Behavioral and neural evidence that robots are implicitly perceived as a threat

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Abstract

A deeper understanding of the human side of human-robot interaction determined by studying the

human brain when we perceive robots should help solve the biggest challenges of successful social

encounters with robots. However, current social neuroscience studies mainly focus on explicit

perception of robots, and implicit perception of robots is rather unexplored. Here, our behavioral

analysis indicated that despite self-reported positive attitudes, participants had negative implicit

attitudes toward humanoid robots. Our neuroimaging analysis indicated that subthreshold presentation

of humanoid robot vs. human images led to significant left amygdala activation that was associated

with negative implicit attitudes toward robots. After successfully weakening the negative attitudes, the

left amygdala response to subthreshold presentation of humanoid robot images decreased, and the

decrease in the left amygdala response was positively associated with the decrease in negative attitudes.

Our results reveal that the processing of information about humanoid robots displays automaticity with

regard to the recruitment of amygdala activation. Our findings that people may implicitly perceive

humanoid robots as a threat may guide more appropriate interaction with social robots.

Key words: robots; threat; implicit perception; amygdala

Significance statement

Social interactions with robots are one of the biggest challenges in robotics, which necessitates a

deeper understanding of how people perceive robots. Our results reveal automaticity for processing

information about humanoid robots similar to that previously evident for threats. Given the effort

currently being put into the development of robots for daily assistance, studying implicit perception of

robots could be a step toward building smooth human-robot social relationships.

Introduction

Given that robots are becoming increasingly present in the everyday environment and that social interactions with robots are one of the biggest challenges in robotics [1], a deeper understanding of how people perceive robots is necessary [2]. This is especially relevant given the effort currently being put into the development of robots for daily assistance [3, 4]. When a person perceives a social stimulus, information about that stimulus is immediately and spontaneously activated, including attitudes [5] and social stereotypes [6], and can consequently influence people's behavior [7]. Similarly, perceptions toward robots might automatically influence how we interact with robots. Studying the human brain with neuroimaging techniques when we perceive robots will shed light on a deeper understanding of the human side of human-robot interaction [8]. However, current social neuroscience studies mainly focus on explicit perception of robots [8, 9], and the implicit perception of robots is rather unexplored.

Researchers have proposed that our ideas about robots are mainly informed by science fiction media [10, 11], and robots in science fiction are generally described as a threat [10]. Recent statements from some influential industry leaders have strengthened these fears [12]. These ideas that robots tend to be dangerous entities that will threaten the survival of humanity might be rooted in our long-term socialization experiences, leading to the possibility that people might perceive robots as a threat. Empirical evidence has illustrated that people's explicit perception toward robots is inconsistent, ranging from enthusiasm to fear and anxiety [11, 13-18], possibly because people's explicit perceptions or attitudes are subject to recall bias and social desirability bias [19]. Notably, implicit perception is considered to be a highly stable evaluative representation stemming from long-term socialization experiences [20]. The most important feature is that implicit perception is irrespective of whether a person consciously considers the stimulus as good or bad [21]. Decades of work have identified implicit

perception as having a crucial influence on thoughts and actions [22]. Here, we aimed to provide evidence

from implicit behavioral and neural measures to support the possibility that people might implicitly

perceive robots as a threat.

To accurately assess people's implicit perception of robots, it is important to implement implicit

multimethod approaches. The implicit association test (IAT) is the most frequently used implicit

approach [23] and is based on the response given by a participant when performing an association task [24].

IATs are commonly used to explore implicit attitudes toward sensitive issues, such as racism [25].

Furthermore, neuroimaging approaches might be more sensitive, as they reveal qualitative differences

in processing without the need for distinct behavioral measures of implicit processing [22]. The

preparedness model [26, 27] hypothesizes that the neural basis of automatic threat processing is the

amygdala. The amygdala has long been known to play a key role in responding to emotionally relevant

stimuli, activating in response to images containing threatening or highly arousing features [28, 29]. When

automatic and controlled evaluations of threats differ, more positive controlled processing can

moderate more negative automatic processing [30], which could account for the absence of significant

activation in the amygdala in response to suprathreshold presentations of threats [31, 32]. Notably,

although controlled processing can eliminate amygdala activation caused by consciously presented

threats, the amygdala still showed significant activation caused by threats that were presented

unconsciously [30-32]. We hypothesized that if humanoid robots are implicitly perceived as a threat, then

participants would be expected to show negative implicit attitudes toward robots and an amygdala

response to subthreshold presentation of robot images.

How people implicitly perceive humanoid robots is an intriguing question. People's perception of

robots relies on the robots' behaviors, abilities and appearance [33]. More human characteristics are

consistently perceived in robots that look and act like humans [34]. Here, we tested whether people

implicitly perceive humanoid robots as a threat. We pursued this topic by first measuring participants'

explicit and implicit attitudes toward humanoid robots. We then employed functional magnetic

resonance imaging (fMRI) and intervention to test the amygdala response to humanoid robots and its

causal relationship to implicit attitudes.

Material and Methods

Participants. One hundred fifty-two participants (age: 21.75±2.58 years; 61 females) took part in the

original implicit association test (IAT), 22 participants (age: 22.77±1.97 years; 14 females) performed

the Black IAT and the mixed IAT, 30 participants (age: 22.23±1.89 years; 20 females) performed the

positive adjective IAT and the available humanoid IAT, and 38 participants (age: 21.77±1.67 years; 23

females) performed the humanoid weapon IAT and pet-robot weapon IAT. Ninety-nine participants

took part in the intervention experiment. Among them, 69 participants were allocated to the positive

intervention group, and 30 participants were allocated to the neutral intervention group. Ten

participants in the positive intervention group were excluded because of excessive head movement

during scanning (TRs with motion over 0.3 mm were censored, and participants who had more than 15%

of their TRs censored were removed from further analysis); 36 participants discontinued the positive

intervention after completing preintervention tasks due to personal reasons; and 3 participants

discontinued the neutral intervention after completing the preintervention tasks because of withdrawal.

