1   Self-referencing prioritizes moral character on perceptual matching

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16   Abstract

17   Evidence for the prioritization of moral information in cognitive processes is mixed. We

18   examined this question using a series of eleven experiments where participants first learned

19   associations between moral characters and geometric shapes and then performed simple

20   speed tasks. In the first six experiments, we tested and validated prioritized responses to

21   good characters over bad and neutral characters. To pin down the processes that are

22   critical to the prioritization effects, in the remaining five experiments, we examined two

23   opposing hypotheses: the valence hypothesis suggests that a general positivity bias towards

24   all underpins the effects, while the self-binding account posits that self-referencing, rather

25   than other-referencing is the fundamental driver of the effects. The data support the latter.

26   Together, these results show a robust prioritization effect of good character through

27   self-referencing processes, indicating the innate connection between morality and oneself

28   and how humans use self-reference to explore the world and learn morality.

29   *Keywords:* Perceptual matching, self positivity bias, primacy of morality, Bayesian

30   hierarchical models

31   Word count: X

³²   Self-referencing prioritizes moral character on perceptual matching

³³   **Introduction**

³⁴   Morality is central to human life (Haidt & Kesebir, 2010). Thus, gathering

³⁵ information about morality efficiently and accurately is crucial for individuals to navigate

³⁶ the social world (Brambilla, Sacchi, Rusconi, & Goodwin, 2021). The importance of

³⁷ morality naturally leads to the hypothesis that morality-related information is prioritized in

³⁸ information processing, especially when attentional resources are limited. This hypothesis

³⁹ is plausible because a large volume of studies has reported that valuable stimuli are

⁴⁰ prioritized, e.g., threatening stimuli (e.g., Ohman, Lundqvist, & Esteves, 2001), rewards

⁴¹ (B. A. Anderson, Laurent, & Yantis, 2011; Sui & Humphreys, 2015a), or self-related stimuli

⁴² (Sui & Rotshtein, 2019). Consistent with this hypothesis, a few studies reported a

⁴³ prioritization effect of negative moral information in visual processing: negative moral trait

⁴⁴ words (Fiske, 1980; Gantman & Van Bavel, 2014; Ybarra, Chan, & Park, 2001) and faces

⁴⁵ associated with bad behaviors (E. Anderson, Siegel, Bliss-Moreau, & Barrett, 2011;

⁴⁶ Eiserbeck & Abdel Rahman, 2020) attracted more attention and were responded faster.

⁴⁷   However, evidence for this negative moral bias effect is mixed. First, the opposite

⁴⁸ effect was also reported. For example, Shore and Heerey (2013) found that faces with

⁴⁹ positive interaction in a trust game were prioritized in the pre-attentive process. Also,

⁵⁰ Abele and Bruckmueller found faster responses to moral words were not moderated by

⁵¹ valence (Abele & Bruckmüller, 2011). Second, the robustness of the negative moral bias

⁵² effect is questioned, a direct replication study failed to support the conclusion that faces

⁵³ associated with bad social behaviors dominate visual awareness (eg., Stein, Grubb,

⁵⁴ Bertrand, Suh, & Verosky, 2017). Third, the prioritization effect of morality might be

⁵⁵ confounded with other factors, such as the priming effect (Firestone & Scholl, 2015, 2016b;

⁵⁶ Jussim, Crawford, Anglin, Stevens, & Duarte, 2016) or differences between lexical

⁵⁷ characteristics (Larsen, Mercer, & Balota, 2006). As a result, while the importance of

morality is widely recognized and there is initial evidence for a negative moral bias,

whether moral information is prioritized in perceptual processing is still an open question.

Here, we conducted a series of well-controlled experiments to examine the

prioritization effect of morality and its potential mechanisms. To eliminate the priming

effect and other potential confounding factors, we employed a task where participants first

acquired moral meanings of geometric shapes during the instruction phase and then

performed a simple perceptual matching task. The instruction-based associative learning

task is based on the fact that humans can rapidly learn based on verbal instructions (e.g.,

Cole, Braver, & Meiran, 2017). This instruction-based associative learning task is widely

used in aversive learning, value-based learning, and other tasks (Atlas, 2023; Cole et al.,

2017; Deltomme, Mertens, Tibboel, & Braem, 2018). Unlike previous studies relies on faces

or words (e.g., Bortolon & Raffard, 2018; Yaoi, Osaka, & Osaka, 2021), stimuli in the

current study are geometric shapes, whose moral meanings were acquired right before the

perceptual matching task. By counter-balancing associations between shapes and valence

of moral characters across different participants, we controlled the effect of these shapes on

the matching task. Also, in the matching task, we repeatedly present a few pairs of shapes

and labels to participants, the results can not be explained by semantic priming

(Unkelbach, Alves, & Koch, 2020), which is the center of the debate on previous results

(Firestone & Scholl, 2015, 2016a; Gantman & Bavel, 2015, 2016; Jussim et al., 2016).

Finally, we conducted a series of control experiments and established that moral content,

rather than other factors such as familiarity of stimuli, drove the prioritization effects.

To pin down the factors that are central to the prioritization effects, two competing

hypotheses were examined. One is the valence-based account, suggesting that a general

positivity bias towards all underpins the prioritization effects. In fact, the account has been

applied to explain not only positivity biases but also negativity biases. For example, the

negative bias toward moral information was explained by a threat detection mechanism

which might be general for all negative information (e.g., Fiske, 1980). The positive bias

toward moral information, on the other hand, was explained by the positive valence of the stimuli because the stimuli imply potential benefits (Shore & Heerey, 2013). However, these explanations often ignore the fact that valence is subjective *per se* (Juechems & Summerfield, 2019). That is, being related to a person is the premise of a stimulus or outcome being of value to the person. The subjective value is "a broader concept that refers to the personal significance or importance that a person assigns to a particular stimulus or outcome" and when the outcome is affective or emotional, researchers refer to it as "valence", i.e., positive or negative (Carruthers, 2021). The subjectivity of valence leads to an alternative explanation: self-binding account (Sui & Humphreys, 2015b). The self-binding account suggests that merely associating with the self can prioritize stimuli in perception, attention, working memory, and long-term memory (Sui & Humphreys, 2015b; Sui & Rotshtein, 2019), especially for positive information (Hu, Lan, Macrae, & Sui, 2020). According to the self-binding account, the prioritization of good character is a result of spontaneous self-referencing.

