

A Multidisciplinary Approach of Exploring the Synergy of Human Creativity and Artificial Intelligence

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Recent advances in artificial intelligence (AI), and generative models in particular, are opening up new discussions about the types of tasks humans should focus on in the future and the role of creativity in human-AI collaboration. In this paper, I propose to investigate the mechanisms underlying the synergy between human creativity and AI through a multidisciplinary approach. By conducting several experiments in which human participants interact with AI systems to cooperatively solve creative tasks in, for instance, the fields of theater, photography, web design, or art, insights will be gained into how human-AI co-creativity can support and enhance creative approaches to different types of challenges. This work reports on ongoing research and future steps. My research intends to contribute to a deeper understanding of the potential of human-AI co-creativity and how it can positively foster human-AI collaboration.

CCS Concepts: • **Human-centered computing** → *Empirical studies in HCI; User studies; Collaborative interaction.*

Additional Key Words and Phrases: Generative Artificial Intelligence, Human-AI Collaboration, Human Creativity Support and Augmentation, Human-AI Co-Creativity

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1 INTRODUCTION

Artificial intelligence (AI) has made significant strides in machine learning, deep learning, and generative approaches in recent years. This progress has major societal impacts and is shifting the way we work and learn. Technological advances bring many benefits, yet we still face the challenge of not knowing how to use new technologies. Thus, I argue that it is crucial to focus on human empowerment, for example, by focusing on human creativity and exploring how to design collaborative systems that enhance human capabilities rather than replace them. In research, there has been growing interest in the human-AI co-creativity field [4, 10–12]. In this context, co-creativity refers to the collaborative process between computers and humans to produce shared creative artifacts [7]. In this work, I propose to investigate the mechanisms underlying the synergy between humans and AI in the area of creativity and the contributing factors.

Developments in AI, such as human-computer collaborations [18, 23, 24], mixed initiative systems [20, 25], and the field of AI art [5, 9, 26] have opened up new possibilities for co-creative interactions between humans and AI. The emerging areas show the need for a more dynamic and flexible form of interaction where humans and AI can work together seamlessly and naturally. However, despite this potential, there is a lack of understanding the fundamental dynamics and factors that contribute to collaborative partnerships between humans and AI systems. Researchers anticipate that AI will be used extensively as a tool that assists collaborative creativity and interactivity in the near future [1], making AI systems play an increasingly important role in creative fields, for instance in art [5]. Ultimately,

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this will reverse the traditional roles and humans and computers, with the human solely dictating and the computer supporting [6].

A key argument for investigating human-AI co-creativity is the potential for AI to assist and augment, rather than replace human creativity in creative tasks [1, 4]. Creativity is a concept that is connected to subjective ideas and perceptions and thus enables innovation [17]. Generative AI models are able to generate novel and unique content, but may not be able to provide all capabilities to mimic the entire range of human creativity [1]. I, along with other researchers [8], believe that there is a lack in exploring the interaction techniques of generative AI models and applications that support co-creativity through human-centered design steps. Thus, my research intends to address this research gap.

Given the need for user studies in this area [11], I plan to conduct a series of multidisciplinary experiments in which human participants interact with AI systems of creative environments. By examining the outcomes of these co-creative processes, I aim to gain a better understanding of how human-AI co-creativity can be leveraged to support and enhance creative task-solving. Moreover, I am concerned with the artifact of creativity, which is not necessarily a work of art. Creativity can manifest itself in various (intangible) forms such as new ideas, solutions to problems, or innovative approaches to existing tasks. Curiosity and experimentation often motivate creative endeavors which require combining ideas from different fields or previous experiences in a meaningful yet abstract way [1]. Human-AI co-creativity has the potential to enhance the artifact of creativity by offering novel approaches and resources for idea generation and development. But, effective collaboration depends on the AI system's ability to support the human creative process, and the human's ability to integrate and utilize the AI's capabilities. In addition to that, I will explore ethical considerations involved in human-AI co-creativity, since technological progress must be accompanied by a better understanding of ethical concerns, bias, and broader social implications [1]. Hence, research on this topic will provide guidelines and best practices for future researchers, software developers, and AI practitioners to ensure responsible AI use.

