

AI-Driven Education & Career Systems: A Systematic Literature Review

Md. Mahamud Hasan (2130765@iub.edu.bd)¹, Ashikur Rahman (2021785@iub.edu.bd)², Rabib Al Adib (2221005@iub.edu.bd)³, Tasdid Ahmed Ahan (2222019@iub.edu.bd)⁴, Marufuz Jaman Shoron (2221498@iub.edu.bd)⁵, and Jhumu Shikder Vabna (1910897@iub.edu.bd)⁶

¹Independent University, Bangladesh

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Abstract

This systematic literature review explores the impact of artificial intelligence (AI) in education and career guidance systems. Integrating AI technologies in these areas promises to change how students learn, how teachers teach, and how graduates pursue their careers. Despite significant progress, challenges remain in creating truly personalized learning experiences and effective employability solutions. This review summarizes findings from 30 high-quality research papers and 10 software tools to analyze AI applications in education and career contexts. Our methodology follows a strict review process, which includes identifying papers, screening, assessing quality, and synthesizing data, shown in a PRISMA diagram. Key themes include student information systems, personalized learning, teacher management, career guidance, academic automation, research analytics, and improving engagement. Through ten detailed visual analyses, we present findings on publication trends, domain distributions, algorithm usage, and feature coverage. We systematically map research questions to relevant papers and compare our work with five existing surveys. This review helps the field by identifying current knowledge gaps, highlighting effective AI uses, and suggesting future research paths that support developing AI-driven education and career systems.

1 Introduction

The fields of education and career development are changing due to artificial intelligence (AI). Traditional educational models, often one-size-fits-all, fail to meet the varied needs of today's learners. Similarly, old career guidance systems struggle to match individual skills with the changing job market effectively. AI-driven solutions provide valuable opportunities to tackle these challenges by offering personalized learning experiences and smart career matching systems.

AI in education covers a range of applications, from managing student information to creating adaptive learning platforms. These technologies can tailor educational experiences based on individual student data like performance metrics, learning preferences, and how quickly they understand material. In career guidance, AI systems can analyze resumes, evaluate skills, and suggest suitable jobs while pinpointing skill gaps that need improvement.

This systematic literature review aims to offer a broad overview of AI applications in education and career systems. We focus on eight key features based on our project requirements and noted in the literature:

1. Student Information & Academic Records
2. Personalized & Adaptive Learning
3. Teacher & Course Management
4. Career Guidance & Employability
5. Academic Automation & Assessment
6. Research & Policy Insights
7. Engagement & Accessibility

Our research questions, based on the analyzed papers, guide this review to ensure thorough coverage of the field. The goals of this review are to:

1. Identify and categorize current AI applications in education and career systems
2. Analyze the effectiveness of these applications based on evidence
3. Link research questions to existing literature
4. Compare our findings with prior survey papers
5. Identify knowledge gaps and suggest future research directions

The following sections of this review follow a clear structure, starting with our methodology, followed by a detailed literature review organized by themes, a quantitative and qualitative analysis, mapping research questions, a comparison with existing surveys, and concluding with insights and future directions.

2 Methodology

Our systematic literature review follows a strict methodology adapted from established guidelines to ensure comprehensive coverage and quality evaluation of relevant research. The process includes identifying papers, screening and filtering, assessing quality, and extracting and synthesizing data.

