Self-relevance modulates the priorization of the good character in perceptual matching

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Abstract

Moral character is central to social perception and moral psychology, yet how these information is process is unclear. Both negativity effect and positivity effect are possible. Here, we examined these two possibilities using social associataive learning task, where participants first associat social conceptions with neutral stimuli and then perform cognitive tasks (e.g., matching judgment). Across 9 experiments (N = 4XX, trials = XXX), we found a robust positivity effect: shapes associated with good character were prioritized, as compared to shapes associated with neutral or bad characters. Crucially, good character effect was robust when it referred to the self but weak or non-exist when it referred to a stranger. Moreover, even when identity or moral character information became task-irrelevant, participants were still sensitive to good character: the good-self combination facilitate the process while the good-other combination slowed down the process. Together, these results suggested that good character is prioritized in social associative learning task and this effect was due a spontaneous self-referential processing.

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

Word count: X

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Alternative title: The good person is me: Spontaneous self-referential may explain the prioritization of good moral character

Introduction

[quotes about moral character]

[Morality is central to social life, moral character is the central of morality] **People experience a substantial amount of moral events in everyday life (e.g., Hofmann, Wisneski, Brandt, & Skitka, 2014) and judging the moral character of people is indispensable part of these events**. Whether we are the agent, target, or a third party of a moral event, we always judge moral behaviors as "right" or "wrong," and by doing so, we judge people as "good" or "bad" (Uhlmann, Pizarro, & Diermeier, 2015). Moral character is so important in social life that it is a basic dimension in our social evaluation (Goodwin, 2015; Goodwin, Piazza, & Rozin, 2014) and that a substantial part of people's conversation are gossiping others' moral character (or, reputation) (e.g., Dunbar, 2004). These moral character information may help us to evaluate our in-group members and distinguish out-group members (Ellemers, 2018). Crucially, moral character central for self-concept and identity too. A positive moral character is viewed as the core feature of identity (e.g., Strohminger, Knobe, & Newman, 2017).

[Two possibilities about moral character] Given the importance of moral character and limited cognitive resources to process all the information in a social world, will people deferentially process the information related to moral character and prioritize information with certain moral character? One possibility is that 'bad' character are prioritized. This possibility is consistent with early studies in impression formation which found that negative traits are weighted more in overall impression (N. H. Anderson, 1965; Fiske, 1980; Skowronski & Carlston, 1987). This idea also seemed to consistent with the more general idea that “bad is stronger than good” (Baumeister, Bratslavsky, Finkenauer, & Vohs, 2001; Pratto & John, 1991). A few studies provided evidence for this possibility. For example, E. Anderson, Siegel, Bliss-Moreau, and Barrett (2011) asked participants to associate faces with different behaviors (e.g., negative and neutral behaviors from both social and nonsocial domains) and then perform a binocular rivalry task, where a face and a building were presented to each eye. Participants were required report the content of their visual awareness by pressing buttons. The results revealed that faces associated negative social behaviors dominated participants’ visual awareness longer than faces associated with other types of behaviors (but see Stein, Grubb, Bertrand, Suh, & Verosky, 2017). Similarly, Eiserbeck and Abdel Rahman (2020) combined associative learning with attention blink paradigm, where neutral faces were associated with sentences about neutral or negative trust behaviors. They also found that neutral faces associated with negative behavior were processed preferentially.

It is also possible, however, that the negativity effect in impression formation may not be able to generalized to perceptual process, instead, positive moral characters are prioritized (see recent reviews, Pool, Brosch, Delplanque, & Sander, 2016; Unkelbach, Alves, & Koch, 2020). Unkelbach, Alves, and Koch (2020) pointed out that bad is not necessarily stronger than good in all aspects of information processing. Sometimes, good is stronger than bad. For example, when participants are asked to classify words as good or bad, positive trait words are classified faster than negative words (Bargh, Chaiken, Govender, & Pratto, 1992). Similarly, in a lexical decision task, participants judge positive words faster than negative words (Unkelbach et al., 2010). Also, Anisfeld and Lambert (1966) found that positive words are easier to associate with nonsense word-like strings, and this advantage in associative potential also appeared in implicit association test (IAT) (Anselmi, Vianello, & Robusto, 2011). Direct evidence for positivity effect of moral character also exist: Shore and Heerey (2013) found that faces with positive interaction in a trust game were prioritized in pre-attentive process.

Here, we attempted to distinguish these two possibilities by a social associative learning task in which physical features had minimal influences — participants performed a perceptual matching task after associated different moral characters (good, neutral, and bad) with different geometric shapes. If there is a positivity effect, there should be an advantage for shapes associated with good character over shapes associated with neutral or bad shapes. If there is a negativity effect, the advantage should be occur on shapes associated with bad characters. The first four experiments and two additional follow-up experiments provided strong evidence for good character effect in the current paradigm.

The positivity effect consistent with previous studies where positivity effect of social trait words were found (Anselmi, Vianello, & Robusto, 2011; Bargh, Chaiken, Govender, & Pratto, 1992; Unkelbach et al., 2010). However, the effect could not be explained by the similarity hypothesis (Unkelbach, Alves, & Koch, 2020) because we only used three stimuli. There are two possibility explanations. The first one is the value-based attention account, which suggests that stimuli that are valuable to us are prioritized (B. A. Anderson, 2019). In our experiments, the good character label "good person" may represent an indirect but valuable stimuli because, in social life, a good other is usually more valuable than an bad other (Abele & Wojciszke, 2007). Another possibility is derived from social categorization theory, which suggested that we automatically categorize others as in-group or out-group (Turner, Hogg, Oakes, Reicher, & Wetherell, 1987). Moral character is an important criterion for social categorization (DeScioli, 2016; McHugh, McGann, Igou, & Kinsella, 2019). However, the above four experiments could not distinguish between these two possibilities, because "good person" could both be rewarding and be categorized as in-group member. Given that both rewarding stimuli (e.g., Sui, He, & Humphreys, 2012) and in-group information (Enock, Hewstone, Lockwood, & Sui, 2020) are prioritized when using social associative learning paradigm, we further tested these two possibilities in new experiments.