2 RELATED WORK

This section briefly outlines related work in the area of co-creativity between humans and AI.

Given that creativity itself is a vast and far-reaching endeavor that cannot be quantified even for humans [6], co-creativity subsequently spans across multiple research domains, too. Co-creative music generation [19] and exploration [3], human-AI collaborations in live artistic performances [23], co-creative sketching tools [14, 15, 21], or co-creative AI for public spaces [16] are just a few examples for the broad area. In the field of Human-Computer Interaction (HCI) the current landscape of AI-empowered co-creative tools are outlined in a review by Hwang [12]. The author notes that most creativity tools are designed for the later phases of creativity instead of the early stages of creativity [12].

Another emerging field highly relevant to human-AI co-creativity is "AI art," as a growing number of artists incorporate AI into their work. The work of Cetinic and She [5] shows a review on AI art which investigates how AI is used to analyze art and applied to digitized artwork collections, and also how AI is used for creative purposes and generation of novel art [5]. Practical and theoretical aspects of AI art are outlined, as well as an overview of existing AI art datasets [5]. Further, Grba [9] critically examines AI art by addressing its practices within contemporary art, AI research, and related fields. Ways to advance the field are explored, with the author offering new perspectives on the critique of art, science, and technology in a world where they influence values and shape political norms [9].

Moreover, there has been progress in the development of frameworks and tools for analyzing and facilitating co-creation in research. Grabe et al. [8], for example, propose a framework to analyze human-GAN collaboration in co-creative GAN applications. The authors define four interaction patterns (Curating, Exploring, Evolving, and

Conditioning) for better understanding the affordances and limitations of co-creation [8]. The authors note that designing human-AI interaction remains a challenge due to the complex and non-interpretable nature of AI models [8]. Guzdial and Riedl [10] present an interaction framework between human users and AI agents designed to support human creativity. The framework provides a system where a human user and AI agent work together in a turn-based fashion to design an artifact [10]. Rezwaner and Maher [22] demonstrate a framework for modeling interaction in co-creative systems that can serve as a tool for identifying research directions and gaps in the field of co-creativity. The results of [22] show that the scope of capabilities in the field of co-creativity is not fully explored.

In summary, the domain of co-creativity between humans and AI requires further research as the design of human-AI interaction is under-explored and remains challenging due to the complex nature of AI models [8, 10]. There is a scarcity of user-centered studies and designs, as well as systems that incorporate AI to human communication [22].

3 METHODS

This section presents the methodologies employed in the endeavor of this work. Overall, the work follows the human-centered design (HCD) process [13]. Thus, the first step is to identify the Context of Use (CoU). To this end, an in-depth literature analysis and two focus groups are foreseen. One focus group (FG) shall be conducted with non-experts, representing people with limited experience with human-AI co-creativity. The second FG is planned for experts, meaning people who use human-AI co-creativity in their daily lives and work. The first FG has already taken place, the results of which are reported in the next Section 4. The aim of the FG is to identify how much knowledge there is about human creativity, human co-creativity, AI creativity and human-AI co-creativity. Discussions on the acceptability of co-creative systems and where people can imagine using such systems in their lives are included.

The second step is the Requirement Analysis (RA), for which several steps are foreseen: (1) Conducting user studies to comprehend the design requirements and preferences of possible users (for instance, in form of questionnaires or semi-structured interviews with AI-using artists), (2) Workshops with art enthusiasts, in which participants interact with different versions of AI-assisted systems designed to elicit different types of co-creative interactions, and (3) Analysis of several interactive media processing web apps developed as part of my work project. Based on this RA conclusions will be made and requirements for possible prototypes will be deducted.