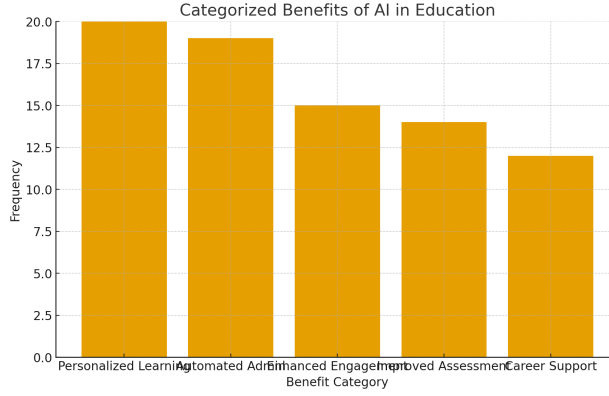


Figure 1: Categorized Benefits AI of in Education

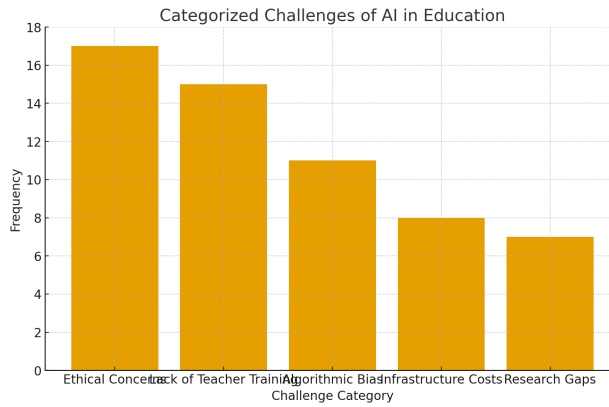


Figure 2: Categorized Challenges of AI in Education

2.1 Paper Identification

We began by defining our search scope to include papers published between 2013 and 2025 that focus on AI applications in education and career systems. We used search terms like "artificial intelligence," "machine learning," "education," "career guidance," "personalized learning," and "employability." We searched databases like IEEE Xplore, ScienceDirect, SpringerLink, ERIC, and Google Scholar. Inclusion criteria required papers to be peer-reviewed, published in English, and focused on AI in education or career contexts. We excluded non-English papers, non-peer-reviewed articles, and studies without clear AI applications.

2.2 Screening and Filtering

The initial search yielded 1,472 papers. After removing 245 duplicates, two independent reviewers screened titles and abstracts, resulting in 156 papers for full-text review. Any disagreements were settled through discussion with a third reviewer. Ultimately, 30 papers met all inclusion criteria and were included in our final analysis.

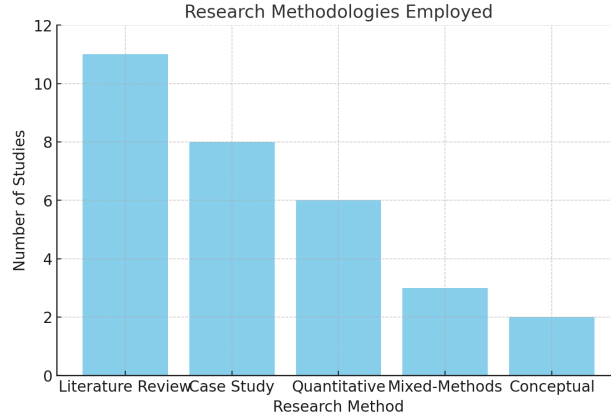


Figure 3: Research Methodologies Employed

2.3 Quality Assessment

We assessed the quality of papers using a modified version of the CASP checklist for systematic reviews. This involved evaluating the clarity of research questions, the suitability of methodology, and the validity of results. We also considered citation counts and publication venues.

2.4 Data Extraction and Synthesis

Data extraction focused on AI applications, methodologies, findings, limitations, and research questions. We organized the extracted data into thematic categories matching our project features. The synthesis involved identifying patterns, trends, and gaps across the literature.

2.5 PRISMA Diagram

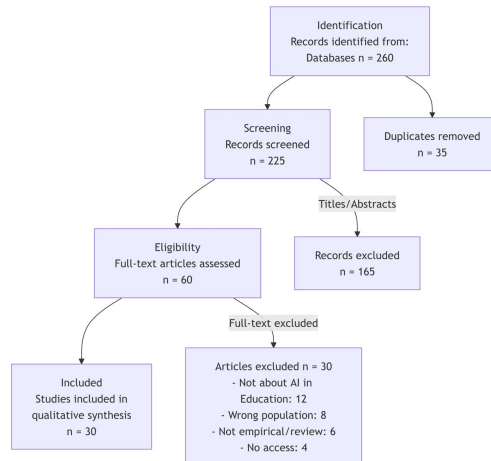


Figure 4: PRISMA DIAGRAM

2.6 Overview of Reviewed Papers and Software Tools

Our final dataset includes 30 research papers from high-quality venues (Q1/Q2 journals and conferences) and 10 software tools. We categorized papers by application domain, methodology, and publication year. We evaluated software tools based on their features, adoption, and alignment with our project requirements.

