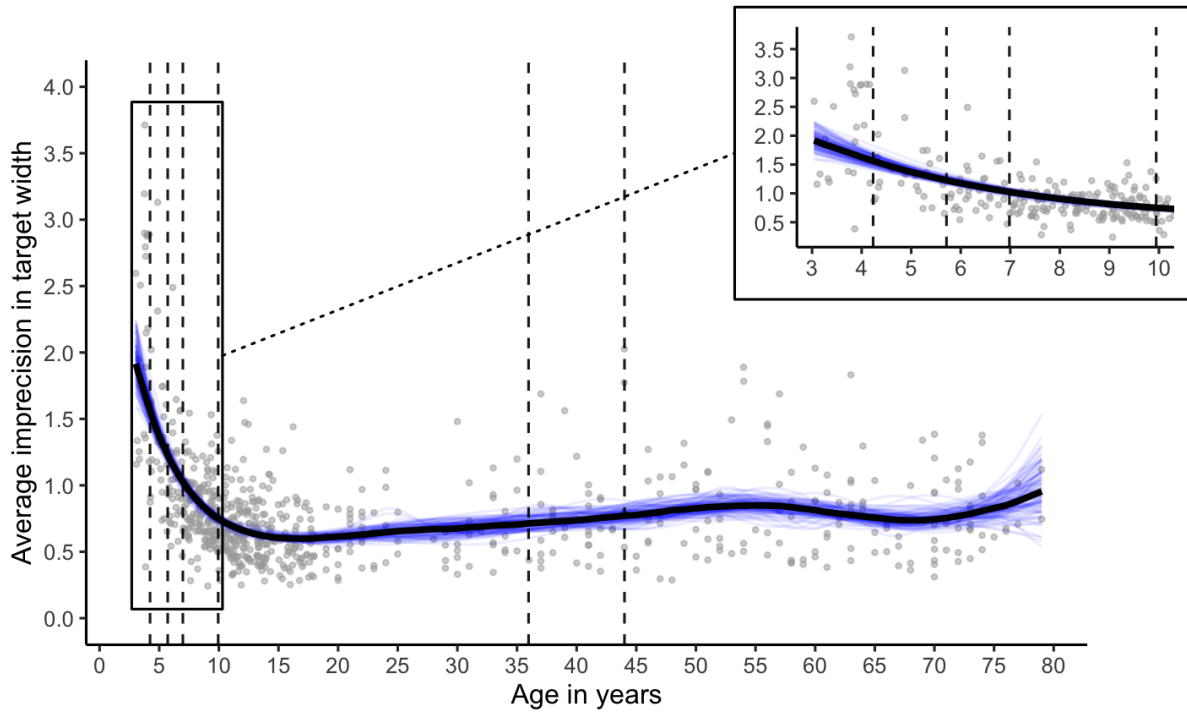
**Table S2**

*Model comparisons ToM tasks.*

Model	WAIC	SE_WAIC	Weight	ELPD_DIFF	SE_ELPD_DIFF
Magnet mean (scaled) + Perspective-taking aggregate (scaled)	200.83	16.16	0.92	0.00	0.00
Magnet mean (scaled)	206.51	16.92	0.05	-2.84	2.64
Magnet mean (scaled) + ToM aggregate (scaled)	208.51	16.79	0.02	-3.84	2.38
Perspective-taking aggregate (scaled)	212.21	15.42	0.00	-5.69	2.48
Null model with Age (scaled)	218.72	15.96	0.00	-8.95	3.35
ToM aggregate (scaled)	220.52	15.83	0.00	-9.85	3.09

