

Empathetic Conversational Artificial Intelligence Systems: Recent Advances and New Frontiers

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1 Title

Empathetic Conversational Artificial Intelligence Systems:
Recent Advances and New Frontiers

2 A two-sentence tutorial description

Recently, empathetic conversational artificial intelligence (ECAI) systems exhibiting emotional and social intelligence have attracted increasing attention from the natural language processing (NLP) community. Recognizing users' affective states (emotions, sentiment, etc.) and behavior and responding accordingly is key to successful communication. While it is straightforward for humans, instilling such human-like capabilities into conversational systems is a challenging task. Since ECAI systems encompass a wide spectrum of human attributes, it becomes difficult for beginners to get started. Therefore, this tutorial aims to present a comprehensive review of empathetic conversational AI systems. The scope of the tutorial includes the background, evolution, datasets, techniques including the state-of-the-art approaches, evaluation methods, and recent trends in ECAI systems. Finally, it points out a few limitations and shortcomings of the existing techniques to facilitate future research.

3 A two-paragraph tutorial description

The primary objective of developing conversational artificial intelligence (CAI) systems is to satisfy users' concerns and requests by imitating human communication. Recent years have witnessed the modeling of empathy in conversational systems to encourage pro-social conduct and strengthen the sense of social bonding. This tutorial intends to review empathetic conversational AI (ECAI) systems, whose goal is not inherently to respond to diverse requests from users but also to boost the perception and expression of emotional states and personal preferences. In this tutorial, we first introduce conceptual models of empathy [Davis and others, 1980; Goleman, 1995] and briefly present the relevance of empathy in CAI systems [McTear *et al.*, 2016].

Afterward, we discuss notable research works on ECAI systems. Specifically, we focus on the affective and social dimensions of the empathetic conversational systems. Emotion identification facilitates the selection of relevant empathetic responses [Rashkin *et al.*, 2019]. A few studies have

identified sentiments in user messages to respond empathetically [Chen and Nakamura, 2021; Lahoz-Beltra and López, 2021]. Following emotion-based generation, we quickly summarize the works that adhere to recognizing users' emotions/sentiments as a de-facto step towards empathetic response generation in CAI systems [Zhou *et al.*, 2018; Firdaus *et al.*, 2021a; Zheng *et al.*, 2021; Firdaus *et al.*, 2021b; Madasu *et al.*, 2022; Majumder *et al.*, 2022]. However, incorporating emotions and/or sentiments in the generated responses could only partially serve as the solution to empathetic response generation.

A more comprehensive ECAI system should have access to other information as well, such as emotion cause, user's intent, persona, social conduct (politeness), and external knowledge to enhance the sense of empathy in the generated responses. We intensively highlight several popular works that focus on these aspects to build better ECAI systems [Li *et al.*, 2021; Gao *et al.*, 2021; Xie and Pu, 2021; Zheng *et al.*, 2021; Zhong *et al.*, 2020; Firdaus *et al.*, 2020d; Saha and Ananiadou, 2022; Firdaus *et al.*, 2022d; Golchha *et al.*, 2019; Firdaus *et al.*, 2020c; Mishra *et al.*, 2022b; Firdaus *et al.*, 2022b; Mishra *et al.*, 2022a; Firdaus *et al.*, 2022a; Mishra *et al.*, 2023; Yoo and Jeong, 2021; Sabour *et al.*, 2022; Liu *et al.*, 2022; Li *et al.*, 2022]. In particular, we discuss the datasets, techniques and/or state-of-the-art (SOTA) approaches, and evaluation methods presented in these works.

Lately, multimodal information (a combination of features from different modalities - text, images, audios, and videos) has shown to be beneficial for emotion analysis in CAI systems [Poria *et al.*, 2018; Firdaus *et al.*, 2020b]. Hence, we also discuss the studies that have incorporated multimodal information in the empathetic response generation framework and have obtained significantly better performance compared to models with unimodal information [Tavabi *et al.*, 2019; Firdaus *et al.*, 2020a; Firdaus *et al.*, 2022c]. Further, a new line of research on ECAI systems has shown that empathetic CAI systems are capable of social influence (affect the users' emotional and cognitive responses, contributing to changes in their beliefs, attitudes, and behaviors through empathetic conversations). The key social influence tasks include persuasion and psychotherapy [Wang *et al.*, 2019; Althoff *et al.*, 2016]. We briefly present the noteworthy study on this emerging trend in ECAI systems [Samad *et al.*, 2022; Mishra *et al.*, 2022c; Sharma *et al.*, 2020; Sharma *et al.*,

85 2021].

