

## **WIP: Gen AI in Engineering Education and the Da Vinci Cube**

**Tammy Mackenzie, The Aula Fellowship**

EcoTech CEO, inventor, MBA, human rights activist, philosopher, and researcher of the intersections between strategic management, institutions, and systems theories.

**Dr. Lisa D. McNair, Virginia Tech**

Lisa D. McNair is Professor of Engineering Education and Director of Arts and Education at the Institute for Creativity, Arts and Technology (ICAT) at Virginia Tech. She is an executive committee member for a2ru and an editorial board member for Ground Works journal. Her research and teaching interests include developing interdisciplinary project-based learning experiences, building networks between university, industry, and community sectors, and expanding engagement in science, engineering, arts, and design. McNair's current projects include building the Interdisciplinary Projects (IDPro) program and a 3D manufacturing module series in undergraduate engineering at Virginia Tech, framing the da Vinci Cube innovation model, and co-designing tools with communities for collaboration on Alaska housing issues. She earned a PhD in Linguistics at the University of Chicago, and an M.A. and B.A. in English at the University of Georgia.

**Rubaina Khan, University of Toronto**

Rubaina Khan is a research associate in Queens University and a doctoral candidate in the Department of Curriculum, Teaching, and Learning at the Ontario Institute for Studies in Education (OISE), University of Toronto with a collaborative specialization in Engineering Education. She also has an M.A. in Curriculum and Pedagogy from OISE and an M. Sc. in Computer Control and Automation from the Nanyang Technology University in Singapore. She worked as a senior research engineers at MIT developing navigation technologies for underwater robotics to model and predict environmental issues in the coastal regions of Singapore. As always interested in education, this led her to take up a position as a lecturer at Singapore Polytechnic. She spent the next five years developing interdisciplinary engineering design courses, designing state-of-the-art classrooms, and delivering activities to promote engagement and motivation in the classroom. Her current research interest lies at the intersection of engineering design education, learning communities, and professional identity formation.

**Mr. Animesh Paul, University of Georgia**

Animesh was born in Tripura, India, and raised in a liberal modern "brown" military upbringing. He prefers the pronouns "He/They" and considers himself a creative, sanguine, and outgoing individual. He graduated with a bachelor's degree in Technology focusing on Electronics and Electrical Engineering from KIIT University. He is now a part of the Engineering Education Transformation Institute as a Ph.D. student under the advisement of Dr. Racheida Lewis. His research is in Engineering Education, focusing on equity, inclusion in the classroom, and easing student transition to the workforce catering to STEM graduates.

**Dr. Sreyoshi Bhaduri, Private Corporation**

Dr. Sreyoshi Bhaduri is an AI scientist. Currently, she spearheads innovative research in applying generative AI to solve complex supply chain logistics and operations challenges. Her expertise spans applied statistics and natural language processing, with a PhD from Virginia Tech and specialized training in Responsible AI from MILA. Sreyoshi has been recognized as a Graduate Academy for Teaching Excellence (VTGrATE) Fellow, a Global Perspectives Program (GPP) Fellow, and was inducted in the Bouchet Honor Society in 2017. Sreyoshi is committed to demystifying and democratizing generative AI solutions and bridging the gap between theoretical research and practical applications using AWS technologies.

# Work-in-Progress: Gen AI in Engineering Education and the Da Vinci Cube

## **Abstract**

As generative AI (GenAI) tools rapidly transform the engineering landscape, a critical question emerges: Are current educational innovations adequately preparing engineers for the socio-technical challenges of the future? This work-in-progress paper presents two key contributions. First, we build on prior work presenting a systematic review of over 160 scholarly articles on GenAI implementations in engineering education, revealing a predominant focus on enhancing technical proficiency while often neglecting essential socio-technical competencies. Second, we apply an emerging framework—the da Vinci Cube (dVC)—to support engineering educators in critically evaluating GenAI-driven innovations. The dVC framework extends traditional models of innovation by incorporating three dimensions: the pursuit of knowledge, consideration of use, and contemplation of sentiment. Our analysis suggests that while GenAI tools can improve problem-solving and technical efficiency, engineering education must also address ethical, human-centered, and societal impacts. The dVC framework provides a structured lens for assessing how GenAI tools are integrated into curricula and research, encouraging a more holistic, reflective approach. Ultimately, this paper aims to provoke dialogue on the future of engineering education and to challenge the prevailing assumption that technical skill development alone is sufficient in an AI-mediated world.

## **1 Introduction**

We take as our starting premise that engineers have a responsibility to society, and consequently, that engineering educators have a responsibility to convey this to their engineering students. However, there are few measures for determining how these responsibilities that span socio-technical boundaries can be integrated into research in engineering education (henceforth, EE). The swift evolution of Artificial Intelligence (AI) technologies, notably

Generative AI (GenAI) and Large Language Models (LLMs), unlocks unprecedented opportunities in education. While existing reviews have successfully surveyed the opportunities of and challenges for GenAI in higher education (e.g., [1], [2], [3], [4], [5]), a significant knowledge gap still remains in engineering education. Specifically, we are investigating how researchers are approaching questions about the role of GenAI in engineering education.

In the past few decades, the post-secondary education landscape has evolved dramatically through massification initiatives to respond to growing societal demands on engineers and to increase participation [6]. As such, post-secondary institutions are dealing with issues like resource constraints around designing curriculum and evaluation strategies [7] [8] [9] [10], ensuring equitable and inclusive access to learning [11] [12] [13], and providing flexible pathways for fresh graduates [14] [15] [16]. Significantly, given public financing cuts on education around the world [17], and the repercussions from the recent pandemic [18] [19], institutions are finding ways to innovate using GenAI as a learning technology. The use of machine learning and natural language processing in engineering education research and classrooms is not new [20] [21] [22] [23]. The widespread potential of GenAI in post-secondary education provides opportunity for major innovations in teaching and learning.

Researchers in engineering education have found that GenAI can be used to improve administrative tasks and backlogs [24], to facilitate resource allocation [25], and to enable higher engagement with students through personalized mentoring [26]. These promising use-cases have led to large-scale investments, disrupting traditional ways of learning and tutoring [27].

However, GenAI has raised concerns among educators and institutions, primarily dealing with plagiarism [28], renegotiating the role of instructors [29], and ethical concerns, for example around using student data [30]. These tensions have led researchers to beseech developers to create solutions that speak to systemic barriers [31], employ transparent methodologies [32], and co-design with educators [33]. In recent research in EE, the issues discussed above are prevalent along with calls for EE programs to ensure that their graduates have the technical skills to develop products and processes embedded in complex systems that work seamlessly [34]. Further, these systems must be developed with sustainable mindsets and use ethical design methodologies [35]. However, such sophisticated teaching and learning expectations are not always adapted to the realities of large classroom sizes and budget constraints. Therefore, GenAI holds both promise and challenges for engaging learners from various backgrounds with dynamic, personalized, and effective tools [36]. Further research

is needed to explore the performance of GenAI tools in teaching epistemic content [37] and in the context of the professional skills needed in industry [38], and to engage in the critical thinking required to solve "wicked" problems [39]. Lastly, EE training needs to show the implications of GenAI and engineering work in a society [40] that requires development of human values [41], empathy development [42], lifelong learning [43] and taking on responsibility for sustaining our planet [44]. To map early patterns in EE-based research on GenAI, we frame the findings of a systematic literature review within the dVC framework that foregrounds the dimension of Contemplation of Sentiment. Our work in progress paper thus offers a comprehensive summary of key innovations and motivating perspectives.

## 2 Methods

### 2.1 Building on a Systematic Literature Review

We chose a systematic review to highlight the range of applications, key pedagogical approaches, and motivations for GenAI integration in engineering curricula and classrooms. The review [45] uses the Search-Screen-Appraise approach [46] as visualized in Figure 2, to set strict inclusion criteria and engage in a precise study selection process. Initial searches were conducted across prominent databases: ERIC, PsycINFO, CINAHL, Web of Science, and Engineering Village, chosen for their comprehensive coverage of education and engineering literature. Keywords including "Engineering," "Education," and "GenAI" were used in each database to ensure relevance and comprehensiveness. A systematic review tool ([47]) was used first, to identify and exclude 40 duplicates from a total of 347 initial search results, resulting in 307 unique records for further evaluation. Abstracts and titles were further screened to identify papers specifically relevant to EE and discussing GenAI. Results of this work are reported in detail elsewhere (blinded for review).

### 2.2 da Vinci Cube Framework

The da Vinci Cube framework is a novel approach that extends the traditional Pasteur's Quadrant model [48] by introducing a third axis: Contemplation of Sentiment [49–51]. In development through a qualitative user study, the new framework is designed for exploring the drivers of innovation and incorporating the crucial role of emotion, human-centered, and ethical considerations in decision-making processes and actions of individuals and organizations.

Stokes's Pasteur's Quadrant expands two axes - basic research and applied research - into a model that highlights the possibility and prevalence of basic and applied research overlaps (as in Louis Pasteur's work). However, this approach still neglects the complex aspects that drive human behavior and decision-making. The da Vinci Cube framework addresses this limitation by incorporating a third axis that explicitly considers sentiment and visually expands the quadrant into a cube. This new model recognizes that emotions, values, and ethics play a vital role in innovation - shaping organizational culture, product design, hiring practices, and individual development. By acknowledging and analyzing this third component and driver of innovation, the da Vinci Cube framework provides a more comprehensive and nuanced understanding of the motivations that drive innovation by individuals and organizations.

Our team recognized the potential of applying the da Vinci Cube framework to enhance our systematic review. We applied its principles to explore patterns across the included papers in terms of pursuit of knowledge, consideration of use, and contemplation of sentiment. Specifically, we sought to investigate to what degree the innovations in engineering education addressed the emotional and ethical aspects of teaching and learning in relation to technological advances and skills development.

### 3 Results

Our systematic review (details: [45]) revealed a diverse and rapidly evolving landscape (see Figure 1). 45 papers focused on Coding Assistance, with a significant emphasis on programming or code correction while a notable subset explored pedagogical approaches to teaching coding or software engineering. Design Methodology was the second prominent theme, with 39 papers addressing various aspects such as Context Awareness, Conceptual Design Aids, Technological Skill Development, Design Feedback, and Ethical Standards. Additionally, the review uncovered 33 Position papers offering perspectives on the integration of AI in engineering education, alongside smaller clusters of research specific to Personalization (21), Writing Assistance (7), and other Miscellaneous topics in the domain (17), collectively shedding light on the multifaceted potential of AI in enhancing engineering education.

At a high level, applying the da Vinci Cube model reveals that while many of the papers in the largest category of Coding Assistance are motivated by "quest for knowledge" (basic research of developing new programming and identifying needs for code correction), such knowledge acquisition was often done with "consideration of use" (applied research in

pedagogical contexts). Likewise, research in the category of Design Methodology explored and innovated ways of creating new technologies for a range of contexts (knowledge and use), which entailed consideration of ethical standards. When proposing and creating ethical guidelines, researchers were operating in an area that required "contemplation of sentiment." Finally, in the categories of position papers and miscellaneous, researchers considered student and teacher motivation, components needed for personalized learning to work, and other topics that required contemplation of sentiment. Importantly, none of the research in any categories engaged in one sole axis but rather involved multiple and intersecting drivers.

