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# Social reactions toward people vs. computers: How mere labels shape interactions

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## 7 Abstract

8 What criteria afford a machine the status of a social agent? In this investigation, the mere label  
9 identifying an oral interviewer as human or computer was sufficient to affect participants' responses  
10 toward the interviewer during an online interview for a competitive mock job. Participants' impres-  
11 sions of the interviewer and self-reported emotional reactions to the interview were unaffected by the  
12 interviewer's identity. Despite this invariance, however, participants exhibited more interpersonal  
13 displays when the interviewer was identified as human. Overall, these results show that participants  
14 engaged in heightened impression management strategies (deferral to, or attempts to engage or  
15 appease) with the "human" interviewer. The computer interviewer did not merit equivalent social  
16 status.

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## 19 1. Introduction

20 Adults have been shown to apply the same social norms and rules of etiquette toward  
21 computers as they do toward other humans (De Laere, Lundgren, & Howe, 1998; Laut-  
22 enschlager & Flaherty, 1990; Nass, Moon, Fogg, Reeves, & Dryer, 1995; Nass, Steuer,  
23 Henriksen, & Dryer, 1994; Reeves & Nass, 1996; Wenger, 1991). For instance, they will  
24 evaluate a computer tutor more favorably when in its presence than not, providing evi-  
25 dence for politeness (Nass, Moon, & Carney, 1999), and this can occur even without con-  
26 scious awareness (Nass & Moon, 2000).

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27 How does our interaction with machines compare with normal social commerce? Many  
28 of these studies of Social Reactions to Communication Technology (SRCT) did not sup-  
29 plement their observations with a human control group, making direct comparison impos-  
30 sible. Studies that have employed a human control group have tended to show that  
31 computer stimuli do not meet the standard for human interaction, eliciting considerably  
32 weaker interpersonal behavior than do human stimuli (Bonito, Burgoon, & Bengtsson,  
33 1999; Lee & Nass, 2002; Martin & Nagao, 1989). This appears to be true even when  
34 the features of the stimuli are identical, and only the identity label is changed. For exam-  
35 ple, in immersive virtual environments, anthropomorphic virtual agents elicit a higher  
36 threshold of social influence when they are believed to be controlled by a human rather  
37 than by a computer (Bailenson, Blascovich, Beall, & Loomis, 2003). Similarly, Schecht-  
38 man and Horowitz observed increased interpersonal behavior when the ostensible interac-  
39 tants were human rather than computerized. This effect was found using a visually  
40 impoverished task in which equipotent interactants had to cooperate through a text-based  
41 interface (2003).

42 Nonetheless, many if not most human–computer and human–human interactions occur  
43 under conditions of *unequal* stature or power (e.g., teacher–student, doctor–patient, super-  
44 visor–employee, officer–soldier, interviewer–interviewee), when the evaluations of the  
45 superordinate member of the dyad have direct repercussions for the subordinate. Under  
46 these conditions, the subordinate must not only engage in the usual social rules, but also  
47 must respond directly to the desirable or undesirable feedback the superordinate supplies.  
48 This forces the question: Do we respond differently when we are praised (or damned) by a  
49 computer rather than another human?

50 One clue derives from studies of impression management, which show that subordinates  
51 display strong interpersonal behaviors toward authority figures (Burgoon, Johnson, &  
52 Koch, 1998). Naturally, this literature makes no prediction regarding computerized author-  
53 ities. We reasoned that, although power asymmetry by itself might not be sufficient to elicit  
54 anthropomorphism of the computer, receiving evaluative feedback (whether desirable or  
55 undesirable) from that authority may strengthen interpersonal reactions. For instance, if  
56 participants feel rejected by the authority figure, perhaps their emotional reaction will be  
57 strong enough to provoke social responses toward that authority even when doing so is gra-  
58 tuitous. Due to the scant literature on this topic, this prediction remained speculative.

59 To gauge the impact of labeling and subordination on SRCT, we asked participants to  
60 take part in individual mock job interviews, in which they answered questions that were  
61 delivered over an audio link. The interviewer at the other end of the link was stipulated  
62 to be either a human speaking over a microphone or a vocalizing computer program. This  
63 arrangement allowed determining whether the putative identity of the remote interviewer  
64 would affect participants' interpersonal responses during a mediated exercise of asymmet-  
65 ric power. Following the mock interview, the interviewer (whether "computer" or  
66 "human") informed the participant whether he/she was accepted for or rejected from  
67 the job. This second manipulation was designed to indicate whether evaluative feedback  
68 would moderate any effects of the interviewer's alleged identity.

69 We examined social reactions by measuring nonverbal behavior (particularly facial dis-  
70 plays), evaluative remarks about the interviewer, and emotion-report before and after the  
71 interview. We used nonverbal, cognitive, and emotional measures because they are well-  
72 established indicators of social influence (Argyle, 1972; Fridlund & Russell, *in press*; Ken-  
73 don, 1981; Lang, 1968; Mandler, Mandler, Kremen, & Sholiton, 1961).

## 74 2. Method

### 75 2.1. Design

76 This experiment consisted of two phases. In Phase 1, participants were told they would  
77 be interviewed for a mock job by either a *human* or an artificially intelligent *computer*;  
78 these constituted the two categories of *interviewer identity* (ID). During the interview, par-  
79 ticipants' verbal and nonverbal displays were recorded and measured. In Phase 2 of the  
80 experiment, the participants from each ID category were informed that they were either  
81 accepted for or rejected from the job. This manipulation comprised a  $2 \times 2$  factorial  
82 design, counterbalancing ID with *interviewer feedback* (FB). After receiving feedback, par-  
83 ticipants' emotion self-reports and impressions of the interviewer were ascertained. All  
84 conditions were between subjects with equal numbers of males and females in each of  
85 the four conditions, and the order of conditions was counterbalanced across participants.

### 86 2.2. Participants

87 Participants consisted of psychology students from the University of California, Santa  
88 Barbara (20 male, 20 female), who received introductory psychology course credit for their  
89 participation. Age ranged from 17 to 23 yr ( $M = 20$  yr). The entire research protocol was  
90 approved by the UC Santa Barbara Committee on Research.

