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Perceived Emotional Intelligence in Virtual Agents

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Abstract

In March 2016, several online news media reported on the inadequate emotional capabilities of interactive virtual assistants. While significant progress has been made in the general intelligence and functionality of virtual agents (VA), the emotional intelligent (EI) VA has yet been thoroughly explored. We examine user's perception of EI of virtual agents through *Zara The Supergirl*, a virtual agent that conducts question and answering type of conversational testing and counseling online. The results show that overall users perceive an emotion-expressing VA (EEVA) to be more EI than a non-emotion-expressing VA (NEEVA). However, simple affective expression may not be sufficient enough for EEVA to be perceived as fully EI.

Author Keywords

Emotional intelligence; emotions; virtual agent speech and language interaction; interactive dialog system.

ACM Classification Keywords

H.5.m. Information interfaces and presentation (e.g., HCI): Miscellaneous.

Introduction

According to Miner et al.'s research published in March 2016, various world-renowned virtual assistants e.g. Siri (Apple), Google Now, S Voice (Samsung), and



Figure 1. The resting state of emotion-expressing virtual agent (EEVA).



Figure 2. An example of EEVA's positive expression.



Figure 3. An example of EEVA's negative expression.

Cortana (Microsoft) provided rather impersonal and inconsistent responses to participants' questions on mental health, domestic violence, and emergencies [19]. As a service intended to mimic human interaction, this scenario beckons a need for empathy and emotional intelligence (EI) to deal with affective information appropriately. Various studies have shown that the way humans react to virtual agents (VA) resembles how they react to real people [16, 20]. Could empathetic agents be used to handle stressful situations and would people perceive them as EI?

Research in affective computing introduced emotions to computers, robots, and VA [3, 9, 11, 15, 16]. Research shows this can alleviate user frustration with computers [16], influence behavioral changes [15], enhance learning in children [9], and endow complex emotions to VAs [3]. A healthcare VA like SimSensei [6] needs a way to ensure empathetic and sensible services are provided to users. However, none of these techniques serves to understand how people perceive EI in VA equipped with emotional expression and recognition. Thus, this research gap leaves room for novel solutions. In our research, we focused on examining EI in VA.

Zara the Supergirl

In this abstract, we present Zara the Supergirl [8, 9], an interactive dialogue system rendered in the form of a virtual agent. It can recognize user emotions, express simple emotions, and respond user emotions detected. The system runs online or locally with a simple UI that shows the virtual agent, agent's question, and user's response. The agent starts by introducing her name, her task, and what to expect from the conversation. The agent analyzes user emotions through speech, text, and facial expressions captured real time [8, 9]. The agent expresses emotions via animations with body language and choice of words. The agent has simple affective expressions (see Figures 2 & 3) for positive (while saying things like "Fantastic," "That is awesome" etc.) and negative (while making comments like "So horrible," "Awful," etc.) emotions. To account for emotionally neutral answers from users, the agent also has a resting pose (see Figure 1) and a list of neutral feedback (i.e. "I see," "Okay").

It is important to clarify at this point that the agent does not manifest the ability to feel empathy or other emotions. Instead, it is designed with minimal elements to convey to users a sense of empathy, or "an affective response more appropriate to someone else's situation" [14]. A way humans express empathy is mimicry [2]. Mimicry happens when an observer makes a reaction similar to the observed [2]. Mimicry can also increase empathy in humans during virtual human interactions [12]. Therefore, the virtual agent (VA) in this study is designed to reciprocate user's sentiment by expressing the same sentiment. This follows the law of attraction that can increase its likability to humans [18].

The VA in this study plays the role of an assistant to evaluate the stress levels of current university students through a 5-7min, one-on-one conversation. The dialog has five subtopics modeled after Depression Anxiety Stress Scales (DASS) [13] in a manner of increased intimacy [1]. It starts with general topics (i.e. hobbies and travel), then personal topics (i.e. friends and families, current study and future plans), and finally a generic appraisal on students' wellbeing (i.e. exercise and downtime). Each topic has 2-3 open-ended short



Figure 4. NEEVA's resting state remains constant during the dialog.

questions and ~ 15 seconds to respond. Users receive an appraisal of their stress level at the end.