To distinguish the value-based account and the social categorization explanations, we introduced the identity (self- vs. other-referential) of moral character as an addition independent variable in exp 3a, 3b, and 6b. Now moral valence is orthogonal to the identity. In this case, the identity of moral character information become salient and participants are less likely to spontaneously categorize a good-other as an extension of self, but the value of good-person still exists. If the positivity effect was driven by social categorization theory, then participants prioritize good-self but not good-other. If the value-based attention theory is true, then, both good-self and good-other are prioritized, or maybe good-other are even more prioritized.

Although the introduction of self- and other-referential processing provided evident that value-based account can not explain the good-character effect, it might introduce the good-self effect, i.e., the good-self is prioritized over all the other stimuli. This effect, if true, may suggest underlying mechanisms other than social-categorization. For example, the moral true self account. Moral true self view suggested that moral self if the true self (Strohminger, Knobe, & Newman, 2017). Therefore, even good-self can be viewed as categorized to in-group, it can also be viewed as the core of the self and it is the anchor of all the other effects.

To test the moral true self view and the social-categorization account, we designed two complementary experiments. In experiment 4a, participants only learned the association between self and other, the words "good-person," "neutral person," and "bad person" were presented as task-irrelevant stimuli, while in experiment 4b, participants learned the associations between "good-person," "neutral-person," and "bad-person," and the "self" and "other" were presented as task-irrelevant stimuli. These two experiments can be used to distinguish the moral-self view and social categorization" account. If moral-self view is true, then, in both experiments, good-self will show advantage over all other stimuli, and there will be no other effects. More specifically, in experiment 4a, where only the self-referential processing is task-relevant, there will be advantage for good as task-irrelevant condition than when bad or neutral character as task-irrelevant for the self conditions, while there is no other effects; in experiment 4b, in the good condition, there will be an advantage for self as task-irrelevant condition over other as task-irrelevant condition, and no other effects. If social categorization is true, then, the prioritization effect will depends on whether the stimuli can be categorized as the same group of good-self. More specifically, in experiment 4a, there will be good effect in self conditions, this prediction is the same as the moral self-view; it predicts a reverse good effect in other condition because good and other a conflict in terms of social-categorization, this prediction is different from the "good-self" anchor account; however, for experiment 4b, it predicts no identity effect in the good-person condition because both self and other are in the good group.

[Good self in self-reported data] As an exploration, we also collected participants' self-reported psychological distance between self and good-person, bad-person, and neutral-person, moral identity, moral self-image, and self-esteem. All these data are available (see Liu et al., 2020). We explored the correlation between self-reported distance and these questionnaires as well as the questionnaires and behavioral data. However, given that the correlation between self-reported score and behavioral data has low correlation (Dang, King, & Inzlicht, 2020), we didn't expect a high correlation between these self-reported measures and the behavioral data.

Disclosures

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. To have a better estimation of the effect size, we included the data from unreported data in our three-level models (experiment 5, 6a, 6b) (See Table S1 for overview of these experiments).

All participant 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

Design and Procedure

This series of experiments studied the perceptual process of moral character, using the social associative learning paradigm (or tagging paradigm)(Sui, He, & Humphreys, 2012), in which participants first learned the associations between geometric shapes and labels of person with different moral character (e.g., in first three studies, the triangle, square, and circle and good person, neutral person, and bad person, respectively). The associations of the shapes and label were counterbalanced across participants. After remembered the associations, participants finished a practice phase to familiar with the task, in which they viewed one of the shapes upon the fixation while one of the labels below the fixation and judged whether the shape and the label matched the association they learned. When participants reached 60% or higher accuracy at the end of the practicing session, they started the experimental task which was the same as in the practice phase.

The experiment 1a, 1b, 1c, 2, 5, and 6a 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 studies and found the prioritization of stimuli associated with good-person. To confirm that it is the moral character that caused the effect, we further conducted experiment 1b, 1c, and 2. More specifically, experiment 1b used different Chinese words as labels to test whether the effect only occurred with certain words. Experiment 1c manipulated the moral valence indirectly: participants first learned to associate different moral behaviors with different Chinese names, after remembered the association, they then performed the perceptual matching task by associating names with different shapes. Experiment 2 further tested whether the way we presented the stimuli influence the effect of valence, by sequentially presenting labels and shapes. Note that part of participants of experiment 2 were from experiment 1a because we originally planned a cross task comparison. Experiment 5 was designed to compare the effect size of moral character and other importance social evaluative dimensions (aesthetics and emotion). Different social evaluative dimensions were implemented in different blocks, the moral character blocks shared the design of experiment 1a. Experiment 6a, which shared the same design as experiment 2, was an EEG experiment which aimed at exploring the neural correlates of the effect. But we will focus on the behavioral results of experiment 6a in the current manuscript.

