Let's Talk it through, anew: Promises and Pitfalls of Customisable Conversational Reflection Support

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ABSTRACT

As modern lifestyles are becoming increasingly stressful and ever more hectic with multiple stimuli constantly competing for our attention, Affective Disorders (ADs) such as anxiety and depression are on the rise. Consequently, due to the burgeoning demand for counselling and therapeutic services, many people who suffer from ADs are struggling to timely access the professional support that they require. To address this problem, voice-enabled Conversational Agents (CAs) have been recently proposed as tools for supporting self-reflection and providing assistance in managing a range of ADs through synthetic voices. However, despite their therapeutic potential, CAs offer a very limited choice when it comes to selection and personalisation of synthetic voices used. The goal of this paper is two-fold: (1) it discusses the potential benefits that a CA's voice customisation can bring to enhance user engagement and promote long term self-reflection, and (2) it offers reflection on the corresponding challenges associated to this approach.

CCS CONCEPTS

• **Human-centered computing** → **Human computer interaction (HCI)**; *Auditory feedback*.

KEYWORDS

self-reflection; conversational agents; behavioural change; synthetic speech

1 INTRODUCTION

> "Between stimulus and response there is a space.
> In that space is our power to choose our response.
> In our response lies our growth and our freedom." [16]
> —*Viktor E. Frankl*

Self-reflection has proven to be a highly effective tool in reducing stress levels [9, 37] and helping people to manage their social anxiety [3, 35]. Thanks to the proliferation of smartphones and

other types of mobile devices, digital interventions are currently used to support a growing number of users in improving their well-being [10, 15, 20, 22], enhancing productivity and focus [4, 18], and making them more physically active [23, 36, 39]. However, regardless of its potential to bring about positive behavioural change, self-reflection is a highly-complex and challenging task which requires dedication and long-term commitment. According to Nahum-Shani et al., who coined the term 'Just-in-Time Adaptive Interventions' (JITAIs), an effective digital intervention needs to adapt to the ever-changing state of the user and account for the interaction context in order to successfully capture attention and elicit desirable behaviour [31]. The key factors that determine individuals' adherence to JITAIs are engagement and fatigue [31], and ensuring a proper balance between the provision of the automated support and personal volition and agency of the user [34]. Indeed, recent research shows that users value mental health applications that are straightforward to use, offer a variety of options to chose from, and allow for customisation of the functionalities provided [2].

In this position paper, building on previous research on self-reflection tools, we will discuss how the application of personalised and customisable, synthetic speech technology can help to increase adherence to digital interventions by increasing engagement, reducing fatigue, and giving users the ability to exercise their agency. We will also consider challenges posed by this approach and make some recommendations for the future research agenda for self-reflection CAs. The idea of providing users with more control over how they receive the CA's support is based on the view advocated by Mobbs et al. who consider 'free will' as a fundamental prerequisite in self-reflection and self-improvement, since people need to believe that they are in full control of their actions in order to strife for change [30].

2 SUPPORTING SELF-REFLECTION IS A CHALLENGING TASK

Research indicates that digital interventions tend to become less effective over time and ultimately lead to abandonment after just several uses [17]. In the context of eHealth applications, this phenomenon is known as 'the law of attrition' [13]. In order to reduce the likelihood of abandonment and increase adherence to digital interventions, a person needs to perceive them as useful and trustworthy [21]. What is more, since processing digital interventions requires engagement of cognitive resources, it is crucial to adjust the scope and style of the message, as well as its delivery frequency so that the user is receptive of it [14]. Otherwise, in the long term, too high frequency of messages that lack variability can lead to intervention fatigue [11], causing participants disengage from the intervention due to boredom or cognitive overload.

Given that signal intervention fatigue can be detrimental to engagement [6], it is crucial to deliver interventions in a way that is engaging. This can be achieved by varying the form of content delivery [26] and giving the user more control regarding how the supportive process is being delivered [31]. The main challenge is to determine *how* to best incorporate participant-determined features to digital interventions to balance the provision of the automated support with personal volition and agency. Here, we propose that providing more control over synthetic speech used by the self-reflection CA could help to achieve this goal.

3 POTENTIAL OF SPEECH IN SELF-REFLECTION

Research indicates that even non-interactive learning materials such as audio recordings can increase perception of social presence and, in turn, result in an improved learning performance [29]. This effect is linked to a phenomenon called 'social facilitation' [38], where performance is improved as a consequence of the felt social presence. Research also shows that feedback provided through media materials helps to establish a stronger social presence of the instructor [1]. Human voices were also identified as triggers for increased social presence [24]. Thanks to recent technological improvements, the quality of synthetic speech is currently sufficient for credible and effective delivery of verbal information [8]. This, in turn, provides an opportunity to reliably convey social cues during self-reflection and improve intervention adherence.

