

The Future of Adaptive Learning Systems in Education

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Abstract

The rapid advancement of artificial intelligence (AI) and data-driven technologies has revolutionized education, paving the way for adaptive learning systems that personalize instruction based on individual learner needs. Traditional one-size-fits-all teaching methods often fail to accommodate diverse learning paces, styles, and abilities, necessitating more dynamic and responsive educational solutions. Adaptive learning systems leverage AI algorithms, real-time analytics, and machine learning to deliver customized learning experiences, enhancing student engagement and knowledge retention. This study aims to evaluate the effectiveness and future potential of adaptive learning technologies in education by analyzing their impact on learner outcomes, engagement, and instructional efficiency. A mixed-methods research approach was employed, integrating experimental assessments, student surveys, and instructor interviews across various educational institutions. Findings indicate that adaptive learning systems significantly improve student performance by tailoring instructional content, providing real-time feedback, and optimizing learning pathways. Statistical analysis revealed strong correlations between adaptive learning features and enhanced academic achievement. The study concludes that adaptive learning will play a transformative role in the future of education, requiring continuous refinement, ethical considerations, and policy development for broader implementation. Future research should explore long-term impacts and best practices for integrating AI-driven education across diverse learning environments.

Keywords: Adaptive Learning, Artificial Intelligence, Personalized Learning



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INTRODUCTION

The rapid evolution of digital technologies has profoundly influenced the landscape of education, leading to the emergence of adaptive learning systems that personalize instruction based on individual student needs. Traditional education models often adopt standardized teaching approaches that fail to accommodate variations in learning paces, cognitive abilities, and knowledge retention among students (Fadele dkk., 2022; Haningsih dkk., 2022). The integration of artificial intelligence (AI), machine learning, and real-time data analytics has facilitated the development of adaptive learning technologies capable of tailoring content delivery, providing instant feedback, and optimizing learning trajectories. These advancements have transformed the way educators approach student engagement, fostering more personalized and efficient learning experiences. The increasing reliance on digital platforms and online learning environments has further accelerated the adoption of adaptive learning systems, making their role in the future of education more critical than ever.

Educational institutions worldwide are exploring innovative strategies to improve student learning outcomes, and adaptive learning has emerged as a promising solution to address diverse learner needs. Research suggests that adaptive learning enhances student motivation, facilitates knowledge retention, and provides educators with valuable insights into student progress (Cohen & Enayat, 2023; McCrea & Al-Najdi, 2012). Unlike conventional learning environments where instruction follows a rigid structure, adaptive learning systems dynamically adjust content difficulty, instructional pace, and assessment methods based on real-time learner interactions. The implementation of AI-driven learning environments offers the potential to bridge educational gaps by ensuring that students receive personalized instruction tailored to their strengths and weaknesses. As adaptive learning systems continue to evolve, it becomes imperative to assess their long-term impact on educational efficacy and scalability.

The integration of adaptive learning in educational settings has been met with both enthusiasm and skepticism, raising important questions about its effectiveness, accessibility, and ethical implications (Göl, 2012; Mudhofi dkk., 2025). While some educators and researchers advocate for its potential to revolutionize personalized instruction, others highlight concerns related to data privacy, algorithmic bias, and the role of human educators in AI-enhanced learning environments. The necessity for empirical research examining the effectiveness, challenges, and future potential of adaptive learning systems remains a pressing issue in the field of education technology. Understanding how these systems shape learning experiences and contribute to academic achievement will provide valuable insights for policymakers, instructional designers, and educators.

The primary issue this study seeks to address is the effectiveness and future potential of adaptive learning systems in enhancing education (Attaufiqi dkk., 2024; Mudhofi dkk., 2025). Despite significant advancements in AI-driven educational tools, questions remain regarding the extent to which adaptive learning improves student performance, engagement, and long-term knowledge retention. Many existing adaptive learning platforms claim to offer personalized learning pathways, but the variability in their implementation and effectiveness raises concerns about consistency and scalability. The absence of standardized metrics for evaluating adaptive learning outcomes further complicates efforts to assess its impact on education. This research aims to provide empirical evidence on how adaptive learning contributes to student success and instructional efficiency.

One of the critical challenges in adaptive learning implementation is the accessibility of these technologies across different educational settings. While some institutions have successfully integrated AI-driven learning platforms, others face barriers related to infrastructure, digital literacy, and financial constraints (Rasit dkk., 2025; Shahnahpur, 2021). The digital divide remains a persistent issue, with students from underprivileged backgrounds often lacking access to the necessary technological resources to benefit from adaptive learning systems. Examining the factors that influence the accessibility and adoption of adaptive learning is essential for ensuring its equitable implementation in diverse educational environments. Addressing these disparities will help maximize the benefits of adaptive learning across various socio-economic and academic contexts.

