- The good person is me: Spontaneous self-referential process prioritizes moral character in
- 2 perceptual matching
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Abstract

Moral character is central to social evaluation and moral judgment. However, whether moral character information is prioritized in perceptual decision-making was debated. Here 19 we investigated the effect of moral character on perceptual decision-making through an 20 associative learning task. Participants first learned associations between different geometric 21 shapes and moral characters and then performed a simple perceptual matching task. 22 Across five experiments (N = 192), we found a robust prioritization effect of good 23 character-related information, i.e., participants responded faster and more accurately to shapes that were associated with good characters than shapes associated with neutral or bad characters. We then examine whether the prioritization of good character was due to valence alone or an interaction bewteen valence and self-reference. Data from three experiments (N = 108) demonstrated that the prioritization effect of good character was robust when the good character referred to the self but weak or non-exist when it referred 29 to others. Additional two experiments (N = 104) further revealed that the mutual 30 facilitation between good character and self-reference occurred even when one of them was 31 task-irrelevant. Together, these results suggested a spontaneous self-referential process as a 32 mechanism of the prioritization effect of good character. 33

Keywords: Perceptual decision-making, Self positivity bias, moral character

Word count: X

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Introduction

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Is moral information prioritized in perception? This question has evoked much heat a 39 few years ago but remains unsolved. On the one hand, morality is a basic dimension in 40 social evaluation (Dunbar, 2004; Ellemers, 2018; Goodwin, 2015; Goodwin, Piazza, & 41 Rozin, 2014), this importance should grant moral information more salient than morally neutral information and thus prioritized when the attentional resource is limited. This logic is similar to other stimuli that are also important to humans, e.g., threatening stimuli (e.g., Ohman, Lundqvist, & Esteves, 2001), rewards (B. A. Anderson, Laurent, & Yantis, 2011), or self-related stimuli (Sui & Rotshtein, 2019). Indeed, previous studies reported bad characters are prioritized in visual processing (E. Anderson, Siegel, Bliss-Moreau, & Barrett, 2011; Eiserbeck & Abdel Rahman, 2020), suggesting that bad people are detected faster than neutral or good people. On the other hand, there is evidence against the view that morally bad information is prioritized in perception. First, researchers reported positive bias in processing moral-related information. For example, Shore and Heerey 51 (2013) found that faces with positive interaction in a trust game were prioritized in the pre-attentive process. Second, the negative bias in perceiving moral information is not 53 robust (Stein, Grubb, Bertrand, Suh, & Verosky, 2017). Third, the mechanism underlying the reported negative bias in processing moral-related information is debated (Firestone & 55 Scholl, 2015, 2016b; Jussim, Crawford, Anglin, Stevens, & Duarte, 2016). In short, while the importance of morality is widely recognized, whether moral information is prioritized in 57 perceptual decision-making is still an open question. Here we manipulated the moral character by an associated learning task and investigated whether immediately acquired moral character information is prioritized in a perceptual matching task.

If moral character information is indeed prioritized, the next question is how?

Previous studies explain the effect based on valence. For example, the negative bias toward moral information is explained by aligning moral information with affective stimuli and threat detection was supposed to be the potential mechanism (B. A. Anderson et al., 2011). The positive bias toward moral information, on the other hand, is explained by 65 value-based attention (Shore & Heerey, 2013). However, these explanations often ignore the fact the value is subjective per se (Juechems & Summerfield, 2019). Merely associating 67 with the self can prioritize the stimuli in perception, attention, working memory, and long-term memory (Sui & Humphreys, 2015 for more details; see Sui & Rotshtein, 2019). Here, we explicitly included self-relevance in our experimental design and tested whether the prioritization of moral character is modulated by self-relevance. We adopted an 71 associative learning task, or self-tagging task, which has been widely used in studying the self-relevance effect. It is based on the well-established fact that humans can quickly learn the associations between symbols via language and change subsequent behaviors accordingly. This associative learning is widely used in aversive learning and value-based learning (Atlas et al., 2022; Deltomme, Mertens, Tibboel, & Braem, 2018). By explicitly instructing participants on which moral character is self-referencing and which is not, we can test whether the prioritization of moral character is by valence per se or by the self-referential of moral valence.