For experiment 3a, 3b, and 6b, we included self-reference as another within-subject variable in the experimental design. For example, the experiment 3a directly extend the design of experiment 1a into a 2 (matching: match vs. nonmatch) 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). The 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), experiment 6b were conducted in two days: the first day participants finished perceptual matching task as a practice, and 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 separate the self-referential trials and other-referential trials. That is, participants finished two different types of block: in the self-referential blocks, they only responded to good-self, neutral-self, and bad-self, with half match trials and half nonmatch trials; in the other-reference blocks, they only responded to good-other, neutral-other, and bad-other.

Experiment 4a and 4b were design to explore the mechanism underlying the prioritization of good-self. In 4a, we only used two labels (self vs. other) and two shapes (circle, square). To manipulate the moral character, we added the moral-related words within the shape and instructed participants to ignore the words in the shape during the task. In 4b, we reversed the role of self-reference and moral character in the task: participant learned three labels (good-person, neutral-person, and bad-person) and three shapes (circle, square, and triangle), and the words related to identity, "self" or "other," were presented in the shapes. As in 4a, participants were told to ignore the words inside the shape during the task.

E-prime 2.0 was used for presenting stimuli and collecting behavioral responses. For participants recruited in Tsinghua University, they finished the experiment individually in a dim-lighted chamber, stimuli were presented on 22-inch CRT monitors and their head were fixed by a chin-rest brace. The distance between participants' eyes and the screen was about 60 cm. The visual angle of geometric shapes was about $3.7^{\circ} \times 3.7^{\circ}$, the fixation cross is of $0.8^{\circ} \times 0.8^{\circ}$ visual angle at the center of the screen. The words were of $3.6^{\circ} \times 1.6^{\circ}$ visual angle. The distance between the center of the shape or the word and the fixation cross was 3.5° of visual angle. For participants recruited in Wenzhou University, they finished the experiment in a group consisted of $3 \sim 12$ participants in a dim-lighted testing room. Participants were required to finished the whole experiment independently. Also, they were instructed to start the experiment at the same time, so that the distraction between participants were minimized. The stimuli were presented on 19-inch CRT monitor. The visual angles are could not be exactly controlled because participants' chin were not fixed.

In most of these experiments, participant were also asked to fill a battery of questionnaire after they finish the behavioral tasks. All the questionnaire data are open (see, dataset 4 in Liu et al., 2020). See Table S1 for a summary information about all the experiments.

Data analysis

We used the `tidyverse` of r (see script `Load_save_data.r`) to preprocess the data. Results of each experiment 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 estimate the posterior distributions of parameters for statistical inference, therefore provided uncertainty in estimation (Rouder & Lu, 2005). Second, BHM, where generalized linear mixed models could be easily implemented, can use distributions that fit the distribution of real data instead of using normal distribution for all data. Using appropriate distributions for the data will avoid misleading results and provide better fitting of the data. For example, Reaction times are not normally distributed but right skewed, and the linear assumption in ANOVAs is not satisfied (Rousselet & Wilcox, 2019). Third, BHM provided an unified framework to analyze data from different levels and different sources, avoid the information loss when we need to combine data from different levels.

We used the `r` package `BRMs` (Bürkner, 2017), which used Stan (Carpenter et al., 2017) for the BHM analyses. We estimated the over-all effect across experiments with similar experimental design, instead of using a two-step approach where we first estimate parameters, e.g., d' for each participant, and then use a random effect model meta-analysis to synthesize the effect (Goh, Hall, & Rosenthal, 2016).

Accuracy. We followed practice of previous studies (Hu, Lan, Macrae, & Sui, 2020; Sui, He, & Humphreys, 2012) and used signal detection theory approach to analyze the accuracy data. More specifically, the match trials are treated as signal and the non-match trials are noise. As we mentioned above, we estimated the sensitivity and criterion of SDT by BHM (Rouder & Lu, 2005). Because the BHM can model different level's data using a single unified model, we used a three-level HBM to model the moral character effect, which include five experiments: 1a, 1b, 1c, 2, 5, and 6a. Similarly, we modeled experiments with both self-referential and moral character with a three-level HBM model, which includes 3a, 3b, and 6b. For experiment 4a and 4b, we used two-level models for each separately. However, we could compare the posterior of parameters directly because we have full posterior distribution of parameters.

We used the Bernoulli distribution to model the accuracy data. For a single participant, we assume that the accuracy of ith trial is Bernoulli distributed (binomial with 1 trial), with probability p_i that $y_i = 1$.

$$y_i \sim Bernoulli(p_i)$$

and the probability of choosing "match" p_i at the ith trial is a function of the trial type:

$$\Phi(p_i) = \beta_0 + \beta_1 IsMatch_i$$

therefore, the outcomes y_i are 0 if the participant responded "nonmatch" on the ith trial, 1 if they responded "match." We then write the generalized linear model on the probits (z-scores; Φ, "Phi") of ps. Φ is the cumulative normal density function and maps z scores to probabilities. In this way, the intercept of the model (β_0) is the standardized false alarm rate (probability of saying 1 when predictor is 0), which we take as our criterion c. The slope of the model (β_1) is the increased probability of responding "match" when the trial type is "match," in z-scores, which is another expression of d'. Therefore, c = -zHR = $-\beta_0$, and $d' = \beta_1$.