To test the valence account and self-binding account in the prioritization effect of good character, we manipulated self-relevance and instructed participants on which moral character is self-referencing and which is not. We then tested whether the prioritization of moral character is by valence or by the associations between self-relevance and moral valence. The results revealed that the prioritization effect only occurred when shapes of good characters referred to the self of participants. We confirmed these results in the subsequent experiments, where shapes of good characters did not explicitly refer to the self or others but were merely presented with labels of the self or others. Together, these data revealed a mutual facilitation effect of good character and the self, suggesting a spontaneous self-referential process as a novel mechanism underlying the prioritization of good character in perceptual matching.

Disclosures

110   We reported all the measurements, analyses, and results in all the experiments in the

111 current study. Participants whose overall accuracy was lower than 60% were excluded from

112 analyses. Also, accurate responses with less than 200ms reaction times were excluded from

113 the analysis. These excluded data can be found in the shared raw data files (see

114 https://doi.org/10.5281/zenodo.8031086). This manuscript is prepared with r pakcage

115 `papaja` (Aust & Barth, 2022).

117   All the experiments reported were not pre-registered. Most experiments (1a ~ 4b,

118 except experiment 3b) reported in the current study were first finished between 2013 to

119 2016 at Tsinghua University, Beijing, China. Participants in these experiments were

120 recruited from the local community. To increase the sample size of experiments to 50 or

121 more (Simmons, Nelson, & Simonsohn, 2013), we recruited additional participants from

122 Wenzhou University, Wenzhou, China, in 2017 for experiments 1a, 1b, 4a, and 4b.

123 Experiment 3b was finished at Wenzhou University in 2017 (See Table 1 for an overview of

124 these experiments).

125   All participants received informed consent and were compensated for their time.

126 These experiments were approved by the ethics board in the Department of Psychology,

127 Tsinghua University.

General methods

Design and Procedure

130   This series of experiments used the perceptual matching paradigm (or self-tagging

131 paradigm, see Sui, He, and Humphreys (2012)), in which participants first learned the

132 associations between geometric shapes and labels of different moral characters (e.g., in the

133 first three studies, the triangle, square, and circle for shapes and Chinese words for "good

person", "neutral person", and "bad person", respectively). The associations of shapes and labels were counterbalanced across participants. The paradigm consists of a brief learning stage and a test stage. During the learning stage, participants were instructed about the association between shapes and labels. Participants started the test stage with a practice phase to familiarize themselves with the task, in which they viewed one of the shapes above the fixation while one of the labels below the fixation and judged whether the shape and the label matched the association they learned. If the overall accuracy reached 60% or higher at the end of the practicing session, participants proceeded to the experimental task of the test stage. Otherwise, they finished another practices sessions until the overall accuracy was equal to or greater than 60%. The experimental task shared the same trial structure as in the practice.

Experiments 1a, 1b, 1c, 2, 5, and 6a were designed to explore and confirm the effect of moral character on perceptual matching. All these experiments shared a 2 (matching: match vs. mismatch) by 3 (moral character: good vs. neutral vs. bad person) within-subject design. Experiment 1a was the first one of the whole series of studies, which aimed to examine the prioritization of moral character and found that shapes associated with good character were prioritized. Experiments 1b, 1c, and 2 were to confirm that it is the moral character that caused the effect. More specifically, experiment 1b used different Chinese words as labels to test whether the effect was contaminated by familiarity. Experiment 1c manipulated the moral character indirectly: participants first learned to associate different moral behaviors with different Chinese names, after remembering the association, they then associated the names with different shapes and finished the perceptual matching task. Experiment 2 further tested whether the way we presented the stimuli influenced the prioritization of moral character, by sequentially presenting labels and shapes instead of simultaneous presentation. Note that a few participants in Experiment 2 also participated in Experiment 1a because we originally planned a cross-task comparison. Experiment 5 was designed to compare the prioritization of good

character with other important social values (aesthetics and emotion). All social values

had three levels, positive, neutral, and negative, and were associated with different shapes.

Participants finished the associative learning task for different social values in different

blocks, and the order of the social values was counterbalanced. Only the data from moral

character blocks, which shared the design of experiment 1a, were reported here.

Experiment 6a, which shared the same design as Experiment 2, was an EEG experiment

aimed at exploring the neural mechanism of the prioritization of good character. Only

behavioral results of Experiment 6a were reported here.

Experiments 3a, 3b, and 6b were designed to test whether the prioritization of good

character can be explained by the valence account or by the self-binding account. For this

purpose, we included self-reference as another within-subject variable. For example,

Experiment 3a extended Experiment 1a into a 2 (matching: match vs.mismatch) by 2

(reference: self vs. other) by 3 (moral character: good vs. neutral vs. bad) within-subject

design. Thus, in Experiment 3a, there were six conditions (good-self, neutral-self, bad-self,

good-other, neutral-other, and bad-other) and six shapes (triangle, square, circle, diamond,

pentagon, and trapezoids). Experiment 6b was an EEG experiment based on Experiment

3a but presented the label and shape sequentially. Because of the relatively high working

memory load (six label-shape pairs), participants finished Experiment 6b in two days. On

the first day, participants completed the perceptual matching task as a practice, and on the

second day, they finished the task again while the EEG signals were recorded. We only

focus on the first day's data here. Experiment 3b was designed to test whether the effect

found in Experiments 3a and 6b is robust if we separately present the self-referencing trials

and other-referencing trials. That is, participants finished two types of blocks: in the

self-referencing blocks, they only made matching judgments to shape-label pairs that

related to the self (i.e., shapes and labels of good-self, neutral-self, and bad-self), in the

other-referencing blocks, they only responded to shape-label pairs that related to the other

(i.e., shapes and labels of good-other, neutral-other, and bad-other).

188   Experiments 4a and 4b were designed to test whether the self and the good character

189 bind spontaneously. In Experiment 4a, participants were instructed to learn the association

190 between two shapes (circle and square) with two labels (self vs. other) in the learning

191 stage. In the test stage, they were instructed only respond to the shape and label during

192 the test stage. However, we presented the labels of different moral characters in the shapes

193 and instructed participants to ignore these labels when making matching judgments. If the

194 self and good character bind together spontaneously, then the mere presence of good

195 character will facilitate the response to shapes associated with the self. In the Experiment

196 4b, we reversed the role of self and moral character in the task: Participants learned

197 associations between three moral labels (good-person, neutral-person, and bad-person) and

198 three shapes (circle, square, and triangle) and made matching judgments about the shape

199 and label of moral character, while words related to identity, "self" or "other", were

200 presented within the shapes. As in Experiment 4a, participants were told to ignore the

201 words inside the shape during the perceptual matching task. In the same vein, if the self

202 and good character bind together spontaneously, then the mere presence of the self will

203 facilitate the response to shapes associated with good character.