We address these questions by investigating how immediately acquired moral
character information modulates the processing of neutral geometric shapes in a perceptual
matching task. Unlike previous studies relies on faces or words as materials, stimuli used in
the social associative task are geometric shapes, which acquire moral meaning before the
perceptual matching task. Moreover, associations between shapes and different labels of
moral characters are counter-balanced between participants, thus eliminating confounding
effects by stimuli. Also, because we only used a few stimuli and they were repeatedly
presented during the task, 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). We examined whether participants' performance in the perceptual matching task was altered 90 by the immediately acquired moral character of the shapes — in particular, whether the 91 shapes associated with good or bad character are prioritized. We found a robust effect that 92 shapes associated with good character are prioritized in the perceptual matching task. In a 93 series of control experiments, we confirmed that moral content drove the prioritization effect, instead of other factors such as familiarity. In the subsequent experiments, we further tested whether the prioritization of moral character was caused by the valence of moral character alone or the interaction between valence and self-referential processing and found that only shapes associated with both good character and the self are prioritized, suggesting spontaneous moral self-referential as a novel mechanism underlying prioritization of good character in perceptual decision-making.

Disclosures

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We reported all the measurements, analyses, and results in all the experiments in the current study. Participants whose overall accuracy lower than 60% were excluded from analysis. Also, the accurate responses with less than 200ms reaction times were excluded from the analysis. These excluded data can be found in the shared raw data files.

All the experiments reported were not pre-registered. Most experiments (1a ~ 4b,
except experiment 3b) reported in the current study were first finished between 2013 to
2016 in Tsinghua University, Beijing, China. Participants in these experiments were
recruited in the local community. To increase the sample size of experiments to 50 or more
(Simmons, Nelson, & Simonsohn, 2013), we recruited additional participants in Wenzhou
University, Wenzhou, China, in 2017 for experiment 1a, 1b, 4a, and 4b. Experiment 3b was
finished in Wenzhou University in 2017 (See Table 1 for overview of these experiments).

All participants received informed consent and compensated for their time. These

experiments were approved by the ethic board in the Department of Psychology, Tsinghua University.

General methods

17 Design and Procedure

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This series of experiments used the social associative learning paradigm (or 118 self-tagging paradigm, see Sui, He, and Humphreys (2012)), in which participants first 119 learned the associations between geometric shapes and labels of different moral characters 120 (e.g., in the first three studies, the triangle, square, and circle and Chinese words for "good 121 person", "neutral person", and "bad person", respectively). The associations of shapes and 122 labels were counterbalanced across participants. The paradigm consists of a brief learning 123 stage and a test stage. During the learning stage, participants were instructed about the 124 association between shapes and labels. Participants started the test stage with a practice 125 phase to familiarize themselves with the task, in which they viewed one of the shapes above 126 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 130 accuracy was equal to or greater than 60%. The experimental task shared the same trial 131 structure as in the practice. 132

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. nonmatch) 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. 140 Experiment 1c manipulated the moral character indirectly: participants first learned to 141 associate different moral behaviors with different Chinese names, after remembering the 142 association, they then associate the names with different shapes and finished the 143 perceptual matching task. Experiment 2 further tested whether the way we presented the 144 stimuli influence the prioritization of moral character, by sequentially presenting labels and 145 shapes instead of simultaneous presentation. Note that a few participants in experiment 2 146 also participated in experiment 1a because we originally planned a cross-task comparison. 147 Experiment 5 was designed to compare the prioritization of good character with other 148 important social values (aesthetics and emotion). All social values had three levels, 149 positive, neutral, and negative, and were associated with different shapes. Participants 150 finished the associative learning task for different social values in different blocks, and the 151 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 153 shared the same design as experiment 2, was an EEG experiment aimed at exploring the 154 neural mechanism of the prioritization of good character. Only behavioral results of 155 experiment 6a were reported here. 156

