




A Critical Glance at Adaptive Learning Systems Using Artificial Intelligence: A Systematic Review and Qualitative Synthesis of Contemporary Research Literature

Sayfa | 3519

Yapay Zeka Kullanılan Uyarlamalı Öğrenme Sistemlerine Eleştirel Bir Bakış: Çağdaş Araştırma Literatürünün Sistematik Olarak İncelenmesi ve Nitel Sentezi

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Geliş tarihi - Received: 31 July 2024
Kabul tarihi - Accepted: 28 December 2024
Yayın tarihi - Published: 28 December 2024



Abstract. This study aims to critically examine the current research on AI-powered adaptive learning systems by synthesizing studies to identify trends, gaps, and challenges. It also explores the applications, benefits, challenges, and future directions of these systems in education. The research design employs a systematic review and qualitative thematic synthesis following PRISMA guidelines. Data collection involves a comprehensive literature search across prominent databases such as ERIC, JSTOR, IEEE Xplore, Google Scholar, Web of Science, and Scopus. Inclusion criteria focus on peer-reviewed articles and high-quality grey literature from the past ten years. Data analysis includes coding and thematic mapping to integrate findings into a comprehensive narrative, ensuring rigor through triangulation, peer debriefing, and reflexivity. The findings reveal significant themes related to the role of AI in adaptive learning, including machine learning algorithms, natural language processing, and AI-driven data analysis. Applications of adaptive learning systems are demonstrated in personalized learning pathways, adaptive assessments, intelligent tutoring systems, and support for diverse learning needs. Case studies highlight the effectiveness of these systems in enhancing student engagement and learning outcomes. This study provides a comprehensive overview of the potential of AI-powered adaptive learning systems to transform education. It identifies significant benefits such as improved learning outcomes, increased engagement, scalability, and cost-effectiveness. The study also addresses challenges like data quality, ethical considerations, and institutional resistance, providing a balanced view of the current landscape. AI-powered adaptive learning systems have innovative potential in personalizing and improving educational experiences. While the benefits are significant, addressing challenges related to data quality, ethical considerations, and educator support is crucial. Future research should focus on long-term impacts, ethical implications, and integrating emotional and social learning to create a holistic educational environment.

Keywords: Adaptive learning systems, Artificial intelligence, Educational technology, Systematic review, Qualitative synthesis.

Öz. Bu çalışma, yapay zeka (AI) destekli uyarlamalı (adaptive) öğrenme sistemleri hakkındaki güncel araştırmaları eleştirel bir şekilde inceleyerek, eğilimler, boşluklar ve zorlukları belirlemek amacıyla sentez yapmayı amaçlamaktadır. Aynı zamanda bu sistemlerin eğitimdeki uygulamalarını, faydalarını, zorluklarını ve gelecekteki yönelimleri de sunulmaktadır. Araştırma tasarımı, PRISMA yönergelerini izleyerek sistematik bir inceleme ve nitel tematik sentez yöntemini kullanmıştır. Veri toplama süreci, ERIC, JSTOR, IEEE Xplore, Google Scholar, Web of Science ve Scopus gibi saygın veri tabanlarında kapsamlı bir literatür taramasını içermektedir. Dahil edilme kriterleri, son on yıldaki hakemli makaleler ve yüksek kaliteli gri literatüre odaklanmıştır. Veri analizi, bulguları kapsamlı bir anlatıya entegre etmek için kodlama ve tematik haritalama yöntemini kullanmıştır. Bulgular, uyarlamalı öğrenme sistemlerinde AI'nin rolüyle ilgili makine öğrenimi (machine learning), doğal dil işleme (natural language processing) ve AI destekli veri analizi gibi önemli temaları ortaya çıkarmıştır. Uyarlamalı öğrenme sistemlerinin uygulamaları, kişiselleştirilmiş öğrenme, uyarlamalı ölçme-değerlendirme süreçleri, akıllı öğretim sistemleri ve özel öğrenme ihtiyaçlarının desteklenmesi gibi alanlarda görülmektedir. Vaka çalışmaları, bu sistemlerin öğrenci katılımını ve öğrenme sonuçlarını artırmadaki etkinliğini vurgulamaktadır. Bu çalışma, AI destekli uyarlamalı öğrenme sistemlerinin eğitimi geliştirmeye yönelik potansiyeline dair kapsamlı bir genel bakış sunmaktadır. Öğrenme sonuçlarının iyileştirilmesi, artan katılım, ölçeklenebilirlik ve maliyet etkinliği gibi önemli faydalar göze

Meylani, R. (2024). A critical glance at adaptive learning systems using artificial intelligence: A systematic review and qualitative synthesis of contemporary research literature. *Western Anatolia Journal of Educational Sciences*, 15(3), 3519-3547.

DOI. 10.51460/baebd.1525452

Batı Anadolu Eğitim Bilimleri Dergisi, (2024), 15 (3), 3519-3547.
Western Anatolia Journal of Educational Sciences, (2024), 15 (3), 3519-3547.
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çarpılmaktadır. Bu çalışma ayrıca veri kalitesi, etik konular ve kurumsal direnç gibi zorluklara da değinerek mevcut duruma dengeli bir bakış açısı sunmaktadır. AI destekli uyarlamalı öğrenme sistemleri, eğitim deneyimlerini kişiselleştirme ve iyileştirme konusunda yenilikçi bir potansiyele sahiptir. Faydalar önemli olsa da, veri kalitesi, etik konular ve eğitmen desteği ile ilgili zorlukların ele alınması kritik öneme sahiptir. Gelecekteki araştırmalar, uzun vadeli etkileri ve etik sonuçları ele almalı, duygusal ve sosyal öğrenmenin entegrasyonuna odaklanarak bütüncül bir eğitim ortamı oluşturmayı hedeflemelidir.

Anahtar Kelimeler: Uyarlamalı öğrenme sistemleri, Yapay zeka, Eğitim teknolojisi, Sistemik inceleme, Nitel sentez.



Genişletilmiş Türkçe Özet

Giriş. Yapay zeka (AI) destekli uyarlamalı öğrenme sistemleri (ALS), eğitimde bireyselleştirilmiş yaklaşımlar sunarak öğrenci performansını, katılımını ve öğrenme sonuçlarını iyileştirme potansiyeline sahiptir (Zhao, 2023). Bu sistemler, öğrenci ile ilgili verileri analiz ederek bireysel ihtiyaçlara göre öğrenme yöntemlerini dinamik biçimde uyarlamaktadır (Choi & McClenen, 2020). AI, makine öğrenim (ML) algoritmaları, doğal dil işleme (NLP) ve yapay sinir ağlarını kullanarak öğrenci performans desenlerini tanımakta ve gerçek zamanlı ayarlamalar yapabilmektedir (Wittal et al., 2022). Bu çalışma, AI destekli ALS üzerine mevcut araştırmaları eleştirel bir şekilde incelemeyi ve bu alandaki eğilimleri, boşlukları ve zorlukları ortaya çıkarmayı amaçlamaktadır (Khan et al., 2022).

