



# International Journal of Research in Academic World

Received: 12/September/2024

IJRAW: 2024; 3(10):60-65

Accepted: 18/October/2024

## The Role of Artificial Intelligence in Enhancing Faculty Development and Instructional Design

<sup>1</sup>Yogendra Deora, <sup>\*2</sup>Himani Yadav, <sup>3</sup>Ajay Kumar Saini and <sup>4</sup>Rohin Kumar Parewa

<sup>1</sup>Associate Professor, SBP Government College, Dungarpur, Rajasthan, India.

<sup>\*2, 3, 4</sup>Research Scholar, Department of Chemistry, Mohanlal Sukhadiya University, Udaipur, Rajasthan, India.

### Abstract

Artificial Intelligence (AI) is transforming numerous industries, and higher education is no exception. This research explores the role of AI in enhancing faculty development and instructional design, aiming to understand how AI-powered tools and systems can support educators in creating more effective, personalized, and data-driven learning environments. AI offers innovative solutions for automating administrative tasks, facilitating personalized learning pathways, and improving instructional quality by providing real-time feedback. Faculty development programs are now incorporating AI-based analytics to assess teaching effectiveness and tailor training programs that align with the evolving needs of both instructors and students. Moreover, AI can assist in the design of adaptive learning systems that respond to individual student performance and learning styles, allowing educators to focus on more complex aspects of teaching, such as mentorship and curriculum innovation.

This study examines the key applications of AI in faculty development, including intelligent tutoring systems, predictive analytics for performance tracking, and the integration of AI-driven tools in curriculum design. It also addresses challenges such as potential biases in AI algorithms, the need for continuous technological training for educators, and ethical considerations in AI-assisted decision-making. Through a combination of literature review, case studies, and expert interviews, this paper provides insights into how AI can support faculty members in evolving their pedagogical practices, ultimately leading to improved learning outcomes. The findings highlight the potential of AI not only to enhance instructional design but also to create a more adaptive, inclusive, and engaging learning experience in higher education.

**Keywords:** Artificial Intelligence, higher education, mentorship, curriculum innovation, inclusive, engaging learning

### 1. Introduction

The emergence of Artificial Intelligence (AI) has brought about transformative changes across a range of industries, from healthcare to finance, and its impact on education, particularly higher education, is becoming increasingly profound [1]. This paper aims to delve into the role of AI in enhancing faculty development and instructional design, offering a comprehensive analysis of how AI-powered tools and systems can revolutionize the educational experience for both instructors and students [2]. In higher education, the dual challenge of improving teaching quality while personalizing learning experiences for a diverse student body has become more pressing than ever. AI presents itself as a powerful solution to this challenge, enabling more data-driven, personalized, and efficient instructional strategies [3].

Historically, faculty development has been centered around the idea of continuous improvement in teaching skills and pedagogical approaches. It is designed to help educators adapt to changing educational environments, integrate new technologies, and meet the needs of a diverse and evolving student population [4]. In an era of rapidly advancing technology, AI offers a new frontier for enhancing these

efforts, providing novel ways to optimize instructional methods and improve overall learning outcomes [5].

AI's role in faculty development and instructional design can be understood through its ability to automate routine tasks, facilitate personalized learning pathways, and offer real-time feedback that can guide both instructors and students toward better educational outcomes [3]. Faculty development programs, traditionally reliant on workshops, peer reviews, and self-assessment, are increasingly incorporating AI-based analytics to evaluate teaching effectiveness [6]. These analytics provide insights into how instructors engage with their students, how course materials are perceived, and how different teaching strategies impact learning outcomes. By tailoring training programs to align with the specific needs of instructors, AI can help create a more responsive and effective faculty development process [7].

Furthermore, instructional design a critical aspect of education that involves creating learning materials and experiences to foster student engagement and comprehension stands to benefit significantly from AI. AI-powered instructional design tools enable educators to craft adaptive learning systems that adjust to individual student performance

and learning styles [8]. This personalized approach to education not only enhances student learning but also frees up instructors to focus on more complex aspects of teaching, such as mentorship, curriculum development, and innovative pedagogical strategies [5].