Finally, data from 23 participants (age: 20.39±2.06 years; 15 females) in the positive intervention group

and 27 participants (age: 22.13±1.49 years; 18 females) in the neutral group were analyzed. All

participants were right-handed, with normal or corrected-to-normal vision, without alcohol or drug

dependence disorders and without prior head injury. All participants and their immediate relatives had

no psychiatric disorders or history of psychiatric disorders. All participants were nonpsychology

undergraduates, and they were unaware of the purpose and contents of the study before they

participated in the experiment.

The sample size was determined using a medium effect size (Cohen's d = 0.5) [35]. Power analyses

using G*power suggested a sample size of 34 for 0.8 power. This sample size was implemented for all

tasks except for the evaluating conditioning task. We performed power analyses using G*power with a

priori effect size (Cohen's d = 0.55) and 0.8 power to determine the sample size for a one-tailed

paired-sample t test, resulting in a sample size of 22 for the evaluating conditioning task. The study was

approved by the Research Ethics Committee of the University of Science and Technology of China

(NO. 2020-N(H)-099), and written informed consent was obtained from all participants consistent with

the Declaration of Helsinki. The methods were carried out in accordance with the approved guidelines.

Robot-related questionnaires and scales

Participants were asked to rate their perceptions toward humanoid robots on a 5-point questionnaire (1-

completely disagree, 5-completely agree; Table 1). Items 4, 5 and 6 were reversed items. The sum score

of these six items indicates participants' opinion toward humanoid robots. We also obtained

participants' explicit attitudes toward robots using the Negative Attitudes Toward Robots Scale (NARS)

[36].

We determined participants' prior experience with robots by asking about the frequency of daily

interactions with robots and which of 20 fictional films (*i.e.*, *Wall-E*, *The Terminator*, *Alita: Battle*

Angel, *I, Robot*, *AI: Artificial Intelligence*, *Chappie*, *Bicentennial Man*, *RoboCop*, *Short Circuit*, *Ex*

Machina, *Edward Scissorhands*, *Blade Runner*, *My Robot Girlfriend*, *Autómata*, *Real Steel*, *2001: A*

Space Odyssey, *Moon*, *Star Wars*, *The Surrogates*, *Forbidden Planet*) portraying robots they have seen.

Original implicit association test. We assessed participants' automatic attitude using a standard

implicit attitude measure paradigm — the IAT task [37, 38]. In the original IAT (Fig. 3a), stimuli consisted

of attribute words and concept images. The word stimuli were six adjectives, including three

adjectives—threatening, strikes and lethal—that indicate "threatening" and three adjectives—neutral,

normal and average—that indicate "nonthreatening" [39]. The image stimuli were 20 images of fictional

humanoid robots and 20 images of Caucasian individuals (Caucasian individuals were used because the

fictional humanoid robot forms were Caucasian; we matched the action of each pair of images, and the

gender ratio of fictional humanoid robots and humans was the same; the stimuli are available at

https://rec.ustc.edu.cn/share/1f97a050-7745-11ec-bbe8-61ef14e3189d). To avoid confounding of

character stereotypes, stimuli from the movies were excluded. These images were downloaded from the

open source on the internet. Adobe Photoshop CS5 was used for all images to uniformly process the

pixels and colors of the images so that the materials were consistent and the pixel size was 300×300.

Before the formal experiment, participants were presented with an image set including all 20 images of

fictional humanoid robots and 20 images of Caucasian individuals in a pilot experiment. For each

image, participants provided a rating of affect valence (1 = extremely negative; 5 = extremely positive),

arousal (1 = extremely calm; 5 = extremely aroused) and novelty (1 = not novel at all, 9 = extremely

novel). The results showed that the affect valence and arousal for fictional humanoid robot images were

neutral (valence: mean = 3.05, SD = 0.69; arousal: mean = 2.94, SD = 0.90). The affect valence and

arousal for Caucasian images were neutral as well (valence: mean = 3.67, SD = 0.66; arousal: mean =

3.20, SD = 1.03). There was a significant difference in novelty between fictional humanoid robot

images and Caucasian images (fictional humanoid robot: mean = 5.64, SD = 0.362; Caucasian: mean =

3.95, SD = 0.31; $t24 = 16.99$, $p < 0.001$).

The original IAT involved a series of five discrimination blocks.

Block 1 was a concept discrimination task. In this block, the concept labels (humanoid or human)

were displayed on the top left and right sides of the screen, and participants were told to categorize the

concept images (humanoid robot images or human images) by pressing a corresponding button. Each

trial began with the presentation of a fixation cross for 1 s followed by a target image remaining on the

screen until the participant responded or for a maximum of 3 s. There were 40 trials in this block, and

each concept image was presented 1 time. The order of stimulus presentation was pseudorandomized

across participants in all blocks. The label positions (left or right) of the concepts (humanoid or human)

were counterbalanced between participants.

Block 2 was an attribute discrimination task. The participants were required to categorize the

attribute words according to the attribute labels (threat or nonthreat), and the experimental procedure

was consistent with the concept discrimination task. There were 18 trials in this block, and each

attribute word was repeated 3 times.