Yöntem. Bu araştırma, sistematik bir inceleme ve nitel tematik sentez yöntemi kullanılarak yapılmıştır. PRISMA (Preferred Reporting Items for Systematic Reviews and Meta-Analyses) yönergeleri, çalışmanın şeffaflığı ve tekrarlanabilirliği için dikkatle ve özenle izlenmiştir. Veri toplama süreci, ERIC, JSTOR, IEEE Xplore, Google Scholar, Web of Science ve Scopus gibi saygın akademik veri tabanlarında kapsamlı bir literatür taraması yapılarak gerçekleştirilmiştir. Arama terimleri olarak "uyarlamalı öğrenme sistemleri", "yapay zeka", "kişiselleştirilmiş öğrenme", "makine öğrenimi" ve ilgili anahtar kelimeler kullanılmıştır. Dahil edilme kriterleri, son on yıldaki hakemli makalelere ve yüksek kaliteli gri literatüre odaklanmıştır. AI bileşeni olmayan veya ampirik veri içermeyen çalışmalar hariç tutulmuştur. İlk olarak başlıklar ve özetler taranmış, ardından tam metin incelemeleri yapılmıştır. Metodoloji, Critical Appraisal Skills Programme (CASP) kontrol listesi kullanılarak değerlendirilmiştir.

Veri analizi, standart bir form kullanılarak veri çıkarımına dayanmaktadır. Bu form ile çalışmanın ayrıntıları, araştırmanın hedefleri, metodolojiler, bulgular, çıkarımlar ve zorluklar toparlanmıştır. İlk kodlama seti, araştırma soruları ve literatür incelemesi temel alınarak geliştirilmiş, güvenilirlik kodlamanın iki farklı zamanda yapılmasıyla sağlanmıştır. Ana temalar belirlendikten sonra, sürekli karşılaştırma yoluyla rafine edilmiş ve tematik haritalar kullanılarak görselleştirilmiştir. Sentez bölümünde, bireysel çalışma bulguları kapsamlı bir anlatıya entegre edilmiş, teorik çerçeve içinde yorumlanmış, uygulama, politikalar ve gelecekteki araştırmalar için önerilerle birlikte sunulmuştur.

Bulgular. Yapılan sistematik inceleme ve nitel sentez, AI'nın uyarlamalı öğrenme üzerindeki rolü, AI destekli veri analizi ve karar verme süreçleri ile ilgili önemli temalar ve alt temaları ortaya koymuştur. Temel bulgular, ML algoritmalarının ve NLP'nin kişiselleştirilmiş ve uyarlanabilir eğitim deneyimlerini sağlamak için nasıl kullanıldığını ve kullanılabilirliğini göstermektedir. AI destekli ALS'lerin öğrenci performansını artırma, öğrenci katılımını teşvik etme ve öğrenme sonuçlarını iyileştirme potansiyelini vurgulayan vaka çalışmaları da bu bulguları desteklemektedir.

ML algoritmaları, uyarlamalı öğrenmede temel bir rol oynamaktadır. Bu algoritmalar, verileri analiz ederek desenleri tanımlar ve öğrenci ihtiyaçlarına göre eğitim içeriğini uyarlamak için veri odaklı kararlar alır. Bu süreç, öğrencilerin ihtiyaçlarına, tercihlerine ve performans düzeylerine dinamik olarak uyum sağlar, bu sayede öğrenme etkinliğini iyileştirir (Zhao, 2023). NLP, öğrenciler ve öğrenme platformları arasında etkileşimi mümkün kılar. NLP teknolojisi, insan dilini yorumlayarak ve kullanarak gerçek zamanlı geri bildirim sağlamak, soruları yanıtlamakta ve içeriği etkileşimli bir



şekilde sunmaktadır. Bu da öğrencilerin katılımını ve anlama düzeyini artırmaktadır (Tapalova & Zhiyenbayeva, 2022).

AI destekli kişiselleştirme, öğrenme sistemlerinin her bir öğrencinin öğrenme stiline, hızına ve yeterlilik seviyesine göre içerik, hız ve değerlendirmeler açısından uyarlanmasına olanak tanır. AI destekli veri analizi, büyük miktarda veriyi toplar, analiz eder ve yorumlar. Veri analiz süreçleri, öğretim stratejilerini yönlendirir, gerçek zamanlı ayarlamalar yapar ve kişiselleştirilmiş öğrenme yöntemlerini optimize eder, bu sayede öğrenme daha etkin seviyelere yükselir (Tapalova & Zhiyenbayeva, 2022).

Uyarlamalı öğrenme sistemleri, eğitimde devrim yapmıştır. Öğrenciye özel öğrenme yolları, öğrencinin hızına ve ihtiyaçlarına göre uyarlanarak kişiselleştirilmiş öğrenme deneyimleri sunulmaktadır. Bu sistemler, AI algoritmalarını kullanılarak değerlendirmeleri de özelleştirmekte ve sürekli destek sağlamak için hedef odaklı geri bildirimler sunmaktadır. Kişiselleştirilmiş öğrenme yöntemleri, bireysel öğrenme ihtiyaçlarını karşılamak için içerik ve hızı özelleştirmekte ve eğitimdeki verimliliği artırmaktadır (Vanbecelaere et al., 2020; Rugaiyah, 2023; Vadivel et al., 2023; Rukadikar & Khandelwal, 2023).

ALS, öğrencilerin kendi hızlarında ve öğrenme tercihlerine göre içerikte ilerlemelerine olanak tanımaktadır. Araştırmalar, özelleştirilmiş yöntemler aracılığıyla öğrenme verimliliğinde önemli artışlar meydana geldiğini göstermiştir (Ranga, 2017; Raj & Renumol, 2022). Bu sistemler, performans verilerini analiz etmek ve içeriği gerçek zamanlı olarak ayarlamak için AI kullanır. Uyarlamalı değerlendirmeler ve geri bildirimler, sürekli değerlendirme ve kişiselleştirilmiş destek sağlayarak öğrenme sonuçlarını sürekli bir biçimde iyileştirmektedir (Rohm et al., 2021; Barianos et al., 2022).

Akıllı öğretim sistemleri (ITS), AI kullanarak bireysel kullanıcıların aldığı eğitim ve desteği uyarlayarak yanıtları değerlendirmekte, ipuçları sağlamakta ve öğrencilerle diyaloga girerek öğrenmeyi teşvik etmektedir (Jakobsche et al., 2023; Huang et al., 2023). AI destekli öğretmenler ve sanal asistanlar, AI algoritmalarını kullanarak özelleştirilmiş içerik, gerçek zamanlı geri bildirim ve etkileşimler sunmaktadır (He, 2023; Del Olmo-Muñoz et al., 2023). ALS, engelli öğrenciler için hedef odaklı destek sistemleri ile değişik ve çeşitli ihtiyaçlara uygun kapsayıcı eğitim ortamları da meydana getirmektedir (López-Gavira et al., 2021; Wilke et al., 2023).

Uyarlamalı öğrenmede oyunsallaştırma unsurlarının eklenmesi, katılımı ve motivasyonu artıran bir bakışa etmektedir. Puanlar, seviyeler ve ödüller gibi özellikler, etkileşimli, oyun benzeri öğrenme deneyimleri sunabilmektedir (Hallifax et al., 2020). Simülasyonlar, sanal gerçeklik ve çoklu ortam içerikleri gibi etkileşimli unsurların entegrasyonu, bilgiyi kavrama ve hatırlama düzeylerini artıran uygulamalı öğrenme fırsatları yaratmaktadır (Ionică et al., 2022). Gerçek dünya uygulamaları ve güncel araştırmalar, ITS ve oyunsallaştırmanın öğrenci katılımını ve öğrenme sonuçlarını iyileştirmedeki başarısını vurgulamaktadır. Karşılaştırmalı çalışmalar, uyarlamalı öğrenme ortamlarının geleneksel yöntemlere göre üstünlüğünü göstermektedir.

