Effects of Social Cues on Robot's Gaze and Head Rotation on Users' Perception

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ABSTRACT

The social cues embodied by social robots may greatly affect people's impressions of them, thereby affecting the human-robot interaction (HRI) experience. This study focuses on the impact of two social cues on participants' impressions in human societies, i.e., robot eyes blinking and head rotating. The conclusion was generated as follows: (1) The social cues of social robots in specific scenes included rotating head or blinking eyes, can improve participants' perception of robots on anthropomorphism, animacy, likeability and perceived intelligence. (2) The social cues of robots will attract the attention of visitors to a certain extent, leading them to follow the robot on the tour. (3) At the same time, the ability to attract attention is limited. On the one hand, social robots need to have more social cues to bring people a feeling of being alive and intelligent. On the other hand, according to different usage scenarios, the attributes of social cues should also be different.

Keywords: Social cues, Social robot, Human-robot interaction, User experience

INTRODUCTION

Currently, service robots or social robots have been applied in many situations, such as hospitals, shopping malls, and hotels, etc. In anthropological and sociological research, it has been found that social cues, especially nonverbal social cues (NVSCs), can affect a series of different human reactions and subjective judgments (Gilboa-Schechtman and Shachar-Lavie, 2013). In general, the perception of NVSCs can be divided into faces (expressions, gaze and head tilt) (Todorov et al. 2013), voices and bodies.

In the field of robotics, social cues also dynamically bring huge differences in the experience of human-robot interaction (HRI). The interaction between humans and robots is usually dynamic, therefore, the social cues displayed by the robot should belong to a dynamic HRI. For example, social robot speech (Torre and White, 2021), facial expression changes (Danev et al. 2017), extended gaze (Fiore et al. 2013), head rotation or tilt (Mara and Appel, 2015), Even the different personalities displayed (Salem et al. 2015; Robert, 2018). These are no different from the characteristics of real human beings.

Study shows that tilt robot head has been found to look more humanlike and cutter than upright robot head (Mara and Appel, 2015), even if the social robots were static. Studies have also shown that more complex and detailed social cue feedback, such as head tilting and nodding, can lead to



Figure 1: Service Robots Provider in UAE. (Brain Quest, 2018).



Figure 2: Navigation robot for the experiment.

participants' more positive feedback (Liu et al. 2012). Besides, social gaze or eyes interaction has a significant positive effect to HRI (Zaga et al. 2017). Appropriate robot eye gaze improves task efficiency and conveys robot emotions (Ruhland et al. 2015). Despite this, there is still little research on the impact of social cues displayed by robots in specific dynamic scenarios on participants' perceptions.

METHODS

The purpose of this study was to explore the influence of social cues shown on service robots on human perception and evaluation. The experiment was conducted with a 2×2 mixed two-way ANOVA design, the research variables were robot eyes (i.e., eyes blinking vs no eyes blinking) and the robot head rotation (i.e., head rotating vs no head rotating). The robot eyes were the within-subjects factor, the head rotation was the between-subjects factor.

Participants and Materials

A total of 24 participants were invited by convenience sampling (12 males and 12 females); the number of participants with smart robot experience was 45.8%. In addition, the main age distribution of the participants was 19-30 years old (75%).

The experiment was designed as a virtual guide scene, and the service robot and scene are presented on the computer screen in the form of virtual animation. The participants needed to sit in front of the computer and observe the three scenes in turn. Participants were randomly assigned to one of the groups in the experiment. We referred to the service robot in Figure 1 and designed the experimental robot which was shown in Figure 2. The video was divided into 3 parts, including scene1, scene2 and scene3 (see Figure 3). Among them, the scene1 content of each group of videos was the same. As the introduction of each video, the robot will welcome the participants to visit the research institute.

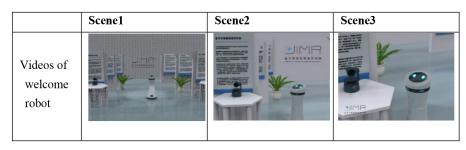


Figure 3: Introduction videos of service robot (introducing Jinhua Intelligent Manufacturing Research Institute).

Procedure

The experimental process of this study was as follows:

Step1: Fill in the basic information. The participant needed to complete the basic personal information.

Step2: Calibration of the eye tracker. Then they needed to calibrate the eye tracking device after sitting down in front of the screen in the fixed position.

Step3: Observe the videos of virtual robot to visit the exhibition hall of a research institute. The participant needed to follow the guidance of the virtual robot to visit the exhibition hall of a research institute. They did not need to operate, just watch the video, while the eye-tracking device records their eye movements. Each participant needed to watch 2 videos in turns.

Step4: Subjective evaluation questionnaire. At the end of each video, the experimenter would ask them to fill out a subjective evaluation questionnaire (Bartneck, Croft and Kulic, 2008) pertinent to this video. A 5-point Likert Scale was adopted for the questionnaire in which 1 represented most disagree and 5 indicated most agree.

It should be noted that in the treatment of the experimental group, the eyes and head rotation of the robot would move according to the voice of the robot, as if the robot would see the operation of the participant (Kelley, 1984).