A non-emotion expressing version (NEEVA) of the agent that has all abovementioned functionalities acts as the control of this experiment. Here we shorthand the empathetic interactive dialogue system as EEVA EEVA responds to participant's answers with a short positive, negative, or neutral emotive interjection while

MSCEIT-Based Perceived EI Questionnaire					
Br.	Questions				
PE	Convey a sense that it listen s openly to participant's emotions				
	Convey a sense that it pays attention to user's moods during the conversation				
	Identify user's emotions_correctly				
	Discern between different emotions				
	Tell apart the degree of emotions present				
UsE	Convey a sense that the agent can feel what the user is feeling				
	Convey a sense that the agent understand user's point of view (POV)				
	Respond in a way that make the user feel sad				
	Respond in a way that make the user feel happy				
	Respond in a way that make the user feel that they are understood				
UnE	Convey a sense that it can be emotionally self-aware and insightful				
	Display some knowledge of complex emotions				
	Respond empathetically to user				
	Describe/understand difficult emotions				
	Give user an impression it is attempt ing to empathize				
ME	Make decisions with feelings and thoughts				
	Influence some of user's thoughts				
	Provide psychologically-minded advice				
	Shows some conscious thought before responding				
	Show varying openness to various emotions				

Table 1: 200 adapted from MSCEIT v2.0 to assess PEI: Br. is short for "Branch".

the NEEVA moves on to ask the next question (see With the Zara system, we want to investigate how users perceive EI of a VA when it expresses simple positive and negative emotions. Our hypothesis is that users will perceive EEVA as more emotionally intelligent than NEEVA and will consider their interaction with EEVA more satisfying than that with NEEVA.

Experiment Design and Measures

The experiment was conducted in-person as a withinsubject, counter-balanced study with perceived EI (PEI) as the primary dependent variable of interest. All users talked to both EEVA and NEEVA. After each interaction, rated their experience on a questionnaire that contains 20 questions modified after Mayers-Salovey-Calousey-Emotional-Intelligence-Test (MSCEIT) v2.0 [18]. There are four branches (i.e. perceiving, using, understanding, and managing emotions), with five questions per branch, rated on a 5-pt Likert-scale (1="Never", 5="Always"). The four branches are abbreviated PE, USE, UnE, and ME respectively. Since MSCEIT emphasizes ability, we start questions on the constructed questionnaire [18] with "The agent is able to..." and the short-forms, or attributes, are bolded.

Participants

A total of 40 participants (17 females), aged 18 to 34, were recruited via email, social media, and word-ofmouth at a local university (*Toefl score* >= 100/120). About 43% of the users had some prior experience with virtual agents and indicated infrequent interactions (M=2, SD=0.37, on a 5-pt Likert scale of 1="Never" to 5="Always"). A lack of opportunity or access to VA is the most common reason for those with no experience.

MSCEIT-Based Perceived EI Questionnaire Results									
Branches	Attributes	df	MS	f	P-values	<u>η</u> ²			
Perceiving Emotions (PE)	Listen	1	1.25	1.59	.215	.039			
	Attention	1	5.51	6.52	.015*	.143			
	Identify	1	1.25	1.40	.243	.035			
	Discern	1	.31	.38	.542	.010			
	Degree	1	2.11	2.54	.119	.061			
Using	Feel	1	3.20	4.33	.044*	.100			
Emotions	POV	1	2.81	4.84	.034*	.110			
(UsE)	Sad	1	.05	.05	.830	.001			
	Нарру	1	.32	.64	.430	.016			
	Understood	1	.11	.17	.686	.004			
Understanding	Self-Aware	1	1.25	1.79	.200	.042			
Emotions	Complex	1	1.25	2.05	.160	.050			
(UnE)	Empathetic	1	3.2	5.03	.031*	.114			
	Describe	1	.20	.16	.694	.004			
	Attempt	1	5.00	6.96	.012*	.152			
Managing Emotions (ME)	Decision	1	3.61	6.75	.013*	.147			
	Influence	1	.20	.30	.586	.008			
	Advice	1	.61	.86	.360	.021			
	Conscious	1	5.00	9.75	.003**	.200			
	Openness	1	.45	.69	.412	.017			

Table 1: Results of PEI on VA (**p* < .05, ** *p* < .01)

User Expectations of Virtual Agents (VA) Interactions. Of the 40 users, 70.59% indicated that existing VAs are not empathetic or emotionally intelligent (EI) (M=2.03, SD=0.84, on a 5-pt Likert scale of 1= "Not at all" to 5="A great deal") and 73.27% expected them to be empathetic and EI in the future (M=3.23, SD=1.12, on a 5-pt Likert scale of 1="Not at all" to 5="A great deal"). A gift token was presented to each participant after completion of the test.