The following subsections provide descriptions of categories identified in the systematic literature review and organized as categories in Figure 1.

### **3.1 Coding Assistance**

Our review uncovered a significant concentration of research on GenAI based coding assistance applications in engineering education. We found three broad sub-categories with 45 papers focussed on Coding Assistance. Half of the papers were focused on GenAI based programming correction, with authors focused on the degree of correctness of the AI solutions (e.g., [52], [53], [54]), creation of customized and ready to use programming exercises (e.g., [55]) or even attempting to distinguish between human-generated or AI code (e.g., [56–58]) Next, a handful of authors were focused more on bringing changes to pedagogical approaches in the age of Generative AI, focusing their papers on strategies to enhance teaching and assessing coding or prompting among learners. Examples from this sub-group include one designing a system to help students learn how to write effective prompts ([59]) and a Human Centered AI approach to understand how post-primary students in Ireland engage with GAI tools ([60]). Finally, the third sub-group of papers were ideologically focused with little or no empirical results, reporting on surveys, perspectives and positions or report outs from outreach workshops (e.g., [61], [62], [63]). While these research directions hold promise, our review revealed a predominant focus on evaluating code accuracy, with a majority of paper abstracts reporting on the successes and limitations of different AI applications in augmenting coding practices, rather than focusing on the necessary pedagogical approaches needed to augment classrooms in preparation for such disruptive technology. Although this preliminary surge of emphasis on technical proficiency is anticipated, we hope that future research will expand its scope to investigate the importance of teaching students to responsibly leverage AI in coding, considering crucial aspects like fairness, accountability, and transparency. As the field continues to evolve, it is essential to prioritize not only

technical competence but also ethical awareness and responsible AI integration in engineering education, ensuring that future engineers are equipped to harness AI's potential while mitigating its risks.

### **3.2 Design Methodology**

39 papers included in this review are categorized as Design Methodology papers. These papers discuss the uses of GenAI tools to teach broad design thinking, human-computer interaction, and engineering design. The papers elaborate the use of GenAI tools at various stages of the design process, including exploring alternative designs, understanding design contexts, and expanding awareness of regulatory design codes and regulations. For example, [64] used chatbots to generate personas to mimic real people and potential users of the designs to be created by students. Few papers presented case studies on LLMs can be leveraged for complex queries, interdisciplinary approaches to engineering design, and context awareness. Some studies showed how GenAI can be used to expedite design thinking, like in generating conceptual designs in mechanical engineering [65], making ethical choices during prototyping in time-sensitive situations such as hackathons [66], and learning disciplinary skills needed for design projects through personalized learning [67]. Lastly, a handful of papers explore how GenAI tools can give timely, relevant, and epistemic feedback during design. One example is the use of ChatGPT to analyze progress reports, instrumental to team collaborations, by recommending readability improvements and clarifying complex ideas [68].

### **3.3 Positions**

Our review found 33 position papers revealing diverse viewpoints on its integration, ethical considerations, and potential applications of GenAI in EE. Specifically, these papers are where authors argue their stance on or against the use of GenAI in EE, highlighting critical discussions often overlooked by the broader education or AI community. While one paper [69] advocated for enhancing GenAI in Data Science through prompt engineering, another [70] emphasized understanding AI's influence on student projects in software engineering. Notable other contributions include papers discussing mixed student experiences with ChatGPT in aviation education [71], specifically addressing trust in AI for programming tasks [72], and pointing out technical limitations of GPT models in educational distribution systems [73]. Non-empirical studies included in these review examined the promise and ethical considerations of GenAI [74] and advocated for a balance between benefits and risks. Others [75] discuss the transformative potential of AI in education and its ethical challenges, or reflected on conversational AI's broad impacts on research and policy, stressing responsible

use [76], highlighting the need for more assessments of GenAI in engineering education [77], proposing future scenarios for AI in software development, emphasizing productivity and ethical concerns [78], and suggesting a co-evolutionary approach to GenAI in human creativity [79]. These positionality related studies collectively underscore the transformative potential of GenAI across different engineering disciplines, advocating for responsible integration, addressing quality, privacy, and equitable access, and highlight the need for ongoing dialogue within the AI community to ensure a balanced and ethically grounded approach.

### **3.4 Personalization**

21 of the reviewed studies have specifically investigated GenAI’s potential for personalized learning experiences. Most studies explore AI’s potential for personalization and adaptive learning, proposing frameworks and systems to tailor educational experiences to individual needs (e.g., [80–82]. AI-enhanced assessment and feedback is another prominent area, with researchers investigating the use of large language models for answering assessment questions and providing formative feedback ( [83–85]. The integration of AI into various educational tools and platforms, including chatbots, virtual reality labs, and career guidance systems, is also a focus (e.g., [86–88]). Some studies examine how AI tools are changing students’ information-seeking and learning behaviors ( [89,90]. Ethical considerations and challenges, such as ensuring equitable access to AI technologies and addressing potential misuse, are recurring concerns discussed across papers in this category (e.g., [90–92]). Finally, many researchers highlight the need for future work, including long-term studies on learning outcomes, addressing current AI limitations, and developing best practices for AI integration in education (e.g., [83,84,91]). This body of research thus demonstrates both the significant potential of AI to transform education with an emphasis on personalization while also highlighting the need for careful implementation and ongoing research to carefully address challenges and ethical concerns.

### **3.5 Writing Assistance**

7 studies explored the use of ChatGPT for generating or scoring text-based content in EE. Similar to the papers under Coding Assistance, although significantly fewer, the papers related to Writing Assistance followed largely from overall industry trends around text generation, essay writing, and consequent scoring. These papers are sub-classified into 4 papers related more specifically to auto-generation and scoring of essays specific to a prompt versus using generative capabilities towards developing instructional tools. Notably, researchers developed an effective three-step prompting process (write, curate, verify or WCV) for teachers to



generate quality scenarios efficiently [93]. Some studies ([94] [95]) also provide guidelines for implementing the WCV approach in educational settings, demonstrating the potential of GenAI to enhance teaching and learning experiences in higher education. In another study [96], students showed increased motivation, improved learning performance, and positive attitudes towards the AI-generated scenarios, aligning with findings on a similar study [97] related to AI-augmented learning.

### **3.6 Miscellaneous**

A section of papers were grouped under the Miscellaneous label. More than half of these 17 papers were concerned with comparing human and AI outputs across domains. One example [98] compared the results of various LLM responses to mechanical engineering exam questions while another [99] performed a similar test on computer engineering exam questions. Most of these papers tended to find mixed results depending on the evaluation frameworks selected. This is to say, humans and AIs do not perform similarly on all metrics, and so in some cases, the LLM can be found to outperform humans, for example, in applying heuristics, while humans may beat the LLMs on math (e.g., [100], [101]). Authors across these papers recommend that educators think deeply about the critical choices on when to use these tools and also underline that the technology is changing very quickly. Other studies under the miscellaneous label emphasized connecting LLMs with engineering, but came from other non-engineering fields. We excluded most such studies if they didn't also apply to engineering. What remains are 3 case studies or experiments in human-ai co-creation. These papers have very small samples are indications of good reasons for human-ai collaborations, elaborating on things like efficiency in summarizing data and collaborations that fail. For example, in one interesting case, because human biases and AI biases both exist but don't necessarily match each other, the authors [102] demonstrate how co-creation requires building up mutual understanding, in a way very different from a usual relationship between an artist and their tools.

As an example of intersecting drivers, our initial findings of the papers showed that engineering educators found immediate relevance of GenAI applications in scaffolding the learning process, especially of technical skills, through just-in-time feedback and epistemic guidance. These findings align with the premise that AI-enhanced personalized learning systems providing feedback and guidance can be tailored to individual students based on progress, learning preferences, demographics, and interests. However, for these tools to find mass relevance across engineering institutions, researchers note that just-in-time value to what educators need and aspire to have in their classrooms is necessary. Therefore, it is

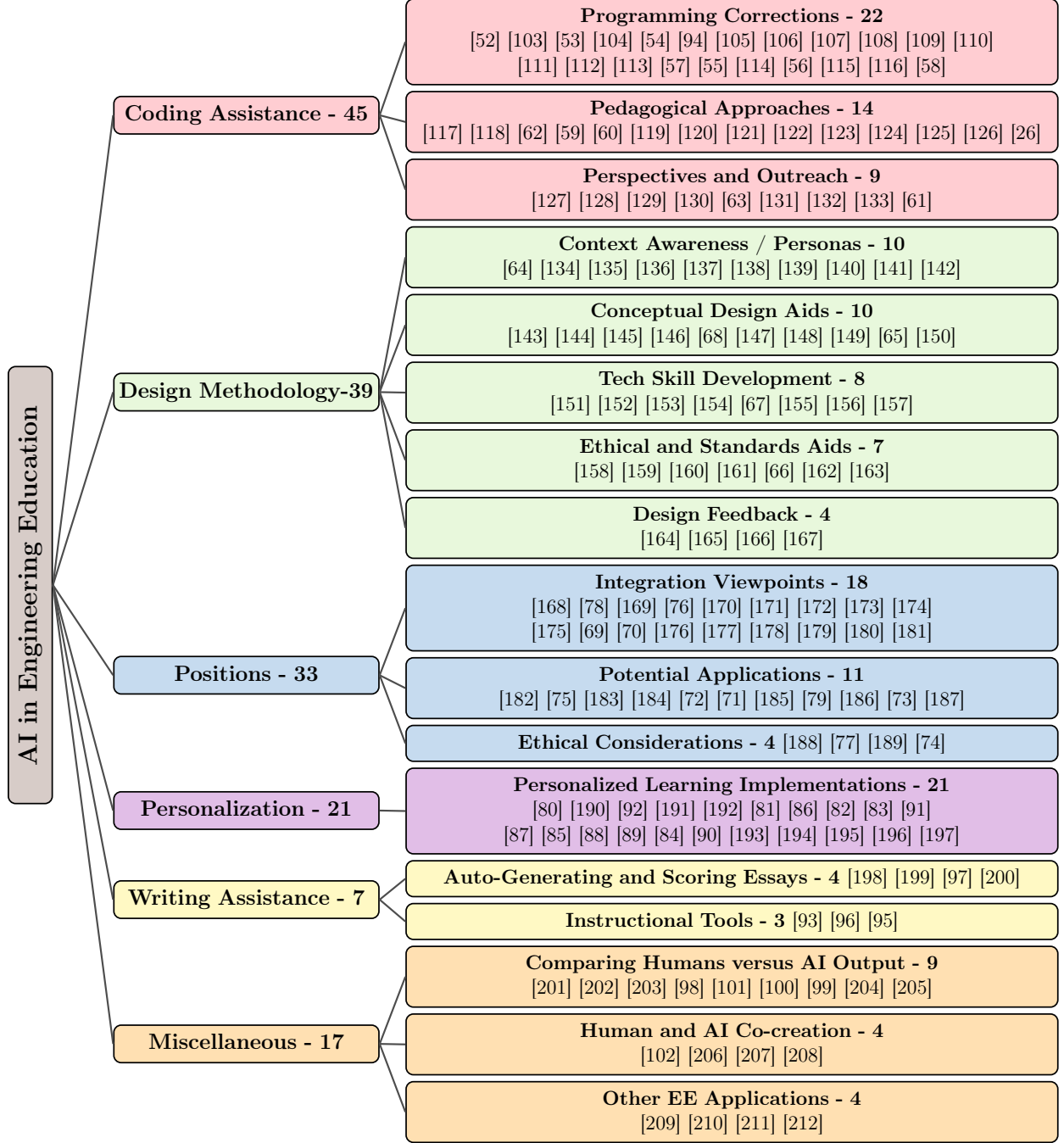


Figure 1: This taxonomy categorizes 162 papers describing the use of GenAI in Engineering Education (EE) into six broad objectives: Coding Assistance, Design Methodology, Positions, Personalization, Writing Assistance, and Miscellaneous

essential that our community focus on designing tools and learning systems that close the feedback loop between students and educators with meaningful outcomes related to ethics, societal values, and other considerations.