Experiments 3a, 3b, and 6b were designed to test whether the prioritization of good 157 character can be explained by the valence effect alone or by an interaction between the 158 valence effect and self-referential processing. To do so, we included self-reference as another 159 within-subject variable. For example, experiment 3a extended experiment 1a into a 2 (matching: match vs. nonmatch) by 2 (reference: self vs. other) by 3 (moral character: 161 good vs. neutral vs. bad) within-subject design. Thus, in experiment 3a, there were six 162 conditions (good-self, neutral-self, bad-self, good-other, neutral-other, and bad-other) and 163 six shapes (triangle, square, circle, diamond, pentagon, and trapezoids). Experiment 6b 164 was an EEG experiment based on experiment 3a but presented the label and shape 165

sequentially. Because of the relatively high working memory load (six label-shape pairs), 166 participants finished experiment 6b in two days. On the first day, participants completed 167 the perceptual matching task as a practice, and on the second day, they finished the task 168 again while the EEG signals were recorded. We only focus on the first day's data here. 169 Experiment 3b was designed to test whether the effect found in experiments 3a and 6b is 170 robust if we separately present the self-referential trials and other-referential trials. That is, 171 participants finished two different types of blocks: in the self-referential blocks, they only 172 made matching judgments to shape-label pairs that related to the self (i.e., shapes and 173 labels of good-self, neutral-self, and bad-self), in the other-referential blocks, they only 174 responded to shape-label pairs that related to the other (i.e., shapes and labels of 175 good-other, neutral-other, and bad-other). 176

Experiments 4a and 4b were designed to further test the interaction between valence 177 and self-referential process in prioritization of good character. In experiment 4a, 178 participants were instructed to learn the association between two shapes (circle and square) 179 with two labels (self vs. other) in the learning stage. In the test stage, they were instructed 180 only respond to the shape and label during the test stage. To test the effect of moral 181 character, we presented the labels of moral character in the shapes and instructed 182 participants to ignore the words in shapes when making matching judgments. In the 183 experiment 4b, we reversed the role of self and moral character in the task: Participants learned associations between three labels (good-person, neutral-person, and bad-person) 185 and three shapes (circle, square, and triangle) and made matching judgments about the 186 shape and label of moral character, while words related to identity, "self" or "other", were 187 presented within the shapes. As in 4a, participants were told to ignore the words inside the 188 shape during the perceptual matching task. 189

90 Stimuli and Materials

We used E-prime 2.0 for presenting stimuli and collecting behavioral responses. Data 191 were collected from two universities located in two different cities in China. Participants 192 recruited from Tsinghua University, Beijing, finished the experiment individually in a 193 dim-lighted chamber. Stimuli were presented on 22-inch CRT monitors and participants 194 rested their chins on a brace to fix the distance between their eyes and the screen around 195 60 cm. The visual angle of geometric shapes was about $3.7^{\circ} \times 3.7^{\circ}$, the fixation cross is of 196 $0.8^{\circ} \times 0.8^{\circ}$ visual angle at the center of the screen. The words were of $3.6^{\circ} \times 1.6^{\circ}$ visual 197 angle. The distance between the center of shapes or images of labels and the fixation cross 198 was of 3.5° visual angle. Participants from Wenzhou University, Wenzhou, finished the 199 experiment in a group consisting of $3 \sim 12$ participants in a dim-lighted testing room. They 200 were instructed to finish the whole experiment independently. Also, they were told to start 201 the experiment at the same time so that the distraction between participants was 202 minimized. The stimuli were presented on 19-inch CRT monitors with the same set of 203 parameters in E-prime 2.0 as in Tsinghua University, however, the visual angles could not 204 be controlled because participants' chins were not fixed. 205

In most of these experiments, participants were also asked to fill out questionnaires after finishing the behavioral tasks. All the questionnaire data were open (see, dataset 4 in Liu et al., 2020). See Table 1 for a summary of information about all the experiments.

$_{209}$ Data analysis

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

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

We used the Bayesian hierarchical model (BHM, or Bayesian generalized linear mixed models, Bayesian multilevel models) to model the reaction time and accuracy data because BHM provided three advantages over the classic NHST approach (repeated measure

ANOVA or t-tests). First, BHM estimates the posterior distributions of parameters for statistical inference, therefore providing uncertainty in estimation (Rouder & Lu, 2005). 216 Second, BHM, where generalized linear mixed models could be easily implemented, can use 217 distributions that fit the distribution of real data instead of using the normal distribution 218 for all data. Using appropriate distributions for the data will avoid misleading results and 219 provide a better fitting of the data. For example, Reaction times are not normally 220 distributed but are right skewed, and the linear assumption in ANOVAs is not satisfied 221 (Rousselet & Wilcox, 2020). Third, BHM provides a unified framework to analyze data 222 from different levels and different sources, avoiding information loss when we need to 223 combine data from different experiments. 224 We used the r package BRMs (Bürkner, 2017), which used Stan (Carpenter et al., 225