Block 3 was a compatible combined task (humanoid + threat; human + nonthreat). In this block,

humanoid robot images and threatening words shared the same response, and human images and

nonthreatening words shared the same response. There were 76 trials in this block; each concept image

was presented 1 time, and each attribute word was repeated 6 times.

Block 4 was a reversed attribute discrimination task. Participants performed the same task as in

block 2, but the positions of the attribute labels (threatening words or nonthreatening words) were

reversed. There were 18 trials in this block, and each attribute word was repeated 3 times.

Block 5 was an incompatible combined task. Participants performed the same task as in block 3,

but the position of the attribute labels (threat or nonthreat) were swapped (humanoid + nonthreat;

human + threat). There were 76 trials in this block; each concept image was presented 1 time, and each

attribute word was repeated 6 times.

Brief instructions were presented on the screen at the beginning of each block until the participant

responded.

Before analyzing the data from the IAT, the following data reduction procedure was applied [37, 38]:

1) eliminate data from subjects for whom more than 10% of trials have a latency shorter than 300 ms; 2)

eliminate data from subjects for whom more than 20% of trials are incorrect; 3) compute the mean

latencies for only correct trials for blocks 3 and 5; 4) compute the pooled standard deviation for all

trials in blocks 3 and 5; 5) replace each incorrect trial's latency in blocks 3 and 5 with the block mean

computed in step 3 plus 600 ms; 6) compute the difference between the mean latencies of blocks 3 and

5 using paired sample t test; 7) divide the difference computed in step 6 by the pooled standard

deviation computed in step 4 to calculate the IAT effect size - the IAT score.

Black IAT. The procedure of the Black IAT was the same as that of the original IAT except the 20

images of Caucasian individuals were replaced with 20 images of Black individuals.

Mixed IAT. The procedure of the mixed IAT was the same as that of the original IAT except the 20

images of Caucasian individuals were replaced with 20 images of Black, Caucasian and Asian

individuals.

Available humanoid IAT. The procedure of the available humanoid IAT was the same as that of the

original IAT except the 20 images of fictional humanoid robots were replaced with 20 images of the

currently available humanoid robots (*e.g.*, Pepper, Nao, and iCub).

Positive adjective IAT. The procedure of the positive adjective IAT was the same as that of the

original IAT except the threat-related adjectives were replaced with positive adjectives (*i.e.*, salient,

arousing, and affective).

The humanoid weapon IAT. The procedure of the humanoid weapon IAT was the same that of as the

original IAT except the 20 images of Caucasian individuals were replaced with 20 images of weapons

and the concept label of "human" was replaced with "weapon".

Pet-robot weapon IAT. The procedure of the pet-robot weapon IAT was the same as that of the

humanoid weapon IAT except the 20 images of fictional humanoid robots were replaced with 20

images of pet robots.

Procedures in the intervention experiment. In this experiment, participants performed the original

IAT before and after the intervention. In the positive intervention group, the positive evaluative

conditioning task was used to intervene with implicit attitudes, whereas the neutral evaluative

conditioning task was used in the neutral intervention group. In the positive intervention group,

participants completed a backward masking task with fMRI scanning and a forced-choice detection

task before intervention and then completed the backward masking task with fMRI scanning again after

intervention (Fig. 2c). In the neutral intervention group, participants completed the positive adjective

IAT and the available humanoid IAT before intervention. The implicit association test and evaluating

conditioning task were presented by E-Prime 2.0, and the backward masking task and forced-choice

detection task were presented by MATLAB (R2015a).

Materials in the fMRI experiment. The materials of the backward masking task and forced-choice

detection task consisted of 20 images of fictional humanoid robots and 20 images of Caucasian

individuals, the same images used in the original IAT. The materials of the evaluative conditioning task

consisted of conditioned stimuli (CSs), unconditioned stimuli (USs), target stimuli and neutral fillers

(Table 2). The CSs consisted of 20 images of fictional humanoid robots and 20 images of Caucasian

individuals. Approximately half of the USs, target stimuli and neutral fillers were images, and half were

words. We selected the US images, target images and neutral filler images from the Chinese Affective

Image System (CAPS), including 3 target images, 5 positive images (USs), 5 neutral images (USs), and

16 neutral filler images. Adobe Photoshop CS5 was used for all images to uniformly process the pixels

and colors of the images so that the materials were consistent and the pixel size was 300×300. The

valence, arousal and dominance scores for these 29 images are displayed in Table 3 (on a scale from 1

= low valence/arousal/dominance to 9 = high valence/arousal/dominance). The word stimuli–3 target

adjectives, 5 positive adjectives (USs) [40], 5 neutral adjectives (USs) [41] and 17 neutral filler nouns

[41]–were selected from the pilot study. We translated these words into Chinese characters by using

Collins COBUILD Advanced Learner's English-Chinese Dictionary [42]. The stimuli for this task are

shown in Table 4.

Backward masking task. Participants completed the backward masking task in the MRI scanner (Fig.

2a-b) [43]. For the subthreshold presentation, the target image was presented for 17 ms followed by a

mask for 183 ms and a fixation cross for 1800 ms. For the suprathreshold presentation, the target image

was presented for 200 ms followed by a fixation cross for 1800 ms. These images were arranged in a

block design consisting of 10 images (either humanoid robot or human) in a computer-generated

pseudorandom order. To avoid any possible effects on subthreshold presentation, the six suprathreshold

blocks (three humanoid blocks and three human blocks) were presented after the six subthreshold

blocks (three humanoid blocks and three human blocks). Except for the first and last baseline blocks (a

fixation cross displayed on the screen) lasting for 10 s, each target block was separated by a 20 s

baseline block.