### 91 2.3. Measures

92 In order to assess participants' responses, observations of three types were conducted –  
93 nonverbal behavior, self-reported emotion, and evaluative impressions of the interviewer.  
94 Nonverbal behaviors were tallied by blind judges, and included judgments about the pres-  
95 ence (or absence) of facial displays and other behaviors indicating social responsiveness.  
96 Nonverbal behaviors that were assessed included facial displays such as smiles, frowns,  
97 and “yuck” faces, and adjunctive behavior such as silence-fillers, self-manipulations  
98 (e.g., nose-scratches), and signs of politeness and embarrassment. All nonverbal behaviors  
99 were rated using a binary scale (“present” or “absent”).<sup>1</sup> (See Appendix A for definitions  
100 of these measures.)

101 Participants' emotional reactions were assessed via self-report questionnaire, adminis-  
102 tered before and after the experimental manipulation, here referred to as the pre-stimulus  
103 and post-stimulus emotion forms. The questionnaire included 7-point Likert-type ques-  
104 tions asking the participant how he/she felt at that moment. These items included ques-  
105 tions like “To what extent do you feel happy?,” and “To what extent do you feel  
106 tired?” Additionally, we sought to obtain a general motivational factor (perceived “enthu-  
107 siasm”) without asking their motivation directly (and reactively). So, a composite factor  
108 was constructed to gauge how “enthusiastic” the participant was before and after the  
109 interviewer. This was achieved by combining scores from the items “excited,” “curious,”  
110 and “alert” (tired reverse-coded). Assessing this aspect of the participant's motivation by

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<sup>1</sup> Eight additional exploratory nonverbal measures were coded by the blind judges which were excluded from analysis on the basis of insufficient power ( $<.45$  for each item). This was not surprising as the measures were exploratory only and unvalidated. Details about these measures available upon request.

111 pooling several related responses should have augmented its reliability, although admit-  
112 tedly it was neither psychometrically constructed nor factorially pure. (See Appendix A  
113 for complete list of measures.)

114 Participants' evaluative impressions of the interviewer were also assessed via self-report  
115 questionnaire, administered after receiving feedback about their acceptance for or rejection  
116 from the job. This questionnaire, referred to as the impressions form, was designed  
117 to gauge on a 7-point Likert-type scale the extent to which participants judged the inter-  
118 viewer to be competent and enjoyable. "To what extent did the interviewer capture your  
119 strengths?" and "To what extent do you think the interviewer was sociable?" (See Section  
120 3 for complete list of measures.)

121 In addition to our three categories of observations, we also collected data about partici-  
122 pants' demographic profiles, their level of experience with interviews, and their level of expe-  
123 rience with computers. These were obtained on a form labeled the personal information  
124 form. Finally, we measured the extent to which participants believed the identity of the inter-  
125 viewer was consistent with the ID condition to which they were assigned. This manipulation  
126 check was administered in the form of a questionnaire called the post-interview form.

#### 127 2.4. *Stimuli and apparatus*

128 Interview stimuli included 10 questions, prerecorded by a male human voice, played in  
129 a single order to the participant, who was seated in front of the computer keyboard, CRT  
130 monitor, and speakers. These questions included topics such as "How did you adjust to  
131 your first year of college?" and "What kinds of activities have you been interested in out-  
132 side of school?" (The interview questions are available upon request.)

133 Although the interviewer was stipulated to be either human or computer, in actuality the  
134 same prerecorded stimulus was used for both – and was altered to be believable in either  
135 case. To accomplish this, the voice recording was passed through a preset digital filter that  
136 distorted the vocal quality and was pretested for ambiguity as to the source (human or com-  
137 puter) ( $n = 23$ ).<sup>2</sup> All audio editing was implemented using Melodyne 1.5 (Celemony) and  
138 Wavelab 4.0 (Steinberg) with the Orange Vocoder plug-in (Prosoniq). Melodyne was used  
139 to filter out prosodic variation in the audio clips. Wavelab was then used to apply the preset  
140 distortion filter on these clips. (The vocoder specifications are available upon request.)

141 During the interview, the computer monitor displayed only the appropriate question  
142 number (question 1, question 2, etc.) for each interview question. A small lapel micro-  
143 phone detected the participant's voice. Video from a covert camera, as well as the micro-  
144 phone audio, were fed to a VHS video recorder, which taped the participant's entire  
145 session. The videotape allowed us to record information about the participant's facial dis-  
146 plays, gesturing, and posturing.

#### 147 2.5. *Procedure*

148 After reading and signing a general information form, the participant completed the  
149 personal information form and the pre-stimulus emotion form at a standard desk space.

<sup>2</sup> Due to a limited sample-size, statistical differences between vocal stimuli were non-significant, however means for each level were in the predicted direction.

150 Following completion of these forms, participants were situated in front of a 17-in desktop  
151 computer monitor and asked to complete an online "Prescreening Questionnaire"  
152 designed to collect demographic data relevant to a mock job-hiring interview. The ques-  
153 tionnaire was used to lend extra face validity to the ensuing interview.

154 Upon completion of this online questionnaire, participants received standard instruc-  
155 tions explaining that they would be asked to answer questions about themselves in a mock  
156 interview. The study, they were told, concerned the effectiveness of Internet-based job  
157 interviews, which required the interviewer to be elsewhere (ostensibly, a nearby lab).

158 Depending on the experimental condition, participants were further told that the inter-  
159 viewer would be either a live human being or an artificially intelligent, vocal computer pro-  
160 gram. Each interview condition was accompanied by a unique justification for the  
161 interviewer's voice. When the interviewer was purportedly human, the justification was  
162 that the speaker's voice would be filtered in real time to simulate limited-bandwidth Inter-  
163 net transmission. When the interviewer was ostensibly a computer, interviewers were told  
164 that the speaker's voice was generated by an advanced computer application that special-  
165 ized in speech production and analysis.