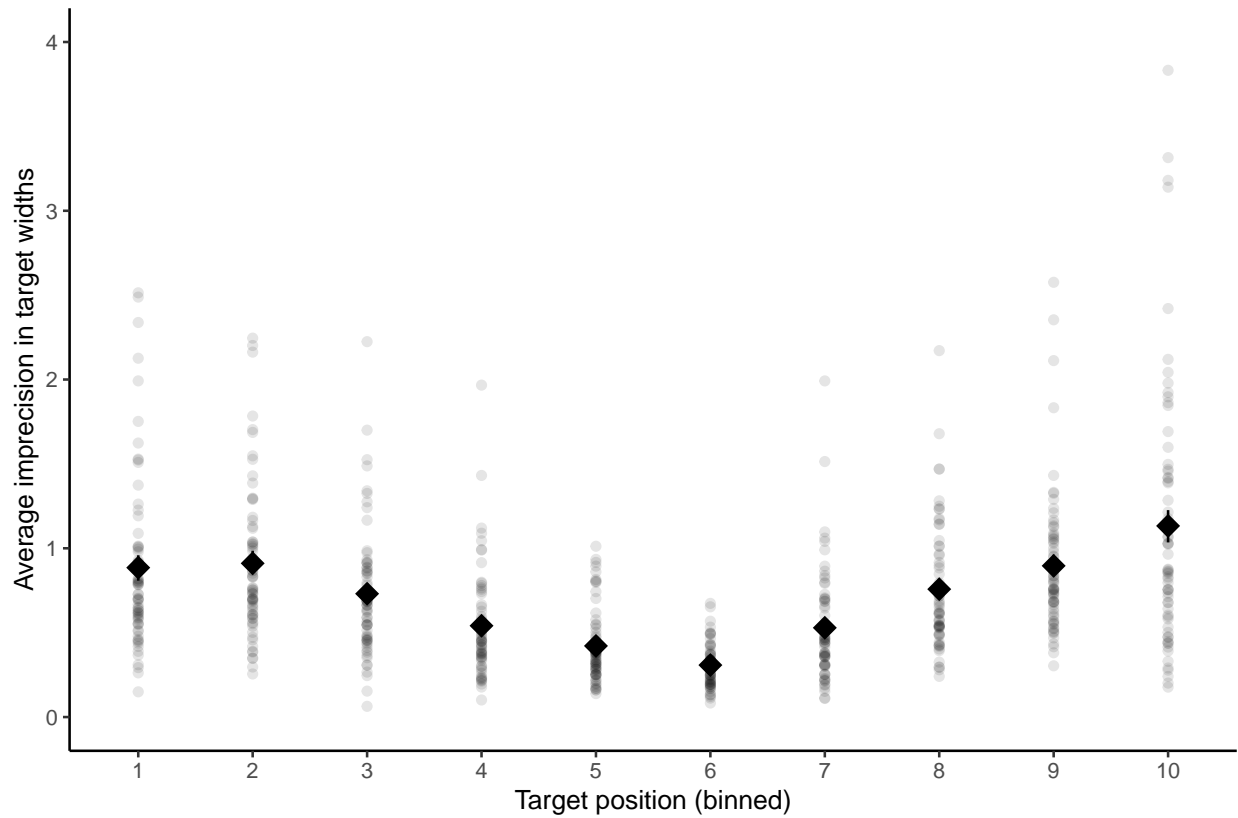
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Supplementary Figures