## 4 Limitations

Integrating the in-development da Vinci Cube framework into our systematic review was challenging in terms of practical implementation. The model's initial testing is being conducted using qualitative analysis of 60-minute transcripts of interviewees exploring the model with a researcher. Translating the early data into an instrument that operates reliably in a systematic task will require further development. For example, our team implemented a reflection process to develop a rubric for future assessment that provides consistency on the linguistic level of analysis, subject of analysis, and interpretation of the sentiment construct. Linguistically, the model does not lend itself to analysis at the lexical level, but works well when a subject focus is employed. For example, while words or phrases such as "feel" or "we hope to" lack relevance when lifted out of context, using the model to focus on components of research studies - such as articulation of problem statements, methodologies, discussion of results, and conclusions - revealed drivers of research that more readily aligned with the axes of Quest for Knowledge, Consideration of Use, and Contemplation of Sentiment. As we further employ the model as a way to characterize patterns of research in a particular field focus, we also expect to refine the meaning and usage of the construct of "sentiment" itself, which will be necessary to distinguish and explore intersections between sentiment, use, and knowledge in different contexts.

## 5 Conclusion

The integration of Generative AI (GenAI) in Engineering Education (EE) presents both unprecedented opportunities and significant challenges that demand careful consideration. Our research highlights the need for a comprehensive, multi-disciplinary approach that bridges technical innovation with human-centered educational practices. Several critical areas emerge as priorities for future research, including the investigation of how GenAI impacts the development of engineering identity, addressing algorithmic bias and equity concerns in AI-generated educational content, and examining the implications of technological access disparities. The path forward requires systematic user experience research with both educators and students, focused on developing adaptive, personalized learning environments that respond to individual needs, while evaluating the effectiveness of GenAI-enhanced educational interventions.

Our ongoing work will focus on expanding the research corpus to enhance the comprehensiveness of our analysis, refining the da Vinci Cube framework application to engineering

education, and developing a practical rubric for analyzing GenAI innovations that considers basic and applied approaches to research as well as drivers that may fall into the evolving construct of "contemplation of sentiment," such as dimensions of emotion, empathy, human-centric design, and ethical considerations.

The future of engineering education lies in the thoughtful integration of GenAI technologies with evidence-based pedagogical practices, requiring intentional research that ensures inclusive, effective implementation while maintaining focus on core educational objectives. This approach will be crucial in preparing engineers who can navigate and shape an increasingly complex socio-technological landscape. To achieve these goals, future research will necessitate interdisciplinary expertise including education policy, education research, learning sciences, AI, and ethics, among others. A significant emphasis must be placed on developing and testing personalized GenAI-powered learning environments that adapt to individual students' needs and abilities, while simultaneously addressing the ethical implications and overall impact on engineering education. Through continued investigation and framework development, we aim to equip educators and researchers with tools to evaluate and implement GenAI solutions that enhance learning outcomes while addressing ethical considerations and maintaining human-centered educational values.

Our ultimate goal remains to provide a comprehensive framework for understanding the evolving landscape of GenAI in engineering education, ensuring that this technological advancement serves to strengthen rather than diminish the human elements of engineering education. The development of an easy-to-use rubric built on the da Vinci Cube framework will facilitate the systematic analysis of GenAI innovations, enabling researchers and practitioners to better understand and navigate the complex interplay between drivers of innovation in engineering education, including knowledge, use, and sentiment.

## References

- [1] L. Chen, P. Chen, and Z. Lin, "Artificial intelligence in education: A review," *Ieee Access*, vol. 8, pp. 75 264–75 278, 2020.
- [2] T. K. Chiu, Q. Xia, X. Zhou, C. S. Chai, and M. Cheng, "Systematic literature review on opportunities, challenges, and future research recommendations of artificial intelligence in education," *Computers and Education: Artificial Intelligence*, vol. 4, p. 100118, 2023.

- [3] P. Denny, J. Prather, B. A. Becker, J. Finnie-Ansley, A. Hellas, J. Leinonen, A. Luxton-Reilly, B. N. Reeves, E. A. Santos, and S. Sarsa, “Computing education in the era of generative ai,” *Communications of the ACM*, vol. 67, no. 2, pp. 56–67, 2024.
- [4] E. Latif, G. Mai, M. Nyaaba, X. Wu, N. Liu, G. Lu, S. Li, T. Liu, and X. Zhai, “Agi: Artificial general intelligence for education,” *arXiv preprint arXiv:2304.12479*, 2023.
- [5] J. Prather, P. Denny, J. Leinonen, B. A. Becker, I. Albluwi, M. Craig, H. Keuning, N. Kiesler, T. Kohn, A. Luxton-Reilly *et al.*, “The robots are here: Navigating the generative ai revolution in computing education,” in *Proceedings of the 2023 Working Group Reports on Innovation and Technology in Computer Science Education*, 2023, pp. 108–159.
- [6] P. G. Altbach, *Responding to massification: Differentiation in postsecondary education worldwide*. Springer, 2017.
- [7] J. Arvanitakis, “Massification and the large lecture theatre: from panic to excitement,” *Higher Education*, vol. 67, pp. 735–745, 2014.
- [8] C. Carrico, H. M. Matusovich, and S. Bhaduri, “Board 164: Engineering interventions in my science classroom: What’s my role?” in *2023 ASEE Annual Conference & Exposition*, 2023.
- [9] C. D. Edwards, B. Peterson, S. Bhaduri, C. J. McCall, and D. S. Özkan, “Work in progress: Coloring outside the lines-exploring the potential for integrating creative evaluation in engineering education,” in *2023 ASEE Annual Conference & Exposition*, 2023.
- [10] M. Soledad, J. Grohs, S. Bhaduri, J. Doggett, J. Williams, and S. Culver, “Leveraging institutional data to understand student perceptions of teaching in large engineering classes,” in *2017 IEEE Frontiers in Education Conference (FIE)*. IEEE, 2017, pp. 1–8.
- [11] D. Dias, “Has massification of higher education led to more equity? clues to a reflection on portuguese education arena,” *International Journal of Inclusive Education*, vol. 19, no. 2, pp. 103–120, 2015.
- [12] T. Chowdhury, C. M. Pee, S. Bhaduri, and R. Ott, “Do we even belong? results from tracing experiences of women who are new graduates to inform practitioners at organizations,” in *37th Annual Society of IO Psychology Conference (SIOP)*, 2022.

- [13] W. C. Lee, B. D. Lutz, H. M. Matusovich, and S. Bhaduri, "Student perceptions of learning about diversity and its place in engineering classrooms in the united states," *International Journal of Engineering Education*, vol. 37, no. 1, pp. 147–162, 2021.
- [14] D. E. Neubauer, K.-H. Mok, J. Jiang *et al.*, *The sustainability of higher education in an era of post-massification*. Routledge London, UK, 2018.
- [15] K. H. Mok and J. Jiang, "Massification of higher education and challenges for graduate employment and social mobility: East asian experiences and sociological reflections," *International Journal of Educational Development*, vol. 63, pp. 44–51, 2018.
- [16] C. Carrico, H. M. Matusovich, and S. Bhaduri, "Preparing engineering students to find the best job fit: Starting early with the career development process," in *2023 ASEE Annual Conference & Exposition*, 2023.
- [17] S. J. Mgaiwa, "The paradox of financing public higher education in tanzania and the fate of quality education: The experience of selected universities," *Sage Open*, vol. 8, no. 2, p. 2158244018771729, 2018.
- [18] K. Pakala and S. Bhaduri, "Opportunities from disruption: How lifelong learning helped create more connected classrooms," 2022.
- [19] J. Grodotzki, S. Upadhya, and A. E. Tekkaya, "Engineering education amid a global pandemic," *Advances in Industrial and Manufacturing Engineering*, vol. 3, p. 100058, 2021.
- [20] S. Bhaduri, "Nlp in engineering education-demonstrating the use of natural language processing techniques for use in engineering education classrooms and research," 2018.
- [21] Y. Gamiieldien, R. McCord, and A. Katz, "Utilizing natural language processing to examine self-reflections in self-regulated learning," *Available at SSRN 4487795*, 2023.
- [22] S. Bhaduri and T. Roy, "A word-space visualization approach to study college of engineering mission statements," in *2017 IEEE Frontiers in Education Conference (FIE)*. IEEE, 2017, pp. 1–5.
- [23] S. Bhaduri, M. Soledad, T. Roy, H. Murzi, and T. Knott, "A semester like no other: Use of natural language processing for novice-led analysis on end-of-semester responses on students' experience of changing learning environments due to covid-19," in *2021 ASEE Virtual Annual Conference Content Access*, 2021.