Tartışma ve Sonuç. AI destekli ALS'ler, eğitim deneyimlerini kişiselleştirerek ve iyileştirerek eğitimde dönüşüm ve gelişim potansiyeline sahiptir. Ancak, veri kalitesi, etik hususlar ve öğretmen desteği ile ilgili zorlukların ele alınması önemlidir. Bu sistemler, öğrenci performansını artırmada ve öğrenme süreçlerini daha verimli hale getirmede önemli bir rol oynar. Gelecekteki araştırmalar, etik ve uzun vadeli etkilerle birlikte, duygusal ve sosyal öğrenme kavramlarının bu sistemlere entegrasyonu üzerine odaklanmalıdır. Sonuç olarak bu çalışma, AI destekli ALS'lerin eğitimdeki potansiyelini ve

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mevcut durumunu dengeli bir şekilde ortaya koymakta, eğitim uygulamaları ve politikaları için değerli öneriler sunmaktadır.



Introduction

Background and significance

Adaptive Learning Systems (ALS) are educational platforms utilizing Artificial Intelligence (AI) to customize learning experiences based on individual student needs, preferences, and abilities (Zhao, 2023). These systems are significant for their ability to personalize learning paths, enhance student engagement, and improve learning outcomes (Tapalova & Zhiyenbayeva, 2022). By adjusting content delivery, pacing, and assessments, ALS provide a more tailored and effective educational approach (Choi & McClenen, 2020).

AI is crucial to ALS, enabling the analysis of extensive data to identify student performance patterns and make real-time adjustments (Wittal et al., 2022). Machine Learning (ML) algorithms, Natural Language Processing (NLP), and neural networks enhance these systems' adaptability and personalization (Zhou et al., 2023). AI and data analytics empower ALS to deliver customized learning experiences tailored to individual learning styles and paces (Zhao, 2023).

Purpose and scope of the review

This review aims to critically examine the current research on AI-powered ALS, synthesizing studies to identify trends, gaps, and challenges in the field (Khan et al., 2022). It explores the applications, benefits, challenges, and future directions of these systems, providing a comprehensive overview of their potential to transform education (Dai, 2021). Additionally, the review addresses the implications of AI in education, the effectiveness of adaptive learning strategies, and the role of technology in shaping future learning (Khan et al., 2022).

Theoretical Underpinnings

Definition and key features

ALS leverage AI to personalize educational experiences based on individual learner needs, preferences, and abilities (Zhang et al., 2023). They adapt content delivery, pacing, and assessments to each student's unique learning style, knowledge level, and cognitive abilities (Zhang et al., 2023).

Historical development and evolution

ALS' development began with computer-assisted instruction in the 1950s and 1960s, focusing on tailoring educational content to individual learners (Shi et al., 2021). Advances in AI and machine learning have since revolutionized these systems, enhancing their capability to provide personalized and effective learning experiences (Shi et al., 2021). This evolution reflects a shift toward student-centered and adaptive educational approaches, moving away from traditional one-size-fits-all methods (Shi et al., 2021).

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Types of ALS

ALS vary to meet different educational contexts and objectives, including Intelligent Tutoring Systems (ITS), personalized learning platforms, adaptive assessment tools, and gamified learning environments (Rastogi & Sikhwal, 2022). Each type utilizes distinct algorithms, data analytics techniques, and instructional strategies to optimize learning outcomes (Rastogi & Sikhwal, 2022). Understanding these distinctions is crucial for educators and researchers implementing adaptive learning approaches.

Theoretical foundations and pedagogical principles

The foundations of ALS lie in educational psychology, cognitive science, and learning theory, drawing on personalized learning, constructivism, behaviorism, and metacognition (Avraham et al., 2022). Integrating these theoretical frameworks ensures effective and engaging learning experiences aligned with students' cognitive processes, motivations, and goals (Avraham et al., 2022). Key pedagogical principles, including differentiation, scaffolding, and formative assessment, guide the design and implementation of ALS to support diverse learner needs (Avraham et al., 2022).

Methodology

Research design

A systematic review and qualitative thematic synthesis were employed to analyze existing research on AI-driven ALS, following Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA) guidelines for transparency and replicability. This approach allowed for an in-depth examination of how these systems function and impact educational outcomes.

Data collection

The literature search spanned multiple academic databases, including ERIC, JSTOR, IEEE Xplore, Google Scholar, Web of Science, and Scopus, using keywords like "adaptive learning systems," "artificial intelligence," and "personalized learning." Inclusion criteria focused on peer-reviewed articles and high-quality grey literature from the past ten years, while studies without an AI component or lacking empirical data were excluded. Titles and abstracts were initially screened, followed by full-text reviews to ensure relevance and methodological quality, assessed using the Critical Appraisal Skills Programme (CASP) checklist.

Data analysis

Data were extracted using a standardized form, capturing study details, research objectives, methodologies, findings, implications, and challenges. An initial coding set was developed based on research questions and literature review, with reliability ensured by coding at two different times. Themes were identified, refined through constant comparison, and visualized using thematic maps.

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The synthesis integrated individual study findings into a coherent narrative, interpreted within theoretical frameworks, and discussed implications for practice, policy, and future research.

Ensuring rigor and ethical issues

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Trustworthiness was ensured through data source triangulation, peer debriefing, detailed context descriptions, an audit trail, and reflexivity to mitigate researcher bias. No ethical approval was required as the study involved secondary data analysis, but ethical guidelines for qualitative research were adhered to throughout the study.

Findings

The systematic review and qualitative synthesis revealed the themes and subthemes depicted in Figure 1. Table 1 also lists the references consulted for each theme and sub-theme.