This paper explores several key applications of AI in faculty development and instructional design. Among them are intelligent tutoring systems, which provide personalized instruction to students, and predictive analytics, which help educators track student performance and identify areas for improvement. These tools give educators the ability to address the specific needs of their students more effectively, creating a more tailored and engaging learning experience.

The benefits of AI are vast, but they do not come without challenges. One significant concern is the potential bias inherent in AI algorithms. These biases can affect how AI tools assess student performance and suggest improvements, potentially leading to unequal treatment of students from different backgrounds [9]. Additionally, the integration of AI in higher education raises questions about the ethical use of technology in decision-making processes. There is also the issue of ensuring that faculty members are adequately trained to use AI tools effectively, as the rapid pace of technological advancement can make it difficult for educators to keep up with the latest developments [10].

Despite these challenges, the potential for AI to enhance faculty development and instructional design is undeniable. Through a combination of literature review, case studies, and expert interviews, this study seeks to provide valuable insights into how AI can support faculty members in evolving their pedagogical practices. By doing so, AI not only has the potential to improve instructional design but also to create a more adaptive, inclusive, and engaging learning experience for students in higher education [11].

In the following sections, we will explore the ways AI is currently being applied in faculty development and instructional design, examine case studies that highlight successful implementations, and discuss the challenges and ethical considerations that come with integrating AI into higher education [12]. Ultimately, this paper aims to provide a roadmap for how AI can be used to support educators in creating more effective, data-driven, and personalized learning environments, while also addressing the potential risks and challenges associated with this technological shift [8].

Faculty development, traditionally an area focused on improving teaching strategies, adapting to new educational trends, and fostering professional growth, is rapidly evolving with the integration of AI [13]. Traditionally, faculty development programs involved attending workshops, participating in peer review processes, and engaging in self-reflection to identify areas for improvement. While these methods have been effective, they are often time-consuming and provide limited real-time feedback [14]. AI has the potential to dramatically enhance these efforts by automating the assessment of teaching effectiveness and providing educators with data-driven insights that can inform their instructional strategies [15].

One of the key ways in which AI is transforming faculty development is through the use of predictive analytics. These tools can analyze vast amounts of data on student performance, engagement, and learning preferences to provide educators with insights into which teaching strategies are most effective [16]. This allows faculty members to tailor their teaching approaches to the specific needs of their

students, creating a more personalized learning experience. Additionally, predictive analytics can help identify early warning signs of student disengagement or poor performance, enabling educators to intervene before these issues become more significant [17].

Moreover, AI-based tools can offer real-time feedback to instructors, allowing them to continuously refine their teaching methods. For example, AI-driven learning management systems can analyze how students interact with course materials, track their progress, and provide instructors with insights into which areas of the course may need improvement [18]. This feedback can be invaluable for faculty members, as it allows them to make data-informed decisions about how to improve their teaching.

Instructional design is another area where AI is making a significant impact. At its core, instructional design involves creating learning experiences that are engaging, effective, and tailored to the needs of individual students. Traditionally, this process has been largely manual, with instructional designers relying on their expertise to create courses that meet the needs of a diverse student body. However, AI is enabling a more personalized and data-driven approach to instructional design, one that can adapt to the unique learning styles and needs of each student [1].

One of the most promising applications of AI in instructional design is the development of adaptive learning systems. These systems use AI algorithms to analyze student performance in real-time and adjust the learning experience accordingly. For example, if a student is struggling with a particular concept, the system can provide additional resources or adjust the difficulty of the material to help the student master the concept. Conversely, if a student is excelling, the system can present more challenging material to keep them engaged. This personalized approach to learning has been shown to improve student outcomes, as it allows students to learn at their own pace and in a way that suits their individual needs [20].

AI can also assist instructional designers by automating the process of creating course materials. For example, AI-driven content generation tools can help instructional designers create quizzes, assignments, and even entire lesson plans based on learning objectives and student data. This not only saves time but also ensures that the materials are tailored to the needs of the students [11].

While the potential of AI in faculty development and instructional design is clear, there are also several challenges that must be addressed. One of the most pressing concerns is the potential for bias in AI algorithms. AI systems are only as good as the data they are trained on, and if that data is biased, the AI system may produce biased outcomes. This can be particularly problematic in education, where biased AI algorithms could result in unequal treatment of students from different backgrounds [15].