204 **Stimuli and Materials**

205   We used E-prime 2.0 for presenting stimuli and collecting behavioral responses. Data

206 were collected from two universities located in two different cities in China. Participants

207 recruited from Tsinghua University, Beijing, finished the experiment individually in a

208 dim-lighted chamber. Stimuli were presented on 22-inch CRT monitors and participants

209 rested their chins on a brace to fix the distance between their eyes and the screen around

210 60 cm. The visual angle of geometric shapes was about $3.7° \times 3.7°$, the fixation cross is of

211 $0.8° \times 0.8°$ visual angle at the center of the screen. The words were of $3.6° \times 1.6°$ visual

212 angle. The distance between the center of shapes or images of labels and the fixation cross

213 was of $3.5°$ visual angle. Participants from Wenzhou University, Wenzhou, finished the

²¹⁴ experiment in a group consisting of 3 ∼ 12 participants in a dim-lighted testing room. They

²¹⁵ were instructed to complete the whole experiment independently. Also, they were told to

²¹⁶ start the experiment at the same time so that the distraction between participants was

²¹⁷ minimized. The stimuli were presented on 19-inch CRT monitors with the same set of

²¹⁸ parameters in E-prime 2.0 as in Tsinghua University, however, the visual angles could not

²¹⁹ be controlled because participants' chins were not fixed.

²²⁰ In most of these experiments, participants were also asked to fill out questionnaires

²²¹ following the behavioral tasks. All the questionnaire data were open (see, dataset 4 in Liu

²²² et al., 2020). See Table 1 for a summary of information about all the experiments.

Data analysis

²²⁴ We used the `tidyverse` of r (see script `Load_save_data.r`) to preprocess the data.

²²⁵ The data from all experiments were then analyzed using Bayesian hierarchical models.

²²⁶ We used the Bayesian hierarchical model (BHM, or Bayesian generalized linear mixed

²²⁷ models, Bayesian multilevel models) to model the reaction time and accuracy data because

²²⁸ BHM provided three advantages over the classic NHST approach (repeated measure

²²⁹ ANOVA or *t*-tests). First, BHM estimates the posterior distributions of parameters for

²³⁰ statistical inference, therefore providing uncertainty in estimation (Rouder & Lu, 2005).

²³¹ Second, BHM, where generalized linear mixed models could be easily implemented, can use

²³² distributions that fit the data, instead of using the normal distribution for all data. Using

²³³ appropriate distributions for the data will avoid misleading results and provide a better

²³⁴ fitting of the data. For example, Reaction times are not normally distributed but are often

²³⁵ right skewed, and the linear assumption in ANOVAs is not satisfied (Rousselet & Wilcox,

²³⁶ 2020). Third, BHM provides a unified framework to analyze data from different levels and

²³⁷ different sources, avoiding information loss when we need to combine data from different

²³⁸ experiments.

Table 1

Information about all experiments.

ExpID	Time	Location	N	n.of.trials	Self.ref	Stim.for.Morality	Presenting.order
Exp_1a_1	2014-04	Beijing	38 (35)	60	NA	words	Simultaneously
Exp_1a_2	2017-04	Wenzhou	18 (16)	120	NA	words	Simultaneously
Exp_1b_1	2014-10	Beijing	39 (27)	60	NA	words	Simultaneously
Exp_1b_2	2017-04	Wenzhou	33 (25)	120	NA	words	Simultaneously
Exp_1c	2014-10	Beijing	23 (23)	60	NA	descriptions	Simultaneously
Exp_2	2014-05	Beijing	35 (34)	60	NA	words	Sequentially
Exp_3a	2014-11	Beijing	38 (35)	60	explicit	words	Simultaneously
Exp_3b	2017-04	Wenzhou	61 (56)	60	explicit	words	Simultaneously
Exp_4a_1	2015-06	Beijing	32 (29)	30	implicit	words	Simultaneously
Exp_4a_2	2017-04	Wenzhou	32 (30)	60	implicit	words	Simultaneously
Exp_4b_1	2015-10	Beijing	34 (32)	60	implicit	words	Simultaneously
Exp_4b_2	2017-04	Wenzhou	19 (13)	60	implicit	words	Simultaneously
Exp_5	2016-01	Beijing	43 (38)	60	NA	words	Simultaneously
Exp_6a	2014-12	Beijing	24 (24)	180	NA	words	Sequentially
Exp_6b	2016-01	Beijing	23 (22)	90	explicit	words	Sequentially

Note. Stim.for.Morality = How moral character was manipulated; Presenting.order = How shapes & labels were presented. Number in () for N is number of participants are included in the analysis. In the current analysis, we only remain participants' data when they participate the experiment for the first time.

239   We used the `r` package `BRMs` (Bürkner, 2017), which used Stan (Carpenter et al.,

240 2017) as the back-end, for the BHM analyses. We estimated the overall effect across

241 experiments that shared the same experimental design using one model, instead of a

242 two-step approach that was adopted in mini-meta-analysis (e.g., Goh, Hall, & Rosenthal,

243 2016). More specifically, a three-level model was used to estimate the overall effect of

244 prioritization of good character, which included data from five experiments: 1a, 1b, 1c, 2,

245 5, and 6a. Similarly, a three-level HBM model is used for experiments 3a, 3b, and 6b.

246 Results of individual experiments can be found in the supplementary results. For

247 experiments 4a and 4b, which tested the implicit interaction between the self and good

248 character, we used HBM for each experiment separately.

249   For questionnaire data, we only reported the subjective distance between different

250 persons or moral characters in the supplementary results and did not analyze other

251 questionnaire data in the present study, which were described in (Liu et al., 2020).

252   **Response data.**   We followed previous studies (Hu et al., 2020; Sui et al., 2012)

253 and used the signal detection theory approach to analyze the response accuracy. More

254 specifically, the match trials are treated as signals and non-match trials are noise. The

255 sensitivity and criterion of signal detection theory are modeled through BHM (Rouder &

256 Lu, 2005).

257   We used the Bernoulli distribution for the signal detection theory. The probability

258 that the jth subject responded "match" ($y_{ij} = 1$) at the ith trial p_{ij} is distributed as a

259 Bernoulli distribution with parameter p_{ij}:

$$y_{ij} \sim Bernoulli(p_{ij})$$

260 The reparameterized value of p_{ij} is a linear regression of the independent variables:

$$\Phi(p_{ij}) = 0 + \beta_{0j} Valence_{ij} + \beta_{1j} IsMatch_{ij} * Valence_{ij}$$

261 where the probits (z-scores; Φ, "Phi") of ps is used for the regression.