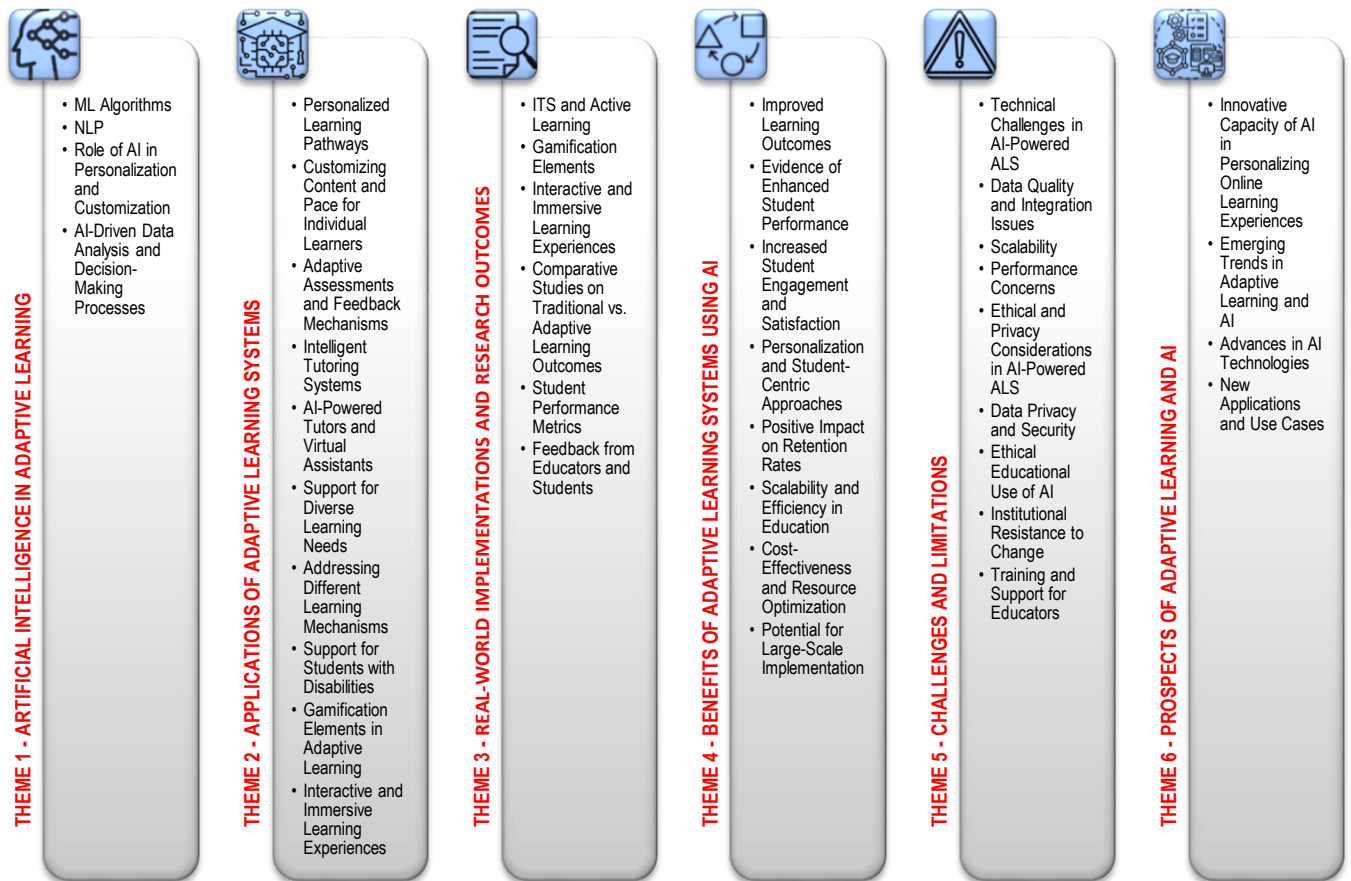


Figure 1. The themes and sub-themes emerging from the qualitative synthesis of the systematic review.



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Table 1.

The themes, sub-themes, number of references and actual references consulted in the study.

| Theme | Sub-theme | Number of References Consulted | Actual References Consulted |
|--|---|--------------------------------|--|
| AI in Adaptive Learning | Machine Learning Algorithms | 1 | Zhao (2023) |
| | NLP | 1 | Tapalova and Zhiyenbayeva (2022) |
| | Role of AI in Personalization and Customization | 1 | Tapalova and Zhiyenbayeva (2022) |
| | AI-Driven Data Analysis and Decision-Making Processes | 1 | Tapalova and Zhiyenbayeva (2022) |
| Applications of ALS | Personalized Learning Pathways | 4 | Vanbecelaere et al. (2020); Rugaiyah (2023); Vadivel et al. (2023); Rukadikar and Khandelwal (2023) |
| | Customizing Content and Pace for Individual Learners | 2 | Ranga (2017); Raj and Renumol (2022) |
| | Adaptive Assessments and Feedback Mechanisms | 5 | Rohm et al. (2021); Barianos et al. (2022); Ntim et al. (2021); Singh et al. (2024); Liu et al. (2022) |
| | ITS | 2 | Jakobsche et al. (2023); Huang et al. (2023) |
| | AI-Powered Tutors and Virtual Assistants | 2 | He (2023); Del Olmo-Muñoz et al. (2023) |
| | Support for Diverse Learning Needs | 1 | López-Gavira et al. (2021) |
| | Addressing Different Learning Mechanisms | 2 | Wilke et al. (2023); Horner and Sugai (2015) |
| | Support for Students with Disabilities | 4 | Yıldız et al. (2022); Bai (2022); Li (2022); Barrientos (2023) |
| | Gamification Elements in Adaptive Learning | 4 | Hallifax et al. (2020); Suresh Babu and Dhakshina Moorthy (2024); Bennani et al. (2022); Bigdeli et al. (2023) |
| | Interactive and Immersive Learning Experiences | 2 | Ionică et al. (2022); Şenocak et al. (2021) |
| Real-World Implementations and Research Outcomes | ITS and Active Learning | 2 | Sharma and Harkishan (2022); Zhao (2023) |
| | Gamification Elements | 2 | Lavoué et al. (2019); Khakpour and Colomo-Palacios (2021) |
| | Interactive and Immersive Learning Experiences | 3 | Linden et al. (2019a); Linden et al. (2019b); Chou et al. (2021) |
| | Comparative Studies on Traditional vs. Adaptive Learning Outcomes | 4 | Ling and Chiang (2022); Nathaniel et al. (2021); Zhao (2023); How and Hung (2019) |
| | Student Performance Metrics | 2 | Palanisamy et al. (2021); Kayed et al. (2022) |
| Benefits of ALS Using AI | Feedback from Educators and Students | 1 | Chen et al. (2020) |
| | Improved Learning Outcomes | 1 | Krouska et al. (2021) |
| | Evidence of Enhanced Student Performance | 2 | Kwak and Mondisa (2021); Linden et al. (2019a); Linden et al. (2019b) |

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| | | | |
|---------------------------------------|--|---|---|
| | Increased Student Engagement and Satisfaction | 2 | Diep et al. (2023); Gupta et al. (2020) |
| | Personalization and Student-Centric Approaches | 1 | Tomkin et al. (2019) |
| | Positive Impact on Retention Rates | 2 | Nikou et al. (2023); Troussas et al. (2021) |
| | Scalability and Efficiency in Education | 1 | Shekahwat (2023) |
| | Cost-Effectiveness and Resource Optimization | 1 | Shekahwat (2023) |
| | Potential for Large-Scale Implementation | 1 | Shekahwat (2023) |
| Challenges and Limitations | Technical Challenges in AI-Powered ALS | 3 | Shi et al. (2015); Issa et al. (2018); Longa et al. (2022) |
| | Data Quality and Integration Issues | 1 | Shi et al. (2015) |
| | Scalability | 1 | Issa et al. (2018) |
| | Performance Concerns | 1 | Longa et al. (2022) |
| | Ethical and Privacy Considerations in AI-Powered ALS | 1 | Akgun and Greenhow (2022) |
| | Data Privacy and Security | 1 | Yu and Yu (2023) |
| | Ethical Educational Use of AI | 1 | Holmes et al. (2022) |
| | Institutional Resistance to Change | 1 | O'Rourke et al. (2022) |
| | Training and Support for Educators | 2 | Amod and Mkhize (2023); Hallelwell et al. (2020) |
| Prospects of Adaptive Learning and AI | Innovative Capacity of AI in Personalizing Online Learning Experiences | 3 | Willis (2024); Gligorea et al. (2023); Kayed et al. (2022) |
| | Emerging Trends in Adaptive Learning and AI | 4 | Aggarwal et al. (2023); Onesi-Ozigagun et al. (2024); Akavova et al. (2023); Akintayo et al. (2024) |
| | Advances in AI Technologies | 3 | Gligorea et al. (2023); Dabingaya (2022); Luo (2023) |
| | New Applications and Use Cases | 3 | Chisom et al. (2024); Li et al. (2024); Alashwal (2024) |

Results

Theme 1 - AI in adaptive learning

AI in adaptive learning utilizes ML algorithms to analyze data, identify patterns, and tailor educational content to meet individual student needs. NLP enhances engagement and comprehension through real-time feedback and interactive interactions. AI-driven personalization and customization, along with advanced data analysis, enable highly individualized educational experiences:

- **ML algorithms:** ML algorithms are fundamental in AI for adaptive learning, enabling systems to analyze data, identify patterns, and make data-driven decisions to tailor educational

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DOI. 10.51460/baebd.1525452



content and experiences. These algorithms dynamically adjust to students' needs, preferences, and performance levels, enhancing learning effectiveness (Zhao, 2023).