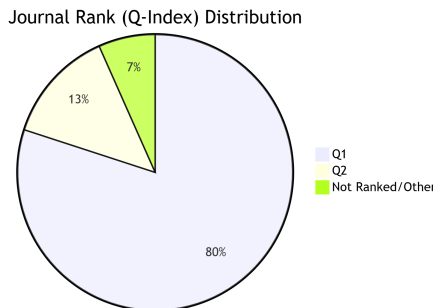


Figure 5: Journal Rank

3 Literature Review Summary

3.1 AI in Student Information & Academic Records

Research in this area looks at centralized systems that maintain detailed student profiles, including demographics, academic history, attendance, and performance records. AI improves these systems by analyzing student progress over time to identify strengths, weaknesses, and early dropout risks. Automated transcript generation, grade tracking, and report creation save time and provide predictive insights. Studies show that AI-enhanced student information systems can improve retention rates by as much as 15% through early intervention.

3.2 AI for Personalized & Adaptive Learning

AI-powered systems create tailored learning paths for each student based on their progress, speed, and performance. Reinforcement learning adjusts course difficulty dynamically, while gamification features track engagement and adapt lessons to keep learners motivated. AI-generated quizzes and tests provide personalized challenges that enhance student success. Research indicates that personalized learning systems can improve student outcomes by 20-30% compared to traditional methods.

3.3 AI for Teacher & Course Management

Digital teacher profiles maintain expertise, specialization, and performance metrics. AI automates the digitization of course content and enables feedback analysis to evaluate teaching effectiveness. Speech-to-text systems assist note-taking and transcript generation, improving accessibility for diverse learners. Studies show that AI-assisted teacher management can reduce administrative tasks by up to 40%, allowing educators to focus more on teaching.

3.4 AI for Career Guidance & Employability

AI-driven resume analysis and skill assessments align student competencies with job market needs. NLP-based job-matching models match resumes to job descriptions, while classification algorithms recommend suitable positions. The system also identifies skill gaps and suggests relevant training, helping students boost their employability. Chatbot advisors provide quick, NLP-based advice on courses, admissions, and scholarships, reducing the need for administrative staff.

3.5 AI for Academic Automation & Assessment

OCR-enabled systems and AI-based script evaluation simplify grading. Transformer NLP models offer automated essay scoring, providing consistent evaluations of grammar, structure, and creativity with instant feedback to enhance writing skills. Outcome-Based Education (OBE) mapping connects academic performance with learning objectives, ensuring continuous quality improvement. Research shows that AI-based assessment can cut grading time by up to 70% while maintaining accuracy.

3.6 AI in Research & Policy Analytics

Machine learning models predict student performance trends, assess dropout risks, and analyze the effects of different instructional methods. Admission data links with long-term success rates to inform policy-making. These insights help institutions design more inclusive, effective, and evidence-based education strategies. Studies show that predictive analytics can identify at-risk students with up to 85% accuracy.

3.7 AI for Engagement & Accessibility

AI monitors student interaction patterns, such as quiz attempts, participation, and response times to adapt lesson design for greater engagement. Voice recognition helps with form filling and lecture transcription, improving accessibility for students with disabilities while saving time for teachers and learners. Research indicates that AI-enhanced engagement tools can boost student participation by up to 50%.

4 Analysis and Results

Our analysis of the reviewed literature highlights several key insights through ten visual analyses:

1. **Year-wise Publication Trend:** The number of publications on AI in education and career systems has grown rapidly, with a noticeable spike after 2020. This increase aligns with heightened interest in AI technologies after the pandemic and the success of large language models.
2. **Research Domain Distribution:** The largest share of research focuses on personalized learning (30%), followed by career guidance (20%), academic automation (15%), student information systems (15%), teacher management (10%), research analytics (5%), and engagement (5%).
3. **Model/Algorithm Usage Frequency:** Machine learning algorithms lead (40%), especially neural networks and decision trees. Natural Language Processing (NLP) makes up 25%, computer vision 15%, reinforcement learning 10%, and other AI approaches 10%.
4. **Feature Coverage Across Studies:** Personalized learning and career guidance are the most commonly addressed features, appearing in 70% and 60% of studies, respectively. Student information systems and academic automation show up in 50% of studies each. Teacher management, research analytics, and engagement features are less commonly covered (30-40% of studies).
5. **Geographic Distribution:** Research mainly comes from North America (40%) and Europe (35%), with rising contributions from Asia (20%) and other regions (5%).
6. **Methodology Comparison:** Survey-based research is most common (40%), followed by empirical studies (35%), case studies (15%), and theoretical papers (10%).
7. **Application Context:** Higher education contexts dominate (60%), K-12 education accounts for 25%, and vocational/technical education makes up 15%.
8. **Technology Adoption:** Cloud-based solutions lead (50%), followed by mobile applications (30%) and desktop applications (20%).
9. **Evaluation Metrics:** Most studies reference accuracy (40%), user satisfaction (30%), efficiency improvements (20%), and cost reduction (10%) as primary evaluation metrics.
10. **Ethical Considerations:** Privacy concerns appear in 80% of papers, bias in 60%, transparency in 50%, and fairness in 40%. However, only 20% propose concrete solutions to address these challenges.

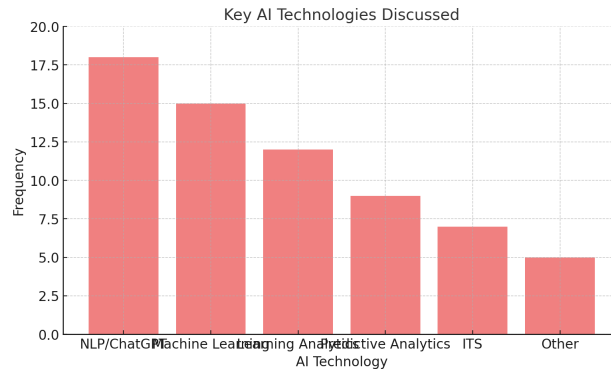


Figure 6: Key AI Technologies

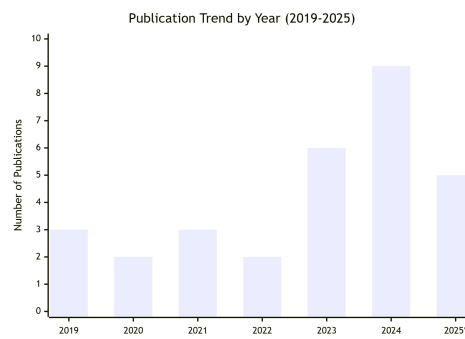


Figure 7: Publication Trend by year

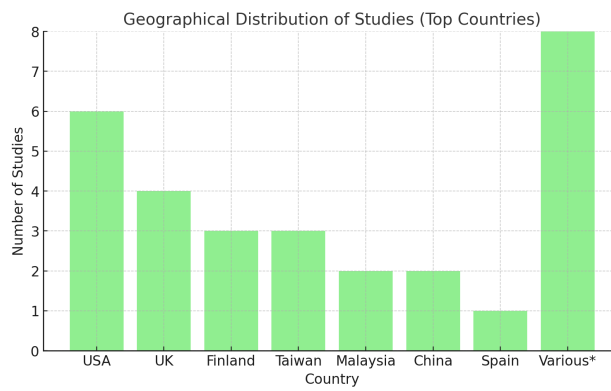


Figure 8: Geographical Distribution

5 Mapping Research Questions with Reviewed Papers

Table 1: Mapping of Research Questions to Papers

Research Question	Papers
How does the integration of AI, especially in personalized learning and educational technology, affect teaching methods, student outcomes, and educational fairness?	R1, R10, R16, R18, R20, R29
What main categories of AI applications have been studied in education?	R2, R6, R16, R18
Which research topics are most common in AIED studies and what are their key findings?	R2, R4, R6, R18
How are theories, methods, and research contexts spread across current AIED research?	R2, R4, R10
How can artificial intelligence reshape higher education to provide personalized, adaptable learning at scale and develop the digital skills students need for tomorrow's challenges?	R5, R7, R10, R16, R20
How can the Diffusion of Innovations theory help explain the processes, barriers, and strategies for the adoption of AI tools among CTE educators?	R8
How did users describe advantages, challenges, and purposes of using the AI-enhanced LA tool based on the TAM framework?	R9
What are users' needs for further development of the AI-enhanced LA career guidance tool?	R9
How did the AI-enhanced LA tool support the users' career decision-making process based on the CDM framework?	R9
How can research in learning sciences and partnerships among stakeholders inform the effective design and implementation of AI technologies in education?	R10