86 Finally, as a conclusion, we draw attention to the limita-
87 tions and shortcomings of contemporary approaches so that
88 it will be simpler for beginners to advance research on ECAI
89 systems.

90 4 Proposed length of the tutorial

91 1/2 day (consisting of one slot or two 1:45h slots)

92 5 Outline of the Tutorial

93 This tutorial is organized as follows:

94 • Conceptual models of empathy (15 minutes)

95 We will introduce the fundamental conceptual models of
96 empathy, including the different types of empathy and
97 the individual differences in the perception of empathy
98 [Davis and others, 1980; Goleman, 1995].

99 • Need for empathy in conversational AI systems (15 100 minutes)

101 Empathy is regarded as a necessary trait, and studies
102 have been undertaken to enhance empathy in humans in
103 a variety of contexts [Spinrad and Gal, 2018]. Computa-
104 tional modeling of empathy helps in better comprehending
105 human relations [Yalcin and DiPaola, 2018]. Thus,
106 we will precisely talk about the works which suggest that
107 the incorporation of empathy in CAI systems could en-
108 hance the user experience and contribute to bridging the
109 human-machine gap [McTear *et al.*, 2016].

110 • Empathy-related concepts in ECAI systems (135 111 minutes)

112 An ideal ECAI system is expected to exhibit emotional
113 and social competence. In this part of the tutorial, we
114 will first introduce the various concepts related to empa-
115 thy and then points out the works that we will cover in
116 detail.

117 – *Emotion/Sentiment*: Emotion/sentiment helps in
118 understanding users’ feelings/attitudes towards an
119 entity. Ever since the introduction of the EM-
120 PATHETICDIALOGUES dataset [Rashkin *et al.*,
121 2019], the interest in building ECAI systems has
122 spiked. There have been works that attempt to iden-
123 tify the emotion and use the predicted emotion label
124 to generate emotionally consistent responses [Zhou
125 *et al.*, 2018; Madasu *et al.*, 2022]. The multiple
126 emotions and corresponding emotional intensity-
127 based empathetic response generation is presented
128 in [Firdaus *et al.*, 2021a]. The key factors of empa-
129 thy expression (communication mechanism, dia-
130 log act, and emotion) are modeled in a hierarchi-
131 cal way in [Zheng *et al.*, 2021] for empathetic
132 response generation. The correlation between sen-
133 timent and emotion has been exploited in [Fir-
134 daus *et al.*, 2021b] for generating empathetic re-
135 sponses. The distinct elements of human communi-
136 cation, *viz.* emotional presence, interpretation, ex-
137 ploration, and sentiment are used to guide the re-
138 sponse generation towards empathy in [Majumder
139 *et al.*, 2022].

– *Emotion cause*: Emotion cause facilitates under-
140 standing the underlying cause of a particular emo-
141 tion for generating more relevant empathetic re-
142 sponses. Emotion cause identification in ECAI sys-
143 tems studies has been an active area of research
144 in recent times due to significant performance im-
145 provements [Li *et al.*, 2021; Gao *et al.*, 2021].
146

– *Intent*: Recognizing the users’ intent in conversa-
147 tional systems is essential for delivering the req-
148 uisite information to the users. A few studies on
149 ECAI systems have attempted to identify users’
150 intent during the conversation. In particular, the
151 works in [Xie and Pu, 2021] identify 8 new in-
152 tents and combine these intents with 32 emotion
153 categories from the EMPATHETICDIALOGUES
154 dataset [Rashkin *et al.*, 2019] to form a hybrid
155 emotion-intent category. This emotion-intent cat-
156 egory data is utilized in another ECAI study to em-
157 phasize its usefulness [Zheng *et al.*, 2021].
158