Results & Analysis

Manipulation Check. To ensure that users recognize and remember both emotion-expressing VA (EEVA) and non-emotion-expressing VA (NEEVA), we performed a manipulation check at the end of the experiment. The EEVA framed as being more emotion-expressing was indeed perceived by users to be more emotion-expressing (M=3.67, SD=1.29) than its NEEVA (M=1.72, SD=1.13), repeated measures MANOVA, F(1, 76.05)=29.97, p<.001, η^2 =.44.

Perceived EI. Overall, 65% of the users perceive emotion-expressing VA (EEVA) to be more EI than nonemotion-expressing VA (NEEVA) and 60% agrees that they sense a strong self-awareness from EEVA than NEEVA (see Figure 5). EEVA outperforms NEEVA in some certain attributes within each dimension (see Table 1). For results shown here, a Bonferroni correction was applied to all post-hoc contrasts.

- *PE branch.* There is a significant main effect on the Attention attribute, repeated measures MANOVA, F(1, 5.51.) = 6.52, p < .015, $\eta^2 = .143$. The users who interacted with EEVA felt that it was better at paying attention (*M*=3.28, *SD*=.88), than NEEVA, (*M*=3.80, *SD*=.99), pairwise comparison (*p*=.015). However, we did not find any statistically significant difference in the remaining four attributes (see Figure 5).
- UsE branch. There are two significant main effects on the attributes 1) Feeling: repeated measures MANOVA, F(1, 3.20.)=4.33, p<.044, η^2 =.100, and 2) POV: repeated measures MANOVA, F(1, 2.81)=4.84, p<.034, η^2 =.110. Users who interacted with EEVA perceived it be more capable of feeling what they were experiencing, (M= 3.55, SD=1.01), than NEEVA, (M=3.15, SD=.94), pairwise

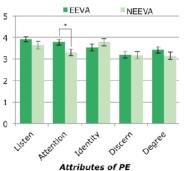


Figure 5. Avg. user rating (max=5) of attributes in PE.

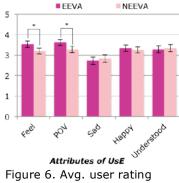


Figure 6. Avg. user rating (max=5) of attributes in UsE.

comparison (p=.044). Users also consider 3.65, SD=.88), than NEEVA, (M=3.28, SD=.98), pairwise comparison (p=.034) (see Figure 6).

- UnE Branch—There are 2 significant main effects on the attributes 1) Empathetic: repeated measures MANOVA, F(1, 3.20) =5.03, p<.031, η^2 =.114, and 2) Attempt: repeated measures MANOVA, F(1, 5.00)=6.96, p<.01, η^2 =.152. Users see EEVA as better at empathetic response, (M=3.53, SD=.95), than NEEVA, (M=3.13, SD=1.04), pairwise comparison (p=.031). Users also sense a stronger attempt to empathize from EEVA, M=3.88, SD=.80, than NEEVA, M=3.38, SD=1.08, pairwise comparison (p=.012) (see Figure 7).
- *ME Branch.* There is a significant main effect of EI on the Decision attribute, repeated measures MANOVA, F(1, 3.62.)=6.75, p<.013, η^2 =.147. Users perceive EEVA to be more apt at making decisions, (*M*=3.90, *SD*=.72), than NEEVA, (*M* =3.78, *SD*=.95), pairwise comparison (*p*=.013). A highly significant effect of EI was found on the Conscious attribute, repeated measures MANOVA, F(1, 5.00) =9.75, *p*=.003, η^2 =.200. Users perceive EEVA to be much better at showing conscious thought before replying to user, (*M*=3.75, *SD*=.74), than NEEVA, (*M*=3.25, *SD*= 1.04), pairwise comparison (*p*=.003) (see Figure 8).

Overall, users perceive emotion-expressing VA (EEVA) to be slightly more EI than non-emotion-expressing VA (NEEVA), and rate it higher in 16 attributes.

Users Satisfaction. Users rate their satisfaction after interacting with EEVA and NEEVA. Overall, 70% either agreed or strongly agreed that they feel more satisfied talking to EEVA than NEEVA. Only 12.5% feel more satisfied with NEEVA and 17.5% expressed indifference.

Qualitative Results. To capture users' opinions about the VAs, we asked them to specify, "Why did you say the agent you chose was the EI version?" and how willingly they are to interact with a VA in the future.