The participant-specific intercepts ($\beta_0 = -zFAR$) and slopes ($\beta_1 = d'$) are described by multivariate normal with means and a covariance matrix for the parameters.

$$\begin{bmatrix} \beta_{0j} \\ \beta_{1j} \end{bmatrix} \sim N(\begin{bmatrix} \theta_0 \\ \theta_1 \end{bmatrix}, \sum)$$

We used the following formula for Experiments 1a, 1b, 1c, 2, 5, and 6a, which have a 2 (matching: match vs. mismatch) by 3 (moral character: good vs. neutral vs. bad) within-subject design:

```
saymatch ~ 0 + Valence + Valence:ismatch + (0 + Valence +
Valence:ismatch | Subject) + (0 + Valence + Valence:ismatch |
ExpID_new:Subject), family = bernoulli(link="probit")
```

in which the `saymatch` is the response data whether participants pressed the key corresponding to "match", `mismatch` is the independent variable of matching, `Valence` is the independent variable of moral character, `Subject` is the index of participants, and `Exp_ID_new` is the index of different experiments. Note that we distinguished data collected from two universities.

For experiments 3a, 3b, and 6b, an additional variable, i.e., reference (self vs. other), was included in the formula:

```
saymatch ~ 0 + ID:Valence + ID:Valence:ismatch + (0 + ID:Valence +
ID:Valence:ismatch | Subject) + (0 + ID:Valence + ID:Valence:ismatch |
ExpID_new:Subject), family = bernoulli(link="probit")
```

in which the `ID` is the independent variable "reference", which means whether the stimulus was self-referencing or other-referencing.

Reaction times. We used log-normal distribution to model the RT data (see https://lindeloev.github.io/shiny-rt/#34_(shifted)_log-normal). This means we need to estimate the posterior of two parameters: μ, and σ. μ is the mean of the `logNormal` distribution, and σ is the disperse of the distribution.

The reaction time of the jth subject on ith trial, y_{ij}, is log-normal distributed:

$$log(y_{ij}) \sim N(\mu_j, \sigma_j)$$

The parameter μ_j is a linear regression of the independent variables:

$$\mu_j = \beta_{0j} + \beta_{1j} * IsMatch_{ij} * Valence_{ij}$$

and the parameter σ_j does not vary with independent variables:

$$\sigma_j \sim HalfNormal()$$

The participant-specific intercepts (β_{0j}) and slopes (β_{1j}) are described by multivariate normal with means and a covariance matrix for the parameters.

$$\begin{bmatrix} \beta_{0j} \\ \beta_{1j} \end{bmatrix} \sim N(\begin{bmatrix} \theta_0 \\ \theta_1 \end{bmatrix}, \sum)$$

The formula used for experiments 1a, 1b, 1c, 2, 5, and 6a, which have a 2 (matching: match vs. non-match) by 3 (moral character: good vs. neutral vs. bad) within-subject design, is as follows:

```
RT_sec ~ 1 + Valence*ismatch + (Valence*ismatch | Subject) +
(Valence*ismatch | ExpID_new:Subject), family = lognormal()
```

in which `RT_sec` is the reaction times data with the second as a unit. The other variables in this formula have the same meaning as the response data.

For experiments 3a, 3b, and 6b, which have a 2 by 2 by 3 within-subject design, the formula is as follows: `RT_sec ~ 1 + ID*Valence + (ID*Valence | Subject) + (ID*Valence | ExpID_new:Subject), family = lognormal()` .

Note that for experiments 3a, 3b, and 6b, the three-level model for reaction times only included the matched trials to avoid divergence when estimating the posterior of the parameters.

304   **Testing hypotheses.**   To test hypotheses, we used the Sequential Effect eXistence

305 and sIgnificance Testing (SEXIT) framework suggested by Makowski, Ben-Shachar, Chen,

306 and Lüdecke (2019). In this approach, we used the posterior distributions of model

307 parameters or other effects that can be derived from posterior distributions. The SEXIT

308 approach reports centrality, uncertainty, existence, significance, and size of the input

309 posterior, which is intuitive for making statistical inferences. We used `bayestestR` for

310 implementing this approach (Makowski, Ben-Shachar, & Lüdecke, 2019).

311   ***Prioritization of moral character.**   We tested whether moral characters are

312 prioritized by examining the population-level effects (also called fixed effect) of the

313 three-level Bayesian hierarchical model of Experiments 1a, 1b, 1c, 2, 5, and 6a. More

314 specifically, we calculated the differences between the posterior distributions of the

315 good/bad character and the neutral character and then tested these posterior distributions

316 with the SEXIT approach.

317   ***Modulation of self-relevance.**   We tested the modulation effect of the

318 self-referencing process by examining the interaction between moral character and the

319 self-referencing process for the three-level Bayesian hierarchical model of Experiments 3a,

320 3b, and 6b. More specifically, we tested two possible explanations for the prioritization of

321 good character: the valence effect alone or an interaction between the valence effect and

322 self-relevance. If the former is correct, then there will be no interaction between moral

323 character and self-relevance, i.e., the prioritization effect exhibits a similar pattern for both

324 self- and other-referencing conditions. Otherwise, there will be an interaction between the

325 two factors, i.e., the prioritization effect exhibits different patterns for self- and

326 other-referencing conditions. To test the interaction, we calculated the posterior

327 distribution of the difference of difference: $(good - neutral)_{self}$ vs. $(good - neutral)_{other}$.

328 We then tested the difference of difference with SEXIT approach.

329   ***Spontaneous binding between the self and good character.**   For data from

330 Experiments 4a and 4b, we further examined whether the self-referencing process is

331 spontaneous (i.e., whether the good character is spontaneously bound with the self). For

332 Experiment 4a, if there exists a spontaneous binding between self and good character,

333 there should be an interaction between moral character and self-relevance. More

334 specifically, we tested the posterior distributions of $good_{self} - neutral_{self}$ and

335 $good_{other} - neutral_{other}$, as well as the difference between these differences with the

336 SEXIT framework. For Experiment 4b, if there exists a spontaneous binding between

337 self-relevance and good character, then, there will be a self-other difference for some moral

338 character conditions but not for other moral character conditions. More specifically, we

339 tested the posteriors of $good_{self} - good_{other}$, $neutral_{self} - neutral_{other}$, and

340 $bad_{self} - bad_{other}$ as well as the difference between them with SEXIT framework.

Results

Prioritization of good character

343   To test whether moral characters are prioritized, we modeled data from Experiments

344 1a, 1b, 1c, 2, 5, and 6a with three-level Bayesian hierarchical models. All these experiments

345 shared similar designs, with a total sample size of 192. Note that for both experiments 1a

346 and 1b, two datasets were collected at different time points and locations, thus we treated

347 them as independent samples. Here we only reported the population-level results of

348 three-level Bayesian models, the results of each experiment can be found in supplementary

349 materials.