166 In both conditions, participants were told that the interviewer alone would generate  
167 questions based on their Prescreening Questionnaire, issue the questions over the com-  
168 puter speakers, analyze their responses, and then judge whether they merited being hired  
169 for the job by comparing his/her answers to those of the others in the study. They were  
170 told that the interviewer followed strict procedures for making this decision, but we could  
171 not inform them of the admission criteria, nor the exact nature of the job, until the end of  
172 the experiment.

173 Participants were told that they would be given 45 s to answer each question. Finally, it  
174 was explained that all answers were fully confidential and anonymous and that they were  
175 not obligated to answer any question that made them uncomfortable.

176 While the participant was completing the online questionnaire, the experimenter left the  
177 room, ostensibly to ready the fictitious "other lab." When the experimenter returned, he/  
178 she affixed the microphone and told participants that it would be transmitting their  
179 answers to the interviewer. The participant was then left in privacy for the duration of  
180 the interview. Following the last interview question, the participant received predeter-  
181 mined feedback telling him/her whether or not he/she got the job. The experimenter then  
182 gave the participant the post-stimulus emotion form, impressions form, and post-interview  
183 form. After their completion, the participant was fully debriefed. Finally, the experimenter  
184 disclosed the covert videotaping and administered a videotape consent form to permit the  
185 use and retention of the videotaped session.

### 186 3. Results

187 *Response coding and reliability.* Audiovisual data were coded by two independent blind  
188 judges who were trained to assess the presence/absence of various nonverbal displays. The  
189 judges recorded the presence or absence of smiles, frowns, "yuck" faces, silence fillers, self-  
190 manipulations, and signs of politeness and embarrassment throughout the interview. Par-  
191 ticipants' self-reported emotion ratings before and after the interview, as well as their  
192 impressions of the interviewer, were coded using 7-point Likert-type scales.

193 For each of the seven dichotomous behavioral measures, a single composite score was  
194 computed by averaging across the individual scores for the 10 interview questions (i.e.,

195 smiles, frowns, yuck faces, silence fillers, self-manipulations, politeness, and embarrass-  
196 ment). All of the composites showed non-normal distributions. Three of them  
197 (smiles, silence fillers, and embarrassment) were successfully normalized using a logarithmic  
198 transformation [ $X = \log_{10}(x + 1)$ ]. The remaining four composite scores could not be  
199 logarithmically corrected and were excluded from further analysis.

200 Interrater reliability (Cronbach's Alpha) was established between the two blind judges.  
201 Smiles and silence fillers were significantly correlated,  $\alpha = .804$  and  $\alpha = .621$ , respectively.  
202 Embarrassment was not reliable between raters,  $\alpha = .061$ , so this measure was excluded  
203 from further analysis.

204 *Effect of interviewer identity.* We first wished to assess if ID had any effect on the two  
205 dichotomous variables, smiles and silence fillers. These did differ with ID,  
206  $F(1,39) = 5.45$ ,  $p < .05$  and  $F(1,39) = 7.94$ ,  $p < .01$ , respectively. Specifically, participants  
207 were more likely both to smile and to exhibit silence-filling behavior when the interviewer  
208 was human (see Figs. 1 and 2; Tables 1 and 2 for means).

209 *Effects of feedback with interviewer identity.* Data were also collected on participants'  
210 emotion ratings. First, we searched for changes in emotion as a function of ID and FB,  
211 controlling for pre-stimulus emotion scores. Thus, we subjected each post-stimulus emo-  
212 tion score to an analysis of co-variance, with the corresponding pre-stimulus emotion  
213 score as the covariate. We found no independent effects of ID and no interactions (All  
214  $F$ -values  $< 3.0$ ,  $p > .05$ ). We did, however, find several effects for feedback type. Partici-  
215 pants were more curious, excited, and "enthusiastic" after receiving desirable feedback  
216 from the interviewer, as compared to undesirable feedback, regardless of interviewer type,  
217  $F(1,39) = 4.91$ ,  $p < .05$ ;  $F(1,39) = 16.01$ ,  $p < .0001$ ;  $F(1,39) = 12.366$ ,  $p < .001$ , respectively.  
218 We also found a marginal effect in which participants tended to be less frustrated after  
219 receiving desirable feedback compared to undesirable feedback,  $F(1,39) = 3.71$ ,  $p = .062$ .  
220 No differences were found in reported happiness, sadness, boredom, anxiety, anger, or  
221 relaxation (All  $F$ -values  $< 2.3$ ,  $p > .05$ ; see Table 3 for means).

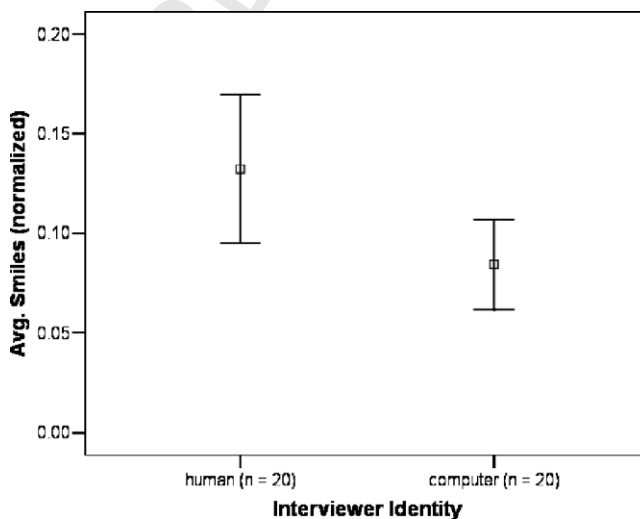


Fig. 1. Average number of smiles toward human and computer interviewers,  $F(1,39) = 5.45$ ,  $p < .05$  (95% CI).