Another pressing concern is the role of educators in AI-driven learning environments. While adaptive learning systems provide personalized instruction, they cannot fully replace human educators' pedagogical expertise, emotional intelligence, and ability to foster meaningful interactions (Azman dkk., 2025; Tuhcic & Topalovic, 2020). Balancing technological innovation with effective teaching practices requires an understanding of how educators can complement AI-driven instruction. This study seeks to explore how adaptive learning can be integrated into traditional and blended learning models while maintaining the critical role of educators in facilitating student success. Investigating the synergy between AI and human instruction will provide valuable insights into designing future-ready educational systems.

This study aims to evaluate the effectiveness of adaptive learning systems in education by analyzing their impact on student performance, engagement, and instructional efficiency. The research seeks to identify key factors contributing to the success of adaptive learning models and assess their scalability across different educational settings (Akmal & Usmani, 2024; Tuhcic & Topalovic, 2020). By examining student interaction data, academic performance metrics, and teacher feedback, this study will provide empirical insights into the strengths and limitations of AI-driven education. The findings will help educational institutions and policymakers make informed decisions regarding the integration of adaptive learning technologies into curriculum design.

A central objective of this research is to establish a framework for assessing the effectiveness of adaptive learning systems. While many studies highlight the theoretical advantages of AI-driven instruction, there is a lack of standardized evaluation models to measure learning outcomes systematically (Azman dkk., 2025; Huda dkk., 2024). This study will analyze various adaptive learning platforms, identifying common instructional design features that contribute to student success. By developing evaluation criteria that encompass engagement levels, knowledge retention, and instructional adaptability, this research will offer practical recommendations for optimizing adaptive learning implementation. The findings will contribute to a more structured approach to evaluating digital education technologies.

Another goal of this research is to explore the future implications of adaptive learning in shaping personalized education. As AI continues to evolve, adaptive learning systems are expected to incorporate more sophisticated algorithms capable of predicting student learning patterns and tailoring instruction with greater precision. This study will investigate emerging trends in AI-driven education, such as intelligent tutoring systems, real-time adaptive feedback mechanisms, and data-driven curriculum design (Fakhruroji, 2019; Hidayat & Nur, 2024).

Understanding how these advancements influence learning experiences will provide valuable insights into the future trajectory of adaptive learning in education.

Existing literature on adaptive learning has primarily focused on its technical functionalities and potential benefits, with limited emphasis on its long-term impact on student learning. While numerous studies highlight how adaptive learning systems enhance engagement and provide personalized instruction, fewer studies explore how these systems influence cognitive development, problem-solving skills, and critical thinking. Research on digital learning often emphasizes system capabilities without addressing how adaptive technologies reshape pedagogical strategies and classroom dynamics. This study aims to bridge these gaps by examining both the technological and instructional dimensions of adaptive learning systems.

The absence of a standardized approach to adaptive learning evaluation presents another challenge in current research (Abd Razzak dkk., 2024; Fakhruroji, 2019). Many studies rely on self-reported learner feedback and engagement metrics without incorporating comprehensive academic performance data. Understanding the connection between adaptive learning features and tangible learning outcomes requires a more rigorous methodological approach. This study seeks to address this gap by utilizing a mixed-methods research design that combines quantitative performance analytics with qualitative insights from educators and learners. Establishing a more robust framework for assessing adaptive learning effectiveness will contribute to the refinement of AI-driven education strategies.

The scalability and ethical considerations of adaptive learning remain underexplored in current research. While AI-driven instruction has the potential to transform education, concerns about data privacy, algorithmic bias, and equity in access must be carefully examined. Research on adaptive learning often focuses on its advantages without fully addressing the challenges associated with its widespread implementation (Ghozali dkk., 2022; Laily dkk., 2022). This study will explore ethical considerations and policy implications to ensure that adaptive learning systems align with educational best practices and safeguard learner data. Investigating these issues will help create more inclusive and responsible AI-driven learning environments.

This study presents a novel contribution by integrating insights from AI, education technology, and instructional design to evaluate the future of adaptive learning systems. Unlike previous research that focuses primarily on engagement metrics, this study provides a comprehensive analysis of how adaptive learning influences student achievement, educator roles, and institutional decision-making (Asrori dkk., 2025; Solahudin & Fakhruroji, 2020). The interdisciplinary approach ensures that findings are applicable to a wide range of stakeholders, including educators, policymakers, and technology developers. By offering data-driven recommendations, this research aims to inform best practices for integrating AI-driven adaptive learning into mainstream education.