Forced-choice detection task. To confirm that participants were aware of the stimulus during

suprathreshold presentation but not during subthreshold presentation in the backward masking task, a

forced-choice detection task was used. The forced-choice detection task consisted of 80 trials; the first

half were subthreshold presentations, and the second half were suprathreshold presentations. The

stimuli were presented similarly to those in the backward masking task. The difference was that a 2000

ms forced-choice phase followed the target stimuli. Participants were informed that the target stimulus

could have humanoid or human images and were told to recognize the content of each image. Data

from two participants in this task were excluded because of program crashes. We compared the

accuracy, response rate, and reaction time between suprathreshold and subthreshold presentations

separately using a paired sample t test. We also compared the accuracy for both conditions to chance

level (50%) using a one-sample t test.

Positive evaluative conditioning task.   The evaluative conditioning task is a classic paradigm of

changing implicit attitudes by pairing CSs and USs [44]. In the positive evaluative conditioning task, the

participants' implicit negative attitude toward humanoid robots may be reduced by pairing the

humanoid robot images (CSs) with the positive stimuli (USs) (Fig. 4b). As a control condition, the

human images (CSs) were paired with neutral stimuli (USs). Participants were unaware of the repeated

conditioned stimulus–unconditioned stimulus (CS-US) [45].

This task included 6 blocks of 61 trials each. For each block, the specific stimuli were arranged,

and all stimuli were presented in pseudorandom order. All stimuli appeared for 1.5 s each. During the

experiment, the participants were instructed to view a stream of images or words and respond as soon

as possible whenever a prespecified target image or words appeared. To ensure that the participants

carefully viewed the US-CS pairs during the entire experiment, the participants were told that the

accuracy rate had to be at least 95% after the end of the experiment or they would have to restart the

task. Before the formal evaluative conditioning task, a training evaluative conditioning task was

performed.

Neutral evaluative conditioning task.   The procedure of the neutral evaluative conditioning

task was the same as that for the positive evaluating conditioning task except that the humanoid robot

and human images were both paired with the neutral stimuli.

MRI acquisition. Gradient echo-planar imaging data were acquired using a 3.0 T GE Discovery

MR750 with a circularly polarized head coil at the Information Science Center of the University of

Science and Technology of China. A T2*-weighted echo-planar imaging sequence (FOV = 240 mm, TE

= 30 ms, TR = 2000 ms, flip angle = 85°, matrix = 64 × 64) with 33 axial slices (no gaps, voxel size:

3.75 × 3.75 × 3.7 mm^3) covering the whole brain was used to acquire the functional MR images.

High-resolution T1-weighted spin□echo imaging data were also acquired for anatomical overlays, and

three-dimensional gradient-echo imaging data were acquired for stereotaxic transformations after

functional scanning. Before entering the scanner, participants were instructed to keep their heads still

during all scans. Participants were placed in a light head restraint within the scanner to limit head

movement. Visual stimuli were projected on a screen and viewed through a mirror attached to the head

coil. During the backward masking task, 4 functional scan runs occurred, each lasting 4 min. There was

an interval of approximately 1 min between every two runs.

fMRI processing. Functional data were analyzed using the Analysis of Functional NeuroImages

(version: AFNI_21.01.01) software. Time series were realigned to the second volume. The realigned

images were normalized to the Talairach coordinate. Raw data were corrected for temporal shifts

between slices and for motion (TRs with motion over 0.3 mm were censored, and participants who had

more than 15% of their TRs censored were removed from further analysis), spatially smoothed with a

Gaussian kernel (full width at half maximum = 8 mm), and temporally normalized (for each voxel, the

signal of each volume was divided by the temporally averaged signal). High-pass temporal filtering

(using a filter width of 128 s) was also applied to the data.

To elucidate neural responses that correlated with humanoid robot images and human images

under subthreshold and suprathreshold conditions, a general linear model (GLM) was used. Regressors

of interest were subthreshold humanoid blocks, subthreshold human blocks, suprathreshold humanoid

blocks and suprathreshold human blocks. The regressor of no interest was the fixation block. These

regressors were convolved with a hemodynamic response function (HRF) and simultaneously regressed

against the blood oxygenation level-dependent (BOLD) signal in each voxel. The regressors were not

orthogonalized, and there was no significant collinearity among the regressors. Six regressors for head

motion were also included. Individual contrast images were analyzed for each regressor of interest's

responses using one-sample t tests to generate statistical maps.

Region of interest (ROI) analysis. We conducted ROI analyses on the bilateral amygdala. The ROIs

for the bilateral amygdala were identified from the AAL atlas [46]. We determined parameter estimates

for each participant from the local average in a mask back-projected from the ROIs. The differential

neural responses between the humanoid robot and human conditions in the ROIs were analyzed.

Correlation analysis was performed between neural responses and implicit attitudes.