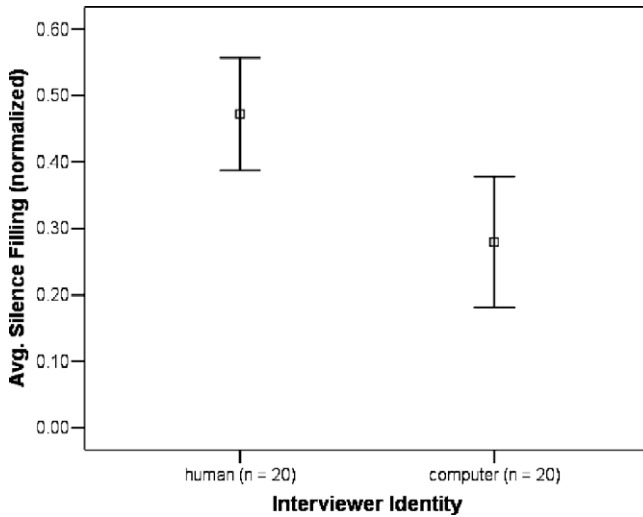


Fig. 2. Silence filling behavior toward human and computer interviewers,  $F(1,39) = 7.94, p < .01$ .

Table 1  
Average smiles toward human and computer interviewers (from 0 to 1)

ID	Mean	SD	N
Human	.132	.084	20
Computer	.084	.050	20

Table 2  
Average silence fillers toward human and computer interviewers (from 0 to 1)

ID	Mean	SD	N
Human	.472	.190	20
Computer	.280	.221	20

222 *Attributions toward interviewer.* We then tried to discern whether ID affected partici-  
 223 pants' impressions of and attributions toward the interviewer. Overall, people rated the  
 224 interviewer more positively when FB itself was desirable, regardless of ID. Specifically,  
 225 people rated the interviewer as more *sociable*,  $F(1,39) = 6.15, p < .05$ , and more *likeable*,  
 226  $F(1,89) = 7.06, p < .05$ , when they were accepted to the mock job. Other ratings related  
 227 to likeability (nice, fun, and casual) suffered from weak statistical power, so we pooled  
 228 them with the above-mentioned factors and found a corresponding effect,  
 229  $F(1,39) = 6.33, p < .03$ , independent of ID.

230 Participants receiving desirable feedback also reported that the interviewer was better  
 231 able to *capture their strengths*,  $F(1,39) = 7.67, p < .01$ , and that the interviewer enabled  
 232 them to *give a more accurate impression*,  $F(1,39) = 8.60, p < .01$ . As above, other ratings  
 233 related to expertise were found to have weak statistical power, and so these latter two fac-  
 234 tors were subsequently pooled with two other factors (competence and objectivity) to cre-  
 235 ate the factor named "expertise." It was found that evaluations of "expertise" also

Table 3  
Emotion ratings pre- and post-interview

Emotion	ID	FB	<i>M</i> (pre-)	<i>M</i> (post-)	<i>n</i>
Curious	Human	Rejection	4.90 (1.370)	3.60 (2.221)	10
		Acceptance	4.90 (1.449)	5.40 (1.838)	10
	Computer	Rejection	5.60 (1.174)	5.00 (1.732)	9
		Acceptance	4.60 (.699)	5.60 (.699)	10
Tired	Human	Rejection	5.00 (1.700)	4.60 (2.221)	10
		Acceptance	4.30 (1.567)	2.80 (1.317)	10
	Computer	Rejection	3.50 (1.650)	2.89 (1.764)	9
		Acceptance	5.40 (1.647)	4.40 (1.838)	10
Happy	Human	Rejection	4.50 (.972)	4.60 (1.350)	10
		Acceptance	4.55 (1.462)	5.10 (1.101)	10
	Computer	Rejection	4.10 (1.287)	4.33 (1.414)	9
		Acceptance	3.90 (.738)	4.90 (.994)	10
Sad	Human	Rejection	1.70 (.949)	2.20 (1.398)	10
		Acceptance	1.70 (.483)	1.50 (.972)	10
	Computer	Rejection	2.60 (1.838)	2.33 (1.225)	9
		Acceptance	2.80 (1.135)	2.60 (1.430)	10
Bored	Human	Rejection	2.90 (1.792)	2.40 (1.647)	10
		Acceptance	2.50 (1.269)	1.80 (.789)	10
	Computer	Rejection	3.30 (1.418)	2.44 (1.130)	9
		Acceptance	2.50 (1.354)	2.10 (.994)	10
Anxious	Human	Rejection	3.10 (1.912)	2.40 (2.066)	10
		Acceptance	3.50 (1.434)	3.10 (2.183)	10
	Computer	Rejection	4.00 (1.333)	3.33 (1.871)	9
		Acceptance	4.10 (1.449)	4.10 (1.853)	10
Excited	Human	Rejection	2.50 (1.269)	1.80 (1.135)	10
		Acceptance	3.80 (1.549)	4.70 (1.767)	10
	Computer	Rejection	3.90 (1.197)	3.33 (1.500)	9
		Acceptance	3.60 (1.350)	4.30 (1.636)	10
Frustrated	Human	Rejection	1.80 (1.476)	2.00 (1.886)	10
		Acceptance	1.70 (.823)	1.30 (.675)	10
	Computer	Rejection	2.00 (1.247)	2.33 (1.414)	9
		Acceptance	3.10 (2.132)	2.40 (1.776)	10
Angry	Human	Rejection	1.60 (1.265)	1.90 (1.912)	10
		Acceptance	1.20 (.632)	1.30 (.675)	10
	Computer	Rejection	1.80 (1.874)	1.33 (.500)	9
		Acceptance	2.00 (1.247)	1.50 (.707)	10
Relaxed	Human	Rejection	5.10 (1.197)	5.20 (1.549)	10
		Acceptance	4.20 (1.687)	4.20 (1.135)	10
	Computer	Rejection	4.90 (1.197)	4.67 (1.500)	9
		Acceptance	4.00 (.943)	4.60 (1.174)	10
“Enthusiastic”	Human	Rejection	3.47 (.789)	2.93 (1.350)	10
		Acceptance	4.13 (1.146)	5.10 (1.089)	10
	Computer	Rejection	4.67 (1.042)	4.48 (1.281)	9
		Acceptance	3.60 (.886)	4.50 (1.189)	10