- [24] J. Huang and K. Huang, “Chatgpt in government,” in *Beyond AI: ChatGPT, Web3, and the business landscape of tomorrow*. Springer, 2023, pp. 271–294.
- [25] P. Budhwar, S. Chowdhury, G. Wood, H. Aguinis, G. J. Bamber, J. R. Beltran, P. Boselie, F. Lee Cooke, S. Decker, A. DeNisi *et al.*, “Human resource management in the age of generative artificial intelligence: Perspectives and research directions on chatgpt,” *Human Resource Management Journal*, vol. 33, no. 3, pp. 606–659, 2023.
- [26] I. Pesovski, R. Santos, R. Henriques, and V. Trajkovik, “Generative ai for customizable learning experiences,” *Sustainability*, vol. 16, no. 7, p. 3034, 2024.
- [27] J. Knox, “Artificial intelligence and education in china,” *Learning, Media and Technology*, vol. 45, no. 3, pp. 298–311, 2020.
- [28] E. Francke and A. Bennett, “The potential influence of artificial intelligence on plagiarism: A higher education perspective,” in *European Conference on the Impact of Artificial Intelligence and Robotics (ECIAIR 2019)*, vol. 31, 2019, pp. 131–140.
- [29] A. Alam, “Should robots replace teachers? mobilisation of ai and learning analytics in education,” in *2021 International Conference on Advances in Computing, Communication, and Control (ICAC3)*. IEEE, 2021, pp. 1–12.
- [30] C. K. Y. Chan, “A comprehensive ai policy education framework for university teaching and learning,” *International journal of educational technology in higher education*, vol. 20, no. 1, p. 38, 2023.
- [31] F. Pedro, M. Subosa, A. Rivas, and P. Valverde, “Artificial intelligence in education: Challenges and opportunities for sustainable development,” 2019.
- [32] M. A. Chaudhry, M. Cukurova, and R. Luckin, “A transparency index framework for ai in education,” in *International Conference on Artificial Intelligence in Education*. Springer, 2022, pp. 195–198.
- [33] K. Holstein, B. M. McLaren, and V. Aleven, “Co-designing a real-time classroom orchestration tool to support teacher-ai complementarity.” *Grantee Submission*, 2019.
- [34] C. Klötzer, J. Weißenborn, and A. Pflaum, “The evolution of cyber-physical systems as a driving force behind digital transformation,” in *2017 IEEE 19th Conference on Business Informatics (CBI)*, vol. 2. IEEE, 2017, pp. 5–14.

- [35] D. G. Broo, O. Kaynak, and S. M. Sait, “Rethinking engineering education at the age of industry 5.0,” *Journal of Industrial Information Integration*, vol. 25, p. 100311, 2022.
- [36] B. Fatahi, H. Khabbaz, J. Xue, and R. Hadgraft, “Generative ai as a catalyst for enhanced learning experience in engineering education,” *Proceedings of the AAEE*, 2023.
- [37] S. J. Zhang, S. Florin, A. N. Lee, E. Niknafs, A. Marginean, A. Wang, K. Tyser, Z. Chin, Y. Hicke, N. Singh *et al.*, “Exploring the mit mathematics and eecs curriculum using large language models,” *arXiv preprint arXiv:2306.08997*, 2023.
- [38] Y. Shen, P. Yu, H. Lu, X. Zhang, and H. Zeng, “An ai-based virtual simulation experimental teaching system in space engineering education,” *Computer Applications in Engineering Education*, vol. 29, no. 2, pp. 329–338, 2021.
- [39] Muthmainnah, P. M. Ibna Seraj, and I. Oteir, “Playing with ai to investigate human-computer interaction technology and improving critical thinking skills to pursue 21st century age,” *Education Research International*, vol. 2022, no. 1, p. 6468995, 2022.
- [40] D. A. Martin, E. Conlon, and B. Bowe, “A multi-level review of engineering ethics education: Towards a socio-technical orientation of engineering education for ethics,” *Science and Engineering Ethics*, vol. 27, no. 5, p. 60, 2021.
- [41] M. V. Huerta, A. R. Carberry, T. Pipe, and A. F. McKenna, “Inner engineering: Evaluating the utility of mindfulness training to cultivate intrapersonal and interpersonal competencies among first-year engineering students,” *Journal of Engineering Education*, vol. 110, no. 3, pp. 636–670, 2021.
- [42] J. Strobel, J. Hess, R. Pan, and C. A. Wachter Morris, “Empathy and care within engineering: Qualitative perspectives from engineering faculty and practicing engineers,” *Engineering Studies*, vol. 5, no. 2, pp. 137–159, 2013.
- [43] G. Guest, “Lifelong learning for engineers: a global perspective,” *European Journal of Engineering Education*, vol. 31, no. 3, pp. 273–281, 2006.
- [44] D. Bairaktarova, “Caring for the future: Empathy in engineering education to empower learning,” pp. 502–507, 2022.
- [45] R. Khan, S. Bhaduri, T. Mackenzie, A. Paul, S. KJ, and I. Sen, “Path to personalization: A systematic review of genai in engineering education,” in *KDD AI4Edu Workshop*, 2024.



- [46] M. Borrego, M. J. Foster, and J. E. Froyd, “Systematic literature reviews in engineering education and other developing interdisciplinary fields,” *Journal of Engineering Education*, vol. 103, no. 1, pp. 45–76, 2014.
- [47] M. Ouzzani, H. Hammady, Z. Fedorowicz, and A. Elmagarmid, “Rayyan—a web and mobile app for systematic reviews,” *Systematic Reviews*, vol. 5, no. 1, p. 210, 2016. [Online]. Available: <http://dx.doi.org/10.1186/s13643-016-0384-4>
- [48] D. E. Stokes, *Pasteur’s quadrant: Basic science and technological innovation*. Brookings Institution Press, 2011.
- [49] B. Knapp, T. Martin, L. D. McNair, and T. Rassi, “The da vinci’s cube: A new model for understanding innovation at the edges of art and science,” in *Integrative Contemporary Art and Science Practices*. Routledge, pp. 81–90.
- [50] M. T. e. a. McNair, Lisa, “da vinci’s cube: Reframing innovation in ece curricula.” in *2024 IEEE Frontiers in Education Conference (FIE)*. IEEE, 2024.
- [51] B. Nepal and J. Mathai, “Initial developments of da vinci’s cube: An expansion of pasteur’s quadrant.” in *Presented at the MIT Undergraduate Research Technology Conference*. IEEE Xplore, 2024.
- [52] I. Azaiz, O. Deckarm, and S. Strickroth, “Ai-enhanced auto-correction of programming exercises: How effective is gpt-3.5?” *INTERNATIONAL JOURNAL OF ENGINEERING PEDAGOGY*, vol. 13, no. 8, pp. 67–83, 2023.
- [53] F. M. Megahed, Y.-J. Chen, J. A. Ferris, S. Knoth, and L. A. Jones-Farmer, “How generative ai models such as chatgpt can be (mis)used in spc practice, education, and research? an exploratory study,” *Quality Engineering*, vol. 36, no. 2, pp. 287 – 315, 2024. [Online]. Available: <http://dx.doi.org/10.1080/08982112.2023.2206479>
- [54] T. Wang, D. V. Diaz, C. Brown, and Y. Chen, “Exploring the role of ai assistants in computer science education: Methods, implications, and instructor perspectives,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2306.03289>
- [55] N. B. D. Ta, H. G. P. Nguyen, and S. Gottipati, “Exgen: Ready-to-use exercise generation in introductory programming courses,” *31st International Conference on Computers in Education, ICCE 2023 - Proceedings*, vol. 1, pp. 104 – 113, 2023.
- [56] M. Hoq, Y. Shi, J. Leinonen, D. Babalola, C. Lynch, and B. Akram, “Detecting chatgpt-generated code in a cs1 course,” *CEUR Workshop Proceedings*, vol. 3487, pp. 53 – 63, 2023.

- [57] M. Hoq, Y. Shi, J. Leinonen, D. Babalola, C. Lynch, T. Price, and B. Akram, “Detecting chatgpt-generated code submissions in a cs1 course using machine learning models,” *SIGCSE 2024 - Proceedings of the 55th ACM Technical Symposium on Computer Science Education*, vol. 1, pp. 526 – 532, 2024. [Online]. Available: <http://dx.doi.org/10.1145/3626252.3630826>
- [58] O. J. Idialu, N. S. Mathews, R. Maipradit, J. M. Atlee, and M. Nagappan, “Whodunit: Classifying code as human authored or gpt-4 generated- a case study on codechef problems,” *arXiv*, 2024. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2403.04013>
- [59] P. Denny, J. Leinonen, J. Prather, A. Luxton-Reilly, T. Amarouche, B. A. Becker, and B. N. Reeves, “Prompt problems: A new programming exercise for the generative ai era,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2311.05943>
- [60] I. Stone, “Exploring the research gap: Generative ai and learning of python programming among post-primary students,” *ACM International Conference Proceeding Series*, pp. 51 –, 2023. [Online]. Available: <http://dx.doi.org/10.1145/3633083.3633099>
- [61] B. A. Tanay, L. Arinze, S. S. Joshi, K. A. Davis, and J. C. Davis, “An exploratory study on upper-level computing students’ use of large language models as tools in a semester-long project,” *arXiv*, 2024. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2403.18679>
- [62] O. Petrovska, L. Clift, and F. Moller, “Generative ai in software development education: Insights from a degree apprenticeship programme,” *ACM International Conference Proceeding Series*, pp. ACM UK SIGCSE; Swansea University – Prifysgol Abertawe; technocamps –, 2023. [Online]. Available: <http://dx.doi.org/10.1145/3610969.3611132>
- [63] K. Hanifi, O. Cetin, and C. Yilmaz, “On chatgpt: Perspectives from software engineering students,” *IEEE International Conference on Software Quality, Reliability and Security, QRS*, pp. 196 – 205, 2023. [Online]. Available: <http://dx.doi.org/10.1109/QRS60937.2023.00028>
- [64] I. Terzic, A. Drobnjak, and I. Boticki, “Designing educational personas using generative ai,” *31ST INTERNATIONAL CONFERENCE ON COMPUTERS IN EDUCATION, ICCE 2023, VOL II*, pp. 961–963, 2023.

- [65] A. S. Gill, “Chat generative pretrained transformer: Extinction of the designer or rise of an augmented designer,” *Proceedings of the ASME Design Engineering Technical Conference*, vol. 3, pp. Computers and Information in Engineering Division; Design Engineering Division –, 2023. [Online]. Available: <http://dx.doi.org/10.1115/DETC2023-116971>
- [66] R. Sajja, C. E. Ramirez, Z. Li, B. Z. Demiray, Y. Sermet, and I. Demir, “Integrating generative ai in hackathons: Opportunities, challenges, and educational implications,” *arXiv*, 2024. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2401.17434>
- [67] T. N. Ngoc, Q. N. Tran, A. Tang, B. Nguyen, T. Nguyen, and T. Pham, “Ai-assisted learning for electronic engineering courses in high education,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2311.01048>
- [68] A. Neyem, J. P. S. Alcocer, oval, M. Mendoza, L. Centellas-Claros, L. A. Gonzalez, and C. Paredes-Robles, “Exploring the impact of generative ai for standup report recommendations in software capstone project development,” *SIGCSE 2024 - Proceedings of the 55th ACM Technical Symposium on Computer Science Education*, vol. 1, pp. 951 – 957, 2024. [Online]. Available: <http://dx.doi.org/10.1145/3626252.3630854>
- [69] Y. Shen, X. Ai, A. G. Soosai Raj, R. J. Leo John, and M. Syamkumar, “Implications of chatgpt for data science education,” *SIGCSE 2024 - Proceedings of the 55th ACM Technical Symposium on Computer Science Education*, vol. 1, pp. 1230 – 1236, 2024. [Online]. Available: <http://dx.doi.org/10.1145/3626252.3630874>
- [70] S. Rasnayaka, G. Wang, R. Shariffdeen, and G. N. Iyer, “An empirical study on usage and perceptions of llms in a software engineering project,” *arXiv*, 2024. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2401.16186>
- [71] S. Wandelt, X. Sun, and A. Zhang, “Ai-driven assistants for education and research? a case study on chatgpt for air transport management,” *JOURNAL OF AIR TRANSPORT MANAGEMENT*, vol. 113, 10 2023.
- [72] M. Rahman and Y. Watanobe, “Chatgpt for education and research: Opportunities, threats, and strategies,” *APPLIED SCIENCES-BASEL*, vol. 13, no. 9, 5 2023.
- [73] R. S. Bonadia, F. Trindade, a C. L., W. Freitas, and B. Venkatesh, “On the potential of chatgpt to generate distribution systems for load flow studies using opendss,”