350   For the d prime, results from the Bayesian model revealed a robust effect of moral

351 character. Shapes associated with good characters ("good person", "kind person" or a

352 name associated with good behaviors) have higher sensitivity (median = 2.45, 95% HDI =

353 [2.24 2.72]) than shapes associated with neutral characters (median = 2.15, 95% HDI =

354 [1.92 2.45]), the difference ($median_{diff} = 0.31$, 95% HDI [0, 0.62]) has a 97.31% probability

355 of being positive (> 0), 94.91% of being significant (> 0.05). But we did not find a

difference between shapes associated with bad characters (median = 2.21, 95% HDI = [2.00

2.48]) and neutral character, the difference ($median_{diff}$ = 0.05, 95% HDI [-0.27, 0.38])

only has a 60.56% probability of being positive (> 0), 49.34% of being significant (> 0.05).

The results from reaction times also found a robust effect of moral character for both

match trials (see figure 1 C) and nonmatch trials (**see supplementary materials**). For

match trials, shapes associated with good characters were faster (median = 583 ms, 95%

HDI = [506 663]) than shapes associated with neutral characters (median = 626 ms, 95%

HDI = [547 710]), the effect ($median_{diff}$ = -44, 95% HDI [-67, -24]) has a 99.94%

probability of being negative (< 0), 99.94% of being significant (< -0.05). We also found

that RTs to shapes associated with bad characters (median = 643 ms, 95% HDI = [564

729]) were slower as compared to the neutral character, the effect ($median_{diff}$ = 17, 95%

HDI [-6, 36]) has a 93.58% probability of being positive (> 0), 93.55% of being significant

(> 0.05).

For the nonmatch trials, we found a similar pattern but a much smaller effect size.

Shapes associated with good characters (median = 657 ms, 95% HDI = [571 739]) were

faster than shapes associated with neutral characters (median = 673 ms, 95% HDI = [589

761]), the difference ($median_{diff}$ = -18, 95% HDI [-27, -8]) has a 99.91% probability of

being negative (< 0), 99.91% of being significant (< -0.05). In contrast, the shapes

associated with bad characters (median = 678 ms, 95% HDI = [592 764]) were slower than

shapes associated with neutral characters, the effect ($median_{diff}$ = 5, 95% HDI [-3, 13])

has a 92.43% probability of being positive (> 0), 92.31% of being significant (> 0.05).

Modulation effect self-referential processing

To test the modulation effect of self-relevance, we also modeled data from three

experiments (3a, 3b, and 6b) with three-level Bayesian models. These three experiments

included 108 unique participants. We focused on the population-level effect of the
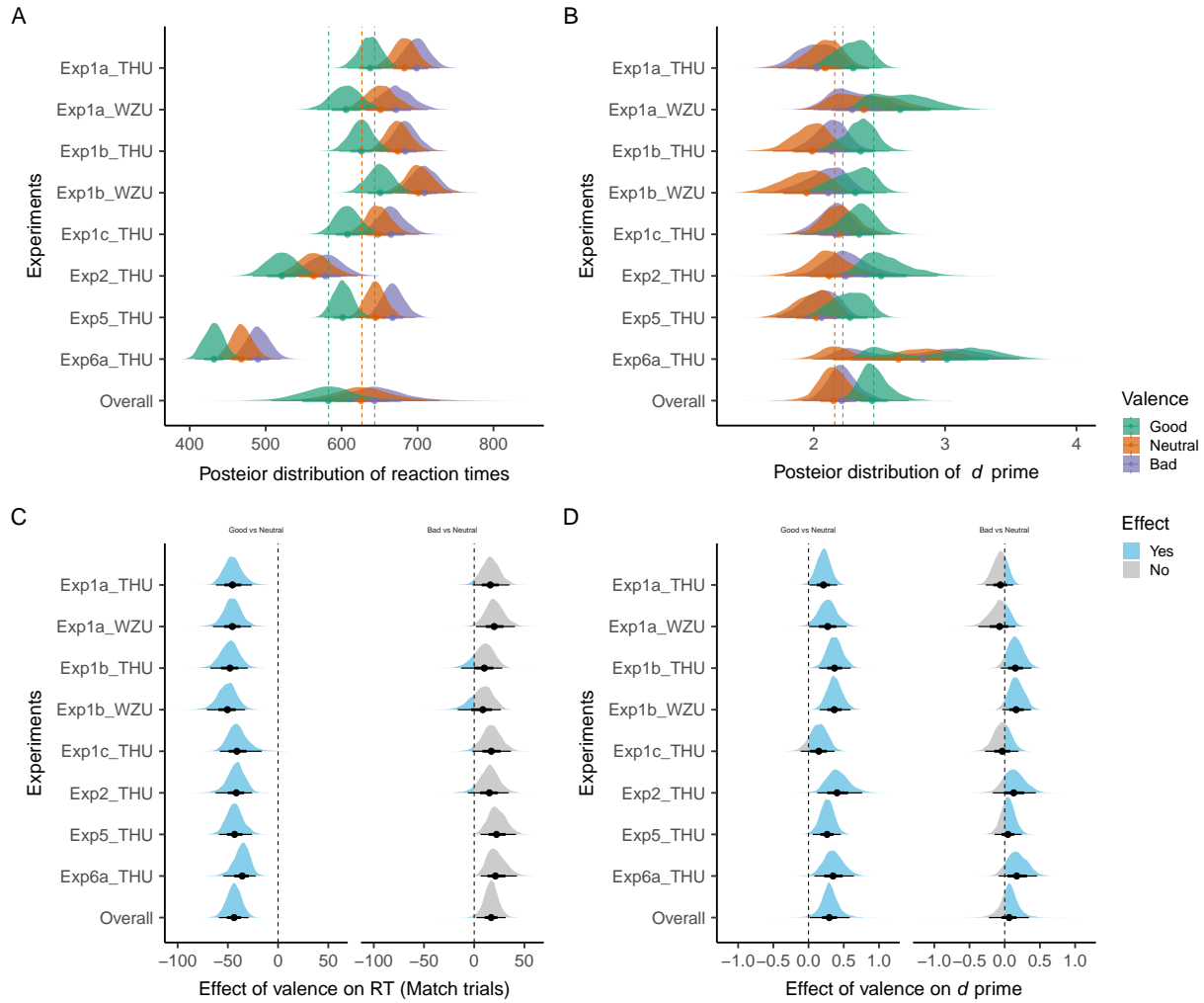
Figure 1. Effect of moral character on perceptual matching. (A) Experimental level (six experiments, with eight independent samples) and population level posterior distributions of RT under different matching conditions; (B) Experimental level and population level posterior distributions of *d*-prime under different conditions; (C) Experimental level and population level posterior distributions of the RT differences between conditions (left, Good vs. Neutral; right, Bad vs. Neutral); (D) Experimental level and population level posterior distributions of the *d*-prime differences between conditions (left, Good vs. Neutral; right, Bad vs. Neutral).