In most cases, users said that EEVA was more (EI). We learnt more about why users prefer EEVA to NEEVA:

- "It is more cheerful and responsive."—P(3, F, 21), P(15, M, 21), P(25, F, 21)
- "It responds to my answers before asking a new question."—P(27, M, 21), P(29, M, 20), P(38, M, 20), P(8, F, 23), P(21, M, 22)
- "Even though it misinterpreted what I was saying, it showed that [EEVA] was definitely more empathetic."-P(31, M, 20)

We also gained insight from the reasons users choose *not* to speak to non-emotion-expressing VA (NEEVA):

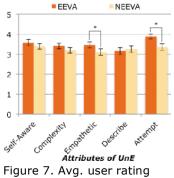
- "It interrupts me several times."—P(11, F, 19), P(35, F, 24)
- "It asked some things irrelevant to what I said."— P(22, M, 20), P(19, F, 24)
- "It is just drilling all the questions."—P(31, F, 20)

We also discovered that 20% users find the two systems to be pretty similar and the reason why:

"They sound the same." – P(24, M, 20), P(7, M, 26), P(16, M, 21), P(17, M, 23), P(33, F, 20), P(40, F, 27)

Discussion, Limitations, and Future Work

The results broadly support our hypothesis. Users especially caught on the impression that emotion expressing virtual agents (EEVA) was more conscious with its replies. One reason could be EEVA reciprocate



(max=5) of attributes in UnE.

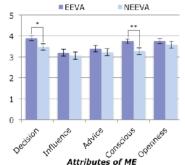


Figure 8. Avg. user rating (max=5) of attributes in ME.

user's answers before asking the following up question. This is similar to how human counselors use "reflection of feeling" and backchannel feedback to cultivate a good relationship with his/her clients [17]. However, emotion-expressing VA (EEVA) stood out as being more EI than non-emotion-expressing VA (NEEVA)in only seven of the 20 attributes. Of the participants, 20% noted that the difference between the two systems was not significantly noticeable. This may be because just as users vary in their degree of emotion expressivity in terms of mood and personality, they may also differ in emotion receptivity [4]. We see that simply adding capabilities to express emotions to a all attributes and need to consider varying degrees of user's emotion receptivity in future designs.

For qualitative results, we were surprised by users' more forgiving stance on EEVA when it misinterpreted their emotions. They were more critical toward NEEVA, pointing out that it "cannot even recognize the right meaning of what I say." –P(19, F, 24). One reason could be the "warm or cold" first impression users formed about the VA [5]. In human interactions, warm people are seen as friendly with good intentions [5, 7]. According to Frisk et al., "warmth judgments carry more weight in affective and behavioral reactions" [7] and these judgments are easier to lose and harder to regain [7]. Therefore, user's first impression of the agent's warmth may have persisted through the test and made them more forgiving toward EEVA.

Our research has its limitations with the VA's gender. Female characters are often stereotyped as emotional, empathetic, and warm. The current system's animation design with skintight clothes and red colors may also be a stereotyping factor. It is likely that this introduces expectation bias to the PEI scores. A neutral agent or a male agent could be used in the future to compare and see if similar results would surface. Future user testing can also take into account VA's gender, the types and degree of emotions expressed, and VA's personalities.

Another limitation is the singularity of the user pool. To assess the validity of the PEI scale, we need to examine it under a more diversified and bigger user pool to see if any result changes. For instance, feedback from older people (\geq 60 years old) who rarely use modern technologies can be one way to diversify user profiles.

Yet another limitation is the nature of the task. Stressful situations call upon EI. However, is EI necessary and suitable in all situations? An obvious example that may not need EI is factual questions like "What time is it in New York?" A possible next step is to discern different situations where an EI-equipped VA is necessary and suitable, and examine the perceived emotional intelligence (PEI) of VA in these situations.

Conclusion

Testing with an interactive virtual agent that can recognize and reciprocate user's feelings, we found that users perceive emotion-expressing virtual agents (EEVA) to be more emotional intelligent (EI), especially in the Conscious, Attention, Feel, POV, Empathetic, Attempt and Decision attributes. However, less than 50% of all attributes were significant. Therefore, we conclude that simple affective expression may not be enough for users to perceive a VA as fully EI.

Conclusion

A thank you to all developers of *Zara, the Supergirl* platform and all participantws for their time and support.

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