RESULTS

The collected data were analyzed using a mixed two-way analysis of variance (ANOVA) in SPSS with the robot head as the within-subjects factor, and the robot eyes as the between-subjects factor. The significant value α was set to 0.05.

The perceptions of service robot scores on Anthropomorphism

The main effect of the robot eyes (F = 12.778, P = 0.002 < 0.01) and robot head (F = 18.772, P = 0.000 < 0.01) were both significant on Anthropomorphism. The robot with eyes blinking (M = 4.28, SE = 0.11) was significantly more humanlike than the robot with no eyes blinking (M = 3.49, SE = 0.16). The robot with head rotating (M = 4.39, SE = 0.11) was significantly more humanlike than the robot with no head rotating (M = 3.49, SE = 0.16) (see Table 1).

Source	SS	df	MS	F	Р	Post Hoc			
Robot eyes	5.672	1	5.672	12.778	0.002**	eyes blinking blinking	>	no	eyes
Robot head	9.630	1	9.630	18.772	0.000**	head rotating rotating	>	no	head
Robot head Robot eyes	0.333	1	0.333	0.650	0.429	0			

Table 1. The mixed two-way ANOVA of Anthropomorphism score.

* Significantly different at $\alpha = 0.05$ level (*P < 0.05).

** Significantly different at $\alpha = 0.01$ level (**P < 0.01).

Table 2. The mixed two-way ANOVA of Animacy score.

Source	SS	df	MS	F	Р	Post Hoc	
Robot eyes					0.103	11	11
Kobot head	9.630	1	9.630	35.007	0.000**	head rotating > no rotating	nead
Robot head Robot eyes	0.630	1	0.630	2.291	0.144		

* Significantly different at $\alpha = 0.05$ level (*P < 0.05).

** Significantly different at $\alpha = 0.01$ level (**P < 0.01).

Table 5. The mixed two-way ANOVA of Likeability score.										
Source	SS	df	MS	F	Р	Post I	Hoc			
Robot eyes	1.021	1	1.021	3.651	0.069					
Robot head	3.101	1	3.101	8.046	0.010*	head	rotating	>	no	head

0.001 0.002

Table 3. The mixed two-way ANOVA of Likeability score.

* Significantly different at $\alpha = 0.05$ level (*P < 0.05).

0.001

Robot head

Robot eyes

** Significantly different at $\alpha = 0.01$ level (**P < 0.01).

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The Perceptions of Service Robot Scores on Animacy

The main effect of the robot head was significant on Animacy (F = 35.007, P = 0.000 < 0.01). The robot with head rotating (M = 4.20, SE = 0.12) was significantly more animacy than that of the robot with no head rotating (M = 3.34, SE = 0.15) (see Table 2).

0.963

rotating

The Perceptions of Service Robot Scores on Likeability

The main effect of the robot head was significant on Likeability (F = 8.046, P = 0.010 < 0.05). The robot with head rotating (M = 4.44, SE = 0.10) was significantly more liked than the robot with no head rotating (M = 3.93, SE = 0.13) (see Table 3).

The Perceptions of Service Robot Scores on Perceived Intelligence

The main effect of the robot head was significant on Perceived Intelligence (F = 5.024, P = 0.035 < 0.05). The robot with head rotating (M = 4.23,

Source	SS	df	MS	F	Р	Post Hoc
Robot eyes	0.563	1	0.563	2.041	0.167	
Robot head	2.253	1	2.253	5.024	0.035*	head rotating > no head rotating
Robot head Robot eyes	0.120	1	0.120	0.268	0.610	C C

Table 4. The mixed two-way ANOVA of Perceived Intelligence score.

* Significantly different at $\alpha = 0.05$ level (*P < 0.05).

** Significantly different at $\alpha = 0.01$ level (**P < 0.01).

Treatm	no eyes blinking,	no eyes blinking, eyes blinking,		eyes blinking,	
ents	no head rotating	no head rotating	head rotating	head rotating	
Heat maps of Scene2 (5s)					
Heat maps of Scene3 (9s)					

Figure 4: The heat maps of four treatments at several key points.

SE = 0.13) was significantly more intelligent than the robot with no head rotating (M = 3.79, SE = 0.12) (see Table 4).

The Heat Maps and Average Fixation Time of Eyetracking Data

The time of four key frames of scene2 and scene3 was selected to analyze the heat maps and average fixation time of different treatments. These nodes were all where the robot blinked or rotated its head.

It could be seen from the heat map that when the robot was explaining, the participants' attention was relatively concentrated. Especially when the robot could both blink and rotate its head, the participants' attention was mostly focused on the robot's head and face compared to the robot without any movement, so the color reaches red (see Figure 4).

The robot was divided into 3 areas of interest (AOI), including AOI001 (the robot's display or face), AOI002 (the middle part that can rotate), and AOI003 (other parts) (see Figure 5).