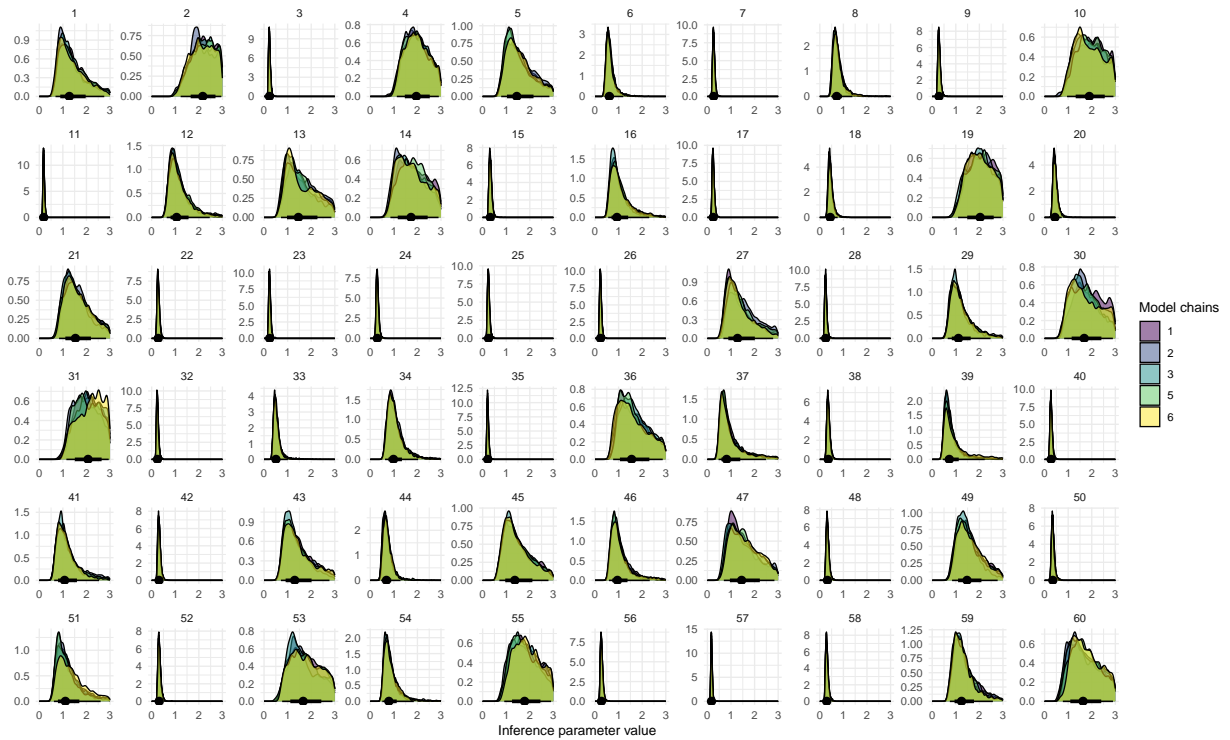
**Figure S1**

*Comparison between models on a trial- or subject-level. Grey dots show data of each trial. Solid lines show the mean predicted developmental trajectory for both models. Line color denotes model structure (yellow: trial-level; black: aggregated on subject-level).*