381 interaction between self-referential processing and moral valence. Also, we examined the

382 differences of differences, i.e., how the differences between good/bad characters and the

383 neutral character under the self-referencing conditions differ from that under

384 other-referencing conditions. The results of each experiment can be found in

385 supplementary materials.

386   For the d prime, we found an interaction between the moral valence and

387 self-relevance: the good-neutral differences are larger for the self-referencing condition than

388 for the other-referencing condition, the difference ($median_{diff} = 0.48$, 95% HDI [-0.62,

389 1.65]) has a 93.04% probability of being positive (> 0), 91.92% of being significant ($>$

390 0.05). However, the bad-neutral differences ($median_{diff} = 0.0087$, 95% HDI [-0.96, 1.00])

391 only have a 51.85% probability of being positive (> 0), 41.29% of being significant ($>$

392 0.05). Further analyses revealed that the prioritization effect of good character (as

393 compared to neutral) only appeared for self-referencing conditions but not

394 other-referencing conditions. The estimated d prime for good-self was greater than

395 neutral-self ($median_{diff} = 0.54$, 95% HDI [-0.30, 1.41]), with a 95.99% probability of being

396 positive (> 0), 95.36% of being significant (> 0.05). The differences between bad-self and

397 neutral-self, good-other and neutral-other, and bad-other and neutral-other are all centered

398 around zero (see Figure 2, B, D).

399   For the RTs of matched trials, we also found an interaction between moral valence

400 and self-relevance: the good-neutral differences were larger for the self- than the

401 other-referencing conditions ($median_{diff} = -148$, 95% HDI [-413, 73]) has a 96.05%

402 probability of being negative (< 0), 96.05% of being significant (< -0.05). However, this

403 pattern was much weaker for bad-neutral differences ($median_{diff} = -47$, 95% HDI [-280,

404 182]) has a 79.91% probability of being negative (< 0) and 79.88% of being significant ($<$

405 -0.05). Further analyses revealed a robust good-self prioritization effect as compared to

406 neutral-self ($median_{diff} = -59$, 95% HDI [-115, -22]) has a 98.87% probability of being

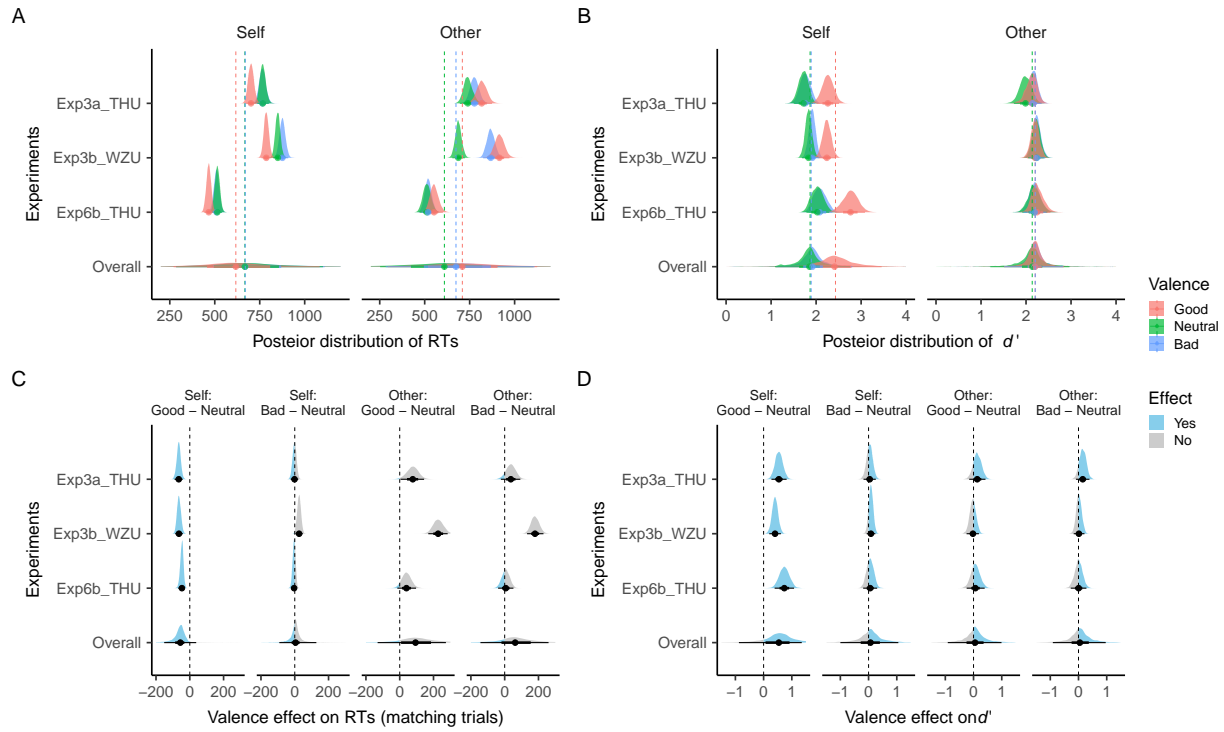407 negative (< 0) and 98.87% of being significant (< -0.05)) and good-other ($median_{diff} =$

Figure 2. Interaction between moral character and self-referential. (A) Experimental level (three experiments) and population level posterior distributions of RT under different conditions; (B) Experimental level and population level posterior distributions of *d*-prime under different conditions; (C) Experimental level and population level posterior distributions of the RT differences between conditions, from left to right: Good-self vs. Neutral-self, Bad-self vs. Neutral-self, Good-other vs. Neutral-other, Bad-other vs. Neutral-other; (D) Experimental level and population level posterior distributions of the *d*-prime differences between conditions, from left to right: Good-self vs. Neutral-self, Bad-self vs. Neutral-self, Good-other vs. Neutral-other, Bad-other vs. Neutral-other.

-109, 95% HDI [-227, -31]) has a 98.65% probability of being negative (< 0) and 98.65% of

being significant (< -0.05)) conditions. Similar to the results of *d'*, we found that

participants responded slower for both good character than for the neutral character when

they referred to others, $median_{diff} = 85.01$, 95% HDI [-112, 328]) has a 92.16%

probability of being positive (> 0) and 92.15% of being significant (> 0.05). A similar

pattern was also found for the bad character when referred to others: bad-other responded

slower than neutral-other, $median_{diff} = 44$, 95% HDI [-146, 268]) has an 80.03%

probability of being positive (> 0) and 79.99% of being significant (> 0.05). See Figure 2.