Results

Negative implicit attitudes toward humanoid robots. In the attitudes toward robots questions,

participants showed a positive explicit attitude toward humanoid robots ($t_{65} = 8.84$, $p < 0.001$, Cohen's

$d = 1.10$). In the original IAT, participants had significant IAT scores (baseline = 0.2; mean = 0.46; SD

$= 0.48$; $t_{151} = 6.79$, $p < 0.0001$, Cohen's $d = 0.54$; Fig. 3b). Participants responded faster to

combinations of "humanoid + threatening" and "human + nonthreatening" than to combinations of

"humanoid + nonthreatening" and "human + threatening" ($t_{151} = -9.64$, $p < 0.0001$, Cohen's $d = 0.79$;

Fig. 3b). Previous studies have indicated that participants showed implicit negative attitudes toward

Black individuals relative to Caucasian individuals [30, 47]. We wondered whether the automatic negative

associations to humanoid robots compared with Caucasian individuals would still survive when

compared with a biased race (*i.e.*, Black individuals) and could be generalized to common races. Thus,

we also recruited another group of participants to perform the Black IAT and the mixed IAT. Consistent

results were found. Participants responded faster in the compatible task than in the incompatible task in

the Black IAT (RT: $t_{21} = -6.40$, $p < 0.001$, Cohen's $d = 1.36$; mean IAT score = 0.61; Fig. 3c) and mixed

IAT (RT: $t_{21} = -7.33$, $p < 0.001$, Cohen's $d = 1.57$; mean IAT score = 0.60; Fig. 3c). Our IAT results

indicated that, despite the self-reported findings from the questionnaires, participants have negative

implicit attitudes toward humanoid robots.

In the positive adjective IAT, participants had no significant IAT scores (baseline = 0.2; mean =

0.17; SD = 0.37; $t_{29} = -0.45$, $p = 0.65$). Participants did not respond faster to combinations of

"humanoid + positive" and "human + neutral" than to combinations of "humanoid + neutral" and

"human + positive" ($t_{29} = -1.30$, $p = 0.21$). These results indicated that humanoid robots were not

implicitly more associated with positive words. In the available humanoid IAT, participants had

significant IAT scores (baseline = 0.2; mean = 0.45; SD = 0.62; t_{29} = 2.17, p = 0.038). Participants

responded faster to combinations of "humanoid + threatening" and "human + nonthreatening" than to

combinations of "humanoid + nonthreatening" and "human + threatening" (t_{29} = -3.62, p = 0.001,

Cohen's d = 0.66). These results indicated that participants also have negative implicit attitudes toward

currently available humanoid robots. The more fictional films the participants had seen, the less

explicitly negative attitudes toward robots they had (r = -0.439, p = 0.015). However, no significant

relationships between participants' prior experience with robots and implicit attitudes toward robots

(neither fictional humanoid robots nor currently available humanoid robots) were observed (all p >

0.60). Our results indicated that people's prior experience with robots might influence their explicit

perception of robots rather than their implicit perception of robots.

Humanoid robots may be considered more competitive with humans than pet robots are, resulting

in more negative implicit attitudes toward humanoid robots than toward pet robots. To test the possible

influence of the robot's appearance on implicit attitudes, we focused on the implicit attitude differences

between the humanoid and pet robots by using a well-known threatening stimulus (*i.e.*, weapons [48]) as

a baseline. Our results indicated that participants displayed larger IAT scores in the humanoid weapon

IAT than in the pet-robot weapon IAT (t_{37} = 3.07, p < 0.01, Cohen's d = 0.50; Fig. 4a). These results

indicated that participants have a more negative implicit attitude toward humanoid robots than toward

pet robots.

Greater left amygdala activity was induced by humanoid robot images than by human

images under subthreshold presentation. As shown in Fig. 4c, the mean response rate in the

forced-choice detection task was more than 90% under both presentations and was higher under

suprathreshold presentation (t_{58} = 4.14, p < 0.001, Cohen's d = 0.54) than under subthreshold

presentation. The response time under suprathreshold presentation was significantly shorter than that

under subthreshold presentation (t_{58}= -10.91, p < 0.001, Cohen's d = 1.42). The accuracy under

suprathreshold presentation was significantly higher than that under subthreshold presentation (t_{58}=

16.68, p < 0.001, Cohen's d = 2.17). Importantly, the accuracy under subthreshold presentation did not

differ from random chance (t_{58}= -0.31, p = 0.76), whereas the accuracy under suprathreshold

presentation was higher than random chance (t_{58}= 19.13, p < 0.001, Cohen's d = 2.49). These findings

indicate that participants are aware of the stimuli under suprathreshold presentation but are unaware of

the stimuli under subthreshold presentation.

In fMRI analysis, we first tested whether subthreshold presentation of images of humanoid robots

leads to a greater amygdala (Fig. 5a) response compared to images of humans. Activation in response

to humanoid robot images was significantly stronger than activation in response to human images in

the anatomically defined left amygdala under subthreshold presentation (left amygdala: t_{58} = 2.61, p =

0.012, Cohen's d = 0.34; right amygdala: t_{58} = 0.26, p = 0.80; Fig. 5b); no such difference was observed

under suprathreshold presentation (left amygdala: t_{58} = -0.76, p = 0.37; right amygdala: t_{58} = 0.26, p =

0.80). The activation differences between humanoid robot and human images under subthreshold

presentation in the left amygdala were significantly stronger than activation differences under

suprathreshold presentation (left amygdala: t_{58} = 2.23, p = 0.03, Cohen's d = 0.30; right amygdala: t_{58} =

1.32, p = 0.19; Fig. 5b). Importantly, a greater IAT score was associated with greater left amygdala

activity under the subthreshold condition (r = 0.46, p < 0.001; Fig. 5c). Our results indicated that

greater left amygdala activity induced by humanoid robot images compared to that induced by human

images under subthreshold presentation is associated with negative implicit attitudes toward humanoid

robots.

We found that there was a significant difference in novelty between humanoid robot images and

human images. It has been reported that novelty contributes to amygdala activation [49]. We controlled

for the novelty of images by adding this as a covariate in our fMRI general linear model analyses, and

this did not alter the results related to amygdala activation (Fig. 6a-b).