– *Persona*: Persona-based research in CAI systems
159 necessitates that the agent exhibits a specific per-
160 sonality while interacting with users. Persona is
161 highly correlated with personality, which in turn
162 influences empathy. A Persona-based Empathetic
163 Conversation (PEC) dataset has been created in
164 [Zhong *et al.*, 2020]. Several other studies have
165 utilized persona-aware techniques for enhancing
166 the performance of empathetic response generation
167 [Firdaus *et al.*, 2020d; Saha and Ananiadou, 2022;
168 Firdaus *et al.*, 2022d]. The results of these ECAI
169 studies indicate that persona influences empathetic
170 conversations more than non-empathetic ones.
171

– *Politeness*: Politeness is an important virtue of hu-
172 mans and to make the CAI agents behave more like
173 a human, it is necessary to induce politeness in such
174 agents. Politeness is vital for developing a cordial
175 and empathetic connection with the users [Mishra
176 *et al.*, 2022b]. Lately, certain studies attempt to
177 build CAI systems that incorporate polite behavior
178 in customer care agent’s responses [Golchha *et al.*,
179 2019; Firdaus *et al.*, 2020c; Mishra *et al.*, 2022a;
180 Firdaus *et al.*, 2022a; Mishra *et al.*, 2023]. The
181 work in [Firdaus *et al.*, 2022b] generates polite re-
182 sponses based on the age and gender of the users to
183 offer a more personalized experience to them. The
184 findings of these studies indicate that politeness im-
185 proves the empathetic nature of the agent.
186

– *External knowledge*: The CAI agents sometimes
187 need to have access to external knowledge for gen-
188 erating more informative and relevant responses.
189 The use of such knowledge sources in ECAI sys-
190 tems research facilitates a more in-depth compre-
191 hension of the user’s circumstances/feelings by
192 deducing the implicit information inherent in the
193 user’s responses. For instance, from the user’s re-
194 sponse “*I have cough issue for the last 10 days.*”,
195 the agent can infer the correct emotional state of the
196 user by accessing the knowledge bases like Con-
197

309 and avoid reproducing or amplifying social inequalities.
310 Research presented ensures that conversational AI is fair
311 to all users, regardless of their race, gender, age, or other
312 characteristics.

313 • **Transparency and Explainability:** Conversational AI
314 systems should be transparent about their capabilities
315 and limitations, as well as how they make decisions.
316 Users should understand how their interactions with
317 conversational AI are being analyzed and used. In this
318 regard, we provide detailed evaluations both automatic
319 and human to make the findings in this tutorial com-
320 pletely transparent to the readers.

321 • **Empathy and Emotional Intelligence:** Conversational
322 AI should be developed with empathy in mind and take
323 into account the user’s emotional state during a conver-
324 sation. AI should not only respond correctly to users’
325 statements but also identify their emotional state and re-
326 spond in an appropriate manner. This ethical considera-
327 tion is the key topic of the tutorial that has been actively
328 considered in the research presented.

329 • **User Empowerment and Control:** Conversational AI
330 should be designed with the user in mind, allowing them
331 to control the conversation and make decisions about
332 what information is shared. Conversational AI should
333 be respectful of users’ autonomy and provide a choice
334 to end the conversation or request additional assistance.
335 The datasets designed for different research topics dis-
336 cussed in this tutorial takes care of user’s autonomy.

337 Overall, ethical considerations for empathetic conversa-
338 tional AI research should prioritize the needs of the users
339 while maintaining privacy, fairness, and transparency. Devel-
340 opers should seek to promote user empowerment, emotional
341 intelligence, and empathy while avoiding harmful biases or
342 discriminatory practices.

343 9 Presenters

344 Priyanshu Priya, Mauajama Firdaus, Kshitij Mishra, and Asif
345 Ekbal. The CVs of all the presenters are attached to this pro-
346 posal.

347 10 Relevant publications from research group

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358 (EACL), May 2-4, Croatia (accepted) (Core - A; h5-
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tions, Neurocomputing, 94, 242-254, Elsevier (IF-5.719; 399
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