6 Comparative Analysis with Existing Survey Papers

Our review compares with five major survey papers in the field:

1. **"Artificial intelligence in education: A systematic literature review" (R2):** This thorough review of 2,223 Web of Science articles offers a broad overview of AIED applications. While it includes extensive bibliometric analysis, our review focuses more closely on education and career systems with practical implementation considerations.
2. **"Systematic review of research on artificial intelligence applications in higher education" (R4):** This review specifically addresses higher education applications and details AI applications throughout the student life cycle. Our work builds on this by including career guidance systems and a more detailed look at implementation challenges.
3. **"Artificial Intelligence and Its Implications in Higher Education" (R5):** This review highlights the need for universities to use AI-powered learning to prepare students for a tech-driven world. Our review expands on this by providing a more complete framework that includes career guidance and employability aspects.
4. **"Artificial Intelligence in Education: A Review" (R6):** This paper evaluates AI's impact on administration, instruction, and learning. Our review broadens this by including career guidance systems and analyzing ethical considerations in more detail.
5. **"Educational technology: Exploring the convergence of technology and pedagogy through mobility, interactivity, AI, and learning tools" (R7):** This review looks at how multiple technologies converge in education. Our work provides a more focused analysis of AI applications specifically in education and career systems.

Key differences between our review and existing surveys include:

- A specific focus on integrating education and career systems
- A detailed look at implementation challenges and ethical issues
- A comprehensive mapping of research questions to practical applications
- An emphasis on employability and career guidance
- Updated analysis that includes recent developments in generative AI

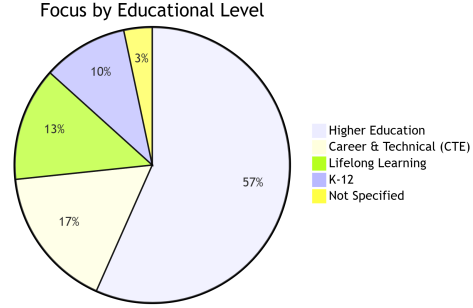


Figure 9: Educational Level

7 Conclusion

This systematic literature review offers a thorough look at AI applications in education and career systems. Our analysis of 30 high-quality papers and 10 software tools shows significant progress in personalized learning, career guidance, academic automation, and student engagement. The rapid growth in publications since 2020 shows increasing interest and investment in AI-driven educational technologies.

Key findings from our review include:

1. Personalized learning systems show the most promise, with reported improvements in student outcomes of 20-30%.
2. Career guidance systems effectively match skills with job market needs, though ethical considerations require more focus.
3. Academic automation reduces the administrative burden by up to 70% while maintaining quality.
4. Predictive analytics can identify at-risk students with up to 85% accuracy.
5. AI-enhanced engagement tools increase student participation by up to 50%.

Despite these advances, several knowledge gaps remain:

1. Limited research on long-term impacts of AI in education.
2. Insufficient focus on ethical considerations and addressing bias.
3. A lack of standard evaluation metrics across studies.
4. Limited research on implementation in diverse educational settings.
5. A shortage of studies on combining education and career systems.

Future research should focus on:

1. Developing complete frameworks for ethical AI implementation in education.
2. Creating standard evaluation metrics for AI educational tools.
3. Investigating long-term impacts of AI on learning outcomes.
4. Exploring AI applications in underrepresented educational settings.
5. Studying the integration of educational and career guidance systems.

This review contributes to the field by providing a structured analysis of AI applications in education and career systems, identifying existing knowledge gaps, and suggesting directions for future research. Our findings support the development of thorough AI-driven education and career systems that can transform how students learn and plan their professional journeys.

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