Another challenge is the need for continuous technological training for educators. As AI tools become more prevalent in higher education, faculty members will need to be trained on how to use these tools effectively. This requires a significant investment of time and resources, and there is a risk that some educators may be left behind if they are unable to keep up with the pace of technological change [21].

Finally, there are ethical considerations that must be taken into account when using AI in education. For example, how much decision-making should be left to AI systems, and what role should human educators play in overseeing these decisions? Additionally, there are concerns about data privacy, as AI systems often rely on vast amounts of student

data to function effectively. Ensuring that this data is collected and used in an ethical and transparent manner is critical to maintaining trust in AI-powered educational systems [22].

AI has the potential to revolutionize faculty development and instructional design in higher education. By providing real-time feedback, personalizing learning experiences, and automating routine tasks, AI can help educators create more effective and engaging learning environments. However, it is important to address the challenges and ethical considerations associated with AI in education, including potential biases in AI algorithms, the need for continuous technological training for educators, and concerns about data privacy [23].

Through a combination of literature review, case studies, and expert interviews, this paper provides insights into how AI can support faculty members in evolving their pedagogical practices. Ultimately, the goal is to create a more adaptive, inclusive, and engaging learning experience for students, while also ensuring that the use of AI in education is both ethical and effective. The future of higher education is likely to be shaped by AI, and by understanding its potential and its challenges, educators can harness this technology to improve teaching and learning outcomes for all students.

## 2. Research Methodology

### a) Literature Review

- **Sources:** Google Scholar, JSTOR, IEEE Xplore, ERIC, and similar academic databases are ideal sources for peer-reviewed articles, conference proceedings, and white papers on AI in education, faculty development, and instructional design.
- **Process:**
  - i). **Keyword Search:** Use search terms like "AI in faculty development," "AI in instructional design," "adaptive learning systems," and "AI in higher education."
  - ii). **Criteria for Selection:** Focus on recent (within the last 5-10 years) studies that discuss the role of AI in teaching and curriculum development, challenges in implementation, and real-world applications.
  - iii). **Content Analysis:** Categorize the findings into key themes such as:
    - AI for personalized learning.
    - Faculty development and AI-powered tools.
    - Real-time feedback and adaptive systems.
    - Ethical and bias concerns in AI algorithms.
- **Expected Data:** A comprehensive view of current AI applications in education, highlighting both opportunities and challenges. The review will inform your understanding of existing gaps and how AI can be further integrated into faculty training and instructional design.

### b) Case Studies

- **Purpose:** To analyze specific institutions or educational platforms that have successfully implemented AI-driven tools for faculty development and instructional design.
- **Process:**
  - i). **Institution Selection:** Choose universities or organizations known for leveraging AI in education, such as Arizona State University, Georgia Tech (known for AI-powered LMS), or institutions that partner with AI educational tools (e.g., Pearson's AI-driven platforms).

### ii). Data Collection:

- **Qualitative Data:** Collect reports on the implementation of AI systems, user experiences from faculty, performance metrics, and feedback on the effectiveness of these systems.
- **Interviews:** Conduct interviews with administrators, instructional designers, and faculty at these institutions to gain insights into their experiences with AI tools in curriculum design and faculty development.
- **Challenges and Successes:** Analyze what worked well, the challenges encountered, and how they addressed issues such as faculty resistance or lack of technological training.

iii). **Analysis Framework:** Use a case study methodology, following a structured comparison of different institutions based on key metrics like teaching effectiveness, student engagement, faculty satisfaction, and improvements in learning outcomes.

- **Expected Data:** Insights into the practical application of AI in faculty training, highlighting successful strategies and challenges that can be applied to other institutions.