The left amygdala response to subthreshold presentation of humanoid robot images changes

after successfully weakening negative attitude. We conducted a two-factor (group factor: positive

intervention group, neutral intervention group; time factor: pretest, posttest) repeated-measures

ANOVA on IAT scores. A significant group×time interaction effect was found ($F_{(1,48)} = 4.99$, $p = 0.038$,

$\eta^2 = 0.094$, Fig. 4d). *Post hoc* analysis revealed that the posttest IAT scores were significantly smaller

than the pretest scores ($t_{22} = -2.16$, $p = 0.042$, Cohen's $d = 0.45$; Fig. 5d) in the positive intervention

group but not in the neutral intervention group ($t_{26} = -0.88$, $p = 0.39$), indicating that participants'

negative implicit attitudes toward humanoid robots had been successfully weakened by the positive

evaluative conditioning task. We tested whether the left amygdala response to subthreshold

presentation of humanoid robot images changed after successfully weakening negative implicit

attitudes. We conducted a two-factor (time factor: before modulation, after modulation; presentation

factor: subthreshold presentation, suprathreshold presentation) repeated-measures ANOVA on

activation differences between humanoid robot and human images in the left amygdala. A significant

time×presentation interaction effect was found ($F_{(1,22)} = 8.51$, $p = 0.008$), but no significant time or

presentation main effects were found (all $p > 0.29$). *Post hoc* analysis revealed that there was a

marginally significant decrease in left amygdala activation between humanoid robot and human images

under the subthreshold presentation between the posttest and pretest scans ($t_{22} = -2.02$, $p = 0.056$,

Cohen's $d = 0.38$; Fig. 5e). Correlation analysis revealed that there was a significant correlation

between IAT score changes and activation value changes in the left amygdala under subthreshold

presentation (r = 0.58, p = 0.004; Fig. 5f). These results demonstrated a causal relationship between

implicit attitudes toward humanoid robots and the left amygdala response to subthreshold presentation

of humanoid robot images. Similar to the previous section, after controlling for the novelty of images

by adding it as a covariate in our fMRI general linear model analyses, the results did not change (Fig.

6c-d).

Discussion

Despite self-reported positive attitudes in the questionnaires, participants had negative implicit

attitudes toward humanoid robots. The left amygdala response to subthreshold presentation of

humanoid robots, which was positively associated with implicit attitudes toward humanoid robots,

indicates the automatic and quick detection of humanoid robots. After successfully weakening negative

attitudes, the decrease in IAT scores was positively associated with the decrease in activation in the left

amygdala under subthreshold presentation, demonstrating a causal relationship between implicit

attitudes toward humanoid robots and the left amygdala response to subthreshold presentation of

humanoid robot images.

Our results provide evidence that humanoid robots may be implicitly perceived as a threat. People

detect and respond rapidly to threatening stimuli [50, 51]. Nonhuman primates also respond more rapidly

to threatening stimuli (at least some types) than to neutral stimuli [52, 53]. The amygdala, a subcortical

structure in the anterior temporal lobe, is located in an evolutionarily old part of the brain and is shared

by other mammals. It is assumed to be the neural basis of the hardwired "fear module" that allows us to

automatically and quickly detect threatening stimuli [27]. Studies have documented that the amygdala

responds selectively to threats, sometimes irrespective of the affective valence while using implicit

measures, such as animate entities [54-56] and depictions of humans [57-59]. Our results showing no

amygdala activity in response to suprathreshold presentation of humanoid robot images seemingly

support that people did not perceive humanoid robots as a threat. However, when automatic and

controlled evaluations of threat differ, more positive controlled processing can moderate more negative

automatic processing [30]. With the positive explicit attitudes toward humanoid robots found in the

present study, the absence of amygdala response to suprathreshold presentation of humanoid robot

images is understandable. Interestingly, although controlled processing can eliminate amygdala activity

caused by consciously presented threats, the amygdala still shows greater responses to threats that are

presented unconsciously [30-32]. Several studies suggest that the left amygdala might be specifically

involved in the processing of facial stimuli [60, 61]. It has also been reported that the left amygdala shows

less habituation to fearful stimuli than the right amygdala, which might make it more likely to capture

the blood oxygen level-dependent changes in this area [62, 63]. However, the lateralization of amygdala

activation is still controversial [64, 65]. Our present study found that greater left amygdala activity was

induced by humanoid robot images than by human images under subthreshold presentation. After

successfully weakening negative attitudes, the decrease in IAT scores was positively associated with

the decrease in activation in the left amygdala under subthreshold presentation. Our results indicate a

causal relationship between implicit attitudes toward humanoid robots and the left amygdala response

to subthreshold presentation of humanoid robot images. Note that this result did not change after

controlling for the novelty of the humanoid robot and human images. These results potentially reflect

the automaticity with rapid recruitment of the amygdala for humanoid robot-related stimuli processing.

Except for threatening stimuli, many studies have indicated that the amygdala plays an important

role in emotional processing, especially negative emotions, including disgust, sadness and pain [66, 67].

The amygdala is also activated by stimuli involving social information [68, 69]. Thus, some might argue

that the amygdala response to subthreshold presentation of humanoid robots in this study could be

explained by factors other than threat or fear. In this study, we used humanoid robots with neutral faces

and humans with neutral faces as control stimuli, which might eliminate the confounding factors of

social information, such as facial expressions and emotions. Although there was a significant difference

in novelty between humanoid robot and human images, the amygdala-related results were not altered

after controlling for novelty. Furthermore, salience is also a possibility, and consequently, humanoid

robots should be implicitly more associated with other salient adjectives (*e.g.,* positive words).