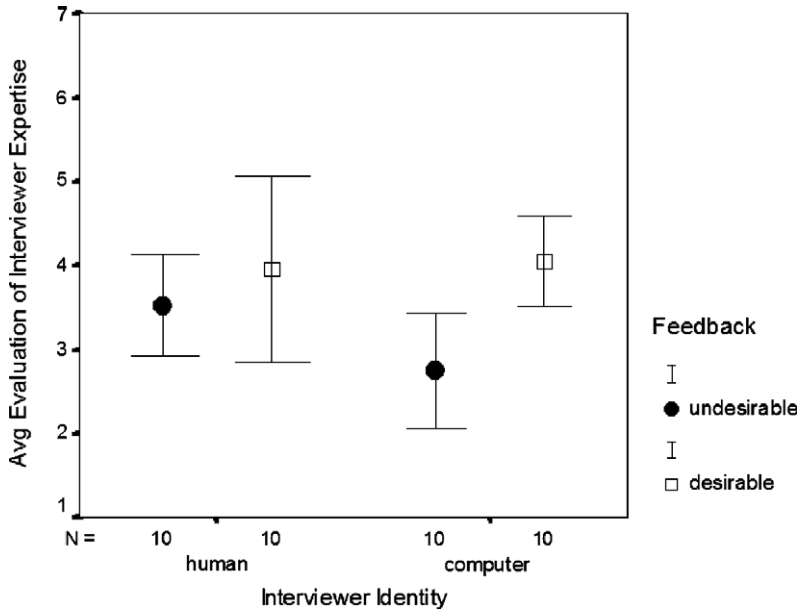


Fig. 3. Evaluations of “Expertise” a function of desirable FB but not ID,  $F(1,39) = 6.45, p < .05$  (95% CI).

236 increased as a function of desirable FB but not as a function of ID,  $F(1,39) = 6.45, p < .05$   
 237 (Fig. 3). Participants who received undesirable feedback from the computer interviewer  
 238 showed marginal tendencies to demote the computer interviewer’s intelligence,  
 239  $F(1,39) = 3.11, p = .086$ , and objectivity,  $F(1,39) = 3.07, p = .089$ , relative to the other  
 240 conditions. The other five items did not differ as a function of ID (see Table 4 for means).  
 241 There is no strong evidence to conclude that the identity of the interviewer had an impact  
 242 on the above impressions of the interviewer. However, because statistical power is low  
 243 ( $<.3$ ), this pattern is vulnerable to Type II error.<sup>3</sup>

244 Finally, we examined whether the beliefs that the participants reported about the inter-  
 245 viewer identity were consistent with the condition to which they were assigned. First, we  
 246 transformed the belief question in each condition into a single continuous variable from 0  
 247 (“don’t believe”) to 1 (“believe”). We subjected this variable to a one-way analysis of var-  
 248 iance, with ID as the independent factor and observed a significant difference in which  
 249 people in the “computer” condition reported a stronger belief in the purported identity  
 250 of the interviewer than did those in the “human” condition,  $F = 18.102, p < .001$ . Cell  
 251 means illustrate that participants in the “computer” condition did indeed believe in the

<sup>3</sup> A series of three-way ANOVAs was also used to test the effects of ID, FB, and participant sex on impressions of the interviewer. These tests showed no affect of ID; however, one yielded a two way interaction in which female participants who received undesirable feedback rated the interviewer as significantly less able to capture their strengths than did females receiving desirable feedback ( $F(2,38) = 4.5, p < .05$ ). We then conducted a series of post hoc tests (Fisher’s LSD, Tukey’s HSD, & Bonferroni) to observe the interaction while accounting for the inflation of  $\alpha$  that results from three-way ANOVAs. Only Fisher’s LSD showed significance, and did so in the predicted direction (All MD  $> 1.8, p < .05$ ), suggesting that the female tendency to demote the interviewer’s expertise on the basis of undesirable feedback occurred regardless of ID. But because statistical power was weak ( $<.1$ ), it remains possible that the interaction was due to chance.

Table 4  
Evaluative impressions of human and computer interviewers after acceptance or rejection

Evaluation	ID	FB	<i>M</i>	<i>n</i>
Competent	Human	Rejection	3.89 (1.364)	9
		Acceptance	3.70 (2.058)	10
	Computer	Rejection	3.40 (1.430)	10
		Acceptance	4.00 (.667)	10
Nice	Human	Rejection	3.22 (1.922)	9
		Acceptance	3.80 (1.687)	10
	Computer	Rejection	3.00 (1.633)	10
		Acceptance	4.20 (1.135)	10
Fun	Human	Rejection	2.00 (1.581)	9
		Acceptance	2.50 (1.958)	10
	Computer	Rejection	2.00 (1.155)	10
		Acceptance	2.70 (1.767)	10
Sociable	Human	Rejection	2.00 (1.500)	9
		Acceptance	2.70 (1.947)	10
	Computer	Rejection	1.70 (.823)	10
		Acceptance	3.40 (1.897)	10
Objective	Human	Rejection	4.33 (1.871)	9
		Acceptance	4.10 (1.370)	10
	Computer	Rejection	3.00 (1.700)	10
		Acceptance	4.50 (1.179)	10
Intelligent	Human	Rejection	4.00 (1.581)	9
		Acceptance	3.20 (2.150)	10
	Computer	Rejection	3.20 (1.989)	10
		Acceptance	4.40 (1.430)	10
Casual	Human	Rejection	3.22 (1.716)	9
		Acceptance	4.20 (2.251)	10
	Computer	Rejection	3.40 (2.171)	10
		Acceptance	4.60 (1.174)	10
Slang	Human	Rejection	4.33 (1.414)	9
		Acceptance	4.00 (2.357)	10
	Computer	Rejection	3.90 (2.283)	10
		Acceptance	3.60 (2.011)	10
Impression	Human	Rejection	3.11 (1.364)	9
		Acceptance	4.40 (1.897)	10
	Computer	Rejection	2.20 (1.687)	10
		Acceptance	4.00 (1.247)	10
Strengths	Human	Rejection	2.11 (.782)	9
		Acceptance	3.60 (2.011)	10
	Computer	Rejection	2.40 (1.174)	10
		Acceptance	3.70 (1.160)	10
Likability	Human	Rejection	2.67 (1.414)	9
		Acceptance	3.30 (1.889)	10
	Computer	Rejection	2.20 (1.229)	10
		Acceptance	4.00 (1.563)	10
“Expertise”	Human	Rejection	3.53 (.837)	9
		Acceptance	3.95 (1.563)	10
	Computer	Rejection	2.75 (.957)	10
		Acceptance	4.05 (.744)	10

252 interviewer's "computer" identity ( $M = .90$ ,  $SD = .308$ ) whereas participants in the  
253 "human condition" confessed some suspicion of the interviewer's "human" identity  
254 ( $M = .35$ ,  $SD = .489$ ).