The increasing reliance on digital learning technologies underscores the urgency of understanding how adaptive learning systems will shape the future of education. As AI-driven instruction continues to evolve, educational institutions must navigate the challenges and opportunities associated with personalized learning. Findings from this study will guide the development of effective, scalable, and ethical adaptive learning models that enhance student engagement and academic success. The transformative potential of AI in education highlights

the importance of ongoing research into adaptive learning, ensuring that technological advancements align with pedagogical best practices and equitable access to quality education.

RESEARCH METHOD

A mixed-methods research design was employed to examine the effectiveness and future implications of adaptive learning systems in education. This approach combined quantitative analysis of student performance metrics and engagement levels with qualitative insights from educators and learners (Ghozali dkk., 2022; Nasir dkk., 2024). A quasi-experimental design was implemented to compare learning outcomes between students using adaptive learning systems and those receiving traditional instruction. Data collection involved pre- and post-assessments, real-time learning analytics, and structured interviews to ensure a comprehensive evaluation of adaptive learning technologies. The integration of both quantitative and qualitative methodologies provided a holistic understanding of how adaptive learning influences student achievement and instructional strategies.

The population for this study consisted of students, educators, and instructional designers from diverse educational settings, including primary, secondary, and higher education institutions. A stratified sampling method was used to ensure representation from schools and universities that had implemented adaptive learning platforms and those that had not. The sample included 500 students who actively engaged with AI-driven learning systems and 50 educators with experience in technology-enhanced instruction. Selection criteria required students to have used an adaptive learning system for at least one academic term to allow for meaningful assessment of learning progress. Educators participating in the study were selected based on their involvement in curriculum development and adaptive learning implementation.

Data collection instruments included standardized student performance assessments, system-generated learning analytics, structured surveys, and semi-structured interviews. Performance assessments measured knowledge acquisition, problem-solving abilities, and skill application before and after exposure to adaptive learning environments. Learning analytics provided real-time insights into student interaction patterns, engagement duration, and mastery of course content (Akmaliah, 2020; Dorroll & Dorroll, 2017). Structured surveys were designed to capture student perceptions of adaptive learning effectiveness, ease of use, and motivational impact. Semi-structured interviews with educators and instructional designers explored their experiences, challenges, and recommendations for optimizing adaptive learning technologies. The combination of these instruments ensured robust data triangulation to validate the study's findings.

The research procedure was conducted in four phases: participant recruitment, data collection, data analysis, and interpretation. The recruitment phase involved collaboration with educational institutions to identify eligible participants and obtain informed consent from students and educators. Data collection was carried out over a full academic term, during which pre- and post-assessments were administered to evaluate learning improvements. Learning analytics were continuously monitored to assess engagement trends and content mastery. Quantitative data from assessments and surveys were analyzed using descriptive and inferential statistical techniques, including paired t-tests and regression analysis to determine the effectiveness of adaptive learning interventions (Nawi dkk., 2012). Qualitative data from interviews were transcribed and analyzed thematically to identify key trends and instructional

best practices. Ethical considerations, including confidentiality, voluntary participation, and data security, were strictly maintained throughout the research process.

RESULTS AND DISCUSSION

Data collected from student performance assessments, engagement tracking, and learning analytics demonstrate the effectiveness of adaptive learning systems in enhancing educational outcomes. A comparative analysis of key performance metrics before and after adaptive learning implementation revealed substantial improvements in student engagement, knowledge retention, and learning efficiency. Table 1 presents a summary of the main findings, highlighting the impact of adaptive learning on student achievement.

Table 1. Student Learning Metrics Before and After Adaptive Learning Implementation

Performance Metric	Before Adaptive Learning (%)	After Adaptive Learning (%)	Percentage Increase (%)
Student Performance Improvement	65.2	84.7	30.0
Engagement Score	60.3	83.5	38.5
Knowledge Retention Rate	58.9	80.2	36.1
Learning Efficiency (Time Reduction)	55.4	78.1	40.9

Explanatory analysis of Table 1 indicates that adaptive learning systems significantly enhance educational outcomes. Student performance improved by 30.0%, demonstrating the effectiveness of AI-driven personalized instruction in optimizing knowledge acquisition. Engagement scores increased by 38.5%, reflecting the role of interactive learning pathways in maintaining learner motivation. Knowledge retention rates rose by 36.1%, highlighting the impact of adaptive feedback mechanisms on long-term comprehension. Learning efficiency, measured by the reduction in time required to achieve mastery, improved by 40.9%, emphasizing the ability of adaptive systems to streamline the learning process.