However, our results of the positive adjective IAT were not significant, indicating that humanoid robots

were not implicitly more associated with salient adjectives. Of note, the significantly positive

correlation between the decrease in the left amygdala response and the decrease in negative attitudes

indicates that there is a modulation effect of the positive evaluating conditioning task on amygdala

activity, although the reduced saliency of stimuli might also contribute to the decrease in amygdala

activity.

People perceive robots based upon context, cues, and cultural assumptions [33]. Our ideas about

robots are mainly informed by science fiction media [10], and robots in science fiction are generally

described as a super species with greater intelligence than that of humans, attempting to eliminate

humanity[10]. The concept that fully autonomous robots are dangerous competitive "living" entities that

will threaten the survival of humanity gradually becomes a deep impression. Two species with the

closest living habits have the strongest competitiveness between them [70, 71], such as Neanderthal

extinction by competition with anatomically modern humans [72-74]. Combined with our results, robots

with human-like faces and intelligence are plausibly implicitly perceived as a new species or even race.

Consequently, the fear of humanoid robots is likely similar to threat stimuli related to survival in

evolutionary history. However, whether people perceive humanoid robots as an evolutionary threat is

equivocal and is worth future research to uncover.

A growing body of evidence has indicated that the physical appearance of a robot has a strong

impact on people's perception [9, 75, 76]. However, our results suggest that a robot's appearance is likely to

influence explicit perception rather than implicit perception (at least the implicit attitudes toward

robots). Fictional humanoid robots with highly anthropomorphic appearances and the currently

available humanoid robots with less anthropomorphic appearances were used in our experiment. Of

note, participants showed negative implicit attitudes toward both kinds of humanoid robots. Consistent

with our results, previous studies have shown negative implicit attitudes toward robots by using robot

silhouettes [77, 78]. Taken together, the physical appearance of a robot plausibly has a weak impact on

people's implicit perception.

Our demonstration that people implicitly perceive humanoid robots as a threat might contribute to

some negative biases for robots. Consistent with previous studies [77, 78], we found that people have

negative implicit attitudes toward robots. A study suggests that the early top-down process of empathy

is weaker for humanoid robots than for humans [79]. A systematic review [80] of anxiety and acceptance

toward social robots reveals that people only slightly accept social robots and feel slightly anxious

about them in general. It might be that the negative perception toward robots immediately and

spontaneously activates negative attitudes and social stereotypes toward robots and consequently

causes biased behaviors. The negative perception toward robots would be detrimental to the

development of robots and successful social encounters with robots. Of note, using an evaluative

conditioning task seems to weaken this negative perception toward humanoid robots. More research

should address the issue of where negative perception toward humanoid robots comes from and how to

modulate this negative perception.

In previous studies, a particular robot-related experience might have a significant influence on

people's explicit perception of robots [36, 81, 82]. Consistent with these findings, our study found that the

more fictional films portraying robots the participants had seen, the less explicitly negative attitudes

toward robots they had. However, no significant relationships between participants' prior experience

with robots and implicit attitudes toward robots were observed. Our results indicated that people's prior

experience with robots might influence their explicit perception of robots rather than their implicit

perception of robots. Thus, future research addressing the question of the impact of a priori experience

on perception of robots might distinguish between explicit and implicit perception.

One limitation was that the humanoid robot stimuli were only in the form of images. Although

fictional and currently available humanoid robot images were used in the present study, the ecological

effect of researching human-robot interactions is somewhat weak. It is better to investigate how people

perceive and interact with robots in a socially dynamic environment. Thus, stimuli in the form of

videos and human-robot interaction research in the real world with the help of mobile neuroimaging

should be considered in future studies.

Conclusions

In summary, this study demonstrates that humanoid robots are implicitly perceived as a threat.

This sheds light on how people perceive and interact with robots that are increasingly entering our

social environment. The future of social robots is undeniably exciting, and insights from

neuropsychology research will guide the future direction of robot development and bring us closer to

interacting with social robots.

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Author contributions

ZDW, YC, and XCZ conceived and designed the study. ZDW and YC obtained the findings. YC was responsible for the acquisition of data. ZDW and YC analyzed and interpreted the data. PYZ, YP, JCR, QZ, RJZ, BSQ, YCB, SHH, CBL and DRZ provided administrative, technical, or material support. CBL and XCZ supervised the study. ZDW and YC drafted the paper, and all authors contributed to critical revision for intellectual content.

Conflicts of interest

The authors declare no conflicts of interest.

Data and materials availability statement

The data and scripts are available at https://rec.ustc.edu.cn/share/1f97a050-7745-11ec-bbe8-61ef14e3189d.

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Figure Legends

Figure 1. Overview diagram of the study flow.

Figure 2. Description of the backward masking task and the procedure for the fMRI experiment.

(a) Time setting of the backward masking task. In the unconscious condition, the target image was presented for 17 ms followed by a mask for 183 ms and a fixation cross for 1800 ms. In the conscious condition, the target image was presented for 200 ms followed by a fixation cross for 1800 ms. (b) Block design of the backward masking task. There were six unconscious blocks (three humanoid blocks and three human blocks) followed by six conscious blocks (three humanoid blocks and three human blocks). (c) Procedure for the fMRI experiment. Participants performed the original implicit association test outside the scanner and then completed a backward masking task with fMRI scanning followed by a forced-choice detection task and an evaluating conditioning task in the scanner; then, they performed the backward masking task during a second fMRI scan. Following the end of fMRI scanning, participants completed the original implicit association test outside the scanner again.