These results suggested that the prioritization of good character is not solely driven

by the valence of moral character. Instead, self-relevance modulated the prioritization of

good character: good character was prioritized only when it referred to the self. When the

moral character referred to others, responses to both good and bad characters were slowed

down.

The link between oneself and good character

Experiments 4a and 4b were designed to test whether the good character and the self

bind together spontaneously. Because these two experiments have different experimental

designs, we model their data separately.

In experiment 4a, where "self" vs. "other" were task-relevant and moral character

were task-irrelevant, we found the "self" conditions performed better than the "other"

conditions for both *d* prime and reaction times. This pattern is consistent with previous

studies (e.g., Sui et al. (2012)).

More importantly, we found evidence that task-irrelevant moral character also played

a role. For shapes associated with "self", *d'* was greater when shapes had a good character

inside (median = 2.82, 95% HDI [2.64 3.03]) than shapes that have neutral character

(median = 2.74, 95% HDI [2.58 2.94]), the difference (median = 0.08, 95% HDI [-0.10,

433   0.27]) has an 81.60% probability of being positive (> 0), 64.33% of being significant ($>$

434   0.05). For shapes associated with "other", the pattern reversed: d prime was smaller when

435   shapes had a good character inside (median = 1.87, 95% HDI [1.70 2.04]) than had neutral

436   (median = 1.96, 95% HDI [1.79 2.14]), the difference (median = -0.09, 95% HDI [-0.25,

437   0.05]) has an 89.03% probability of being negative (< 0), 71.38% of being significant ($<$

438   -0.05). The difference between these two effects (median = 0.18, 95% HDI [-0.06, 0.43]) has

439   a 92.88% probability of being positive (> 0), 85.08% being significant (> 0.05). See Figure

440   3.

441   A similar but more robust pattern was found for RTs in matched trials. For the "self"

442   condition, when a good character was presented inside the shapes, the RTs (median = 633,

443   95% HDI [614 654]) were faster than when a neutral character (median = 647, 95% HDI

444   [628 666]) was inside, the effect (median = -8, 95% HDI [-17, 2]) has a 94.55% probability

445   of being negative (< 0) and 94.50% of being significant (< -0.05). In contrast, when the

446   shapes referred to other, RTs for shapes with good character inside (median = 733, 95%

447   HDI [707 756]) were slower than those with neutral character inside (median = 713, 95%

448   HDI [691 734]), the effect (median = 12, 95% HDI [-4, 28]) has a 93.00% probability of

449   being positive (> 0) and 92.83% of being significant (> 0.05). The difference between these

450   effects (median = -19, 95% HDI [-43, 4]) has a 94.90% probability of being negative (< 0)

451   and 94.88% of being significant (< -0.05).

452   In experiment 4b, where moral characters were task-relevant and "self" vs "other"

453   were task-irrelevant, we found a main effect of moral character: performance for shapes

454   associated with good characters was better than other-related conditions on both d' and

455   reaction times. This pattern, again, shows a robust prioritization effect of good character.

456   Most importantly, we found evidence that task-irrelevant labels, "self" or "other",

457   also played a role. For shapes associated with good character, the d prime was greater

458   when shapes had a "self" inside than with "other" inside ($mean_{diff} = 0.14$, 95% HDI

[-0.05, 0.34]) has a 92.35% probability of being positive (> 0) and 81.80% of being significant (> 0.05). However, the difference did not occur when the target shape where associated with "neutral" ($mean_{diff} = 0.04$, 95% HDI [-0.13, 0.22]) and has a 67.20% probability of being positive (> 0) and 44.80% of being significant (> 0.05). Neither for the "bad" person condition: $mean_{diff} = 0.10$, 95% HDI [-0.16, 0.37]) has a 77.03% probability of being positive (> 0) and 64.62% of being significant (> 0.05).

The same trend appeared for the RT data. For shapes associated with good character, having a "self" inside shapes reduced the reaction times as compared to having an "other" inside the shapes ($mean_{diff} = -55$, 95% HDI [-75, -35]) has a 100% probability of being negative (< 0) and 100.00% of being significant (< -0.05). However, when the shapes were associated with the neutral character, having a "self" inside shapes increased the RTs: $mean_{diff} = 11$, 95% HDI [1, 21]) has a 98.20% probability of being positive (> 0) and 98.15% of being significant (> 0.05). While having "self" slightly increased the RT than having "other" inside the shapes for the bad character: $mean_{diff} = 5$, 95% HDI [-17, 27]) has a 69.45% probability of being positive (> 0) and 69.27% of being significant (> 0.05), See Figure 3.