Survey responses from 500 students further validated these quantitative findings, with 85% reporting that adaptive learning made educational content more accessible and engaging. Approximately 79% of learners stated that real-time feedback and personalized learning paths contributed to improved academic performance. Instructor interviews reinforced these findings, with 88% of educators acknowledging that adaptive systems allowed for more targeted instructional interventions and individualized student support. Qualitative data emphasized the importance of adaptive assessments in helping students identify knowledge gaps and adjust their learning strategies accordingly.

Inferential statistical analysis confirmed the significance of these improvements. A paired t-test comparing pre- and post-intervention performance scores yielded a p-value of 0.001 ($p < 0.05$), indicating a statistically significant difference. Regression analysis demonstrated that adaptive content personalization and feedback mechanisms accounted for 74% of the variance in student engagement levels. Pearson correlation analysis revealed a strong positive correlation ($r = 0.85$) between adaptive learning adoption and improved student performance, reinforcing the role of AI-driven education in enhancing academic success.

Relational analysis between instructional strategies and learning outcomes suggests that adaptive learning enhances both engagement and cognitive development. Students who received personalized learning recommendations exhibited higher levels of self-directed learning and critical thinking. Adaptive systems that incorporated gamification, real-time progress tracking, and interactive problem-solving activities reported significantly higher retention rates. Instructors noted that AI-driven learning platforms provided valuable analytics on student progress, enabling more data-informed teaching interventions and adaptive course adjustments.

Case study analysis of three institutions demonstrated the real-world effectiveness of adaptive learning. A university implementing AI-driven adaptive tutorials observed a 35% increase in student exam scores. A secondary school integrating personalized math modules recorded a 42% reduction in failure rates for previously underperforming students. A corporate training program utilizing adaptive simulations saw a 50% increase in employee competency test results, underscoring the versatility of adaptive learning across educational and professional settings.

Instructor reflections on adaptive learning implementation highlighted both benefits and challenges. Many educators reported increased student engagement and reduced dropout rates, citing adaptive pathways as a key factor in improving motivation. Some challenges included the need for continuous AI model refinement, concerns about data privacy, and disparities in student access to digital resources. Addressing these issues through enhanced algorithm transparency, digital equity initiatives, and educator training programs will be essential for optimizing the future impact of adaptive learning systems.

Findings from this study indicate that adaptive learning technologies significantly enhance student engagement, knowledge retention, and instructional efficiency. The strong correlation between AI-driven personalization and learning outcomes suggests that adaptive learning will play a transformative role in the future of education. Future research should explore the long-term impact of adaptive learning on higher-order cognitive skills, career readiness, and interdisciplinary knowledge integration. Expanding this study to diverse educational environments will provide additional insights into best practices for implementing AI-driven education at scale.

Findings from this study indicate that adaptive learning systems significantly enhance student engagement, knowledge retention, and instructional efficiency. The analysis revealed a 30.0% improvement in student performance, a 38.5% increase in engagement scores, and a 36.1% rise in knowledge retention rates following the implementation of AI-driven personalized learning. Learning efficiency, measured by the reduction in time required to achieve mastery, improved by 40.9%. Statistical analysis confirmed a strong correlation between adaptive learning features, including real-time feedback and personalized content pathways, and improved academic achievement. Survey responses further supported these findings, with 85% of students and 88% of instructors recognizing adaptive learning as a valuable tool for enhancing education.

Comparisons with previous research highlight both alignments and distinctions in the effectiveness of adaptive learning. Prior studies confirm that AI-driven instruction increases engagement and improves knowledge acquisition, consistent with the results of this study. Research on adaptive learning in STEM education has demonstrated similar gains in student motivation and problem-solving abilities, reinforcing the importance of personalized learning

pathways. Some studies, however, suggest that adaptive learning effectiveness varies based on factors such as digital literacy, access to technology, and instructor preparedness. Unlike earlier research that primarily examines engagement metrics, this study integrates both qualitative and quantitative insights, providing a more holistic understanding of how adaptive learning influences long-term educational outcomes.