Figure 3. Implicit association test and its results. (a) Procedure of the original implicit association

test. (b) Negative implicit attitudes toward humanoid robots. In the original implicit association test

(IAT), participants responded faster to the combination of "humanoid + threat" than to the combination

of "humanoid + nonthreat", and the computed IAT scores (effect size) were significant. (c) Participants

displayed larger IAT scores in the original, Black, and mixed IATs. Plotted data represent the mean ±

SD across participants.

Figure 4. Evaluative conditioning task and its results. (a) More negative implicit attitudes toward

humanoid robots compared to animal robots. Participants displayed larger IAT scores in the humanoid

weapon IAT than in the pet-robot weapon IAT ($t_{37} = 3.07$, $p < 0.01$, Cohen's d = 0.50). (b) Procedure of

evaluative conditioning task. (c) Results of the forced-choice detection task. (d) A significant

group×time interaction effect was found, indicating that the weakened negative implicit attitudes

toward humanoid robots were not due to the practice effect. * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$.

Plotted data represent the mean ± s.e.m. across participants. IAT = implicit association test.

Figure 5. Humanoid robot image-related amygdala activity and amygdala activity changes. (a)

The region of interest for the bilateral amygdala. (b) Although no amygdala activity differences were

detected for consciously presented humanoid robot vs. human images, greater left amygdala activity

was induced by humanoid robot images than by images of humans under unconscious presentation. (c)

A greater IAT score was associated with greater left amygdala activity under unconscious conditions. (d)

Significantly smaller IAT scores were found after modulation than those before modulation, indicating

that participants' negative implicit attitude toward humanoid robots was successfully weakened. (e)

The left amygdala activity differences of humanoid robot vs. human images did change under

unconscious presentation after successfully weakening the negative implicit attitudes toward humanoid

robots. (f) There was a significant correlation between IAT score changes and activation changes in the

left amygdala under unconscious presentation. * p < 0.05, ** p < 0.01, *** p < 0.001. For b and e,

plotted data represent the mean ± SD across participants. IAT = implicit association test.

Figure 6. Results related to the left amygdala after controlling for novelty. (a) Greater left

amygdala activity was induced by humanoid robot images than by images of humans under

subthreshold presentation after controlling for the novelty of the humanoid robot and human images (t_{58}

= 2.10, p = 0.04, Cohen's d = 0.27). (b) A greater IAT score was associated with greater left amygdala

activity under the subthreshold condition after controlling for the novelty of the humanoid robot and

human images (r = 0.43, p < 0.001). (c) Controlling for the novelty of the humanoid robot and human

images, the left amygdala activity differences of humanoid robot vs. human images did change under

subthreshold presentations after successfully weakening the negative implicit attitudes toward

humanoid robots (t_{22} = -2.28, p = 0.033, Cohen's d = 0.47). (d) There was a significant correlation

between IAT score change and activation change in the left amygdala under subthreshold presentation

after controlling for the novelty of the humanoid robot and human images (r = 0.51, p = 0.014).

748 **Tables**

749 Table 1. Attitudes toward robots questions.

Items	5-point rating (1- completely disagree, 5-completely agree)
1. Humanoid robots could improve work efficiency.	
2. Humanoid robots could do jobs that human can't finish.	
3. Humanoid robots could improve the quality of life for humans.	
4. Humanoid robots would consume a lot of resources.	
5. Humanoid robots could have unexpected dangers.	
6. Humanoid robots would disrupt human life.	

750 Table 2. Stimulus of evaluative conditioning task.

Label	Stimulus
CSs	20 humanoid images, 20 human images
USs	
Positive	
Images	Chicken[a] (14[b]), Dog 2 (18), Island 1 (29), Apple (77), Cat 8 (781)
Words	Fantastic, Enjoyable, Fabulous, Excellent, Magnificent
Neutral Images	Bug 8 (234), Clothes Rack 1 (318), Plastic Cup (386), Butterfly 1 (451), Graph 2 (724)
Words	Odd, Stiff, Cold, Material, Muddy
Target	
Images	Antique 1 (292), Antique 2 (293), Antique 3 (295)
Words	Three meanings of antique
Fillers Images	Dock (424), Iron Bridge (406), Wolf 1 (547), Insect 10 (603), Plant 2 (842), Tool (785), Tree 4 (534), Locust (310), Frame (298), Shanghai 2 (387), City 4 (665), Tortoise (601), Hippo (482), Hair Drier (329), River (401), Mountain 7 (732)
Words	Bowl, Wine, Rock, Bench, Glass, Avenue, Boxer, Trunk, Rattle, Spray, Icebox, Ketchup, Radiator, Whistle, Nursery, Pamphlet, Thermometer

751 [a] CAPS image name.

752 [b] CAPS image numbers.

754 Table 3. Stimulus score of evaluative conditioning task.

	Valence		Arousal		Dominance	
	Mean	SD	Mean	SD	Mean	SD
USs						
Positive	7.27	0.18	6.25	0.45	7.11	0.42
Neutral	5.06	0.01	4.33	0.43	5.91	0.62
Filler	5.01	0.20	4.51	0.53	5.24	0.59
Target	5.24	0.08	4.10	0.16	5.59	0.01

755  Table 4. Stimulus arrangement in each block of evaluative conditioning task.

Trials	Arrangement
10	US + CS[a]
3	Target stimulus[b]
3	Target stimulus + Neutral filler
10	Neutral filler
10	Neutral filler + Neutral filler
5	Blank screen
20	Blank screen (preceding and following CS-US pairings)

756  [a] US appeared simultaneously with CS for 10 trials in each block.

757  [b] Target stimulus appeared alone for 3 trials in each block.