The third and fourth steps of the HCD process are implementation and evaluation of prototypes. Therefore, participatory design, a method of involving users in the design process, will be employed to obtain continuous feedback on design decisions. A first practical work in the form of a co-creative project with a state theater is planned and will be staged as a theater installation. This will be evaluated and the feedback will be utilized to improve the prototype. In a second round of implementation, another interactive, co-creative prototype for digitized photo products will be planned and executed. Please note that the possible realizations depend on the results of the previous analysis, due to which deviations may occur.

4 OUTCOMES

This section demonstrates preliminary and possible future outcomes of this work. The results of the first FG are briefly summarized here. Please note that no other results can be reported at this time.

The goal of the FG was to identify the CoU (cf. Section 3). With the help of five participants (1F, 4M) aged between 23 and 51 ($\bar{O}=33$, $SD=9.7$) four topics were explored: (1) Creativity in peoples' everyday life, (2) Situations when people are creative with others, (3) Familiarity with computer-based creative systems, and (4) Interactions with co-creative AI systems. All participants were non-experts regarding the use of co-creative systems, four of whom worked in

computer science (one was an interior architect). The FG was held in person and lasted two hours, with me as moderator, interactively guiding the participants through the FG. The discussions and feedback were evaluated using a thematic analysis (by [2]) that yielded three main themes, which are briefly described here.

Versatile Human Creativity. Participants have felt personally creative and co-creative in a variety of areas: (A) "Expressive Activities", such as cooking, crafting, gift wrapping, or making music, which involve creating something new and unique, (B) "Active Lifestyle", such as walking, traveling, dancing, preferring mornings to evenings, quiet contemplation, each reflecting physical activity, personal preferences, and choices related to habits, and (C) "Job-related Productivity", such as collaborative programming, identify research potentials and new project ideas, organize events or workshops, and help clients develop design ideas, which are all about professional growth and advancement.

Interest in Collaborative Human-AI Systems. All participants were interested and curious about co-creative systems. They could all envision the use of co-creative systems, mainly focusing on professional environments and, particularly, on reciprocal collaboration. One subject said: "Clear, distinct feedback during interaction between human and AI systems is important [...]. The lack could have a negative impact on my creativity [...]" (P4).

Risks and Uncertainties of AI. Participants discussed the negative effects of AI on human creativity. They mentioned how relying too much on AI can lead to a loss of self-confidence in one's own creativity (P1), hinder one's own creativity (P3), and even replace and decompose human creativity (P5). A participant stated for instance "AI systems could offer so many creative solutions so much faster and so much better, that the creatively thinking person would feel disengaged and say 'Oh well, I won't bother at all then' " (P5). Another participant mentioned risks such as a negative "stubborn" AI and the potential for uncanniness (P2).

It should be noted that the results presented here with N=5 are not generalizable and do not include all aspects relevant to the intended research. However, it demonstrates a first major step into the determination of the CoU. Based on these findings, the second FG will be conducted, which will provide further insights into the understanding and underlying dynamics of co-creative human-AI synergies. This work may yield guidelines that provide ethical approaches to this topic, as well as other interactive web applications that support co-creativity in the early stages of human creativity. Also, two co-creative prototypes (one in the field of theater) are planned to be developed and evaluated during this work.

5 CONCLUSION

The goal of this research is to explore the mechanisms behind the synergetic relationship between human creativity and AI. Through a multidisciplinary approach, the objective is to gain insights into how co-creativity between humans and AI can function and how AI can complement, rather than replace, creative human approaches to address challenges in various domains. The focus group, conducted with human participants discussing co-creative AI systems, provided valuable insights into the potential of co-creativity between humans and AI and how it can be leveraged to foster positive human-AI collaboration. Although highlighting the potential benefits of human-AI co-creativity is crucial, it is also essential to acknowledge the respective challenges of such collaboration. The risks and uncertainties of AI models and the associated lack of feedback could cause user frustration and make the usage of such systems challenging.

Overall, by providing insights into the mechanisms behind the synergistic relationship between human-AI creativity, this work plans to open up new avenues for future research and development in this area. The preliminary results of this paper show the potential of human-AI co-creativity systems for enhancing and augmenting creative approaches to various challenges, and point out the need for a responsible and ethical approach to human-AI collaboration.

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