We used the r package BRMs (Burkner, 2017), which used Stan (Carpenter et al.,
2017) as the back-end, for the BHM analyses. We estimated the overall effect across
experiments that shared the same experimental design using one model, instead of a
two-step approach that was adopted in mini-meta-analysis (e.g., Goh, Hall, & Rosenthal,
2016). More specifically, a three-level model was used to estimate the overall effect of
prioritization of good character, which included data from five experiments: 1a, 1b, 1c, 2,
5, and 6a. Similarly, a three-level HBM model is used for experiments 3a, 3b, and 6b.
Results of individual experiments can be found in the supplementary results. For
experiments 4a and 4b, which tested the implicit interaction between the self and good
character, we used HBM for each experiment separately.

For questionnaire data, we only reported the subjective distance between different persons or moral character in the supplementary results and did not analyze other questionnaire data, which are described in (Liu et al., 2020).

Response data. We followed previous studies (Hu, Lan, Macrae, & Sui, 2020; Sui
et al., 2012) and used the signal detection theory approach to analyze the response data.

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

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

²⁴² & Lu, 2005).

We used the Bernoulli distribution for the signal detection theory. The probability that the jth subject responded "match" $(y_{ij}=1)$ at the ith trial p_{ij} is distributed as a Bernoulli distribution with parameter p_{ij} :

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

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

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

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

The subjective-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. non-match) 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", ismatch 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. Not 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-referential or other-referential.

Reaction times. We used log-normal distribution

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(https://lindeloev.github.io/shiny-rt/#34_(shifted)_log-normal) to model the RT data.

This means that 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_i is a linear regression of the independent variables:

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

and the parameter σ_i does not vary with independent variables:

$$\sigma_i \sim HalfNormal()$$

The subjective-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:

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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 has the same

meaning as the response data.
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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.

Testing hypotheses.

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Prioritization of moral character. We tested whether moral characters are 291 prioritized by examining the population-level effects (also called fixed effect) of the 292 three-level Bayesian hierarchical model of experiments 1a, 1b, 1c, 2, 5, and 6a. More 293 specifically, we calculated the difference between the posterior distribution of the good/bad 294 character and the neutral character and tested whether the 95% highest density intervals 295 (HDIs) of the difference include zero. If the 95% highest density intervals do not include 296 zero, we infer that there is a population-level difference between the conditions in the test, 297 otherwise, we will infer that there is no evidence for such a difference. Note that for reaction times, we focused on the matched trials as in previous studies.

Modulation of self-referential processing. We tested the modulation effect
ofself-referential processing by examining the interaction between moral character and
self-referential process for the three-level Bayesian hierarchical model of experiments 3a, 3b,
and 6b. More specifically, we tested two possible explanations for the prioritization of good
character: pure valence effect or the valence effect is modulated by self-referential process.

If the former is correct, then there will be no interaction between moral character and

self-referential processing, i.e., the prioritization effect exhibits a similar pattern for both self- and other-referential conditions. On the other hand, if spontaneous self-referential processing account is true, then there will be an interaction between the two factors, i.e., the prioritization effect exhibits different patterns for self- and other-referential conditions.

Spontaneous binding between the self and good character. For on data 310 from experiments 4a and 4b, we further examined whether the self-referential processing 311 for moral characters is spontaneous (i.e., whether the good character is spontaneously bound with self). For experiment 4a, if there exists a spontaneous binding between self and 313 good character, there should be an interaction between moral character and self-referential 314 processing, e.g., the task-irrelevant moral words either facilitate or slows down the response 315 to self- or other-referential conditions. For experiment 4b, if there exists a spontaneous 316 binding between self and good character, then, there will be a self-other difference for some 317 moral character conditions but not for other moral character conditions. 318

Results

Prioritization of good character

To test whether moral characters are prioritized, we modeled data from experiments 321 1a, 1b, 1c, 2, 5, and 6a with three-level Bayesian hierarchical models. All these experiments 322 shared similar designs and can be used for testing the prioritization effect of moral 323 character. The valid and unique sample size is 192. Note that for both experiments 1a and 324 1b, two datasets were collected at different time points and locations, thus we treated them 325 as independent samples. Here we only reported the population-level results of three-level 326 Bayesian models, the detailed results of each experiment can be found in supplementary 327 materials. 328

For the d prime, results from the Bayesian model revealed a robust effect of moral character. Shapes associated with good character ("good person", "kind person" or a name

associated with good behaviors) has higher sensitivity (median = 2.51, 95% HDI = [2.23 2.78]) than shapes associated with neutral character (median = 2.19, 95% HDI = [1.88 2.50]), $median_{diff} = 0.31$, 95% HDI [0.00 0.64], but we did not find differences between shapes associated with bad character (median = 2.25, 95% HDI = [1.94 2.55]) and neutral character, $median_{diff} = 0.05$, 95% HDI [-0.28 0.39].