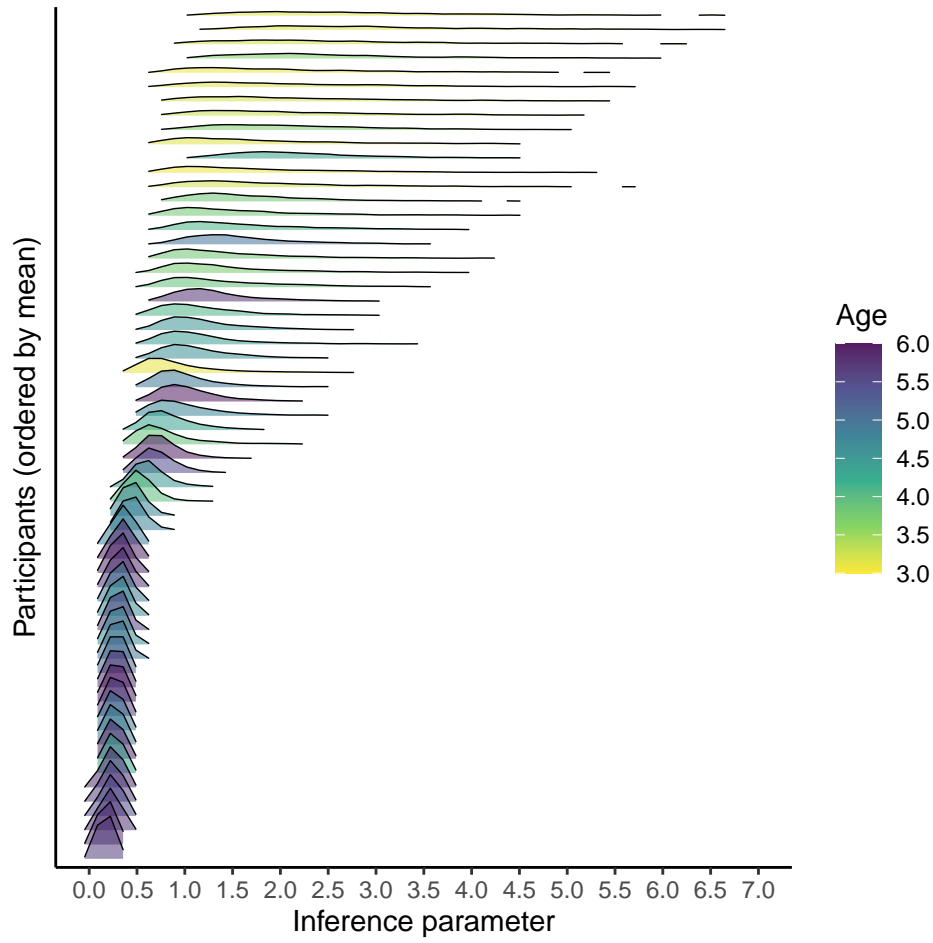


**Figure S2**

*Gaze funnel for adult sample with higher trial number (30 trials). The x-axis shows the target position, binned into 10 sections. The y-axis shows participants' imprecision in target widths.*

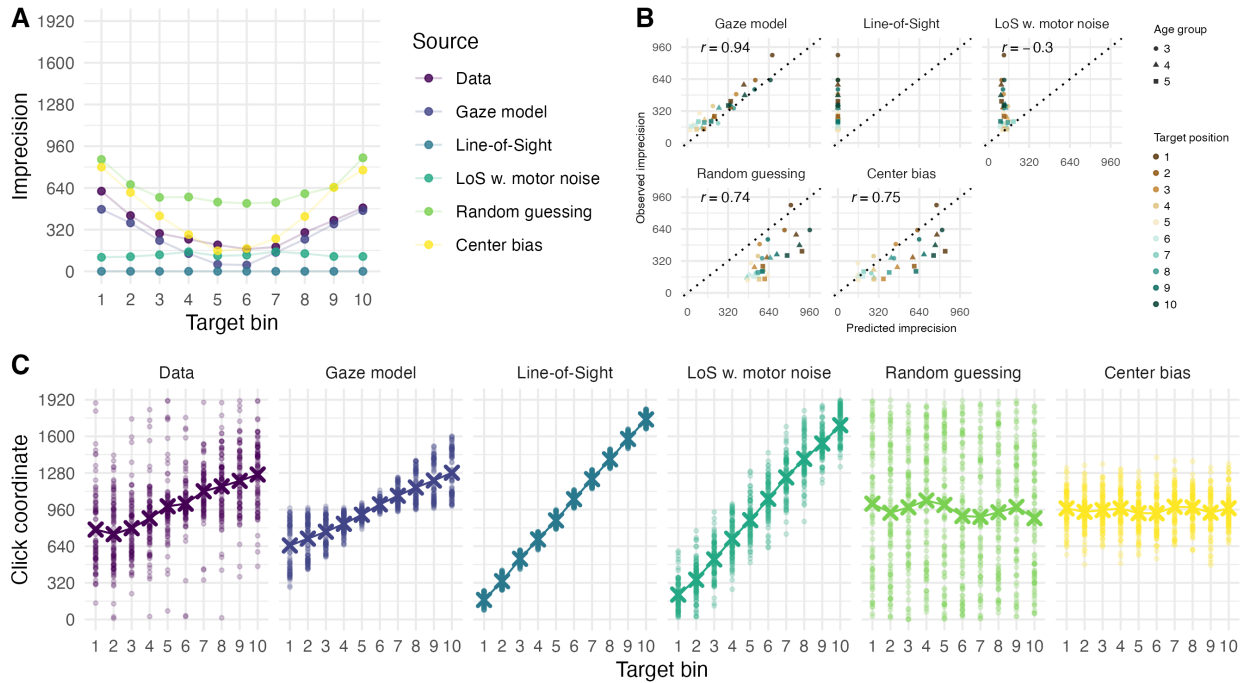


**Figure S3**  
*Gaze model estimates faceted by individual. Density curves show the distribution of the inference parameter by individual.*



**Figure S4**

*Gaze model estimates ordered by descending inference parameter value. Density curves show the distribution of the inference parameter across individuals.*



**Figure S5**

**Predictions of alternative models across target positions.** *A: Predicted U-pattern across target positions by the different models. Note that this is averaged across trials and displays imprecision, i.e., the absolute distance between the target center and the click. B: Correlation between the predicted imprecision and the observed imprecision, by target position and age group. C: Predicted click coordinates (cf. not imprecision) by target position across the different models.*