- **NLP:** NLP facilitates communication between students and learning platforms. It enables systems to interpret and generate human language, providing real-time feedback, answering queries, and delivering content interactively, enhancing student engagement and comprehension (Tapalova & Zhiyenbayeva, 2022).
- **Role of AI in personalization and customization:** AI-driven personalization allows systems to adapt content, pacing, and assessments to match each student's learning style, pace, and proficiency level. This optimization improves learning outcomes, increases engagement, and fosters a more effective learning environment (Tapalova & Zhiyenbayeva, 2022).
- **AI-driven data analysis and decision-making processes:** AI-driven data analysis collects, analyzes, and interprets vast amounts of data to inform instructional strategies and interventions. These processes enable real-time adjustments, personalized learning pathways, and optimized content delivery to maximize learning outcomes (Tapalova & Zhiyenbayeva, 2022).

AI in adaptive learning harnesses ML algorithms to analyze educational data, discern patterns, and personalize content to fit individual student needs. By using natural language processing (NLP), these systems facilitate real-time feedback and interactive engagement, enhancing student comprehension and participation. The role of AI extends to personalizing and customizing educational experiences, optimizing content delivery based on each learner's unique preferences and proficiency levels. Additionally, AI-driven data analysis enables continuous monitoring and decision-making, providing tailored learning pathways and improving overall educational outcomes.

Theme 2 - Applications of ALS

ALS revolutionize education by offering personalized learning pathways that adjust content and pace to each learner's needs. They utilize AI algorithms to customize assessments and provide targeted feedback, ensuring continuous support for student progress:

- **Personalized learning pathways:** These systems provide personalized learning pathways, customizing content and pace for individual learners. They enhance linguistic competence and cater to individual learning needs, showcasing AI's potential in improving proficiency among young learners and in employee education and training (Vanbecelaere et al., 2020; Rugaiyah, 2023; Vadivel et al., 2023; Rukadikar & Khandelwal, 2023).
- **Customizing content and pace for individual learners:** ALS allow students to advance through content at their own pace and according to their learning styles and preferences. Research shows significant boosts in engagement and outcomes through tailored pathways (Ranga, 2017; Raj & Renumol, 2022).
- **Adaptive assessments and feedback mechanisms:** These systems utilize AI to analyze performance data and adjust content delivery in real-time. Adaptive assessments and feedback ensure continuous evaluation and personalized support, leading to improved learning outcomes (Rohm et al., 2021; Barianos et al., 2022; Ntim et al., 2021; Singh et al., 2024; Liu et al., 2022).



- **ITS:** ITS tailor instruction and support to individual users using AI, assessing responses, providing hints, and engaging in dialogue to enhance learning (Jakobsche et al., 2023; Huang et al., 2023).
- **AI-powered tutors and virtual assistants:** These systems leverage AI algorithms to deliver customized content, real-time feedback, and engaging interactions, enhancing student engagement and optimizing outcomes (He, 2023; Del Olmo-Muñoz et al., 2023).
- **Support for diverse learning needs:** ALS provide targeted support for students with disabilities, creating inclusive educational environments that cater to diverse needs (López-Gavira et al., 2021; Wilke et al., 2023; Horner & Sugai, 2015; Yıldız et al., 2022; Bai, 2022; Li, 2022; Barrientos, 2023).
- **Gamification elements in adaptive learning:** Incorporating gamification elements enhances engagement and motivation. Features like points, levels, and rewards create interactive, game-like learning experiences (Hallifax et al., 2020; Suresh Babu & Dhakshina Moorthy, 2024; Bennani et al., 2022; Bigdeli et al., 2023).
- **Interactive and immersive learning experiences:** Integrating interactive elements such as simulations, virtual reality, and multimedia content offers hands-on learning opportunities that enhance comprehension and retention (Ionică et al., 2022; Şenocak et al., 2021).

ALS revolutionize education by creating personalized learning experiences that adjust content and pace according to individual student needs. These systems use AI algorithms to customize assessments and offer targeted feedback, ensuring consistent support for student progress. Personalized learning pathways cater to diverse learning styles, enhancing engagement and outcomes. Adaptive assessments and feedback mechanisms continuously evaluate performance, while AI-powered tutors and virtual assistants provide real-time, interactive support. ALS also accommodate diverse learning needs, incorporating gamification and immersive elements to create engaging, hands-on learning experiences.

Theme 3 - Real-world implementations and research outcomes

Real-world implementations and cutting-edge research highlight the success of ITS and gamification in improving student engagement and learning outcomes. Comparative studies showcase the superiority of ALS over traditional methods, with positive feedback from educators and students:

- **ITS and active learning:** Successful implementations of ITS, like HINTS, have supported word problem-solving skills and promoted active learning methodologies, improving engagement and comprehension (Sharma & Harkishan, 2022; Zhao, 2023).
- **Gamification elements:** Gamified learning activities increase participation, knowledge retention, and performance by creating interactive learning environments (Lavoué et al., 2019; Khakpour & Colomo-Palacios, 2021).
- **Interactive and immersive learning experiences:** Interactive elements in ALS, such as simulations and virtual reality, provide experiential learning opportunities, enhancing engagement and outcomes (Linden et al., 2019a; Linden et al., 2019b; Chou et al., 2021).



- **Comparative studies on traditional vs. adaptive learning outcomes:** Research shows that personalized and adaptive approaches improve student performance, engagement, and satisfaction. Studies demonstrate the effectiveness of AI in personalized learning and adaptive systems (Ling & Chiang, 2022; Nathaniel et al., 2021; Zhao, 2023; How & Hung, 2019).
- **Student performance metrics:** Personalized adaptive learning paths and AI models enhance engagement and outcomes by generating questions and objectives based on performance (Palanisamy et al., 2021; Kayed et al., 2022).
- **Feedback from educators and students:** Educators report increased efficiency and personalized teaching experiences, while students express higher engagement and satisfaction with personalized learning pathways (Chen et al., 2020).

Real-world implementations and contemporary research underscore the efficacy of Intelligent Tutoring Systems (ITS) and gamification in enhancing student engagement and learning outcomes. Comparative studies reveal that ALS outperform traditional methods, with positive feedback from educators and students alike. Successful implementations of ITS, such as HINTS, have improved problem-solving skills and active learning methodologies. Gamified learning activities have been shown to increase participation and retention. Research highlights the superiority of personalized adaptive approaches in improving performance, engagement, and satisfaction, demonstrating the practical benefits of AI in education.