#### 255 4. Conclusion

256 Do people differ in their social reactions to an interviewer depending upon whether the  
257 interviewer is human or a computer? And does the interviewer's judgment matter? Our  
258 results reveal a mixed answer. Participants did not report feeling any happier with the  
259 human interviewer, but they smiled more toward him. They did not think the human  
260 was nicer or more likeable, sociable, or fun, but they made greater efforts to speak to  
261 him. They did not consider the human to be more of an expert than the computer, but they  
262 were more affected by a rejection from the human.

263 Participants' reactions to the interview feedback also yielded effects that were indepen-  
264 dent of the interviewer's identity. Even though all participants were fully aware that the  
265 interview was only a simulation, they were emotionally affected by information about their  
266 acceptance to or rejection from the job: They were more curious, excited, and "enthusias-  
267 tic" after being accepted. Other emotion ratings, however, did not differ. Furthermore,  
268 likeability ratings and evaluations of expertise were strongly predicted by acceptance ver-  
269 sus rejection, particularly for females.

270 Our mixed results follow a consistent pattern. Nonverbal displays (smiling and  
271 silence filling) were consistently responsive to the human. Both self-reported emotion  
272 and impressions of interviewer likeability and expertise were consistently responsive  
273 to the nature of the interview feedback independent of ID. Thus, the behavioral results  
274 complement the findings of [Schechtman and Horowitz \(2003\)](#), and suggest that in a  
275 context of asymmetric power, people are more communicative toward ostensibly human  
276 interactants than toward computer interactants, even when the behavior of both inter-  
277 viewers is identical (For related findings, see [Bailenson et al., 2003](#); [Bonito et al., 1999](#);  
278 [Lee & Nass, 2002](#); [Martin & Nagao, 1989](#)). This pattern was not confirmed by the  
279 emotional and cognitive reactions to the interviewer, which were sensitive to feedback  
280 only.

#### 281 5. Discussion

282 We consider a limited number of ways to explain the consistent, but mixed pattern of  
283 results. First, why did participants smile more to the human interviewer if they did not  
284 report feeling any happier having interacted with him? One explanation for this discrep-  
285 ancy is that the smiles emitted in this context are not expressions of emotion so much  
286 as communications to the interviewer, much as we commonly make expressions to others  
287 on the telephone ([Fridlund, 1994, 1997](#)). An analogous argument explains why partici-  
288 pants exhibited greater effort speaking to the human despite the fact that they did not con-  
289 sider him to be nicer or more likeable, sociable, or fun, than the computer interviewer.  
290 Likeable or not, efforts to fill silences with topic-relevant speech suggest that participants  
291 were operating under a norm of reciprocity in their communications with the interviewer.  
292 That this pattern was not found with computer interviewers is evidence that computers  
293 were not regarded as independent agents of communication that should call upon such  
294 a reciprocity norm.

295 If nonverbal behavior is more responsive to the human interviewer, why do we not find  
296 a similar pattern in participants' reported emotion? Could it be that this equal sensitivity  
297 to both interviewers is evidence of a social reaction toward the computer as well as the  
298 human? Although this interpretation is plausible, it fails because the absolute values of  
299 the means for these emotion ratings converge near the neutral point of the Likert scale.  
300 This suggests that, in absolute terms, participants are not particularly happy, sad, excited,  
301 etc., after interacting with either interviewer. Although they are clearly emotionally  
302 affected by different types of feedback, this is not associated with interviewer identity.  
303 Therefore, by observing solely the emotion ratings, we cannot find evidence that either  
304 interviewer qualifies as an agent of social exchange. Indeed, if participants were emotion-  
305 ally partial to either interviewer, the fact that this was only a *mock* interview may have  
306 been sufficient to motivate them to under-report the magnitude of their emotional  
307 response. Such demand characteristics are a common problem in experimental simula-  
308 tions, and future studies of this kind may overcome this problem by using less reactive  
309 ways of ascertaining emotion (but see [Fridlund & Russell, in press](#)).

310 A related question is that if participants' nonverbal behaviors are "human-biased,"  
311 then why did not they report similar biases in cognitive appraisals of the interviewer?  
312 For example, why was not the human deemed more expert than the computer? Because  
313 evaluations of the interviewer were relatively anonymous, we see no compelling reason  
314 why participants should be motivated to misreport their impressions as they might have  
315 with their emotion ratings. Instead, we believe that participants evaluated both interview-  
316 ers' expertise the same, not because they awarded the computer with humanlike expertise,  
317 but because they felt neither interviewer was particularly expert. Again, the absolute values  
318 of the means suggest this is the case. Rather than a commentary on social reactions toward  
319 the computer, we feel this result signals a ceiling effect for expertise. We suggest that future  
320 studies on this topic take precautions to ensure that the stimuli are designed to accommo-  
321 date to higher ratings of expertise.

322 One reason why emotional and evaluative impressions of the interviewer were insensi-  
323 tive to identity could be that people reported a moderate suspicion that the "human"  
324 interviewer was in fact human. We feel that this is an insufficient explanation for our  
325 effects because if participants were truly suspicious of the interviewer's identity, we would  
326 not have observed the pronounced difference in allocation of smiles and silence-filling  
327 behavior between the two interviewer types. So, even if they were suspicious to some  
328 degree, this did not appear to diminish their special responsiveness to the human inter-  
329 viewer. It may also imply that, even when participants were suspicious of the interactant's  
330 "human" identity, they classified the interactant as human for the purposes of the  
331 interaction.