In our experimental design, there are three conditions for both match and non-match trials, we can estimate the d' and c separately for each condition. In this case, the criterion c is modeled as the main effect of valence, and the d' can be modeled as the interaction between valence and match:

$$\Phi(p_i) = 0 + \beta_0 Valence_i + \beta_1 IsMatch_i * Valence_i$$

In each experiment, we had multiple participants. We can estimate the group-level parameters by extending the above model into a two-level model, where we can estimate parameters on individual level (varying effect) and the group level parameter simultaneously (fixed effect). The probability that the jth subject responded "match" ($y_{ij} = 1$) at the ith trial p_{ij}. In the same vein, we have

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

The the generalized linear model can be re-written to include two levels:

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

We again can write the generalized linear model on the probits (z-scores; Φ, "Phi") of ps.

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

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

For experiments that had 2 (matching: match vs. non-match) by 3 (moral character: good vs. neutral vs. bad), i.e., experiment 1a, 1b, 1c, 2, 5, and 6a, the formula for accuracy in `BRMs` is as follow:

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

For experiments that had two by two by three design, we used the follow formula for the BGLM:

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

In the same vein, we can estimate the posterior of parameters across different experiments. We can use a nested hierarchical model to model all the experiment with similar design:

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

the generalized linear model is then

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

The outcomes y_{ijk} are 0 if participant j in experiment k responded “nonmatch” on trial i, 1 if they responded “match.”

$$\begin{bmatrix} \beta_{0jk} \\ \beta_{1jk} \end{bmatrix} \sim N(\begin{bmatrix} \theta_{0k} \\ \theta_{1k} \end{bmatrix}, \sum)$$

and the experiment level parameter mu_{0k} and mu_{1k} is from a higher order distribution:

$$\begin{bmatrix} \theta_{0k} \\ \theta_{1k} \end{bmatrix} \sim N(\begin{bmatrix} \mu_0 \\ \mu_1 \end{bmatrix}, \sum)$$

in which mu_0 and mu_1 means the population level parameter.

Reaction times. For the reaction time, we used the log normal distribution (https://lindeloev.github.io/shiny-rt/#34_(shifted)_log-normal) to model the data. This means that we need to estimate the posterior of two parameters: μ, σ. μ is the mean of the `logNormal` distribution, and σ is the disperse of the distribution. Although the log normal distribution can be extended to shifted log normal distribution, with one more parameter: shift, which is the earliest possible response, we found that the additional parameter didnt' improved the model fitting and therefore used the logNormal in our final analysis.

The reaction time of the jth subject on ith trial is a linear function of trial type:

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

while the log of the reaction time is log-normal distributed:

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

y_{ij} is the RT of the ith trial of the jth participants.

$$\mu_j \sim N(\mu, \sigma)$$

$$\sigma_j \sim Cauchy()$$

Formula used for modeling the data as follow:

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

or

```
RT_sec ~ ID*Valence*ismatch + (ID*Valence*ismatch | Subject), family =
lognormal()
```

we expanded the RT model three-level model in which participants and experiments are two group level variable and participants were nested in the experiments.

$$log(y_{ijk}) \sim N(\mu_{jk}, \sigma_{jk})$$

y_{ijk} is the RT of the ith trial of the jth participants in the kth experiment.

$$\mu_{jk} \sim N(\mu_k, \sigma_k)$$

$$\sigma_{jk} \sim Cauchy()$$

$$\mu_k \sim N(\mu, \sigma)$$

$$\theta_k \sim Cauchy()$$

Effect of moral character. We estimated the effect size of d' and RT from experiment 1a, 1b, 1c, 2, 5, and 6a for the effect of moral character. We reported fixed effect of three-level BHM that included all experiments that tested the valence effect.

Interaction between moral character and self-referential process. We also estimated the interaction between moral character and self-referential process, which included results from experiment 3a, 3b, and 6b. Using three-level models, we tested two possible explanations for the prioritization of good character: value-based or social categorization based prioritization.

Implicit interaction between valence and self-relevance. In the third part, we focused on experiment 4a and 4b, which were designed to examine two more nuanced explanation concerning the good-self. The design of experiment 4a and 4b are complementary. Together, they can test whether participants are more sensitive to the moral character of the Self (4a), or the identity of the good character (4b).

For the questionnaire part, we are most interested in the self-rated distance between different person and self-evaluation related questionnaires: self-esteem, moral-self identity, and moral self-image. Other questionnaires (e.g., personality) were not planned to correlated with behavioral data were not included. Note that all questionnaire data were reported in (Liu et al., 2020).

Results

Perceptual processing moral character related information

In this part, we report results from five experiments that tested whether an associative learning task, including 192 participants. Note that for both experiment 1a and 1b, there were two independent samples with different equipment, trials numbers and testing situations. Therefore, we modeled them as independent samples. These five experiments revealed a robust effect of moral character on perceptual matching task.

For the *d* prime, we found robust effect of moral character. Shapes associated with good character ("good person," "kind person" or a name associated with morally good behavioral history) has higher sensitivity (median = 2.49, 95% HDI = [2.19 2.75]) than shapes associated with neutral character (median = 2.18, 95% HDI = [1.90 2.48]), $median_{diff}$ = 0.31, 95% HDI [0.02 0.63] , but we did not find differences between shapes associated with bad character (median = 2.23, 95% HDI = [1.94 2.53]) and neutral character, $median_{diff}$ = 0.05, 95% HDI [-0.29 0.37].

For the reaction times, we also found 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 = 578.64 ms, 95% HDI = [508.15 661.14]) than shapes associated with neutral character (median = 623.45 ms, 95% HDI = [547.98 708.24]), $median_{diff}$ = -44.05, 95% HDI [-59.96 -30.43]. We also found that the responses to shapes associated with bad character (median = 640.41 ms, 95% HDI = [559.94 719.63]) were slower as compared to the neutral character, $median_{diff}$ = 17.04, 95% HDI [4.02 29.92]. See Figure 1.