The results from reaction times data 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 character has faster responses (median = 579.03 ms, 95% HDI = [500.20 660.89]) than shapes associated with neutral character (median = 623.59 ms, 95% HDI = [542.83 710.82]), $median_{diff} = -44.19$, 95% HDI [-59.85 -30.36]. We also found that RTs to shapes associated with bad character (median = 640.86 ms, 95% HDI = [561.22 729.99]) were slower as compared to the neutral character, $median_{diff} = 16.85$, 95% HDI [2.82 30.10].

For the nonmatch trials, we found the advantage of good character: Shapes associated with good character (median = 654.16 ms, 95% HDI = [573.12 742.91]) are faster than shapes associated with neutral (median = 671.81 ms, 95% HDI = [588.33 762.65]), $median_{diff} = -17.72$ ms, 95% HDI [-24.58 -11.19]. Similarly, the shapes associated with bad character (median = 676.93 ms, 95% HDI = [590.23 765.67]) was responded slower than shapes associated with neutral character, $median_{diff} = 5.20$ ms, 95% HDI [-0.04 10.66], but the effect size was smaller than the match trials.

Modulation effect self-referential processing

To test the modulation effect of self-referential processing, 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 interaction between self-referential processing and moral valence. Also, we examined

the differences of differences, i.e., how the differences between good/bad characters and the
neutral character under the self-referential conditions differ from that under
other-referential conditions. The detailed results of each experiment can be found in
supplementary materials.

For the d prime, we found an interaction between the moral valence and 360 self-referential processing: the good-neutral differences are larger for the self-referential 361 condition than for the other-referential condition ($median_{diff} = 0.52$; 95% HDI = [-1.54] 362 2.72]). However, this is not the case for the bad-neutral differences ($median_{diff} = -0.01$; 363 95% HDI = $[-1.55 \ 2.37]$). Further analyses revealed that the prioritization effect of good 364 character (as compared to neutral) only appeared for self-referential conditions but not 365 other-referential conditions. The estimated d prime for good-self was greater than neutral-self ($median_{diff} = 0.57$; 95% HDI = [-0.93 2.88]), d prime for good-self was also 367 greater than good-other condition ($median_{diff} = 0.23$; 95% HDI = [-1.31 1.87]). The differences between bad-self and neutral-self, good-other and neutral-other, and bad-other and neutral-other are all centered around zero (see Figure 2, B, D).

For the RTs of matched trials, we also found an interaction between moral valence and self-referential processing: the good-neutral differences were different for the self- and other-referential conditions ($median_{diff} = -155.11$; 95% HDI = [-755.36 395.38]). However, this was not the case for bad-neutral differences ($median_{diff} = -55.63$; 95% HDI = [-604.70 561.00]). Further analyses revealed a robust good-self prioritization effect as compared to neutral-self ($median_{diff} = -57.91$; 95% HDI = [-224.43 41.14]) and good-other ($median_{diff} = -107.21$; 95% HDI = [-369.05 92.95]) conditions. Similar to the results of d', we found that participants responded slower for both good character and bad character than for the neutral character when they referred to others. See Figure 2.

These results suggested that the prioritization of good character is not solely driven by the valence of moral character. Instead, the self-referential processing modulated the prioritization of good character: good character was prioritized only when it was self-referential. When the moral character was other-referential, responses to both good and bad characters were slowed down.