- IEEE Transactions on Power Systems*, vol. 38, no. 6, pp. 5965 – 5968, 2023. [Online]. Available: <http://dx.doi.org/10.1109/TPWRS.2023.3315543>
- [74] A. Johri, E. Lindsay, and J. Qadir, “Ethical concerns and responsible use of generative artificial intelligence in engineering education,” *SEFI 2023 - 51st Annual Conference of the European Society for Engineering Education: Engineering Education for Sustainability, Proceedings*, pp. 2244 – 2253, 2023. [Online]. Available: <http://dx.doi.org/10.21427/OT6R-FZ62>
- [75] A. Bozkurt, “Generative artificial intelligence (ai) powered conversational educational agents: The inevitable paradigm shift,” *Asian Journal of Distance Education*, vol. 18, no. 1, pp. 198–204, 1 2023. [Online]. Available: <https://search.ebscohost.com/login.aspx?direct=true&AuthType=ip,shib&db=eric&AN=EJ1389644&site=ehost-live&scope=site&custid=uga1>
- [76] Y. Dwivedi, N. Kshetri, L. Hughes, E. Slade, A. Jeyaraj, A. Kar, A. Baabdullah, A. Koohang, V. Raghavan, M. Ahuja, H. Albanna, M. Albashrawi, A. Al-Busaidi, J. Balakrishnan, Y. Barlette, S. Basu, I. Bose, L. Brooks, D. Buhalis, L. Carter, S. Chowdhury, T. Crick, S. Cunningham, G. Davies, R. Davison, R. De, D. Dennehy, Y. Duan, R. Dubey, R. Dwivedi, J. Edwards, C. Flavian, R. Gauld, V. Grover, M. Hu, M. Janssen, P. Jones, I. Junglas, S. Khorana, S. Kraus, K. Larsen, P. Latreille, S. Laumer, F. Malik, A. Mardani, M. Mariani, S. Mithas, E. Mogaji, J. Nord, S. O’Connor, F. Okumus, M. Pagani, N. Pandey, S. Papagiannidis, I. Pappas, N. Pathak, J. Pries-Heje, R. Raman, N. Rana, S. Rehm, S. Ribeiro-Navarrete, A. Richter, F. Rowe, S. Sarker, B. Stahl, M. Tiwari, W. van der Aalst, V. Venkatesh, G. Viglia, M. Wade, P. Walton, J. Wirtz, and R. Wright, “"so what if chatgpt wrote it?" multidisciplinary perspectives on opportunities, challenges and implications of generative conversational ai for research, practice and policy,” *INTERNATIONAL JOURNAL OF INFORMATION MANAGEMENT*, vol. 71, 8 2023.
- [77] K. Mustapha, E. Yap, and Y. Abakr, “Bard, chatgpt and 3dgpt: a scientometric analysis of generative ai tools and assessment of implications for mechanical engineering education,” *INTERACTIVE TECHNOLOGY AND SMART EDUCATION*, 2 2024.
- [78] J. Sauvola, S. Tarkoma, M. Klemettinen, J. Riekki, and D. Doermann, “Future of software development with generative ai,” *AUTOMATED SOFTWARE ENGINEERING*, vol. 31, no. 1, 5 2024.

- [79] K. Bohm and L.-M. Schedlberger, “The use of generative ai in the domain of human creations, a case for co-evolution?” *CEUR Workshop Proceedings*, vol. 3598, pp. 90 – 102, 2023.
- [80] Y. Luo and Y. Yang, “Large language model and domain-specific model collaboration for smart education,” *FRONTIERS OF INFORMATION TECHNOLOGY & ELECTRONIC ENGINEERING*, vol. 25, no. 3, pp. 333–341, 3 2024.
- [81] R. Sajja, Y. Sermet, M. Cikmaz, D. Cwiertny, and I. Demir, “Artificial intelligence-enabled intelligent assistant for personalized and adaptive learning in higher education,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2309.10892>
- [82] H. A. Rasheed, C. Weber, and M. Fathi, “Knowledge graphs as context sources for llm-based explanations of learning recommendations,” *arXiv*, 2024. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2403.03008>
- [83] A. Shoufan, “Can students without prior knowledge use chatgpt to answer test questions? an empirical study,” *ACM Transactions on Computing Education*, vol. 23, no. 4, 2023. [Online]. Available: <http://dx.doi.org/10.1145/3628162>
- [84] C. D. Ponte, S. Dushyanthen, and K. Lyons, “"close...but not as good as an educator" - using chatgpt to provide formative feedback in large-class collaborative learning,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2311.01634>
- [85] G. Yadav, Y.-J. Tseng, and X. Ni, “Contextualizing problems to student interests at scale in intelligent tutoring system using large language models,” *CEUR Workshop Proceedings*, vol. 3487, pp. 17 – 25, 2023.
- [86] J. C. Farah, S. Ingram, y, B. Spaenlehauer, F. K.-L. Lasne, and D. Gillet, “Prompting large language models to power educational chatbots,” *Lecture Notes in Computer Science (including subseries Lecture Notes in Artificial Intelligence and Lecture Notes in Bioinformatics)*, vol. 14409, pp. 169 – 188, 2023. [Online]. Available: [http://dx.doi.org/10.1007/978-981-99-8385-8\\_14](http://dx.doi.org/10.1007/978-981-99-8385-8_14)
- [87] D. Ayre, C. Dougherty, and Y. Zhao, “Implementation of an artificial intelligence (ai) instructional support system in a virtual reality (vr) thermal-fluids laboratory,” *ASME International Mechanical Engineering Congress and Exposition, Proceedings (IMECE)*, vol. 8, pp. American Society of Mechanical Engineers (ASME) –, 2023. [Online]. Available: <http://dx.doi.org/10.1115/IMECE2023-112683>

- [88] A. Talib, M. Housni, and M. Radid, “Utilizing m-technologies for ai-driven career guidance in morocco: An innovative mobile approach,” *International Journal of Interactive Mobile Technologies*, vol. 17, no. 24, pp. 173 – 188, 2023. [Online]. Available: <http://dx.doi.org/10.3991/IJIM.V17I24.44263>
- [89] T. Karunaratne and A. Adesina, “Is it the new google: Impact of chatgpt on students’ information search habits,” *Proceedings of the European Conference on e-Learning, ECEL*, vol. 2023, pp. 147 – 155, 2023.
- [90] C. R. Bego, “Using chatgpt for homework: Does it feel like cheating? (wip),” *Proceedings - Frontiers in Education Conference, FIE*, pp. American Society for Engineering Education (ASEE)/Educational Research and Methods Division (ERM); IEEE; IEEE Computer Society; IEEE Education Society –, 2023. [Online]. Available: <http://dx.doi.org/10.1109/FIE58773.2023.10343397>
- [91] B. D. Nye, D. Mee, and M. G. Core, “Generative large language models for dialog-based tutoring: An early consideration of opportunities and concerns,” *CEUR Workshop Proceedings*, vol. 3487, pp. 78 – 88, 2023.
- [92] N. Shoeibi, “Cross-lingual transfer in generative ai-based educational platforms for equitable and personalized learning,” *CEUR Workshop Proceedings*, vol. 3542, 2023.
- [93] S. Bai, D. E. Gonda, and K. F. Hew, “Write-curate-verify: A case study of leveraging generative ai for scenario writing in scenario-based learning,” *IEEE Transactions on Learning Technologies*, vol. 17, pp. 1313–1324, 1 2024. [Online]. Available: <https://search.ebscohost.com/login.aspx?direct=true&AuthType=ip,shib&db=eric&AN=EJ1420427&site=ehost-live&scope=site&custid=uga1>
- [94] B. Cowan, Y. Watanobe, and A. Shirafuji, “Enhancing programming learning with llms: Prompt engineering and flipped interaction,” *ACM International Conference Proceeding Series*, pp. 10 – 16, 2023. [Online]. Available: <http://dx.doi.org/10.1145/3634814.3634816>
- [95] B. Gorer and F. B. Aydemir, “Generating requirements elicitation interview scripts with large language models,” *Proceedings - 31st IEEE International Requirements Engineering Conference Workshops, REW 2023*, pp. 44 – 51, 2023. [Online]. Available: <http://dx.doi.org/10.1109/REW57809.2023.00015>

- [96] B. Eager and R. Brunton, “Prompting higher education towards ai-augmented teaching and learning practice,” *JOURNAL OF UNIVERSITY TEACHING AND LEARNING PRACTICE*, vol. 20, no. 5, 2023.
- [97] M. Bernabei, S. Colabianchi, A. Falegnami, and F. Costantino, “Student’s use of large language models in engineering education: A case study on technology acceptance, perceptions, efficacy, and detection chances,” *Computers and Education: Artificial Intelligence*, vol. 5, 2023. [Online]. Available: <http://dx.doi.org/10.1016/j.caeai.2023.100172>
- [98] J. Tian, J. Hou, Z. Wu, P. Shu, N. Liu, Z. Liu, Y. Xiang, B. Gu, N. Filla, Y. Li, X. Chen, K. Tang, T. Liu, and X. Wang, “Assessing large language models in mechanical engineering education: A study on mechanics-focused conceptual understanding,” *arXiv*, 2024. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2401.12983>
- [99] R. Rodriguez-Echeverria, J. D. Gutierrez, J. M. Conejero, and A. E. Prieto, “Analysis of chatgpt performance in computer engineering exams,” *Revista Iberoamericana de Tecnologias del Aprendizaje*, pp. 1 – 1, 2024. [Online]. Available: <http://dx.doi.org/10.1109/RITA.2024.3381842>
- [100] a. Udias, A. Alonso-Ayuso, I. Sanchez, Hern, S. ez, M. E. Castellanos, R. M. Diez, and E. L. Cano, “The potential of large language models for improving probability learning: A study on chatgpt3.5 and first-year computer engineering students,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2310.05686>
- [101] V. Pursnani, Y. Sermet, M. Kurt, and I. Demir, “Performance of chatgpt on the us fundamentals of engineering exam: Comprehensive assessment of proficiency and potential implications for professional environmental engineering practice,” *Computers and Education: Artificial Intelligence*, vol. 5, 2023. [Online]. Available: <http://dx.doi.org/10.1016/j.caeai.2023.100183>
- [102] C. Rodier, J. Millar, W. Deisinger, and S. J. Hodgson, “Art critically examining generative ai,” *2023 IEEE IFEES World Engineering Education Forum and Global Engineering Deans Council: Convergence for a Better World: A Call to Action, WEEF-GEDC 2023 - Proceedings*, pp. ABET; Consejo de Acreditacion de la Ensenanza de la Ingenieria, A.C. (CACEI); Dassault Systemes; et al.; MathWorks; Quanser –, 2023. [Online]. Available: <http://dx.doi.org/10.1109/WEEF-GEDC59520.2023.10343903>