Theme 4 - Benefits of ALS using AI

AI-driven ALS enhance engagement, motivation, and outcomes by integrating gamification, interactive features, and personalized pathways. They provide scalability, cost-effectiveness, and resource optimization for large-scale implementation:

- **Improved learning outcomes:** ALS leverage gamification, interactive features, and personalized pathways to enhance engagement, motivation, and outcomes (Krouska et al., 2021).
- **Evidence of enhanced student performance:** Gamification elements increase participation and performance by making learning interactive and enjoyable (Kwak & Mondisa, 2021; Linden et al., 2019a; Linden et al., 2019b).
- **Increased student engagement and satisfaction:** Adaptive technologies prepare students for exams, enhance engagement, and provide personalized experiences, improving performance (Diep et al., 2023; Gupta et al., 2020).
- **Personalization and student-centric approaches:** Personalized approaches in ALS create tailored experiences, fostering engagement and deeper understanding (Colchester et al., 2017; Tomkin et al., 2019).
- **Positive impact on retention rates:** Personalized and adaptive approaches increase student satisfaction, engagement, and retention (Nikou et al., 2023; Troussas et al., 2021).
- **Scalability and efficiency in education:** ALS optimize resource utilization and cost-effectiveness, enhancing scalability and efficiency (Shekawat, 2023).



- **Cost-effectiveness and resource optimization:** These systems streamline operations, automate tasks, and optimize content delivery, reducing costs and improving outcomes (Shekawat, 2023).

AI-driven ALS significantly improve engagement, motivation, and learning outcomes by integrating gamification, interactive features, and personalized pathways. These systems enhance scalability, cost-effectiveness, and resource optimization, making large-scale implementation feasible. Personalized learning approaches foster deeper understanding and retention, increasing student satisfaction and performance. The evidence shows that adaptive technologies prepare students better for exams, enhance engagement, and provide tailored educational experiences. These benefits collectively contribute to improved retention rates, efficient resource utilization, and overall enhanced educational efficiency.

Theme 5 - Challenges and limitations

AI-powered ALS face technical challenges, including data quality and integration issues, scalability, and performance concerns. Addressing data privacy, ethical considerations, institutional resistance, and providing adequate training for educators is crucial for successful integration:

- **Technical challenges in AI-powered ALS:** Data quality, integration issues, scalability, and performance concerns impact efficiency and adoption (Shi et al., 2015; Issa et al., 2018; Longa et al., 2022).
- **Ethical and privacy considerations in AI-powered ALS:** Data privacy, security, and ethical AI use are essential to safeguard student information and ensure responsible implementation (Akgun & Greenhow, 2022; Yu & Yu, 2023; Holmes et al., 2022).
- **Institutional resistance to change:** Overcoming resistance requires effective communication, stakeholder engagement, and understanding the benefits of AI-powered technologies (O'Rourke et al., 2022).
- **Training and support for educators:** Providing adequate training and support for educators ensures effective use and integration of AI-powered ALS (Amod & Mkhize, 2023; Hallewell et al., 2020).

AI-powered ALS face several technical and non-technical challenges. Issues such as data quality, integration, scalability, and performance can hinder their effectiveness. Ethical and privacy considerations are paramount, requiring strict data security measures to protect student information. Institutional resistance to adopting new technologies poses another significant challenge, necessitating effective communication and stakeholder engagement. Additionally, providing adequate training and support for educators is crucial to ensure they can effectively integrate and utilize AI-powered ALS in their teaching practices.



Theme 6 - Prospects of adaptive learning and AI

The innovative capacity of AI in personalizing online learning experiences is supported by empirical evidence and practical applications. Emerging trends in adaptive learning demonstrate AI's effectiveness in dynamically adjusting content based on continuous assessments:

- **Innovative capacity of AI in personalizing online learning experiences:** AI personalizes online learning by aligning with learning theories and practical applications like adaptive content delivery and data-driven decision-making (Willis, 2024; Gligorea et al., 2023; Kayed et al., 2022).
- **Emerging trends in adaptive learning and AI:** AI-powered ALS provide personalized experiences by adjusting content based on continuous assessments, promoting engagement and improved outcomes (Aggarwal et al., 2023; Onesi-Ozigagun et al., 2024; Akavova et al., 2023; Akintayo et al., 2024).
- **Advances in AI technologies:** AI integration in primary education fosters emotional and social learning, improving engagement and outcomes (Gligorea et al., 2023; Dabingaya, 2022; Luo, 2023).
- **New applications and use cases:** AI-powered adaptive learning platforms cater to diverse student needs, including special education, improving performance and engagement through personalized experiences (Chisom et al., 2024; Li et al., 2024; Alashwal, 2024).

The future of adaptive learning is bright, with AI playing a pivotal role in personalizing online learning experiences. Empirical evidence supports the innovative capacity of AI to align with learning theories and practical applications, such as adaptive content delivery and data-driven decision-making. Emerging trends indicate that AI-powered ALS will continue to enhance engagement and outcomes by dynamically adjusting content based on continuous assessments. Advances in AI technologies are fostering emotional and social learning, particularly in primary education. New applications and use cases are expanding the reach of adaptive learning, catering to diverse needs, including special education, and improving overall student performance and engagement.

Discussion

Theme 1 - AI in adaptive learning

AI significantly enhances ALS by utilizing ML algorithms to analyze data, identify patterns, and tailor educational content to meet individual student needs. AI-driven personalization and customization enable these systems to adapt content, pacing, and assessments, optimizing learning outcomes and fostering a more effective learning environment (Zhao, 2023). NLP facilitates communication between students and learning platforms, providing real-time feedback and interactive interactions, which enhances student engagement and comprehension (Tapalova & Zhiyenbayeva, 2022). Furthermore, AI-driven data analysis and decision-making processes collect and interpret vast amounts of data, enabling real-time adjustments and personalized learning pathways, thus maximizing educational effectiveness (Tapalova & Zhiyenbayeva, 2022).



Theme 2 - Applications of ALS

ALS revolutionize education by offering personalized learning pathways that adjust content and pace to meet each learner's unique needs. These systems enhance linguistic competence and cater to individual learning needs, improving proficiency among young learners and in employee education and training (Vanbecelaere et al., 2020; Rugaiyah, 2023; Vadivel et al., 2023; Rukadikar & Khandelwal, 2023). By allowing students to advance at their own pace and according to their learning styles, ALS significantly boost engagement and outcomes (Ranga, 2017; Raj & Renumol, 2022). They employ AI algorithms to provide adaptive assessments and feedback, ensuring continuous evaluation and personalized support (Rohm et al., 2021; Barianos et al., 2022). ITS and AI-powered tutors enhance learning by providing customized content, real-time feedback, and engaging interactions (Jakobsche et al., 2023; Huang et al., 2023; He, 2023; Del Olmo-Muñoz et al., 2023). Moreover, these systems support diverse learning needs, including students with disabilities, by creating inclusive educational environments (López-Gavira et al., 2021; Wilke et al., 2023; Horner & Sugai, 2015). Incorporating gamification elements and interactive features such as simulations and virtual reality further enhances student engagement and comprehension (Hallifax et al., 2020; Ionică et al., 2022).

Theme 3 - Real-world implementations and research outcomes

Real-world implementations and research outcomes underscore the success of ALS, particularly ITS and gamification elements, in improving student engagement and learning outcomes. ITS like HINTS have effectively supported word problem-solving skills and promoted active learning, enhancing student comprehension (Sharma & Harkishan, 2022; Zhao, 2023). Gamified learning activities have been shown to increase participation, knowledge retention, and performance by creating interactive learning environments (Lavoué et al., 2019; Khakpour & Colomo-Palacios, 2021). Interactive elements such as simulations and virtual reality provide experiential learning opportunities that enhance engagement and outcomes (Linden et al., 2019a; Linden et al., 2019b; Chou et al., 2021). Comparative studies indicate that personalized and adaptive approaches improve student performance, engagement, and satisfaction compared to traditional methods (Ling & Chiang, 2022; Nathaniel et al., 2021). Metrics show that AI-driven personalized learning paths enhance engagement and learning outcomes (Palanisamy et al., 2021; Kayed et al., 2022). Feedback from educators and students highlights the efficiency and personalized experiences offered by these systems (Chen et al., 2020).