Spontaneous binding between the good character and the self

Experiments 4a and 4b were designed to test whether the good character and self-referential processing 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, albeit weak, that task-irrelevant moral 393 character also played a role. For shapes associated with "self", d' was greater when shapes 394 had a good character inside (median = 2.83, 95% HDI [2.63 3.01]) than shapes that have 395 neutral character (median = 2.74, 95% HDI [2.58 2.95], BF = 4.4) or bad character $(\text{median} = 2.76, 95\% \text{ HDI } [2.56 \ 2.95], \text{ BF} = 3.1), \text{ but this is not the case for self-referential}$ shapes with bad character and neutral character inside. For shapes associated with 398 "other", the pattern reversed: d prime was smaller when shapes had a good character 399 inside (median = 1.87, 95% HDI [1.71 2.04]) than had neutral (median = 1.96, 95% HDI 400 $[1.80 \ 2.14]$) or bad character (median = 1.98, 95% HDI $[1.79 \ 2.17]$) inside. See Figure 3. 401

A similar pattern was found for RTs in matched trials. For the "self" condition, when a good character was presented inside the shapes, the RTs (median = 641, 95% HDI [623 662]) were faster than when a neutral character (median = 649, 95% HDI [631 668]) or bad character (median = 648, 95% HDI [628 667]) were inside. This effect was reversed for the "other" condition: RTs for shapes associated with good character inside (median = 733,

95% HDI [711 754]) were slower than those with neutral character (median = 721, 95%
 HDI [702 741]) or bad character (median = 718, 95% HDI [696 740]) inside.

In experiment 4b, where moral characters were task-relevant and "self" vs "other"
were task-irrelevant, we found a main effect of moral character: performance for shapes
associated with good characters was better than other-related conditions on both d' and
reaction times. This pattern, again, shows a robust prioritization effect of good character.

Most importantly, we found evidence that task-irrelevant labels, "self" or "other", also played a role. For shapes associated with good character, the d prime was greater when shapes had a "self" inside than with "other" inside $(mean_{diff} = 0.14, 95\%)$ credible intervals [-0.02, 0.31], BF = 12.07), but this effect did not occur when the target shape where associated with "neutral" $(mean_{diff} = 0.04, 95\%)$ HDI [-.11, .18]) or "bad" person $(mean_{diff} = -.05, 95\%)$ HDI[-.18, .09]).

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 $(mean_{diff} = -55 \text{ ms}, 95\% \text{ HDI } [-75, -35])$, but this effect did not occur when the shapes were associated neutral $(mean_{diff} = 10, 95\% \text{ HDI } [1, 20])$ or bad $(mean_{diff} = 5, 95\% \text{ HDI } [-16, 27])$ person, See Figure 3.

424 Discussion

Across nine experiments, we explored the prioritization effect of moral character and
the underlying mechanism by a combination of social associative learning and perceptual
matching task. First, we found a robust effect that good character was prioritized in the
shape-label matching task across five experiments. Second, across three experiments, we
found that the prioritization of good character was not solely driven by moral valence
itself, i.e., "good" vs "bad". Instead, this effect was modulated by self-referential
processing: prioritization only occurred when moral characters are self-referential. Finally,

the prioritization of the combination of good character and self occurred, albeit weak, even 432 when either the self- or character-related information was irrelevant to the experimental 433 task (experiment 4a and 4b). In contrast, performance to the combination of good 434 character and "other", explicitly or implicitly, was worse than the combination of neutral 435 character and "other". Together, these results highlighted the importance of the self in 436 perceiving information related to moral characters, suggesting a spontaneous self-referential 437 process when making perceptual decision-making for moral characters. These results are in 438 line with a growing literature on the social and relational nature of perception (Xiao, 439 Coppin, and Bavel (2016); Freeman, Stolier, and Brooks (2020); hafri_perception_2021) 440 and deepened our understanding of mechanisms of perceptual decision-making of moral 441 information.

The current study provided robust evidence for the prioritization of good character in 443 perceptual decision-making. The existence of the effect of moral valence on perception has 444 been disputed. For instance, (E. Anderson et al., 2011) reported that faces associated with 445 bad social behavior capture attention more rapidly, however, an independent team failed to 446 replicate the effect (Stein et al., 2017). Another study by Gantman and Van Bavel (2014) 447 found that moral words are more likely to be judged as words when it was presented 448 subliminally, however, this effect may be caused by semantic priming instead of morality (Firestone & Scholl, 2015; Jussim et al., 2016). In the current study, we found the 450 prioritization effect across five experiments, the sample size of individual experiments and 451 combined provide strong evidence for the existence of the effect. Moreover, the associative 452 learning task allowed us to eliminate the semantic priming effect for two reasons. First, associations between shapes and moral characters were acquired right before the perceptual matching task, semantic priming from pre-existed knowledge was impossible. Second, there 455 were only a few pairs of stimuli were used and each stimulus represented different 456 conditions, making it impossible for priming between trials. Importantly, a series of control 457 experiments (1b, 1c, and 2) further excluded other confounding factors such as familiarity, 458

presenting sequence, or words-based associations, suggesting that it was the moral content that drove the prioritization of good character.