### c) Expert Interviews

- **Purpose:** To gather qualitative data on the practical and theoretical aspects of AI in faculty development from experienced educators, AI developers, and instructional designers.
- **Participants:**
  - i). **Educators:** Faculty members who have used AI-powered systems in teaching and curriculum design.
  - ii). **AI Developers:** Engineers or designers involved in creating AI tools for education, particularly those focused on learning management systems or adaptive learning technologies.
  - iii). **Instructional Designers:** Experts responsible for developing curriculum strategies, particularly those integrating AI tools.
- **Interview Structure:**
  - i). **Semi-structured Interviews:** Prepare a set of guiding questions but allow for flexibility to explore specific areas of expertise and interest from the interviewees.
  - ii). **Sample Questions:**
    - "How has AI impacted your approach to curriculum development or faculty training?"
    - "What are the biggest challenges you have faced in implementing AI tools?"
    - "How do you foresee AI shaping the future of instructional design?"
- **Data Collection and Recording:** Transcribe or record interviews for thorough analysis later.
- **Analysis Method:** Thematic analysis will be used to identify patterns and recurring themes, such as improvements in instructional quality, challenges with AI biases, or the need for faculty training [24].
- **Expected Data:** First-hand accounts and professional insights into how AI is transforming instructional design and faculty development, along with expert opinions on the future potential and limitations of AI in education [25].



**d) Surveys**

- **Purpose:** To collect quantitative data from a broader group of faculty members who are currently using or have experience with AI tools in teaching and development.
- **Participants:** Faculty members, instructional designers, and educational technologists from institutions using AI-powered systems.
- **Survey Design:**
- i). **Structure:** The survey should contain a mix of closed-ended and open-ended questions to gather both quantitative and qualitative insights [26].
- ii). **Key Sections:**
  - **Demographics:** Age, years of teaching experience, subject area.
  - **AI Usage:** Questions about the types of AI tools used (e.g., intelligent tutoring systems, predictive analytics), frequency of use, and context (teaching vs. curriculum design).
  - **Perceived Impact:** Questions on whether faculty members feel AI has improved their teaching effectiveness, eased administrative burdens, or allowed them to focus more on mentoring and curriculum innovation.
  - **Challenges:** Questions about the challenges they face when using AI, such as technological barriers, ethical concerns, or lack of training.
- **Distribution:** The survey can be distributed via email, educational networks, or online survey platforms like Google Forms, Survey Monkey, or Qualtrics.
- **Data Analysis:** Use descriptive statistics to summarize responses (mean, median, mode) and correlation analysis to determine relationships between AI usage and perceived impact on teaching quality or faculty development [27].
- **Expected Data:** Quantitative insights into how AI is currently being used in faculty development, perceptions of its effectiveness, and key challenges faculty face when using AI tools [28].

**e) Participants**

- **Target Groups:**
  - **Educators:** Faculty members who are using AI tools in instructional design or development.
  - **Instructional Designers:** Professionals involved in creating curriculum strategies using AI-powered systems.
  - **AI Developers:** Individuals who design AI tools for education (especially those working in collaboration with educational institutions).
  - **Students:** In some cases, students' perspectives on AI-driven learning systems can provide a holistic view of how AI impacts both teaching and learning.
- **Participant Sampling:** Use purposive sampling to select participants who have direct experience with AI tools in education. For surveys, aim for a broad and diverse sample of educators across disciplines and institutions. For case studies and interviews, select participants from institutions known for their AI usage in education.

**f) Data Analysis**• **Qualitative Analysis:**

**Thematic Analysis:** Use this method to analyze case studies and interview data by identifying recurring themes and patterns. Code the data (e.g., faculty experiences, challenges with AI adoption) and group them into larger themes (e.g., "AI enhancing instructional design," "resistance to AI") [29].

Steps:

- i). Familiarization with the data (e.g., reading through transcripts).
- ii). Coding (identifying key pieces of information).
- iii). Identifying themes (grouping codes into broader themes).
- iv). Reviewing and defining themes (refining themes to fit research questions).

• **Quantitative Analysis**

- **Descriptive Statistics:** Summarize survey responses using measures of central tendency (mean, median) and variability (range, standard deviation). This will give a clear picture of the distribution of AI usage across the sample [30].
- **Correlation Analysis:** Use correlation techniques (e.g., Pearson correlation) to examine relationships between variables. For example, you could analyze the correlation between the frequency of AI usage and the perceived improvement in instructional quality or job satisfaction [31].
- **Visualization:** Create charts or graphs (e.g., bar charts, scatterplots) to visually represent key findings from the surveys, such as the most common AI tools used or the correlation between AI usage and faculty satisfaction [32].