Theme 4 - Benefits of ALS using AI

AI-driven ALS significantly enhance student engagement, motivation, and learning outcomes by integrating gamification, interactive features, and personalized pathways. These systems leverage various strategies to create dynamic and engaging educational experiences, leading to improved student performance and engagement (Krouska et al., 2021; Kwak & Mondisa, 2021). Personalized approaches in ALS foster deeper understanding and satisfaction, contributing to higher retention rates (Colchester et al., 2017; Nikou et al., 2023). These systems optimize resource utilization and cost-effectiveness, making them scalable and efficient for large-scale implementation (Shekahwat,



2023). By streamlining operations and automating tasks, ALS reduce costs and improve educational outcomes, demonstrating their potential for widespread adoption in diverse educational settings (Shekawat, 2023).

Theme 5 - Challenges and limitations

AI-powered ALS face several challenges, including data quality and integration issues, scalability, and performance concerns. Ensuring data privacy and addressing ethical considerations are critical for the responsible implementation of these systems (Shi et al., 2015; Akgun & Greenhow, 2022; Yu & Yu, 2023). Overcoming institutional resistance to change and providing adequate training and support for educators are essential for integrating adaptive learning technologies into educational practices (O'Rourke et al., 2022; Amod & Mkhize, 2023). Addressing these challenges is crucial for the effective and widespread adoption of AI-powered ALS, ensuring they meet educational goals and maintain high standards of data security and ethical use (Holmes et al., 2022).

Theme 6 - Prospects of adaptive learning and AI

The innovative capacity of AI in personalizing online learning experiences is well-supported by empirical evidence and practical applications. AI-powered ALS dynamically adjust content based on continuous assessments, providing personalized experiences that promote engagement and improved outcomes (Willis, 2024; Gligorea et al., 2023; Kayed et al., 2022). Advances in AI technologies, particularly in primary education, foster emotional and social learning, enhancing overall student engagement and outcomes (Gligorea et al., 2023; Dabingaya, 2022; Luo, 2023). New applications in special education demonstrate AI's ability to cater to diverse student needs, improving performance and engagement through personalized learning experiences (Chisom et al., 2024; Li et al., 2024; Alashwal, 2024). As AI continues to evolve, its integration into ALS will likely expand, further revolutionizing educational practices and outcomes.

Gaps in research literature and future research directions

The integration of Artificial Intelligence (AI) into Adaptive Learning Systems (ALS) holds considerable potential for improving educational outcomes. However, critical gaps in the research literature require attention. One prominent issue is the lack of exploration into the long-term effects of ALS on diverse learner populations. Current studies often emphasize short-term benefits and specific demographic groups, neglecting extended impacts on varied populations, including underrepresented groups and those with diverse learning needs (Tang, 2024; Anuyahong, 2023; Bakhmat, 2024). Although research highlights the immediate benefits of AI in personalized learning, understanding its sustained effects remains insufficient. Longitudinal studies are needed to evaluate how ALS influence learning trajectories over time and to determine whether these benefits persist uniformly across different demographics. Such insights are essential for designing inclusive and effective educational technologies (Syaikhudin, 2024; Ansor, 2023; Adeleye, 2024).



Another gap lies in the insufficient integration of emotional and social learning within AI-powered ALS. While some studies suggest AI can enhance emotional intelligence and social learning, empirical evidence is limited. Emotional and social competencies are critical for academic success and overall well-being, yet they are often overlooked in AI educational tools (Anwar, 2023; Wang et al., 2020; Abbas et al., 2023). To address this, future research should focus on embedding emotional and social learning frameworks into ALS by developing algorithms that respond not only to learners' academic needs but also to their emotional states and social interactions. Such advancements could foster supportive and engaging learning environments that promote both cognitive and emotional development (Lui & Leung, 2020; Seo et al., 2021; Alshehri, 2023).

Ethical concerns also represent a significant gap in the literature. Issues like data privacy, algorithmic bias, and fairness are often inadequately addressed despite their relevance to educational equity. Many studies fail to explore the societal and ethical implications of widespread AI adoption in education, such as the potential to reinforce biases or compromise student privacy (Mutia, 2024; Nurjanah, 2024; Crompton & Song, 2021). Future research should prioritize ethical considerations by examining how to design and implement AI systems responsibly. This includes mitigating algorithmic bias and ensuring ethical and transparent handling of student data (Khakurel et al., 2018; Chisom, 2024; Cheong et al., 2022).

Finally, the scalability and cost-effectiveness of ALS in under-resourced educational settings remain underexplored. Financial and infrastructural barriers can limit the adoption of these technologies in low-resource contexts, despite their potential to enhance educational access and quality. Research should investigate scalable models and cost-effective strategies for implementing AI technologies in diverse settings, focusing on partnerships with educational institutions, government agencies, and technology providers (Ezzaim, 2024; Anoir, 2024; Mahmudi, 2023). Addressing these challenges will ensure that the benefits of AI in education reach learners regardless of socioeconomic status (Thuan, 2024).

In conclusion, while AI-powered ALS offer transformative potential, addressing gaps in long-term impact studies, emotional and social learning integration, ethical considerations, and scalability research is essential. These efforts will help create inclusive, effective, and equitable educational technologies that meet the diverse needs of learners worldwide.

Recommendations for policy and practice

To maximize the benefits of ALS using AI in education, several recommendations for policy and practice should be considered. These recommendations aim to address current challenges, leverage existing opportunities, and ensure the ethical and effective implementation of AI-powered ALS:

- **Develop clear and comprehensive AI policies:** Educational institutions and policymakers should establish clear policies that govern the use of AI in ALS. These policies should address data privacy, ethical considerations, transparency, and accountability. Guidelines should be



developed to ensure that AI algorithms are fair, unbiased, and used responsibly to protect students' rights and interests.

- **Ensure data privacy and security:** Protecting student data is paramount. Educational institutions must implement robust data privacy and security measures to safeguard sensitive information. Policies should mandate strict adherence to data protection regulations and require that students and parents are informed about how their data is collected, stored, and used.
- **Promote equity and accessibility:** To ensure that all students benefit from ALS, policies should focus on promoting equity and accessibility. This includes providing adequate resources and support for under-resourced schools, ensuring that adaptive learning technologies are accessible to students with disabilities, and addressing any potential biases in AI algorithms that may disadvantage certain groups of students.
- **Invest in professional development for educators:** Educators play a critical role in the successful implementation of ALS. Policymakers should invest in ongoing professional development programs to train educators on effectively integrating AI-powered adaptive learning tools into their teaching practices. This training should cover the technical aspects of using these systems and strategies for interpreting and acting on the data they provide.
- **Foster collaboration and partnerships:** Collaboration between educational institutions, technology developers, and researchers is essential for the continuous improvement of ALS. Policies should encourage partnerships that facilitate the sharing of best practices, resources, and research findings. Such collaboration helps ensure that adaptive learning technologies remain cutting-edge and relevant to the evolving needs of education.
- **Encourage research and innovation:** Policymakers should support research initiatives that explore new applications and improvements in ALS. Funding should be allocated to studies that investigate the long-term impacts of these systems, explore their integration with emotional and social learning, and address ethical concerns. Encouraging innovation in this field will help develop more effective and holistic learning solutions.
- **Monitor and evaluate implementation:** Continuous monitoring and evaluation are crucial for the successful implementation of ALS. Educational institutions should establish mechanisms to regularly assess the effectiveness of these systems in improving student outcomes. Policies should require transparent reporting on the performance of adaptive learning technologies and their impact on various student demographics.
- **Address ethical considerations:** Ethical considerations must be at the forefront of policy development. This includes ensuring that AI systems are designed and used in ways that promote fairness, transparency, and accountability. Policies should also address the potential psychological impacts of AI on students, ensuring that these technologies enhance rather than hinder their well-being.
- **Support scalable and sustainable solutions:** To broaden the reach of ALS, policies should promote scalable and sustainable solutions. This includes investing in cloud-based platforms that is easily scaled to accommodate larger student populations and ensuring that these systems are cost-effective and resource-efficient.