For the nonmatch trials, we also found the advantage of good character: Shapes associated with good character (median = 653.21 ms, 95% HDI = [574.65 739.57]) are faster than shapes associated with neutral (median = 671.14 ms, 95% HDI = [591.71 760.09]), $median_{diff}$ = -17.65 ms, 95% HDI [-23.85 -10.36]. Similarly, the shapes associated with bad character (median = 676.35 ms, 95% HDI = [599.13 767.76]) was responded slower than shapes associated with neutral character, $median_{diff}$ = 17.04 ms, 95% HDI [4.02 29.92], but the effect size was smaller, (**see supplementary materials**).

Self-referential process modulate prioritization of good character

In this part, we report results from three experiments (3a, 3b, and 6b) that aimed at testing whether the moral valence effect found in the previous experiments is modulated by self-referential processes. These three experiments included data from 108 participants.

Because we have found that a facilitation effect of good character and slow-down effect of bad character in the first part, in this part, we will focus on the whether such effect interact with self-referential factor. In others words, we not only reported differences between good/bad character with neutral character for self-referential and other-referential separately, but also compare the differences between the difference.

For the d prime, we found that an interaction between moral character effect and self-referential, the self- and other-referential difference was greater than zero for good vs. neutral character differences ($median_{diff}$ = 0.51; 95% HDI = [-1.48 2.61]) but not for bad vs. neutral differences ($median_{diff}$ = -0.02; 95% HDI = [-1.85 2.17]). Further analyses revealed that the good vs. neutral character effect only appeared for self-referential conditions but not other-referential conditions. The estimated d prime for good-self was greater than neutral-self ($median_{diff}$ = 0.56; 95% HDI = [-1.05 2.15]), d prime for good-self was also greater than good-other condition ($median_{diff}$ = ; 95% HDI = [ ]). The differences between bad-self and neutral-self, good-other and neutral-other, bad-other and neutral-other are all centered around zero (see Figure 2, B, D).

For the RTs part, we also found the interaction between moral character and self-referential, the self- and other-referential differences was below zero for the good vs. neutral differences ($median_{diff}$ = -105.39; 95% HDI = [-533.16 281.69]) but not for the bad vs. neutral differences ($median_{diff}$ = -9.46; 95% HDI = [-290.72 251.38]). Further analyses revealed a robust good-self prioritization effect as compared to neutral-self ($median_{diff}$ = -47.58; 95% HDI = [-202.88 16.83]) and good-other ($median_{diff}$ = -57.14; 95% HDI = [-991.89 621.29]) conditions. Also, we found that both good character and bad character were responded slower than neutral character when it was other-referential. See Figure 2.

Binding the good and self

In this part, we reported two studies in which the moral valence or the self-referential processing is not task-relevant. We are interested in testing whether the task-relevance modulated the effect observed in previous experiment.

In experiment 4a, where self- and other-referential were task-relevant and moral character are task-irrelevant. We found self-related conditions were performed better than other-related conditions, on both *d* prime and reaction times. This pattern is consistent with previous studies (e.g., Sui, He, and Humphreys (2012)).

More importantly, we found evidence, albeit weak, that task-irrelevant moral character also played an role. For shapes associated with self, *d'* was greater when shapes had a good character inside the shape (median = 2.83, 95% HDI [2.63 3.01]) than shapes that have neutral character (median = 2.74, 95% HDI [2.58 2.95], BF = 4.4) or bad character (median = 2.76, 95% HDI [2.56 2.95], 3.1), but we did not found difference between shapes with bad character and neutral character inside for the self-referential shapes. For shapes associated with other, the results of *d'* revealed a reversed pattern to the self-referential condition: *d* prime was smaller when shapes had a good character inside (median = 1.87, 95% HDI [1.71 2.04]) than had neutral (median = 1.96, 95% HDI [1.80 2.14]) or bad character (median = 1.98, 95% HDI [1.79 2.17]) inside. See Figure 3.

The same pattern was found for RTs. For self-referential condition, when good character was presented as a task-irrelevant stimuli, the responds (median = 641, 95% HDI [623 662]) were faster than when neutral character (median = 649, 95% HDI [631 668]) or bad character (median = 648, 95% HDI [628 667]) were inside. This effect was reversed for other-referential condition: shapes associated with other with good character inside (median = 733, 95% HDI [711 754]) were slower than with neutral character (median = 721, 95% HDI [702 741]) or bad character (median = 718, 95% HDI [696 740]) inside.

In experiment 4b, moral character was the task-relevant factor, and we found that there were main effect of moral character: shapes associated with good character were performed better than other-related conditions, on both *d'* and reaction times.

Most importantly, we found evidence that task-irrelevant self-referential process also played an role. For shapes associated with good person, the *d* prime was greater when shapes had an "self" inside than with "other" inside ($mean_{diff}$ = 0.14, 95% credible intervals [-0.02, 0.31], BF = 12.07, p = 0.92), but this effect did not happen when the target shape where associated with "neutral" ($mean_{diff}$ = 0.04, 95% CI [-.11, .18]) or "bad" person ($mean_{diff}$ = -.05, 95% CI[-.18, .09]).