The robust prioritization of good character found in the current study was 461 incongruent with previous moral perception studies, which usually reported a negativity 462 effect, i.e., information related to bad character is processed preferentially (E. Anderson et 463 al., 2011; Eiserbeck & Abdel Rahman, 2020). This discrepancy may be caused by the 464 experimental task: while in many previous moral perception studies, the participants were 465 asked to detect the existence of a stimulus, the current task asked participants to recognize 466 a pattern. In other words, previous studies targeted early stages of perception while the 467 current task focused more on decision-making at a relatively later stage of information 468 processing. This discrepancy is consistent with the pattern found in studies with emotional 460 stimuli (Pool, Brosch, Delplangue, & Sander, 2016). 470

We expanded previous moral perception studies by focusing on the agent who made 471 the perceptual decision-making and examined the interaction between moral valence and 472 self-referential processing. Our results revealed that prioritization of good character is 473 modulated by self-referential processing: the good character was prioritized when it was 474 related to the "self", even when the self-relatedness was task-irrelevant. By contrast, good 475 character information was not prioritized when it was associated with "other". The 476 modulation effect of self-referential processing was large when the relationship between 477 moral character and the self was explicit, which is consistent with previous studies that only 478 positive aspects of the self are prioritized (Hu et al., 2020). More importantly, the effect 479 persisted when the relationship between moral character and self-information was implicit, suggesting spontaneous self-referential processing when both pieces of information were presented. A possible explanation for this spontaneous self-referential of good character is that the positive moral self-view is central to our identity (Freitas, Cikara, Grossmann, & 483 Schlegel, 2017; Strohminger, Knobe, & Newman, 2017) and the motivation to maintain a 484 moral self-view influences how we perceive (e.g., Ma & Han, 2010) and remember (e.g., 485

Carlson, Maréchal, Oud, Fehr, & Crockett, 2020; Stanley, Henne, & De Brigard, 2019).

Although the results here revealed the prioritization of good character in perceptual 487 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 489 encoding the sensory inputs. To fully account for the nuance of behavioral data and/or 490 related data collected from other modules (e.g., Sui, He, Golubickis, Svensson, & Neil 491 Macrae, 2023), we need computational models and an integrative experimental approach 492 (Almaatouq et al., 2022). For example, sequential sampling models suggest that, when 493 making a perceptual decision, the agent continuously accumulates evidence until the 494 amount of evidence passed a threshold, then a decision is made (Chuan-Peng et al., 2022; 495 Forstmann, Ratcliff, & Wagenmakers, 2016; Ratcliff, Smith, Brown, & McKoon, 2016). In 496 these models, the evidence, or decision variable, can accumulate from both sensory 497 information but also memory (Shadlen & Shohamy, 2016). Recently, applications of 498 sequential sample models to perceptual matching tasks also suggest that different processes 490 may contribute to the prioritization effect of self (Golubickis et al., 2017) or good self (Hu 500 et al., 2020). Similarly, reinforcement learning models also revealed that the key difference 501 between self- and other-referential learning lies in the learning rate (Lockwood et al., 2018). 502 These studies suggest that computational models are needed to disentangle the cognitive 503 processes underlying the prioritization of good character. 504