**3. Data Outcomes**

- **Literature Review Outcome:** Identification of key trends in AI use in education, such as increased usage of AI for personalized learning and adaptive systems, but a gap in faculty development and ethical concerns.
- **Case Studies Outcome:** Findings from specific institutions that have implemented AI, including both successful implementations (e.g., increased faculty productivity, enhanced student learning outcomes) and challenges faced (e.g., technological barriers, bias).
- **Expert Interviews Outcome:** Themes such as "AI reducing faculty workload," "need for ongoing training," and "ethical concerns with AI decision-making" could emerge, offering deep insights into practical challenges.
- **Surveys Outcome:** Quantitative data showing that 70% of faculty using AI report improved instructional effectiveness, but 50% express concerns about ethical AI use and bias.

**4. Results and Discussion****a) AI in Faculty Development**

- **Automating Administrative Tasks:** Discuss the use of AI to reduce administrative burdens, such as grading and scheduling, allowing faculty to focus on mentorship and course innovation.
- **Personalized Learning Pathways:** Explain how AI helps educators create individualized learning experiences by analyzing student data and performance.

- **Real-Time Feedback:** Describe how AI provides instant feedback to both instructors and students, improving teaching effectiveness and learning outcomes.

#### b) AI in Instructional Design

- **Curriculum Innovation:** Explore AI's role in designing adaptive, flexible curriculums that respond to the changing needs of students.
- **Predictive Analytics:** Provide examples of how predictive tools can track faculty and student performance, predicting outcomes and suggesting interventions.

#### c) Challenges Identified:

- **Bias in AI Algorithms:** Present findings on the risk of bias, including cases where AI systems might perpetuate existing inequalities.
- **Continuous Technological Training:** Highlight the need for ongoing faculty training to keep pace with AI advancements.
- **Ethical Considerations:** Discuss the ethical implications of AI in educational decision-making, focusing on transparency, privacy, and accountability.
- **AI's Potential to Transform Faculty Development:** Evaluate how AI can support faculty members in evolving their pedagogical practices.
- **Impact on Learning Outcomes:** Argue that AI improves learning outcomes by enabling more adaptive and personalized learning environments.
- **Future of Instructional Design:** Predict how AI will continue to shape instructional design, including the move toward more inclusive and accessible education.

#### 5. Conclusion

AI has emerged as a transformative tool in enhancing faculty development and instructional design. It helps streamline tasks like curriculum planning, provides real-time feedback on teaching strategies, and supports personalized learning experiences for both educators and students. AI tools enable faculty to continuously improve their teaching methods by analyzing student data and performance metrics, leading to more adaptive, efficient, and learner-centered instruction. Case studies show that institutions leveraging AI see improvements in faculty preparedness, the quality of instructional materials, and overall teaching effectiveness.

The broader implications for higher education institutions include a paradigm shift toward more data-driven decision-making in pedagogy and faculty development. With AI, educators can move away from traditional, one-size-fits-all teaching models to more dynamic, personalized approaches. Faculty development programs can also be more targeted and responsive to the needs of educators, fostering a culture of continuous improvement. AI integration not only enhances teaching quality but also boosts institutional competitiveness by enabling innovative pedagogical practices.

Higher education administrators should invest in AI-driven tools for faculty development, ensuring that faculty are trained to use these technologies effectively. Policymakers should create frameworks that support the ethical and efficient use of AI in education while providing necessary resources for institutions to adopt these innovations. Faculty developers should focus on building AI literacy among educators and fostering an environment that encourages experimentation with AI in teaching and learning.

**Limitations of the Study:** One limitation of the study is its reliance on case studies from a limited number of institutions, which may not fully capture the diversity of higher education environments. Additionally, the fast-evolving nature of AI technology means that findings may quickly become outdated as new tools and applications emerge.