The same trend appeared for the RT data. For shapes associated with good person, with a "self" inside the shape reduced the reaction times as compared with when a "other" inside the shape ($mean_{diff}$ = -55 ms, 95%CI[-75, -35]), but this effect did not occur when the shapes were associated neutral ($mean_{diff}$ = 10, 95% CI [1, 20]) or bad ($mean_{diff}$ = 5, 95%CI [-16, 27]) person. See Figure 3.

Self-reported personal distance

We explored the self-reported psychological distance between different person. Participants were presented a pair of two-person each time, and moved a slide to represent the distance between the pair of two persons. We found that, on average, participants rated self is closest to a neutral person, and then a good person. These two are not different from each other. However, both are closer than the distance between good person and neutral person. On average, participants rated themselves has furthest distance to bad person.

Correlation analysis showed that most psychological distance ratings were positively correlated to each other, but the self-bad and self-good are negatively correlated.

[use the network view to visualize the distance]

See Figure 4 and Figure 5.

Discussion

We human inevitably view others and ourselves in a moral lens. Previous studies assume that the principle "bad is stronger than good" also applies to moral character in information processing and assumed that bad moral character are prioritized. Across nine experiments, we studied the processing of moral character using a social associative learning task, we examined the effect of moral character on a matching task and the mechanism underlying the effect. We find evidence that good character are prioritized in the matching task, regardless of the label used for moral characters or the way stimuli are presented. We documented that this positivity effect was driven by a self-referential processing: prioritization only occur when moral characters are self-referential but not stranger-referential. The prioritization effect only occur when self-referential and good character are combined, whether task-relevant or not. When good character were combined with other, the information process was slowed down.

Our studies contribute to the person perception literature in XXX ways. First, we found the negativity effect in impression formation is not applicable in a perceptual-matching task.

Second, we found that self-referential process is important in the prioritization of good character.

Third, we found that either moral true self view or social-categorization view alone could not explain the results. Combine the moral-self-based categorization may help explaining the findings here.

Moreover, we found discrepancy between self-reported distance and the reaction time, suggesting that self-reported the results may be ….

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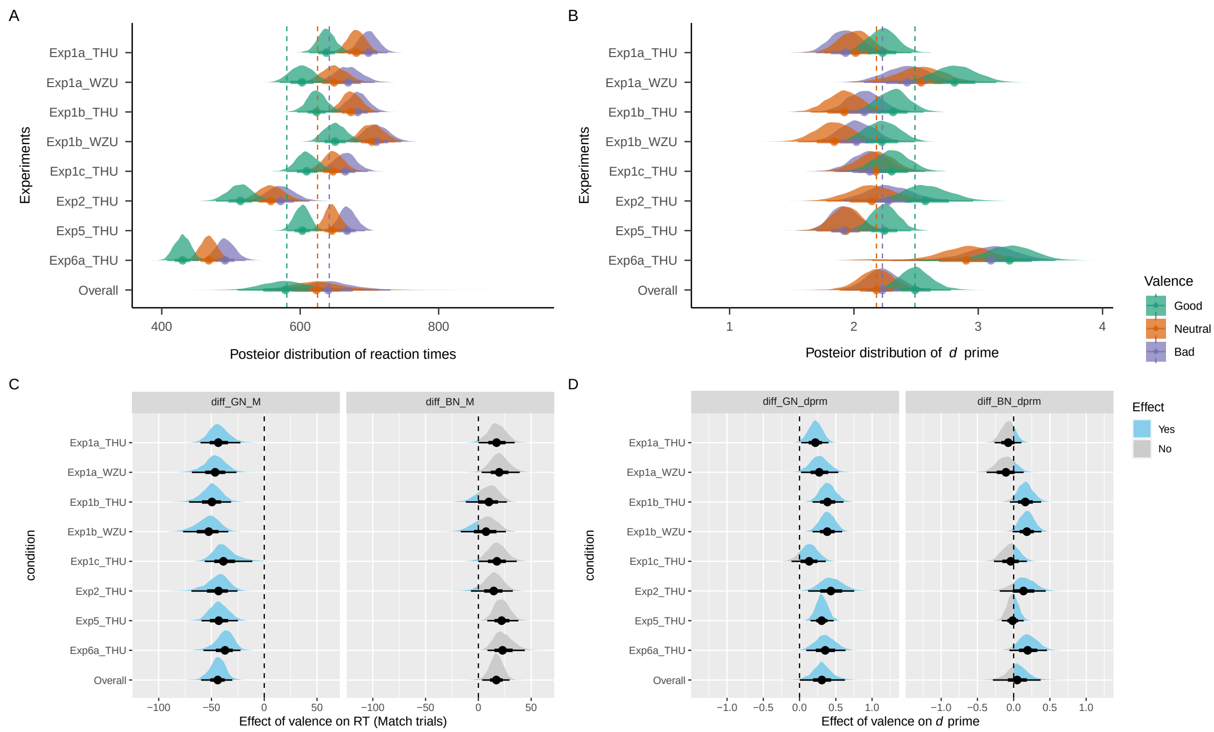

Figure 1. Effect of moral valence on RT and d’