- [103] R. Choudhuri, D. Liu, I. Steinmacher, M. Gerosa, and A. Sarma, “How far are we? the triumphs and trials of generative ai in learning software engineering,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2312.11719>
- [104] N. A. Kumar and A. S. Lan, “Using large language models for student-code guided test case generation in computer science education,” *arXiv*, 2024. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2402.07081>
- [105] C. Glynn, E. Hed, A. Pexa, T. Pohlmann, I. Rahal, and R. Hesse, “Caet: Code analysis and education tutor,” *SIGCSE 2024 - Proceedings of the 55th ACM Technical Symposium on Computer Science Education*, vol. 2, pp. 1656 – 1657, 2024. [Online]. Available: <http://dx.doi.org/10.1145/3626253.3635543>
- [106] P. Denny, V. Kumar, and N. Giacaman, “Conversing with copilot: Exploring prompt engineering for solving cs1 problems using natural language,” *arXiv*, 2022. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2210.15157>
- [107] J. Savelka, A. Agarwal, M. An, C. Bogart, and M. Sakr, “Thrilled by your progress! large language models (gpt-4) no longer struggle to pass assessments in higher education programming courses,” *ICER 2023 - Proceedings of the 2023 ACM Conference on International Computing Education Research V.1*, pp. 78 – 92, 2023. [Online]. Available: <http://dx.doi.org/10.1145/3568813.3600142>
- [108] H. Nguyen and V. Allan, “Using gpt-4 to provide tiered, formative code feedback,” *SIGCSE 2024 - Proceedings of the 55th ACM Technical Symposium on Computer Science Education*, vol. 1, pp. 958 – 964, 2024. [Online]. Available: <http://dx.doi.org/10.1145/3626252.3630960>
- [109] J. Li, Mel, P. H. , Notl, J. S. , A. Storhaug, and J. H. Tysse, “Evaluating the impact of chatgpt on exercises of a software security course,” *International Symposium on Empirical Software Engineering and Measurement*, 2023. [Online]. Available: <http://dx.doi.org/10.1109/ESEM56168.2023.10304857>
- [110] M.-D. Popovici, “Chatgpt in the classroom. exploring its potential and limitations in a functional programming course,” *International Journal of Human-Computer Interaction*, 2023. [Online]. Available: <http://dx.doi.org/10.1080/10447318.2023.2269006>
- [111] N. Dunder, S. Lundborg, J. Wong, and O. Viberg, “Kattis vs chatgpt: Assessment and evaluation of programming tasks in the age of artificial intelligence,” *ACM*



- International Conference Proceeding Series*, pp. 821 – 827, 2024. [Online]. Available: <http://dx.doi.org/10.1145/3636555.3636882>
- [112] M. Kazemitabaar, J. Chow, C. K. T. Ma, B. J. Ericson, D. Weintrop, and T. Grossman, “Studying the effect of ai code generators on supporting novice learners in introductory programming,” *Conference on Human Factors in Computing Systems - Proceedings*, pp. ACM SIGCHI; Apple; Bloomberg; Google; NSF; SIEMENS –, 2023. [Online]. Available: <http://dx.doi.org/10.1145/3544548.3580919>
- [113] P. Abrahamsson, T. Anttila, J. Hakala, J. Ketola, A. Knappe, D. Lahtinen, V. Liukko, T. Poranen, T.-M. Ritala, and M. Setälä, “Chatgpt as a fullstack web developer - early results,” *Lecture Notes in Business Information Processing*, vol. 489, pp. 201 – 209, 2024. [Online]. Available: [http://dx.doi.org/10.1007/978-3-031-48550-3\\_20](http://dx.doi.org/10.1007/978-3-031-48550-3_20)
- [114] B. P. Cipriano and P. Alves, “Gpt-3 vs object oriented programming assignments: An experience report,” *Annual Conference on Innovation and Technology in Computer Science Education, ITiCSE*, vol. 1, pp. 61 – 67, 2023. [Online]. Available: <http://dx.doi.org/10.1145/3587102.3588814>
- [115] T. D. Viet and K. Markov, “Using large language models for bug localization and fixing,” *Proceedings of 2023 12th International Conference on Awareness Science and Technology, iCAST 2023*, pp. 192 – 197, 2023. [Online]. Available: <http://dx.doi.org/10.1109/iCAST57874.2023.10359304>
- [116] J. Zhang, J. Cambronero, S. Gulwani, V. Le, R. Piskac, G. Soares, and G. Verbruggen, “Repairing bugs in python assignments using large language models,” *arXiv*, 2022. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2209.14876>
- [117] A. N. Duc, T. Lønnestad, I. Sundbo, M. R. Johannessen, V. Gabriela, S. U. Ahmed, and R. El-Gazzar, “Generative ai in undergraduate information technology education - insights from nine courses,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2311.10199>
- [118] P. Lauren and P. Watta, “Work-in-progress: Integrating generative ai with evidence-based learning strategies in computer science and engineering education,” *Proceedings - Frontiers in Education Conference, FIE*, pp. American Society for Engineering Education (ASEE)/Educational Research and Methods Division (ERM); IEEE; IEEE Computer Society; IEEE Education Society –, 2023. [Online]. Available: <http://dx.doi.org/10.1109/FIE58773.2023.10342970>

- [119] R. Liu, C. Zenke, C. Liu, A. Holmes, P. Thornton, and D. J. Malan, "Teaching cs50 with ai: Leveraging generative artificial intelligence in computer science education," *SIGCSE 2024 - Proceedings of the 55th ACM Technical Symposium on Computer Science Education*, vol. 1, pp. 750 – 756, 2024. [Online]. Available: <http://dx.doi.org/10.1145/3626252.3630938>
- [120] J. Savelka, A. Agarwal, C. Bogart, and M. Sakr, "From gpt-3 to gpt-4: On the evolving efficacy of llms to answer multiple-choice questions for programming classes in higher education," *Communications in Computer and Information Science*, vol. 2052, pp. 160 – 182, 2024. [Online]. Available: [http://dx.doi.org/10.1007/978-3-031-53656-4\\_8](http://dx.doi.org/10.1007/978-3-031-53656-4_8)
- [121] V. Agarwal, S. Dharmavaram, N. Thureja, Meghna, M. K. Garg, and D. Kumar, "\"which llm should i use?\": Evaluating llms for tasks performed by undergraduate computer science students in india," *arXiv*, 2024. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2402.01687>
- [122] A. D. C. Gutierrez, P. Denny, and A. Luxton-Reilly, "Evaluating automatically generated contextualised programming exercises," *SIGCSE 2024 - Proceedings of the 55th ACM Technical Symposium on Computer Science Education*, vol. 1, pp. 289 – 295, 2024. [Online]. Available: <http://dx.doi.org/10.1145/3626252.3630863>
- [123] J. Rajala, J. Hukkanen, M. Hartikainen, and P. Niemela, "\"call me kiran\" chatgpt as a tutoring chatbot in a computer science course," *ACM International Conference Proceeding Series*, pp. 83 – 94, 2023. [Online]. Available: <http://dx.doi.org/10.1145/3616961.3616974>
- [124] M. Waseem, T. Das, A. Ahmad, P. Liang, M. Fahmideh, and T. Mikkonen, "Chatgpt as a software development bot: A project-based study," *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2310.13648>
- [125] S. E. Wilson and M. Nishimoto, "Assessing learning of computer programming skills in the age of generative artificial intelligence," *Journal of Biomechanical Engineering*, vol. 146, no. 5, 2024. [Online]. Available: <http://dx.doi.org/10.1115/1.4064364>
- [126] A. Ouazaki, K. Bergram, and A. Holzer, "Leveraging chatgpt to enhance computational thinking learning experiences," *2023 IEEE International Conference on Teaching, Assessment and Learning for Engineering, TALE 2023 - Conference Proceedings*, pp. IEEE Education Society; IEEE New Zealand North Section; IEEE Region 10 –, 2023. [Online]. Available: <http://dx.doi.org/10.1109/TALE56641.2023.10398358>

- [127] R. Naik, A. Rajbhoj, M. Patwardhan, and R. K. Medicherla, “Workshop report on generative ai-based software engineering,” *ACM International Conference Proceeding Series*, pp. ABB; Center for Technology Research and Innovation (CTRI) Digital Governance; et al.; Google; IBM; TCS Research –, 2024. [Online]. Available: <http://dx.doi.org/10.1145/3641399.3641437>
- [128] O. Petrovska, L. Clift, F. Moller, and R. Pearsall, “Incorporating generative ai into software development education,” *ACM International Conference Proceeding Series*, pp. 37 – 40, 2024. [Online]. Available: <http://dx.doi.org/10.1145/3633053.3633057>
- [129] M. Daun and J. Brings, “How chatgpt will change software engineering education,” *Annual Conference on Innovation and Technology in Computer Science Education, ITiCSE*, vol. 1, pp. 110 – 116, 2023. [Online]. Available: <http://dx.doi.org/10.1145/3587102.3588815>
- [130] Fern, A. ez, a S., and K. A. Cornell, “Cs1 with a side of ai: Teaching software verification for secure code in the era of generative ai,” *SIGCSE 2024 - Proceedings of the 55th ACM Technical Symposium on Computer Science Education*, vol. 1, pp. 345 – 351, 2024. [Online]. Available: <http://dx.doi.org/10.1145/3626252.3630817>
- [131] Y. Huang, Y. Chen, X. Chen, J. Chen, R. Peng, Z. Tang, J. Huang, F. Xu, and Z. Zheng, “Generative software engineering,” *arXiv*, 2024. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2403.02583>
- [132] Q. Zhang, C. Fang, Y. Xie, Y. Zhang, Y. Yang, W. Sun, S. Yu, and Z. Chen, “A survey on large language models for software engineering,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2312.15223>
- [133] S. MacNeil, P. Denny, A. Tran, J. Kim, J. Leinonen, A. Hellas, S. Bernstein, and S. Sarsa, “Automatically generating cs learning materials with large language models,” *arXiv*, 2022. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2212.05113>
- [134] “Strengths, weaknesses, opportunities, and threats of using chatgpt in scientific research,” *INTERNATIONAL JOURNAL OF TECHNOLOGY IN EDUCATION*, vol. 7, no. 1, pp. 40–58, 2024.
- [135] Y. Dai, S. Lai, C. Lim, and A. Liu, “Chatgpt and its impact on research supervision: Insights from australian postgraduate research students,” *AUSTRALASIAN JOURNAL OF EDUCATIONAL TECHNOLOGY*, vol. 39, no. 4, pp. 74–88, 2023.