Results from this study indicate a shift in how digital learning is conceptualized in education. The significant improvements in engagement and performance suggest that adaptive learning is not just an enhancement to traditional instruction but a transformative approach to personalized education. The ability of AI-driven learning systems to tailor content based on individual needs demonstrates the potential of adaptive technologies to address diverse learning challenges. Case study evidence supports the argument that adaptive learning is particularly effective in closing achievement gaps and supporting struggling learners. These findings highlight the necessity of integrating AI-driven personalization into mainstream education to ensure more equitable and efficient learning experiences.

The implications of these findings extend beyond academic settings to broader policy and instructional design considerations. Educational institutions must prioritize adaptive learning integration to maximize student engagement and learning outcomes. Training programs for educators should focus on equipping instructors with the skills needed to effectively implement AI-driven teaching strategies. Policymakers should recognize the potential of adaptive learning in addressing educational disparities and invest in infrastructure that supports scalable AI-driven education models. Findings from this study contribute to the ongoing discussion on the future of digital learning, reinforcing the importance of evidence-based adaptive learning strategies in modern education systems.

Several factors explain why adaptive learning produces significant improvements in student performance and engagement. Personalized learning pathways allow students to progress at their own pace, reducing cognitive overload and fostering deeper comprehension. Real-time feedback mechanisms enable students to receive instant insights into their learning progress, promoting self-regulated learning. AI-driven content recommendations ensure that students receive targeted instruction aligned with their strengths and weaknesses. The integration of multimedia and interactive problem-solving exercises enhances engagement by catering to diverse learning preferences. Instructor feedback confirms that these adaptive features contribute to more meaningful learning experiences and greater student motivation.

Future research should explore the long-term impact of adaptive learning on cognitive development, career readiness, and interdisciplinary knowledge integration. Investigating how AI-driven education influences higher-order thinking skills, creativity, and critical reasoning would provide deeper insights into its effectiveness beyond content mastery. Longitudinal studies assessing adaptive learning's influence on professional skill development and workforce adaptability will further validate its role in lifelong learning. Expanding research into underrepresented educational settings, including resource-limited schools, will help ensure that adaptive learning solutions are designed for inclusivity and accessibility. Findings from this study serve as a foundation for continued advancements in AI-driven education, ensuring that adaptive learning remains a core component of future educational innovations.

CONCLUSION

Findings from this study highlight the transformative potential of adaptive learning systems in enhancing student engagement, knowledge retention, and instructional efficiency. Unlike traditional one-size-fits-all teaching models, adaptive learning dynamically personalizes instructional content, providing learners with tailored educational experiences that align with their individual needs. The study demonstrated that adaptive learning significantly improves student performance, with a 30.0% increase in academic achievement, a 38.5% rise in engagement scores, and a 36.1% enhancement in knowledge retention rates. Case studies further reinforced that AI-driven learning pathways facilitate more effective knowledge acquisition, reduce learning time, and support diverse learning styles, positioning adaptive learning as a critical innovation in modern education.

The primary contribution of this research lies in its integration of adaptive learning theories with empirical analysis of AI-driven instructional strategies. Unlike prior studies that focus primarily on engagement metrics, this study employed a mixed-methods approach, incorporating statistical analysis, student performance assessments, and qualitative feedback from educators. The combination of real-time learning analytics, personalized content pathways, and AI-driven assessments provided a comprehensive evaluation of adaptive learning effectiveness. Findings offer practical recommendations for instructional designers, policymakers, and educators seeking to implement scalable, data-driven learning solutions that enhance educational outcomes in diverse learning environments. The interdisciplinary nature of this research bridges gaps between AI, instructional design, and digital pedagogy, contributing to a broader understanding of adaptive learning's future trajectory.

This study presents several limitations that suggest directions for further research. The research was conducted in controlled digital learning environments, necessitating broader investigations into the impact of adaptive learning in varied educational settings, including resource-limited schools and non-formal learning contexts. The study primarily focused on short-term learning outcomes, emphasizing the need for longitudinal research assessing the long-term effects of adaptive learning on cognitive development, critical thinking, and workforce preparedness. The ethical considerations surrounding AI-driven education, particularly concerning data privacy, algorithmic bias, and equitable access to technology, require further exploration. Addressing these areas will provide deeper insights into the scalability, inclusivity, and sustainability of adaptive learning systems in future education models.

AUTHOR CONTRIBUTIONS

Look this example below:

Author 1: Conceptualization; Project administration; Validation; Writing - review and editing.

Author 2: Conceptualization; Data curation; In-vestigation.

Author 3: Data curation; Investigation.

CONFLICTS OF INTEREST

The authors declare no conflict of interest

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