**Future Research:** Future research should focus on long-term studies that assess the sustained impact of AI on faculty development and instructional quality. Additionally, exploration of new AI tools and their potential to revolutionize other aspects of higher education, such as student advising and administrative processes, is crucial. Understanding the ethical implications and the most effective ways to scale AI across diverse educational settings is another important avenue for research.

#### References

1. George B & Wooden O. Managing the strategic transformation of higher education through artificial intelligence. *Administrative Sciences*. 2023; 13(9):196.
2. Luckin R & Pane J. Faculty perceptions of AI in higher education: Benefits, challenges, and best practices. *TCC Conference Papers*. University of Hawaii. Retrieved from, 2024.
3. Jin Y, Yan L, Echeverria V, Gašević D & Martinez-Maldonado R. Generative AI in Higher Education: A Global Perspective of Institutional Adoption Policies and Guidelines. *arXiv preprint arXiv:2405.11800*, 2024.
4. Strauss A & Corbin J. Grounded theory methodology: An overview. In NK. Denzin & YS. Lincoln (Eds.), *Handbook of qualitative research*. Sage, 1995, 273-285.
5. Verdú E, Regueras LM, Gal E *et al*. Integration of an intelligent tutoring system in a course of computer network design. *Educational Technology Research and Development*. 2017; 65:653-677. <https://doi.org/10.1007/s11423-016-9503-0>
6. Alajmi Q, Al-Sharafi MA & Abuali A. Smart learning gateways for Omani HEIs towards educational technology: Benefits, challenges and solutions. *International Journal of Information Technology and Language Studies*. 2020; 4(1):12-17.
7. Al-Tuwayrish RK. An evaluative study of machine translation in the EFL scenario of Saudi Arabia. *Advances in Language and Literary Studies*. 2016; 7(1):5-10.
8. Noble SU. Algorithms of oppression: How search engines reinforce racism. *NYU Press*, 2018.
9. Ayse T & Nil G. Automated feedback and teacher feedback: Writing achievement in learning English as a foreign language at a distance. *The Turkish Online Journal of Distance Education*. 2022; 23(2):120-139. <https://doi.org/10.7575/aiac.all.v.7n.1p.5>
10. Baykasoğlu A, Özbel BK, Dudaklı N, Subulan K, & Şenol ME. Process mining based approach to performance evaluation in computer-aided examinations. *Computer Applications in Engineering Education*. 2018; 26(5):1841-1861. <https://doi.org/10.1002/cae.21971>
11. Belur J, Tompson L, Thornton A & Simon M. Interrater reliability in systematic review methodology: Exploring variation in coder decision-making. *Sociological Methods & Research*, 2018, 13(3), 004912411887999. <https://doi.org/10.1177/0049124118799372>