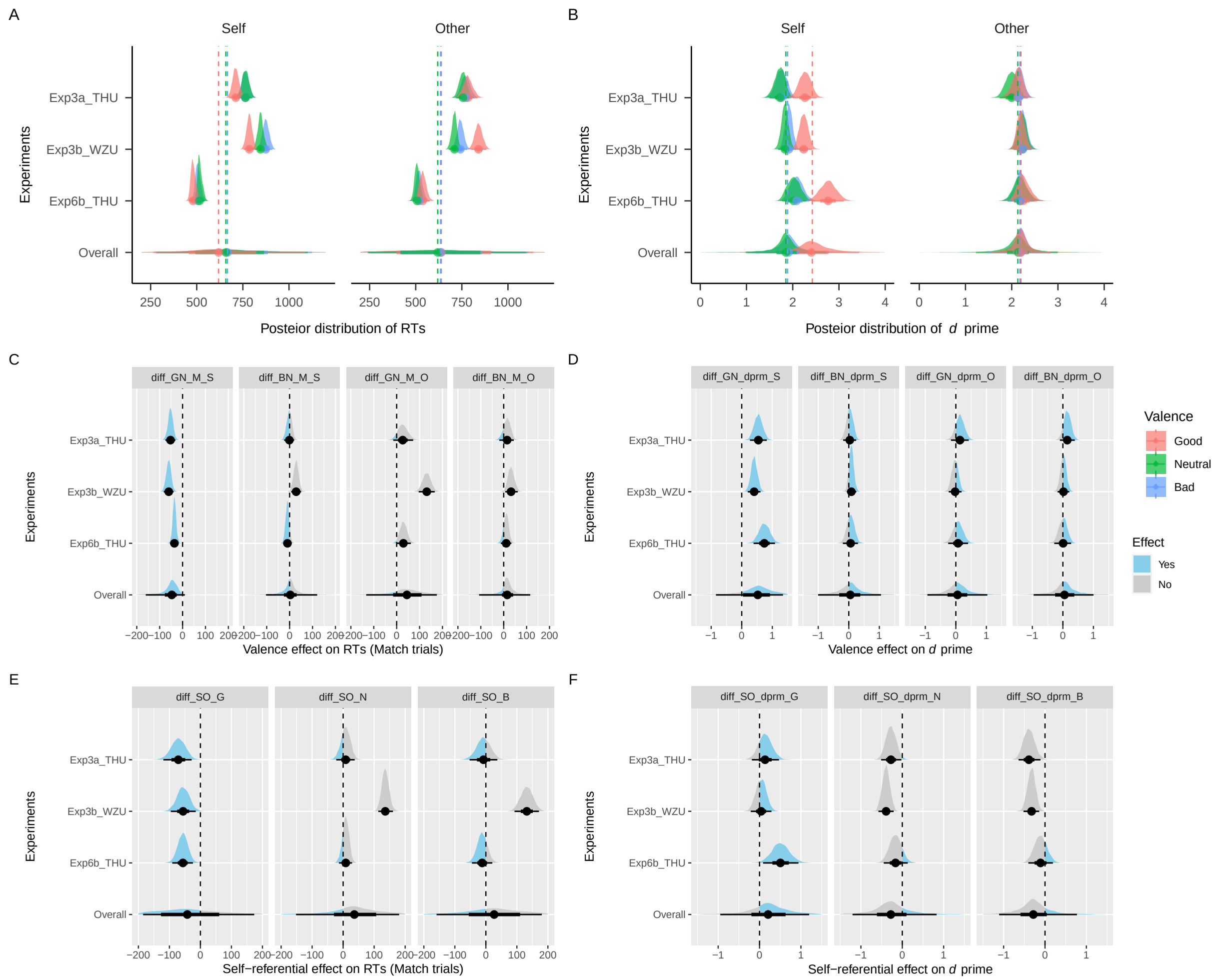

Figure 2. Interaction between moral valence and self-referential

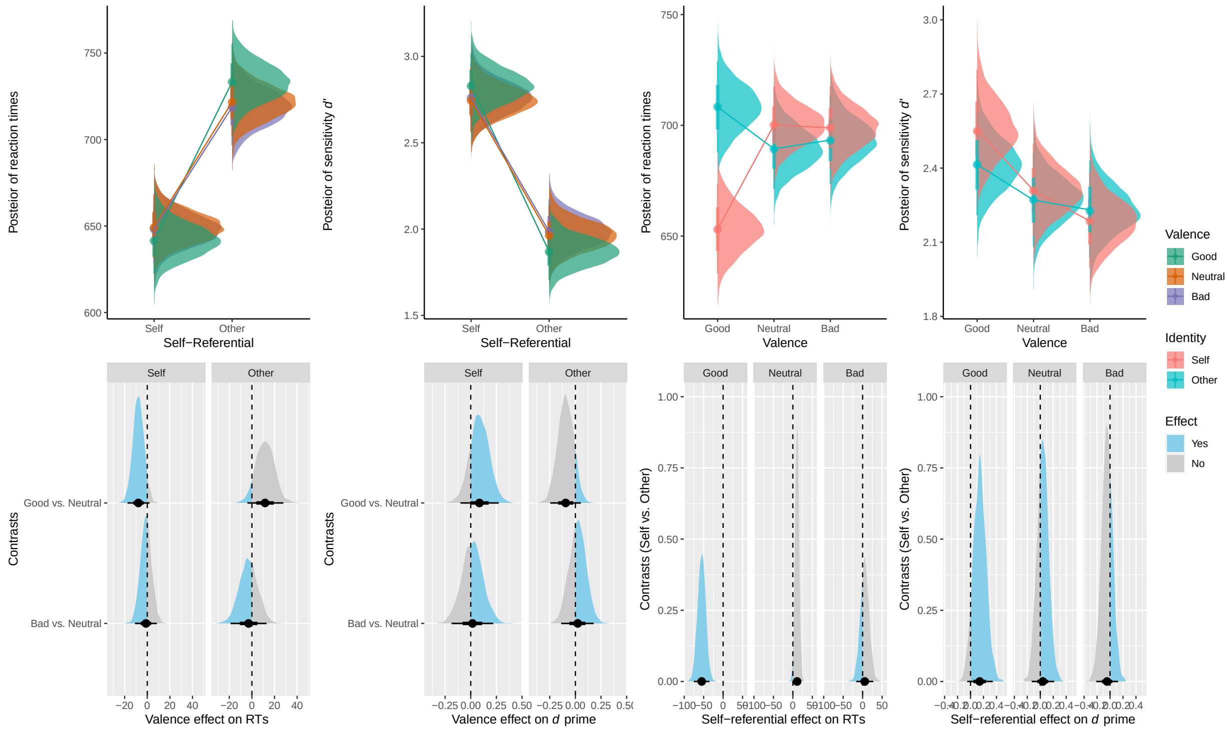

Figure 3. exp4: Results of Bayesian GLM analysis.