- [136] A. Katz, U. Shakir, and B. Chambers, “The utility of large language models and generative ai for education research,” *arXiv*, 2023.
- [137] A. Strzelecki and S. ElArabawy, “Investigation of the moderation effect of gender and study level on the acceptance and use of generative ai by higher education students: Comparative evidence from poland and egypt,” *British Journal of Educational Technology*, vol. 55, no. 3, pp. 1209 – 1230, 2024. [Online]. Available: <http://dx.doi.org/10.1111/bjet.13425>
- [138] C. Cao, Z. Ding, G.-G. Lee, J. Jiao, J. Lin, and X. Zhai, “Elucidating stem concepts through generative ai: A multi-modal exploration of analogical reasoning,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2308.10454>
- [139] R. Zhang, Y. Qiu, and Y. Li, “An empirical study on human-machine collaborative mooc learning interaction empowered by generative ai,” *Proceedings - 2023 International Symposium on Educational Technology, ISET 2023*, pp. 116 – 120, 2023. [Online]. Available: <http://dx.doi.org/10.1109/ISET58841.2023.00031>
- [140] S. Amani, L. White, T. Balart, L. Arora, K. J. Shryock, K. Brumbelow, and K. L. Watson, “Generative ai perceptions: A survey to measure the perceptions of faculty, staff, and students on generative ai tools in academia,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2304.14415>
- [141] C. C. Cao, Z. Ding, J. Lin, and F. Hopfgartner, “Ai chatbots as multi-role pedagogical agents: Transforming engagement in cs education,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2308.03992>
- [142] M. R. Hasan, N. I. Chowdhury, M. H. Rahman, M. A. B. Syed, and J. Ryu, “Analysis of the user perception of chatbots in education using a partial least squares structural equation modeling approach,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2311.03636>
- [143] D. Dalalah and O. Dalalah, “The false positives and false negatives of generative ai detection tools in education and academic research: The case of chatgpt,” *INTERNATIONAL JOURNAL OF MANAGEMENT EDUCATION*, vol. 21, no. 2, 7 2023.
- [144] K. Wang, E. Burkholder, C. Wieman, S. Salehi, and N. Haber, “Examining the potential and pitfalls of chatgpt in science and engineering problem-solving,” *FRONTIERS IN EDUCATION*, vol. 8, 1 2024.

- [145] S. Kim, J. Eun, C. Oh, and J. Lee, "'journey of finding the best query': Understanding the user experience of ai image generation system," *INTERNATIONAL JOURNAL OF HUMAN-COMPUTER INTERACTION*, 2 2024.
- [146] D. N. Ege, H. H. Ovrebo, V. Stubberud, M. F. Berg, C. Elverum, M. Steinert, and H. Vestad, "The trolllabs open hackathon dataset: Generative ai and large language models for prototyping in engineering design," *Data in Brief*, vol. 54, 2024. [Online]. Available: <http://dx.doi.org/10.1016/j.dib.2024.110332>
- [147] P. Fotaris, T. Mastoras, and P. Lameris, "Designing educational escape rooms with generative ai: A framework and chatgpt prompt engineering guide," *Proceedings of the European Conference on Games-based Learning*, vol. 2023, pp. 180 – 189, 2023.
- [148] C. Lemke, K. Kirchner, An, L. arajah, and F. N. Herfurth, "Exploring the student perspective: Assessing technology readiness and acceptance for adopting large language models in higher education," *Proceedings of the European Conference on e-Learning, ECEL*, vol. 2023, pp. 156 – 164, 2023.
- [149] F. Gmeiner, H. Yang, L. Yao, K. Holstein, and N. Martelaro, "Exploring challenges and opportunities to support designers in learning to co-create with ai-based manufacturing design tools," *Conference on Human Factors in Computing Systems - Proceedings*, pp. ACM SIGCHI; Apple; Bloomberg; Google; NSF; SIEMENS –, 2023. [Online]. Available: <http://dx.doi.org/10.1145/3544548.3580999>
- [150] J. Deb, L. Saikia, K. D. Dihingia, and G. N. Sastry, "Chatgpt in the material design: Selected case studies to assess the potential of chatgpt," *Journal of Chemical Information and Modeling*, vol. 64, no. 3, pp. 799 – 811, 2024. [Online]. Available: <http://dx.doi.org/10.1021/acs.jcim.3c01702>
- [151] M. King, A. Abdulrahman, M. Petrovic, P. Poley, S. Hall, S. Kulapatana, and Z. Lamantia, "Incorporation of chatgpt and other large language models into a graduate level computational bioengineering course," *CELLULAR AND MOLECULAR BIOENGINEERING*, vol. 17, no. 1, pp. 1–6, 2 2024.
- [152] Z. Zhong, C. Wijenayake, and C. U. S. Edussooriya, "Exploring the performance of generative ai tools in electrical engineering education," *2023 IEEE International Conference on Teaching, Assessment and Learning for Engineering, TALE 2023 - Conference Proceedings*, pp. IEEE Education Society; IEEE New Zealand North Section; IEEE Region 10 –, 2023. [Online]. Available: <http://dx.doi.org/10.1109/TALE56641.2023.10398370>

- [153] E. Dickey and A. Bejarano, “A model for integrating generative ai into course content development,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2308.12276>
- [154] M.-L. Tsai, C. W. Ong, and C.-L. Chen, “Exploring the use of large language models (llms) in chemical engineering education: Building core course problem models with chat-gpt,” *Education for Chemical Engineers*, vol. 44, pp. 71 – 95, 2023. [Online]. Available: <http://dx.doi.org/10.1016/j.ece.2023.05.001>
- [155] R. Chu and S. C. J. Lim, “Education and training for future engineering teachers in the age of artificial intelligence: A bibliometric analysis,” *2023 IEEE International Conference on Industrial Engineering and Engineering Management, IEEM 2023*, pp. 416 – 420, 2023. [Online]. Available: <http://dx.doi.org/10.1109/IEEM58616.2023.10406630>
- [156] Y. Hu, Y. Goktas, D. D. Yellamati, and C. De Tassigny, “The use and misuse of pre-trained generative large language models in reliability engineering,” *Proceedings - Annual Reliability and Maintainability Symposium*, pp. IEEE –, 2024. [Online]. Available: <http://dx.doi.org/10.1109/RAMS51492.2024.10457630>
- [157] S. Moore, R. Tong, A. Singh, Z. Liu, X. Hu, Y. Lu, J. Liang, C. Cao, H. Khosravi, P. Denny, C. Brooks, and J. Stamper, “Empowering education with llms - the next-gen interface and content generation,” *Communications in Computer and Information Science*, vol. 1831, pp. 32 – 37, 2023. [Online]. Available: [http://dx.doi.org/10.1007/978-3-031-36336-8\\_4](http://dx.doi.org/10.1007/978-3-031-36336-8_4)
- [158] J. Dwighta, “Collaborate, design, and generate cybercrime script tabletop exercises for cybersecurity education,” *31ST INTERNATIONAL CONFERENCE ON COMPUTERS IN EDUCATION, ICCE 2023, VOL II*, pp. 255–264, 2023.
- [159] L. Li, Z. Ma, L. Fan, S. Lee, H. Yu, and L. Hemphill, “Chatgpt in education: a discourse analysis of worries and concerns on social media,” *EDUCATION AND INFORMATION TECHNOLOGIES*, 10 2023.
- [160] K. Bartlett and J. Camba, “Generative artificial intelligence in product design education: Navigating concerns of originality and ethics,” *INTERNATIONAL JOURNAL OF INTERACTIVE MULTIMEDIA AND ARTIFICIAL INTELLIGENCE*, vol. 8, no. 5, 3 2024.

- [161] S. Uddin, A. Albert, A. Ovid, and A. Alsharef, “Leveraging chatgpt to aid construction hazard recognition and support safety education and training,” *SUSTAINABILITY*, vol. 15, no. 9, 4 2023.
- [162] I. Joshi, R. Budhiraja, P. D. Tanna, L. Jain, Deshp, M. e, A. Srivastava, S. Rallapalli, H. D. Akolekar, J. S. Challa, and D. Kumar, “"with great power comes great responsibility!": Student and instructor perspectives on the influence of llms on undergraduate engineering education,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2309.10694>
- [163] M. Schmitt and I. Flechais, “Digital deception: Generative artificial intelligence in social engineering and phishing,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2310.13715>
- [164] R. Gao, N. Thomas, and A. Srinivasa, “Work in progress: Large language model based automatic grading study,” *Proceedings - Frontiers in Education Conference, FIE*, pp. American Society for Engineering Education (ASEE)/Educational Research and Methods Divison (ERM); IEEE; IEEE Computer Society; IEEE Education Society –, 2023. [Online]. Available: <http://dx.doi.org/10.1109/FIE58773.2023.10343006>
- [165] D. Agostini and F. Picasso, “Large language models for sustainable assessment and feedback in higher education: Towards a pedagogical and technological framework,” *CEUR Workshop Proceedings*, vol. 3605, 2023.
- [166] F. Dobslaw and P. Bergh, “Experiences with remote examination formats in light of gpt-4,” *ACM International Conference Proceeding Series*, pp. 220 – 225, 2023. [Online]. Available: <http://dx.doi.org/10.1145/3593663.3593695>
- [167] O. Shaer, A. Cooper, O. Mokryn, A. L. Kun, and H. B. Shoshan, “Ai-augmented brainwriting: Investigating the use of llms in group ideation,” *arXiv*, 2024. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2402.14978>
- [168] V. Kirova, C. Ku, J. Laracy, T. Marlowe, and A. C. Machinery, “Software engineering education must adapt and evolve for an llm (large language model) environment,” *PROCEEDINGS OF THE 55TH ACM TECHNICAL SYMPOSIUM ON COMPUTER SCIENCE EDUCATION, SIGCSE 2024, VOL. 1*, pp. 666–672, 2024.
- [169] A. Bozkurt, “Generative ai, synthetic contents, open educational resources (oer), and open educational practices (oep): A new front in the openness landscape,” *OPEN PRAXIS*, vol. 15, no. 3, pp. 178–184, 2023.