By implementing these recommendations, policymakers and educational practitioners will create a supportive environment for the effective and ethical use of AI-powered ALS. These efforts will help ensure that all students benefit from personalized, engaging, and effective learning experiences tailored to their unique needs and preferences.

Conclusion

Summary of key insights

AI in adaptive learning

- **ML algorithms:** AI in adaptive learning leverages ML algorithms to analyze data, identify patterns, and personalize educational content to suit individual student needs. These algorithms dynamically adjust to the students' needs, enhancing learning effectiveness.
- **NLP:** NLP facilitates communication between students and learning platforms by interpreting and generating human language. This enables real-time feedback, interactive interactions, and enhances student engagement and comprehension.
- **Role of AI in personalization and customization:** AI-driven personalization allows systems to adapt content, pacing, and assessments to match each student's learning style and proficiency level. This approach optimizes learning outcomes, increases engagement, and fosters a more effective learning environment.
- **AI-driven data analysis and decision-making:** AI-driven data analysis collects, analyzes, and interprets vast amounts of data to inform instructional strategies and interventions. These processes enable real-time adjustments and personalized learning pathways to maximize learning outcomes.

Applications of Adaptive Learning Systems (ALS)

- **Personalized learning pathways:** ALS provide customized learning pathways, adjusting content and pace for individual learners. This approach enhances linguistic competence and caters to diverse learning needs.
- **Customizing content and pace for individual learners:** ALS allow students to progress through content at their own pace, according to their learning styles and preferences, significantly boosting engagement and outcomes.
- **Adaptive assessments and feedback mechanisms:** ALS utilize AI to analyze performance data and adjust content delivery in real-time, ensuring continuous evaluation and personalized support, leading to improved learning outcomes.
- **ITS and AI-powered tutors:** ITS tailor instruction to individual users, providing hints and engaging in dialogue to enhance learning. AI-powered tutors deliver customized content and real-time feedback, enhancing student engagement and optimizing outcomes.
- **Support for diverse learning needs:** ALS offer targeted support for students with disabilities, creating inclusive educational environments that cater to diverse needs.



- **Gamification elements in adaptive learning:** Incorporating gamification elements such as points, levels, and rewards enhances engagement and motivation by creating interactive, game-like learning experiences.
- **Interactive and immersive learning experiences:** ALS integrate interactive elements such as simulations, virtual reality, and multimedia content, offering hands-on learning opportunities that enhance comprehension and retention.

Real-world implementations and research outcomes

- **Success of ITS and gamification:** Real-world implementations of ITS and gamification have improved student engagement and learning outcomes. ITS like HINTS have supported problem-solving skills and active learning methodologies, while gamified activities have increased participation and retention.
- **Comparative studies:** Research indicates that personalized and adaptive approaches significantly improve student performance, engagement, and satisfaction compared to traditional methods.
- **Feedback from educators and students:** Educators report increased efficiency and personalized teaching experiences, while students express higher engagement and satisfaction with personalized learning pathways.

Benefits of ALS using AI

- **Improved learning outcomes:** ALS enhance engagement, motivation, and outcomes by integrating gamification, interactive features, and personalized pathways.
- **Enhanced student performance:** Gamification elements increase participation and performance, making learning interactive and enjoyable.
- **Increased student engagement and satisfaction:** Adaptive technologies enhance engagement and provide personalized experiences, improving performance.
- **Personalization and student-centric approaches:** Personalized approaches foster deeper understanding and satisfaction, contributing to higher retention rates.
- **Scalability and efficiency:** ALS optimize resource utilization and cost-effectiveness, enhancing scalability and efficiency.
- **Cost-effectiveness:** These systems streamline operations, automate tasks, and optimize content delivery, reducing costs and improving outcomes.

Challenges and limitations

- **Technical challenges:** Issues such as data quality, integration, scalability, and performance impact the efficiency and adoption of AI-powered ALS.
- **Ethical and privacy considerations:** Ensuring data privacy, security, and ethical AI use is essential to safeguard student information.
- **Institutional resistance:** Overcoming resistance to change requires effective communication and stakeholder engagement.



- **Training and support for educators:** Providing adequate training and support for educators ensures the effective use and integration of AI-powered ALS.

Prospects of adaptive learning and AI

- **Innovative capacity of AI:** AI personalizes online learning by aligning with learning theories and practical applications like adaptive content delivery and data-driven decision-making.
- **Emerging trends:** AI-powered ALS continue to enhance engagement and outcomes by dynamically adjusting content based on continuous assessments.
- **Advances in AI technologies:** AI integration fosters emotional and social learning, particularly in primary education.
- **New applications and use cases:** AI-powered adaptive learning platforms cater to diverse student needs, including special education, improving performance and engagement through personalized experiences.

Limitations

While this review offers comprehensive insights into AI-powered ALS, several limitations must be acknowledged. Firstly, the literature reviewed was confined to studies published in English, potentially excluding valuable research in other languages, which may limit the generalizability to non-English-speaking contexts. Secondly, the review focused on recent peer-reviewed articles, conference papers, and high-quality grey literature from the past ten years, potentially overlooking older research that could provide foundational insights or longitudinal data. Additionally, there was variability in the quality and methodologies of the included studies. Despite using the CASP checklist, the heterogeneity in study designs, sample sizes, and contexts makes it challenging to draw definitive conclusions or compare results. Moreover, the review relies heavily on secondary data, which may not capture the latest advancements or emerging trends in adaptive learning technologies. Rapid technological developments in AI might outpace the research, resulting in a gap between current practice and documented evidence. Finally, while the review highlights benefits and applications, it may not fully capture the nuanced challenges and barriers to implementation, such as resistance to change, the need for extensive teacher training, and the complexities of integrating AI with existing educational infrastructures. These limitations underscore the need for ongoing, high-quality research that includes diverse linguistic and cultural contexts, utilizes robust methodologies, and keeps pace with advancements in AI and educational technologies.

Abbreviations

Adaptive Learning Systems: ALS

Artificial Intelligence: AI

Machine Learning: ML

Natural Language Processing: NLP

Intelligent Tutoring Systems: ITS

Critical Appraisal Skills Programme: CASP

Preferred Reporting Items for Systematic Reviews and Meta-Analyses: PRISMA

Meylani, R. (2024). A critical glance at adaptive learning systems using artificial intelligence: A systematic review and qualitative synthesis of contemporary research literature. *Western Anatolia Journal of Educational Sciences, 15(3), 3519-3547.*

DOI. 10.51460/baebd.1525452



Batı Anadolu Eğitim Bilimleri Dergisi, (2024), 15 (3), 3519-3547.

Western Anatolia Journal of Educational Sciences, (2024), 15 (3), 3519-3547.

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