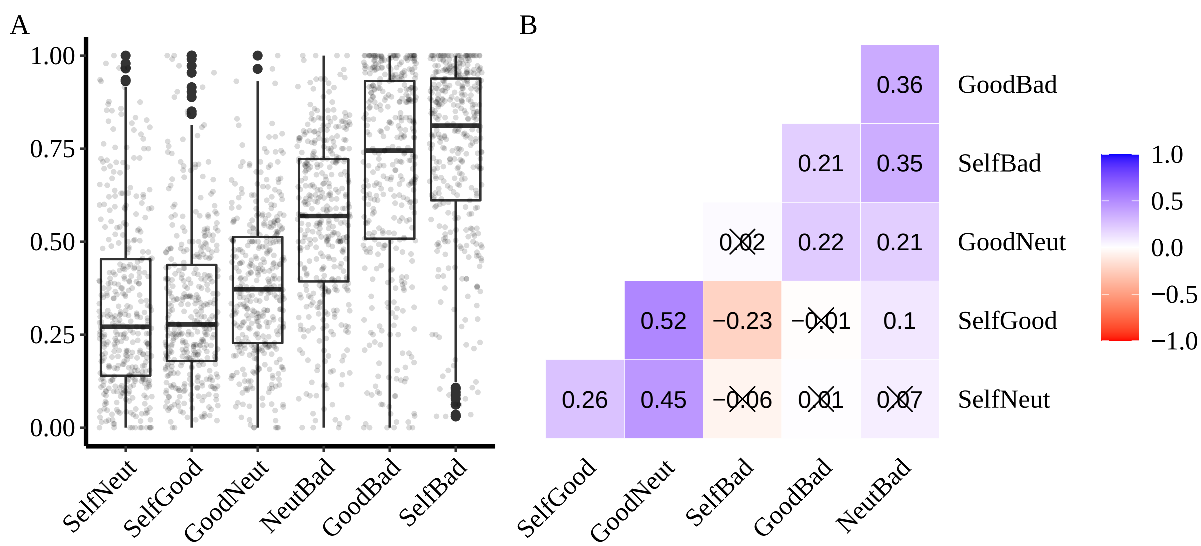

Figure 4. Self-rated personal distance

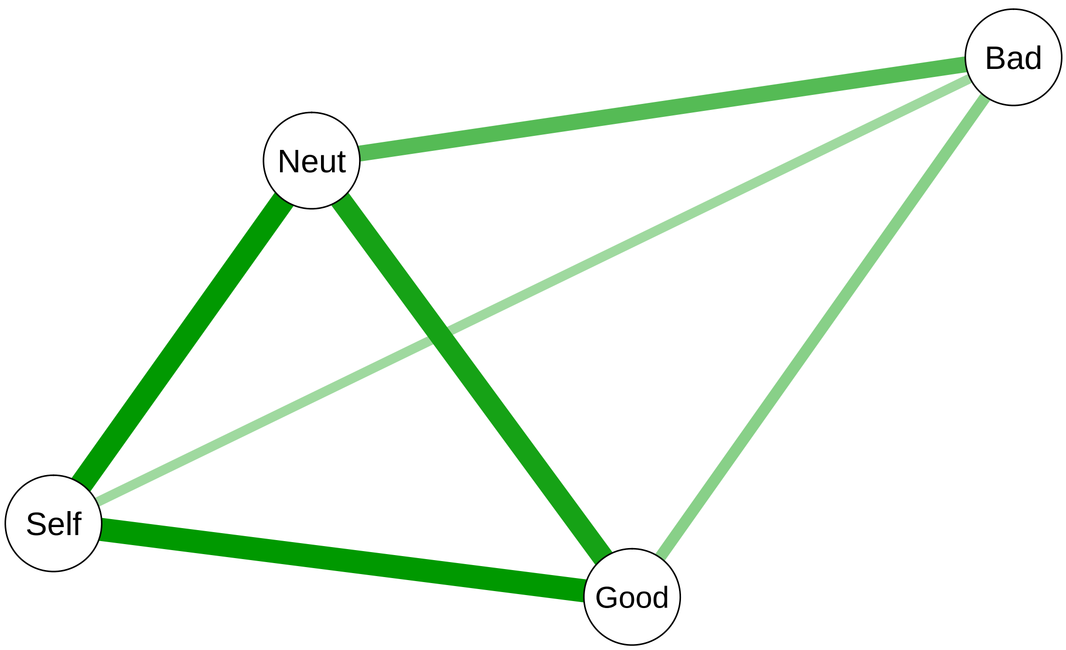

Figure 5. Self-rated personal distance (Network view)