The results of average fixation time showed that the main effect of the robot eyes and robot head were both significant on average fixation of scene2 and scene3 (see Table5 and Table 6). In addition, there was a significant interaction between the robot head and the robot eyes of scene2 (see Figure 6 and Figure 7).



Figure 5: Example of areas of interest (scene2-eyes blinking, head rotating).

Measurements			AOI001 (5s)	AOI001 (12s)
	Robot eyes		1869.65	51.49
Average fixation time		P F	0.000** 142.61	0.000** 140.18
	Robot head		0.000**	0.000**
	Robot head Robot eyes		159.82 0.000**	102.36 0.000**

Table 5. The mixed two-way ANOVA of Average fixation time on scene 2.

* Significantly different at $\alpha = 0.05$ level (*P < 0.05).

** Significantly different at $\alpha = 0.01$ level (**P < 0.01).

It could be seen that at the fifth second of scene2, the robot without head rotating and blinking (M = 59.83) received slightly higher attention than the robot with head rotating and no eyes blinking (M = 48.73). The robot with head rotating and blinking (M = 660.81) received higher attention than the robot without head rotating but blinking (M = 271.19) (see Figure 6). At the twelfth second of scene2, the robot with head rotating but no eyes blinking (M = 736.25) received higher attention than the robot without rotating head and blinking (M = 680.68). The robot with eyes blinking and head rotating (M = 1301.18) received higher attention than the robot without head rotating (M = 592.82) (see Figure 7).

At the ninth second of scene3, the robot with head rotating and without eyes blinking (M = 2510.07) received higher attention than the robot without head rotating and without eyes blinking (M = 731.68). The robot with head rotating and blinking (M = 725.85) received higher attention than the robot without head rotating but eyes blinking (M = 616.12) (see Figure 8). At the fifteenth second of scene3, the robot with head rotating but no eyes blinking (M = 2536.88) received higher attention than the robot without head rotating and without eyes blinking (M = 850.71). The robot with head rotating and eyes blinking (M = 783.66) received higher attention than the robot without head rotating but eyes blinking (M = 629.21) (see Figure 9).

DISCUSSIONS

From the experimental results, it is not difficult to find that the social cues of social robots in specific scenes, such as rotating their heads or blinking their eyes, can improve participants' perception of social robots to a certain extent, such as anthropomorphism, animacy, likeability and perceived intelligence.

Measurements			AOI001 (9s)	AOI001 (15s)
	Robot eyes		644.81	1383.24
			0.000**	0.000**
Amongo firmation time	D . 1 1 1	F	936.94	758.95
Average fixation time	Kobot nead		0.000**	0.000**
	Dahatha d Dahataaa	F	731.79	525.58
	Robot head Robot eyes		0.000**	0.000**

Table 6. The mixed two-way ANOVA of Average fixation time on scene 3.

* Significantly different at $\alpha = 0.05$ level (*P < 0.05).

** Significantly different at $\alpha = 0.01$ level (**P < 0.01).

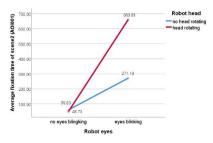


Figure 6: The interaction of Average fixation time of scene2 (AOI001-5s).

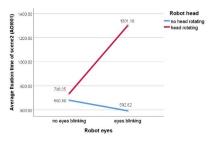


Figure 7: The interaction of Average fixation time of scene2 (AOI001-12s).

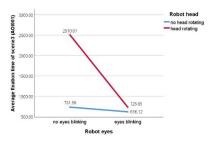


Figure 8: The interaction of Average fixation time of scene3 (AOI001-9s).

This study sets up a social robot simulation in an institute's exhibition hall, and it is believed that if we apply the scenario to other similar scenarios, the results will be similar.

At the same time, the social cues embodied by these robots will also attract the attention of visitors to a certain extent, leading them to follow the robot

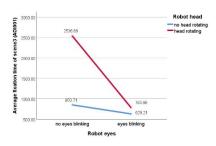


Figure 9: The interaction of Average fixation time of scene3 (AOI001-15s).

on the tour. Of course, the results also show that the ability to attract attention is limited. On the one hand, social robots need to have more social cues to bring people a feeling of being alive and intelligent. On the other hand, according to different usage scenarios, the attributes of social cues should also be different. For example, directing visitors to pay more attention to the exhibits or the exhibition hall environment, etc.

CONCLUSION AND FUTURE IMPLICATIONS

The generated results are as follows: (1) The social cues of social robots in specific scenes included rotating head or blinking eyes, can improve participants' perception of robots on anthropomorphism, animacy, likeability and perceived intelligence. (2) The social cues of robots will attract the attention of visitors to a certain extent, leading them to follow the robot on the tour. (3) At the same time, the ability to attract attention is limited. On the one hand, social robots need to have more social cues to bring people a feeling of being alive and intelligent. On the other hand, according to different usage scenarios, the attributes of social cues should also be different.

This study also has some limitations. There are many social cues, and this study only involves eyes blinking and head rotating. In addition, the usage scenarios of robots can be more universal, which will be reflected in further research. The results of this research can be used as a pre-test experiment to further explore the improvement of HRI interaction by social cues.

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