332 Generally speaking, our behavioral measures appear to paint a picture that only a  
333 human merits social, communicative exchange, and that mere labels about an interactant's  
334 identity are sufficient to produce this effect – even when participants may be suspicious of  
335 that interactant's identity. Our cognitive and emotional measures supplement this conclu-  
336 sion, indicating that some social influence between human and computer can emerge by  
337 receiving trivial feedback about one's acceptance to or rejection from a mock job, indepen-  
338 dent of interviewer identity.

339 It has been argued elsewhere that self-report measures by themselves are oftentimes  
340 inadequate because participants can be poor judges of their own thoughts and feelings  
341 ([Bailenson et al., 2004](#)). Consequently, self-report measures are most appropriate in con-

342 junction with behavioral measures, which can be less obtrusive and more sensitive to more  
343 automatic social signals (Bailenson et al., 2004; Cook & Campbell, 1979). Heeding this  
344 advice, this experiment may lend its strongest support to the argument that identity label-  
345 ing alone can be sufficient to elicit rich social responses toward ostensibly human inter-  
346 viewers in a way that is not matched by SRCT. However, evaluative feedback from the  
347 interviewer can elicit some rich impressions and emotional reports that do not necessarily  
348 discriminate between either identity.

349 Taken together, our results can be interpreted in terms of impression management,  
350 whereby the individual defers to and attempts to engage and appease others because oth-  
351 ers' evaluations bear social consequences (Baumeister, 1982b; Kacmar & Carlson, 1999).  
352 In this case, the audio computer interviewer did not appear to merit the social status  
353 achieved by an ostensible human except in regard to some evaluative components of  
354 the interaction. It may be that a stimulus labeled "human" can afford to be lacking  
355 in a host of other diagnostic cues, such as naturalistic vocal and visual cues, whereas  
356 a stimulus labeled "computerized" must advertise much richer audiovisual attributes  
357 in order to elicit the same level of social reactions so easily directed toward ostensible  
358 humans. This sensitivity change is consistent with established models of social influence  
359 (Blascovich, 2002).

360 Many questions, however, remain unanswered, such as how these interpersonal  
361 responses may differ in contexts in which the participant has more power than the inter-  
362 actant, and when communication occurs along more than audio channels. Pursuing these  
363 research avenues using standard social psychological methods may be quite informative  
364 about various facets of human–computer interaction.

## 365 6. Uncited references

366 Allport (1935), Baumeister (1982a), Booth-Kewley, Edwards, and Rosenfeld (1992),  
367 Burgoon et al. (2000), Gelman, Durgin, and Kaufman (1995), Marakas, Johnson, and Pal-  
368 mer (2000), Moon (2000), Moreno, Mayer, Spires, and Lester (2001), Resnik and Lam-  
369 mers (1985), Parise, Kiesler, Sproull, and Waters (1999).

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## 373 Appendix A

374 Emotion form

375 To what extent are you feeling each of the following right now?

376 Circle a number between 1 and 7. (1 = not at all, 4 = somewhat, 7 = very much):

377 Curious	Tired
381 Happy	Sad
384 Bored	Anxious
386 Excited	Frustrated
388 Angry	Relaxed

## 390 Impressions form

391 *Circle a number between 1 and 7. (1 = not at all, 4 = somewhat, 7 = very)*

- 392 • To what extent did you feel that the interviewer was competent?  
 393 • To what extent did the questions enable you to give an accurate impression of yourself?  
 394 • To what extent did the questions chosen capture your strengths?  
 395 • To what extent did you feel that the interviewer's decision was objective?  
 396 • To what extent did you feel that the interviewer was intelligent?  
 397 • To what extent did you feel that the interviewer was nice?  
 398 • To what extent did you feel that the interviewer was fun?  
 399 • To what extent did you feel that the interviewer was sociable?  
 400 • To what extent did you like the interviewer?  
 401 • To what extent was your relationship with the interviewer casual and informal?  
 402 • To what extent did you feel comfortable using slang words in front of the interviewer?  
 403

## 404 Ratings from blind judges

405 *For each interview question, rate the presence or absence of:*

- 406 • Smiles – self-explanatory.  
 407 • Frowns – self-explanatory.  
 408 • “Yuck” faces – displays of disgust.  
 409 • Silence fillers – statements produced after a conclusive pause in speech.  
 410 • Self-manipulation – e.g., picking nails, twirling hair, biting lips, scratching face.  
 411 • Signs of politeness – e.g., saying “thank you” or “your welcome.”  
 412 • Signs of embarrassment – e.g., blushing, nervous laughter.

## 413 References

- 414 Allport, G. W. (1935). *Attitudes. A handbook of social psychology*. MA: Clark University Press (pp. 798–844).  
 415 Argyle, M. (1972). Non-verbal communication in human social interaction. In R. A. Hinde (Ed.), *Nonverbal*  
 416 *communication* (pp. 243–269). New York: Cambridge University Press.  
 417 Bailenson, J., Aharoni, E., Beall, A., Guadagno, R., Dimov, A., Blascovich, J. (2004). Comparing behavioral and  
 418 self-report measures of agents' social presence in immersive virtual environments. In *Proceedings of the 7th*  
 419 *annual international workshop on PRESENCE* (2004, Valencia, Spain).  
 420 Bailenson, J., Blascovich, J., Beall, A., & Loomis, J. (2003). Interpersonal distance in immersive virtual  
 421 environments. *Personality & Social Psychology Bulletin*, 29(7), 819–833.  
 422 Baumeister, R. (1982a). Self-esteem, self-presentation, and future interaction: a dilemma of reputation. *Journal of*  
 423 *Personality*, 50(1), 29–45.  
 424 Baumeister, R. (1982b). A self-presentational view of social phenomena. *Psychological Bulletin*, 91(1), 3–26.  
 425 Blascovich, J. (2002). Social influence within immersive virtual environments. In R. Schroeder (Ed.), *The social*  
 426 *life of avatars* (pp. 127–145). Springer-Verlag.  
 427 Bonito, J., Burgoon, J., Bengtsson, B. (1999). The role of expectations in human–computer interaction. In  
 428 *Proceedings of the international ACM SIGGROUP conference on supporting group work*.  
 429 Booth-Kewley, S., Edwards, J., & Rosenfeld, P. (1992). Impression management, social desirability, and  
 430 computer administration of attitude questionnaires: does the computer make a difference? *Journal of Applied*  
 431 *Psychology*, 77(4), 562–566.  
 432 Burgoon, J., Bonito, J., Bengtsson, B., Cederberg, C., Lundeberg, M., & Allspach, L. (2000). Interactivity in  
 433 human–computer interaction: a study of credibility, understanding, and influence. *Computers in Human*  
 434 *Behavior*, 16(6), 553–574.  
 435 Burgoon, J. K., Johnson, M. L., & Koch, P. T. (1998). The nature and measurement of interpersonal dominance.  
 436 *Communication Monographs*, 65(4), 308–335.