Discussion

In this study, we investigated the primacy of morality in cognitive processes through systematically manipulating the factors that are central to the information processing of morality. First, we found a robust prioritization of good character in response times and *d'* scores for the shape-label matching tasks across experiments. Second, to pinpoint the underlying processes of the effect, the analyses revealed that a self-referencing process was the fundamental driver of these effects, consistent with the self-binding account; that is, when a stimulus refers to the self, activation of self-representation enhances the binding of external input with internal knowledge through which self-related information can be integrated and optimized. The valence account, on the other hand, which posits that the
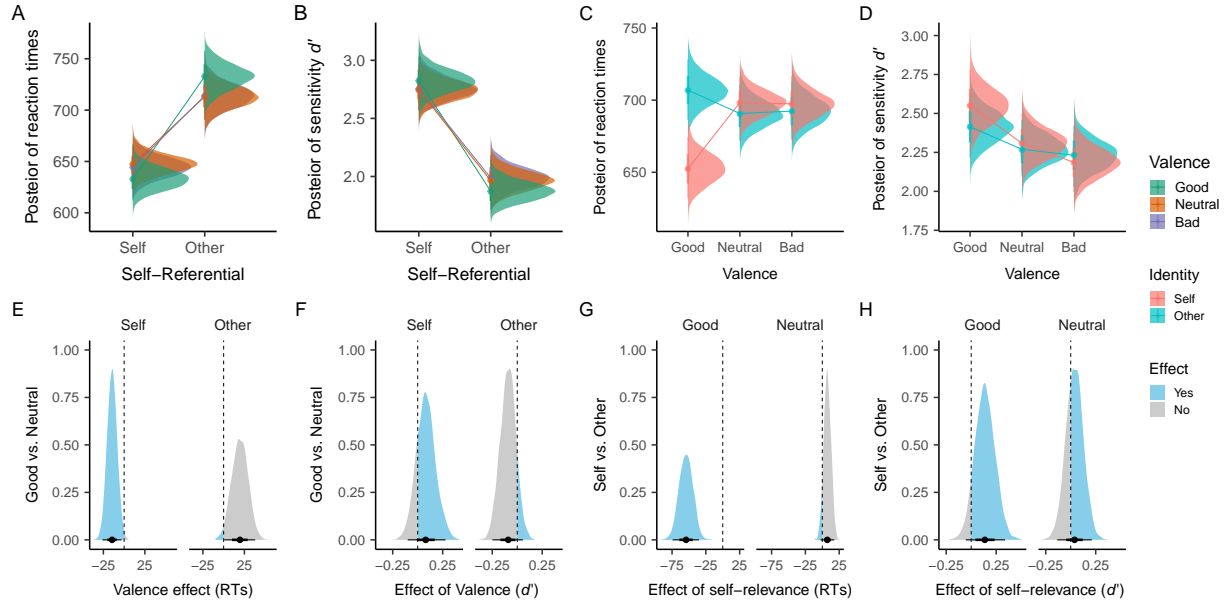
Figure 3. Implicit binding between self and good characters. (A) Posterior distributions of RT under different conditions of Experiment 4a; (B) Posterior distributions of *d*-prime under different conditions of Experiment 4a; (C) Posterior distributions of RT under different conditions of Experiment 4b; (D) Posterior distributions of *d*-prime under different conditions of Experiment 4b; (E) Posterior distributions of the RT differences between good character and neutral character when self (left) and other (right) were presented inside the shapes; (F) Posterior distributions of the *d*-prime differences between good character and neutral character when self (left) and other (right) were presented inside the shapes; (G) Posterior distributions of the RT differences between self- and other-referencing conditions when good character (left) and neutral character (right) were presented inside the shapes; (H) Posterior distributions of the *d*-prime differences between self- and other-referencing conditions when good character (left) and neutral character (right) were presented inside the shapes.

prioritization effect was derived from a general positivity bias towards all (self and others), was not supported by the findings. Importantly, the prioritization effects emerged regardless of whether the relationship between moral character and oneself was task relevant. Collectively, participants tend to attribute moral character to themselves rather than others, leading to prioritized responses to self perceived moral character in decision making.

The current study provided robust evidence for the prioritization of good character in perceptual decision-making. Though the primacy of morality has been argued in social psychology, whether morality is prioritized in information processing has been disputed. For instance, E. Anderson et al. (2011) reported that faces associated with bad social behavior capture attention more rapidly, but an independent team failed to replicate the effect (Stein et al., 2017). In another study, Gantman and Van Bavel (2014) found that moral words are more likely to be judged as words when it was presented subliminally. But this effect may be caused by semantic priming instead of morality (Firestone & Scholl, 2015; Jussim et al., 2016). To overcome this issue, we employed a shape-label matching task to eliminate the semantic priming effect for two reasons. First, associations between shapes and moral characters were acquired during the instruction phase, semantic priming from pre-existed knowledge was impossible (Lee, Martin, & Sui, 2021). Second, there were only a few pairs of stimuli that were used and each stimulus represented different conditions, making it impossible for priming between trials. Importantly, a series of control experiments (1b, 1c, and 2) excluded other confounding factors such as familiarity, presenting sequence, or words-based associations, suggesting that it was the moral content that drove the perceived prioritization of good character. These results are in line with a growing literature on the social and relational nature of perception (Hafri & Firestone, 2021; Xiao, Coppin, & Bavel, 2016).

The prioritization of good character found in the current study was incongruent with previous moral perception studies, which typically reported a negativity bias, i.e.,

information related to bad character is processed preferentially (E. Anderson et al., 2011; Eiserbeck & Abdel Rahman, 2020). This discrepancy may result from different task types employed: while in many moral perception studies, the participants were asked to detect the existence of a stimulus, the current task asked participants to judge the associations between a shape and a person. In other words, previous studies targeted the early stages of perception, while the current task focused more on perceptual decision-making, consistent with previous work (Sui & Humphreys, 2013). This discrepancy is consistent with the positivity bias in studies with emotional stimuli (Pool, Brosch, Delplanque, & Sander, 2016).

The current study expanded previous moral perception studies by testing a novel account that self-referencing processing is the critical driver of the effects. Our results revealed that prioritization of good character is modulated by self-relevance: good character was prioritized when it was referred to oneself. In contrast, good character information was not prioritized when it was referred to others. The modulation effect of self-relevance was amplified when the relationship between moral character and oneself was explicit, consistent with previous studies that only positive aspects of the self are prioritized (Hu et al., 2020). More importantly, the effect persisted even when the relationship between moral character and oneself was task-irrelevant, indicating an implicit self-referencing process emerged from presenting good character and self-related information in the same display. A possible explanation for this spontaneous self-referencing of good character is that the positive moral self-view is central to our identity (Freitas, Cikara, Grossmann, & Schlegel, 2017; Strohminger, Knobe, & Newman, 2017) and the motivation to maintain a moral self-view influences how we perceive (e.g., Ma & Han, 2010) and remember (e.g., Carlson, Maréchal, Oud, Fehr, & Crockett, 2020; Stanley, Henne, & De Brigard, 2019), with implications for the quality of life and wellbeing.

Although the results here revealed the prioritization of good character in perceptual decision-making, we did not claim that the motivation of a moral self-view *penetrates*

perception. The perceptual decision-making process involves processes more than just

encoding the sensory inputs (Scheller & Sui, 2022). To fully account for the nuance of

behavioral data and/or related data collected from other modules (e.g., Sui, He,

Golubickis, Svensson, & Neil Macrae, 2023), we may need computational models and an

integrative experimental approach (Almaatouq et al., 2022). For example, sequential

sampling models suggest that, when making a perceptual decision, the agent continuously

accumulates evidence until the amount of evidence passes a threshold, and then a decision

is made (Chuan-Peng et al., 2022; Forstmann, Ratcliff, & Wagenmakers, 2016; Ratcliff,

Smith, Brown, & McKoon, 2016). In these models, the evidence, or decision variable, can

accumulate from both sensory information but also memory (Shadlen & Shohamy, 2016).

Recently, applications of sequential sample models to perceptual matching tasks also

suggest that different processes may contribute to the prioritization effect of self

(Golubickis et al., 2017) or good self (Hu et al., 2020). Similarly, reinforcement learning

models revealed that the initial discrimination between self- and other-referencing learning

lies in the learning rate (Lockwood et al., 2018). These investigations suggest that

computational models are required to disentangle the cognitive processes underlying the

prioritization of good character.

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