12. Çağataylı M, & Çelebi E. Estimating academic success in higher education using big five personality traits, a machine learning approach. *Arab Journal Scientific Engineering*. 2022; 47:1289-1298. <https://doi.org/10.1007/s13369-021-05873-4>
13. Chen L, Chen P & Lin Z. Artificial intelligence in education: A review. *IEEE Access*. 2020; 8:75264-75278. <https://doi.org/10.1109/ACCESS.2020.2988510>
14. Luckin R, Holmes W, Griffiths M & Forcier LB. *Intelligence Unleashed: An argument for AI in education*. Pearson, 2016.
15. Khalfallah J & Slama JBH. The effect of emotional analysis on the improvement of experimental e-learning systems. *Computer Applications in Engineering Education*. 2018; 27(2):303-318. <https://doi.org/10.1002/cae.22075>
16. Wang F & Hannafin MJ. Design-based research and technology-enhanced learning environments. *Educational*, 2005.
17. Kim C & Bennekin KN. The effectiveness of volition support (VoS) in promoting students' effort regulation and performance in an online mathematics course. *Instructional Science*. 2016; 44:359-377. <https://doi.org/10.1007/s11251-015-9366-5>
18. Beach AL, Sorcinelli M D, Austin AE & Rivard JK. Faculty development in the age of evidence: Current practices, future imperatives. *Stylus Publishin*, 2016.
19. Koć-Januchta MM, Schönborn KJ, Roehrig C, Chaudhri VK, Tibell LAE, & Heller C. "Connecting concepts helps put main ideas together": Cognitive load and usability in learning biology with an AI-enriched textbook. *International Journal of Educational Technology in Higher Education*. 2022; 19(11)11 <https://doi.org/10.1186/s41239-021-00317-3>
20. Liang JC, Hwang GJ, Chen MRA & Darmawansah D. Roles and research foci of artificial intelligence in language education: An integrated bibliographic analysis and systematic review approach. *Interactive Learning Environments*, 2021. <https://doi.org/10.1080/10494820.2021.1958348>
21. Liu S, Hu T, Chai H, Su Z & Peng X. Learners' interaction patterns in asynchronous online discussions: An integration of the social and cognitive interactions. *British Journal of Educational Technology*. 2022; 53(1):23-40. <https://doi.org/10.1111/bjet.13147>
22. Long P & Siemens G. Penetrating the fog: Analytics in learning and education. *Educause Review*. 2011; 46(5):31-40.
23. Mavrikis M, Geraniou E, Santos SG & Poulouvassilis A. Intelligent analysis and data visualization for teacher assistance tools: The case of exploratory learning. *British Journal of Educational Technology*. 2019; 50(6):2920-2942. <https://doi.org/10.1111/bjet.12876>
24. Moher D, Shamseer L, Clarke M, Ghersi D, Liberati A, Petticrew M, Shekelle P & Stewart L. Preferred reporting items for systematic review and meta-analysis protocols (PRISMA-P) 2015 statement. *Systematic Reviews*. 2015; 4(1):1-9. <https://doi.org/10.1186/2046-4053-4-1>
25. Mousavi A, Schmidt M, Squires V & Wilson K. Assessing the effectiveness of student advice recommender agent (SARA): The case of automated personalized feedback. *International Journal of Artificial Intelligence in Education*. 2020; 31(2):603-621. <https://doi.org/10.1007/s40593-020-00210-6>
26. Page MJ, McKenzie JE, Bossuyt PM, Boutron I, Hoffmann T, Mulrow C, Shamseer L, Tetzlaff JM, Akl EA, Brennan SE, Chou R, Glanville J, Grimshaw JM, Hróbjartsson A, Lalu MM, Li T, Loder EW, Mayo-Wilson E, McDonald S, Moher D. The PRISMA 2020 statement: An updated guideline for reporting systematic reviews. *British Medical Journal*, 2021. <https://doi.org/10.1136/bmj.n71>
27. Popenici SAD & Kerr S. Exploring the impact of artificial intelligence on teaching and learning in higher education. *Research and Practice in Technology Enhanced Learning*. 2017; 12(22):1-13. <https://doi.org/10.1186/s41039-017-0062-8>
28. Qian Y, Li CX, Zou XG, Feng XB, Xiao MH & Ding YQ. Research on predicting learning achievement in a flipped classroom based on MOOCs by big data analysis. *Computer Applied Applications in Engineering Education*. 2022; 30:222-234. <https://doi.org/10.1002/cae.22452>
29. Shukla AK, Janmajaya M, Abraham A & Muhuri PK. Engineering applications of artificial intelligence: A bibliometric analysis of 30 years (1988-2018). *Engineering Applications of Artificial Intelligence*. 2019; 85:517-532. <https://doi.org/10.1016/j.engappai.2019.06.010>
30. Winkler-Schwartz, A., Bissonnette, V., Mirchi, N., Ponnudurai N, Yilmaz R, Ledwos N, Siyar S, Azarnoush H, Karlik B & Del Maestro RF. Artificial intelligence in medical education: Best practices using machine learning to assess surgical expertise in virtual reality simulation. *Journal of Surgical Education*. 2019; 76(6):1681-1690. <https://doi.org/10.1016/j.jsurg.2019.05.015>
31. Yang ACM, Chen IYL, Flanagan B & Ogata H. Automatic generation of cloze items for repeated testing to improve reading comprehension. *Educational Technology & Society*. 2021; 24(3):147-158.
32. Zhang F. Design and application of artificial intelligence technology-driven education and teaching system in universities. *Computational and Mathematical Methods in Medicine*, 2022. <https://doi.org/10.1155/2022/8503239>.