- 437 Cook, T. D., & Campbell, D. T. (1979). *Quasi-experimentation: Design & analysis issues for field setting*. Boston,  
438 MA: Houghton Mifflin Co.
- 439 De Laere, K., Lundgren, D., & Howe, S. (1998). The electronic mirror: human-computer interaction and change  
440 in self appraisals. *Computers in Human Behavior*, 14(1), 43-59.
- 441 Fridlund, A. J. (1994). *Human facial expression: An evolutionary view*. San Diego, CA: Academic Press.
- 442 Fridlund, A. J. (1997). The new ethology of human facial expressions. In J. A. Russell & J. Fernandez-Dols  
443 (Eds.), *The psychology of facial expression* (pp. 103-129). Cambridge: Cambridge University Press.
- 444 Fridlund, A. J., Russell, J. A. (in press). The functions of facial expressions: what's in a face? In V. Manusov & M.  
445 L. Patterson (Eds.), *The sage handbook of nonverbal communication*. Thousand Oaks, CA: Sage.
- 446 Gelman, R., Durgin, F., & Kaufman, L. (1995). Distinguishing between animates and inanimates: not by motion  
447 alone. In D. Sperber, D. Premack, & A. J. Premack (Eds.), *Causal cognition: A multidisciplinary debate*  
448 (pp. 150-184). Oxford: Oxford University Press.
- 449 Kacmar, K. M., & Carlson, D. (1999). Effectiveness of impression management tactics across human resource  
450 situations. *Journal of Applied Social Psychology*, 29(6), 1203-1315.
- 451 Kendon, A. (1981). Introduction: current issues in the study of "nonverbal communication. In A. Kendon (Ed.),  
452 *Nonverbal communication, interaction, and gesture* (pp. 1-53). Paris: Mouton.
- 453 Lang, P. J. (1968). Fear reduction and fear behavior: problems in treating a construct. In: J. M. Shlien (Ed.),  
454 *Research in psychotherapy* (Vol. 1, pp. 90-102).
- 455 Lautenschlager, G., & Flaherty, V. (1990). Computer administration of questions: more desirable or more  
456 socially desirable? *Journal of Applied Psychology*, 75(3), 310-314.
- 457 Lee, E., & Nass, C. (2002). Experimental tests of normative group influence and representation effects in  
458 computer-mediated communication: when interacting via computers differs from interacting with computers.  
459 *Human Communication Research*, 28(3), 349-381.
- 460 Mandler, G., Mandler, J. M., Kremen, I., & Sholiton, R. D. (1961). The response to threat: relations among  
461 verbal and physiological indices. *Psychological Monographs*, 75(9 whole No. 513), 22.
- 462 Marakas, G., Johnson, R., & Palmer, J. (2000). A theoretical model of differential social attributions toward  
463 computing technology: when the metaphor becomes the model. *International Journal of Human-Computer*  
464 *Studies*, 52(4), 719-750.
- 465 Martin, C., & Nagao, D. (1989). Some effects of computerized interviewing on job applicant responses. *Journal of*  
466 *Applied Psychology*, 74(1), 72-80.
- 467 Moon, Y. (2000). Intimate exchanges: using computers to elicit self-disclosure from consumers. *Journal of*  
468 *Consumer Research*, 26(4), 323-339.
- 469 Moreno, R., Mayer, R. E., Spire, H., & Lester, J. (2001). The case for social agency in computer-based teaching:  
470 do students learn more deeply when they interact with animated pedagogical agents? *Cognition and*  
471 *Instruction*, 19(2), 177-214.
- 472 Nass, C., & Moon, Y. (2000). Machines and mindlessness: social responses to computers. *Journal of Social Issues*,  
473 56(1), 81-103.
- 474 Nass, C., Moon, Y., & Carney, P. (1999). Are people polite to computers? Responses to computer-based  
475 interviewing systems. *Journal of Applied Social Psychology*, 29(5), 1093-1110.
- 476 Nass, C., Moon, Y., Fogg, B., Reeves, B., & Dryer, D. (1995). Can computer personalities be human  
477 personalities? *International Journal of Human-Computer Studies*, 43(2), 223-239.
- 478 Nass, C., Steuer, J., Henriksen, L., & Dryer, D. (1994). Machines, social attributions, and ethopoeia: performance  
479 assessments of computers subsequent to "self-" and "other-" evaluations. *International Journal of Man-*  
480 *Machine Studies*, 40(3), 543-559.
- 481 Reeves, B., & Nass, C. (1996). *The media equation: How people treat computers, television, and new media like real*  
482 *people and places*. New York, NY: Cambridge University Press.
- 483 Resnik, P., & Lammers, B. (1985). The influence of self-esteem on cognitive responses to machine-like versus  
484 human-like computer feedback. *Journal of Social Psychology*, 125(6), 761-769.
- 485 Schechtman, N., Horowitz, L. (2003). Media inequality in conversation: how people behave differently when  
486 interacting with computers and people. In *Proceedings of the CHI 2003 conference on human-computer*  
487 *interaction*. Ft. Lauderdale, FL, April 5-10.
- 488 Parise, S., Kiesler, S., Sproull, L., & Waters, K. (1999). Cooperating with life-like interface agents. *Computers in*  
489 *Human Behavior*, 15(2), 123-142.
- 490 Wenger, M. (1991). On the rhetorical contract in human-computer interaction. *Computers in Human Behavior*,  
491 7(4), 245-262.
- 492