- [170] K. Ahmad, W. Iqbal, A. El-Hassan, J. Qadir, D. Benhaddou, M. Ayyash, and A. Al-Fuqaha, “Data-driven artificial intelligence in education: A comprehensive review,” *IEEE TRANSACTIONS ON LEARNING TECHNOLOGIES*, vol. 17, pp. 12–31, 2024.
- [171] T. K. Chiu, “The impact of generative ai (genai) on practices, policies and research direction in education: A case of chatgpt and midjourney,” *Interactive Learning Environments*, pp. 1–17, 2023.
- [172] S. Morales, E. Planas, R. Clariso, and M. Gogolla, “Generative ai in model-driven software engineering education: Friend or foe?” *Proceedings - 2023 ACM/IEEE International Conference on Model Driven Engineering Languages and Systems Companion, MODELS-C 2023*, pp. 110 – 113, 2023. [Online]. Available: <http://dx.doi.org/10.1109/MODELS-C59198.2023.00034>
- [173] V. D. Kirova, C. S. Ku, J. R. Laracy, and T. J. Marlowe, “Software engineering education must adapt and evolve for an llm environment,” *SIGCSE 2024 - Proceedings of the 55th ACM Technical Symposium on Computer Science Education*, vol. 1, pp. 666 – 672, 2024. [Online]. Available: <http://dx.doi.org/10.1145/3626252.3630927>
- [174] C. Bull and A. Kharrufa, “Generative ai assistants in software development education a vision for integrating generative ai into educational practice, not instinctively defending against it.” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2303.13936>
- [175] J. D. Velasquez, “Chatgpt’s and large language models influence on research, technology, and education: A comprehensive co-word analysis,” *SSRN*, 2023. [Online]. Available: <http://dx.doi.org/10.2139/ssrn.4596285>
- [176] V. Danry, J. Leong, P. Pataranutaporn, T. P. on, Y. Liu, R. Shilkrot, P. Punpongsanon, T. Weissman, P. Maes, and M. Sra, “Ai-generated characters: Putting deepfakes to good use,” *Conference on Human Factors in Computing Systems - Proceedings*, pp. ACM SIGCHI –, 2022. [Online]. Available: <http://dx.doi.org/10.1145/3491101.3503736>
- [177] Y. Dai, A. Liu, and C. P. Lim, “Reconceptualizing chatgpt and generative ai as a student-driven innovation in higher education,” *Procedia CIRP*, vol. 119, pp. 84–90, 2023.
- [178] A. Fan, B. Gokkaya, M. Harman, M. Lyubarskiy, S. Sengupta, S. Yoo, and J. M. Zhang, “Large language models for software engineering: Survey and open problems,” in *2023*



- IEEE/ACM International Conference on Software Engineering: Future of Software Engineering (ICSE-FoSE)*. IEEE, 2023, pp. 31–53.
- [179] C. W. Tan and K. Y. Lim, “Revolutionizing formative assessment in stem fields: Leveraging ai and nlp techniques,” in *2023 Asia Pacific Signal and Information Processing Association Annual Summit and Conference (APSIPA ASC)*. IEEE, 2023, pp. 1357–1364.
- [180] M. Elkhodr, E. Gide, R. Wu, and O. Darwish, “Ict students’ perceptions towards chatgpt: An experimental reflective lab analysis,” *STEM Education*, vol. 3, no. 2, pp. 70–88, 2023.
- [181] C. K. Y. Chan and K. K. Lee, “The ai generation gap: Are gen z students more interested in adopting generative ai such as chatgpt in teaching and learning than their gen x and millennial generation teachers?” *Smart learning environments*, vol. 10, no. 1, p. 60, 2023.
- [182] W. Cain, “Prompting change: Exploring prompt engineering in large language model ai and its potential to transform education,” *TechTrends: Linking Research and Practice to Improve Learning*, vol. 68, no. 1, pp. 47–57, 1 2024. [Online]. Available: <https://search.ebscohost.com/login.aspx?direct=true&AuthType=ip,shib&db=eric&AN=EJ1406409&site=ehost-live&scope=site&custid=uga1>
- [183] S. Okaiyeto, J. Bai, and H. Xiao, “Generative ai in education: To embrace it or not?” *INTERNATIONAL JOURNAL OF AGRICULTURAL AND BIOLOGICAL ENGINEERING*, vol. 16, no. 3, pp. 285–286, 5 2023.
- [184] S. Eldh, “Generative ai is changing how and what we learn,” *IEEE SOFTWARE*, vol. 41, no. 2, pp. 4–5, 3 2024.
- [185] J. Qadir, “Engineering education in the era of chatgpt: Promise and pitfalls of generative ai for education,” *IEEE Global Engineering Education Conference, EDUCON*, vol. 2023, pp. American University of Kuwait (AUK); et al.; IEEE; IEEE Education Society; IEEE Region 8; KIPCO –, 2023. [Online]. Available: <http://dx.doi.org/10.1109/EDUCON54358.2023.10125121>
- [186] Y. Zhang, “Generative ai has lowered the barriers to computational social sciences,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2311.10833>

- [187] S. S. Gill, M. Xu, P. Patros, H. Wu, R. Kaur, K. Kaur, S. Fuller, M. Singh, P. Arora, A. K. Parlikad *et al.*, “Transformative effects of chatgpt on modern education: Emerging era of ai chatbots,” *Internet of Things and Cyber-Physical Systems*, vol. 4, pp. 19–23, 2024.
- [188] D. S. McNamara, “Aied: From cognitive simulations to learning engineering, with humans in the middle,” *International Journal of Artificial Intelligence in Education*, vol. 34, no. 1, pp. 42–54, 3 2024. [Online]. Available: <https://search.ebscohost.com/login.aspx?direct=true&AuthType=ip,shib&db=psyh&AN=2024-58513-005&site=ehost-live&scope=site&custid=uga1>
- [189] B. Chimbga, “Exploring the ethical and societal concerns of generative ai in internet of things (iot) environments,” *Communications in Computer and Information Science*, vol. 1976, pp. 44 – 56, 2023. [Online]. Available: [http://dx.doi.org/10.1007/978-3-031-49002-6\\_4](http://dx.doi.org/10.1007/978-3-031-49002-6_4)
- [190] Y. Takano, T. Tsurube, H. Ueno, and H. Komatsugawa, “A proposal and evaluation of learning advising using a generative ai,” *31ST INTERNATIONAL CONFERENCE ON COMPUTERS IN EDUCATION, ICCE 2023, VOL II*, pp. 872–874, 2023.
- [191] L. White, T. Balart, S. Amani, K. J. Shryock, and K. L. Watson, “A preliminary exploration of the disruption of a generative ai systems: Faculty/staff and student perceptions of chatgpt and its capability of completing undergraduate engineering coursework,” *arXiv*, 2024. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2403.01538>
- [192] S. Sajadi, O. Ryan, L. Schibelius, and M. Huerta, “Wip: Using generative ai to assist in individual performance feedback for engineering student teams,” *Proceedings - Frontiers in Education Conference, FIE*, pp. American Society for Engineering Education (ASEE)/Educational Research and Methods Divison (ERM); IEEE; IEEE Computer Society; IEEE Education Society –, 2023. [Online]. Available: <http://dx.doi.org/10.1109/FIE58773.2023.10343517>
- [193] S. McGill and R. McGill, “Wip: Generative ai as an enhanced study aid in engineering courses,” in *ASEE Mid-Atlantic Section Spring Conference*, 2024.
- [194] F. A. Bravo and J. M. Cruz-Bohorquez, “Engineering education in the age of ai: Analysis of the impact of chatbots on learning in engineering,” *Education Sciences*, vol. 14, no. 5, p. 484, 2024.

- [195] D. E. Salinas-Navarro, E. Vilalta-Perdomo, R. Michel-Villarreal, and L. Montesinos, "Using generative artificial intelligence tools to explain and enhance experiential learning for authentic assessment," *Education Sciences*, vol. 14, no. 1, p. 83, 2024.
- [196] T. Pham, T. B. Nguyen, S. Ha, and N. T. N. Ngoc, "Digital transformation in engineering education: Exploring the potential of ai-assisted learning," *Australasian Journal of Educational Technology*, vol. 39, no. 5, pp. 1–19, 2023.
- [197] F. French, D. Levi, C. Maczo, A. Simonaityte, S. Triantafyllidis, and G. Varda, "Creative use of openai in education: case studies from game development," *Multimodal Technologies and Interaction*, vol. 7, no. 8, p. 81, 2023.
- [198] D. Rowland, "Two frameworks to guide discussions around levels of acceptable use of generative ai in student academic research and writing," *JOURNAL OF ACADEMIC LANGUAGE AND LEARNING*, vol. 17, no. 1, pp. T31–T69, 2023.
- [199] O. Kolade, A. Owoseni, and A. Egbetokun, "Is ai changing learning and assessment as we know it? evidence from a chatgpt experiment and a conceptual framework," *HELIYON*, vol. 10, no. 4, 2 2024.
- [200] A. Coscia, L. Holmes, W. Morris, J. S. Choi, S. Crossley, and A. Endert, "iscore: Visual analytics for interpreting how language models automatically score summaries," *arXiv*, 2024. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2403.04760>
- [201] F. Kieser, P. Wulff, J. Kuhn, and S. Küchemann, "Educational data augmentation in physics education research using chatgpt," *Physical Review Physics Education Research*, vol. 19, no. 2, 1 2023. [Online]. Available: <https://search.ebscohost.com/login.aspx?direct=true&AuthType=ip,shib&db=eric&AN=EJ1409576&site=ehost-live&scope=site&custid=uga1>
- [202] J. Crawford, K. Allen, B. Pani, and M. Cowling, "When artificial intelligence substitutes humans in higher education: the cost of loneliness, student success, and retention," *STUDIES IN HIGHER EDUCATION*, 3 2024.
- [203] J. Schrier, "Comment on "comparing the performance of college chemistry students with chatgpt for calculations involving acids and bases"," *JOURNAL OF CHEMICAL EDUCATION*, 4 2024.
- [204] D. N. Ege, H. H. Ovrebo, V. Stubberud, M. F. Berg, C. Elverum, M. Steinert, and H. Vestad, "The trolllabs open dataset," *SSRN*, 2024. [Online]. Available: <http://dx.doi.org/10.2139/ssrn.4720643>

- [205] M. E. Frenkel and H. Emara, “Chatgpt & mechanical engineering: Examining performance on the fe mechanical engineering and undergraduate exams,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2309.15866>
- [206] S. J. Zhang, S. Florin, A. N. Lee, E. Niknafs, A. Marginean, A. Wang, K. Tyser, Z. Chin, Y. Hicke, N. Singh, M. Udell, Y. Kim, T. Buonassisi, A. Solar-Lezama, o, and I. Drori, “Exploring the mit mathematics and eecs curriculum using large language models,” *arXiv*, 2023.
- [207] K. Owens, “Employing artificial intelligence to increase occupational tacit-knowledge through competency-based experiential learning,” *CEUR Workshop Proceedings*, vol. 3484, pp. 58 – 67, 2023.
- [208] A. O. O. Tzirides, G. Zapata, N. P. Kastania, A. K. Saini, V. Castro, S. A. Ismael, Y.-l. You, T. A. dos Santos, D. Sears Smith, C. O’Brien *et al.*, “Combining human and artificial intelligence for enhanced ai literacy in higher education,” *Computers and Education Open*, vol. 6, p. 100184, 2024.
- [209] G. Kortemeyer, “Performance of the pre-trained large language model gpt-4 on automated short answer grading,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2309.09338>
- [210] A. Martin, “Challenges requiring the combination of machine learning and knowledge engineering,” *CEUR Workshop Proceedings*, vol. 3433, 2023.
- [211] A. Zheng, “Dissecting bias of chatgpt in college major recommendations,” *arXiv*, 2023. [Online]. Available: <http://dx.doi.org/10.48550/arXiv.2401.11699>
- [212] W. Suraworachet, J. Seon, and M. Cukurova, “Predicting challenge moments from students’ discourse: A comparison of large language models to other natural language processing approaches,” *ACM International Conference Proceeding Series*, pp. 473 – 485, 2024. [Online]. Available: <http://dx.doi.org/10.1145/3636555.3636905>