Indeed, recent studies (e.g., [27, 28] acknowledge the usefulness of voice-enabled CAs in supporting self-reflection. Maharjan et al. [28] found that a smart-speaker-based CA 'Sofia' was perceived as significantly more attractive and novel compared to a web-based self-reflection application that did not feature speech. While there was no significant difference in terms of user engagement, Sofia's engagement scores remained consistent over the span of four weeks while the web based app engagement decreased as the trial went on. The authors highlighted that the currently available CAs are lacking personalisation, which in the long term may lead to discontinued use, as the novelty effect fades away [28]. Mharjan et al. [27] have also found that perception of Sofia and the ways in which participants personified this CA differed between the users — with some people exhibiting a positive while others a negative sentiment towards the agent. Therefore, in order to best support a subjective users' experience, individual needs and personal differences need to be accounted for. Accordingly, self-reflection CAs need to 'provide space for human experiences of socialising, connectedness, empathy and compassion, while allowing users to appropriate technology in the ways they see fit' [27]. We believe that giving participants control over type of the voice used by their CA could facilitate the self-reflection and make it more engaging and effective.

Studies with human speech showed that enthusiastic voices are ranked higher in terms of their social and affective qualities and lead to an increased focus and better performance, as compared to calm voices [25]. When it comes to synthetic voices, it was observed that high-quality synthetic speech received higher ratings for credibility and engagement, compared to low quality synthetic speech [5]. A high-quality synthetic speech has also been shown to increase perception of trust [7]. However, it should be noted that while speech has a clear potential to motivate users, the vast majority of

possible voice factors, such as tone, pitch, intonation, etc. have not been fully explored yet in terms of social processes [33]. This opens up new research avenues for evaluating the impact of synthetic speech in the context of digital interventions.

We believe that the HCI community who work on design and development of reflection interventions should consider the following challenging research gaps, in order to make the design of CA more engaging and improve the prospects of long-term adherence.

- ***How to select the voice of the agent to increase user engagement?*** Dubiel et al. [12] demonstrated that the type of speech corpus selected for voice development should be aligned with the intended purpose of its usage.
- ***Should the CA match the conversational style of the user in terms of prosodic qualities of the voice?*** Hoegen et al. [19] found that individuals with High Consideration conversational styles were more likely to trust a CA that matched their conversational style.
- ***How should we pick the personalisation criteria?*** Should personalisation criteria be determined by personality questionnaires, or automatically elicited over the course of the CA use, based on the user's behaviour?
- ***How to prevent habituation and over-reliance?*** Should users be encouraged to change the type of the voice used by CA? If so, at what point should such a recommendation be delivered?

We argue that the discussion of the above points could help in determining the future research agenda for reflection-support CAs, and provide insights if engaging and customisable speech can facilitate the long-term intervention adherence.

4 PITFALLS OF CUSTOMISABLE SPEECH SUPPORT AND THE ROAD AHEAD

Finally, we would like to acknowledge that while customisable self-reflection CAs can bring a range of benefits, there are also several pitfalls that should be considered. First, long-term exposure to CA can result in addiction and in turn harm real-life intimate relationships, as in the case of the AI-driven social chatbot Replica [40]. Second, and related to first point, as the relationship with the CA develops, users may rely on it exclusively and cease to seek a professional support from therapists and counsellors whose expertise may be necessary for the user to properly manage their AD and prevent it for deteriorating. Third, on the point of accessibility, not all users may have access to devices that support speech interaction, or due to hardware limitations can only support low-quality synthetic voices, therefore it is important to consider alternatives that can be used to bridge this digital divide.

Overall, the goal of implementing personalised synthetic speech is to facilitate self-reflection and make it more engaging and less taxing. However, it should be noted that while important, it is not considered to be a replacement for human interaction. Therefore, in addition to voice customisation, the design of voice-enabled CAs should also promote social interactions such as pointing users to relevant online communities and peer groups that could offer users further support. Consequently, this could further promote user agency giving people more choice over the intervention options [32]. The ultimate goal for making CAs more customisable is to provide

support that is user-centred rather than intervention-centered. As posited by Nahum-Shrani et al. [31] while theoretical overviews can guide developers of self-reflection tools, cross-disciplinary collaborations and users evaluations are required to ensure that the technology is appropriately used. This paper provides a speech processing perspective that should be validated through experimentation.

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