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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)	60	NA	words	Simultaneously
Exp_1b_1	2014-10	Beijing	39 (27)	NA	NA	words	Simultaneously
Exp_1b_2	2017-04	Wenzhou	33 (25)	NA	NA	words	Simultaneously
Exp_1c	2014-10	Beijing	23 (23)	NA	NA	descriptions	Simultaneously
Exp_2	2014-05	Beijing	35 (34)	NA	NA	words	Sequentially
Exp_3a	2014-11	Beijing	38 (35)	NA	explicit	words	Simultaneously
Exp_3b	2017-04	Wenzhou	61 (56)	NA	explicit	words	Simultaneously
Exp_4a_1	2015-06	Beijing	32 (29)	NA	implicit	words	Simultaneously
Exp_4a_2	2017-04	Wenzhou	32 (30)	NA	implicit	words	Simultaneously
Exp_4b_1	2015-10	Beijing	34 (32)	NA	implicit	words	Simultaneously
Exp_4b_2	2017-04	Wenzhou	19 (13)	NA	implicit	words	Simultaneously
Exp_5	2016-01	Beijing	43 (38)	NA	NA	words	Simultaneously
Exp_6a	2014-12	Beijing	24 (24)	NA	NA	words	Sequentially
Exp_6b	2016-01	Beijing	23 (22)	NA	explicit	words	Sequentially
Exp_7a	2016-07	Beijing	35 (29)	NA	explicit	words	Simultaneously
Exp_7b	2018-05	Beijing	46 (42)	NA	explicit	words	Simultaneously

Note. Stim of Morality = How moral character was manipulated; Presenting order = how shapes & labels were presented. The data from experiments 7a & 7b, which were reported in Hu et al (2020), are only included in the meta-analysis in supplementary materials.

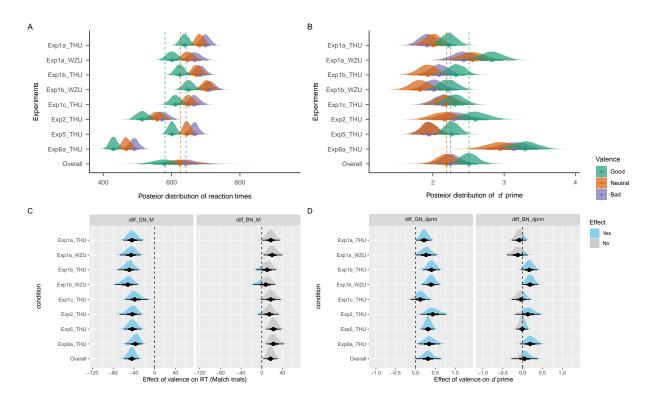
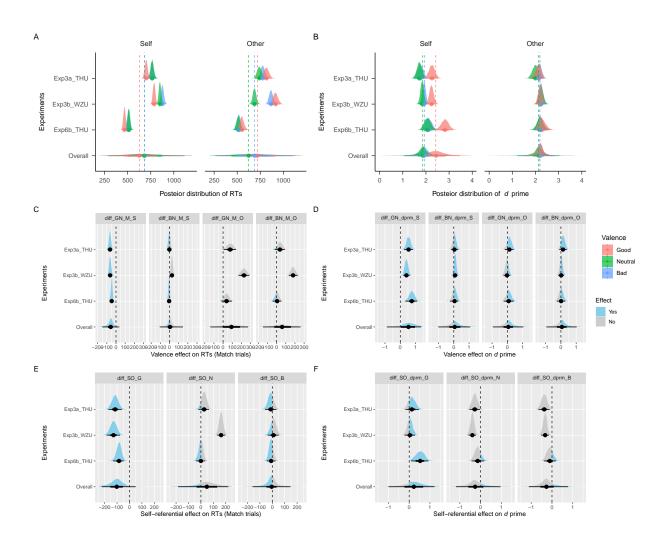


Figure 1. Effect of moral character on RT and \mathbf{d}'



Figure~2. Interaction between moral character and self-referential

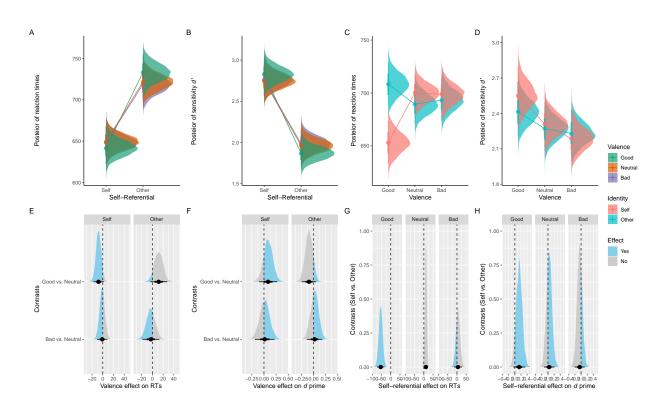


Figure 3. Experiment 4: Implicit binding between good character and the self.