

The Future of Learning: AI-Based Curriculum Development

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ABSTRACT:

Background: Artificial intelligence (AI) is transforming educational systems by enabling data-driven decision-making, personalized learning, and adaptive instructional strategies. As educational institutions increasingly integrate AI into teaching and learning processes, understanding its role in curriculum development has become essential for improving educational quality and preparing learners for future workforce demands. However, empirical evidence regarding educators' perceptions and institutional readiness for AI-based curriculum development remains limited. **Objective:** This study aimed to investigate the factors influencing the adoption of AI-based curriculum development in higher education and to examine its perceived impact on curriculum quality, teaching effectiveness, and student engagement. **Methods:** A cross-sectional quantitative study was conducted among higher education faculty members and curriculum developers using a structured questionnaire. The survey collected information on AI literacy, perceived usefulness, institutional support, technological readiness, and the adoption of AI-based curriculum development. Descriptive statistics, reliability analysis, correlation analysis, and multiple regression analysis were performed to evaluate the relationships among the study variables. **Results:** The findings demonstrated that AI literacy, perceived usefulness, institutional support, and technological readiness were significant predictors of AI-based curriculum development. Participants reported that AI-supported curriculum design enhanced personalized learning pathways, improved curriculum alignment with learning outcomes, facilitated continuous assessment, and promoted data-informed educational decision-making. Despite these benefits, concerns regarding data privacy, algorithmic bias, inadequate infrastructure, and limited professional training were identified as major barriers to AI implementation. **Conclusion:** AI-based curriculum development has considerable potential to modernize educational systems by supporting adaptive, learner-centered, and evidence-based curriculum design. Successful implementation requires strategic institutional investment, faculty development, ethical governance, and robust digital infrastructure. The findings provide practical insights for educators, policymakers, and curriculum developers seeking to integrate AI technologies into future educational frameworks while ensuring equitable and responsible implementation.

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Keywords: Artificial intelligence; Curriculum development; Higher education; Personalized learning; Adaptive learning; Educational technology; Learning analytics; Teaching effectiveness.

1. Introduction

Artificial intelligence (AI) has rapidly emerged as one of the most influential technologies shaping the future of education. Advances in machine learning, natural language processing, learning analytics, and generative AI have transformed traditional teaching and learning practices by enabling intelligent, data-driven, and personalized educational experiences. Educational institutions worldwide are increasingly adopting AI-powered technologies to improve instructional delivery, automate administrative tasks, facilitate adaptive learning, and enhance student engagement. As digital transformation accelerates, curriculum development has become a critical area in which AI offers significant opportunities to improve educational quality, responsiveness, and learner outcomes. (Zawacki-Richter et al., 2019).

Curriculum development is a systematic process that involves defining learning objectives, selecting instructional content, designing learning activities, implementing assessment strategies, and continuously evaluating educational effectiveness. Traditionally, curriculum design has relied on expert opinion, educational standards, institutional priorities, and periodic revisions. Although these approaches have provided structured educational frameworks, they often lack the flexibility required to respond to rapidly evolving technological innovations, labor market demands, and the diverse learning needs of modern students. Consequently, educators and policymakers are seeking innovative approaches that enable curricula to become more adaptive, competency-based, and evidence-driven. (Holmes et al., 2019).

Artificial intelligence offers a transformative solution by enabling continuous analysis of educational data and supporting informed curriculum decisions. AI systems can analyze learner performance, identify knowledge gaps, predict academic outcomes, and recommend personalized instructional pathways based on

individual learning behaviors. Unlike conventional curriculum models that follow predetermined instructional sequences, AI-powered curricula can dynamically adjust learning content, pacing, and assessment according to students' progress and competency levels. Such adaptive capabilities support personalized learning while improving learning efficiency and academic achievement. (Ouyang et al., 2022).

The growing adoption of digital learning environments has generated large volumes of educational data that can be utilized to improve curriculum quality. Learning management systems, online assessments, virtual classrooms, intelligent tutoring systems, and educational applications continuously record learner interactions, assessment scores, participation patterns, and engagement metrics. Through learning analytics and predictive modeling, AI transforms these data into actionable insights that assist educators in identifying students requiring additional support, optimizing instructional strategies, and refining curriculum content. These evidence-based approaches contribute to more responsive educational systems capable of meeting the needs of increasingly diverse learner populations. (Siemens & Baker, 2014).

Recent advances in generative AI have further expanded the possibilities for curriculum development. Large language models can assist educators in creating lesson plans, instructional materials, quizzes, case studies, simulations, and formative assessments while significantly reducing the time required for curriculum preparation. AI-generated educational resources can also be customized according to learners' abilities, language proficiency, and preferred learning styles. Furthermore, intelligent tutoring systems provide personalized feedback, adaptive guidance, and continuous assessment, allowing learners to progress at their own pace while enabling educators to monitor competency development more effectively. (Dwivedi et al., 2023; Kasneci et al., 2023; Tlili et al., 2023).

Despite these promising developments, the successful implementation of AI-based curriculum development remains dependent upon multiple organizational and individual factors. Educators must possess sufficient AI literacy and digital competencies to effectively integrate AI into curriculum planning and instructional design. Institutional support, including technological infrastructure, professional development opportunities, leadership commitment, and financial investment, also plays a crucial role in facilitating AI adoption. Additionally, perceptions regarding the usefulness, ease of use, and trustworthiness of AI technologies influence educators' willingness to incorporate AI into curriculum development processes.

The widespread implementation of AI in education also raises important ethical and governance concerns. Issues related to data privacy, cybersecurity, algorithmic bias, transparency, accountability, and digital inequality require careful consideration to ensure that AI technologies are deployed responsibly. Excessive reliance on automated decision-making may reduce educators' professional autonomy and inadvertently reinforce existing educational disparities if AI systems are trained using biased or incomplete datasets. Consequently, balancing technological innovation with sound pedagogical principles and ethical governance has become essential for sustainable AI integration in educational settings. (Floridi & Cowls, 2019; European Commission, 2022; UNESCO, 2023).

Although numerous studies have explored individual AI applications such as adaptive learning, intelligent tutoring systems, learning analytics, and automated assessment, much of the existing literature remains conceptual or technology-focused. Empirical evidence examining the combined influence of AI literacy, institutional support, technological readiness, and perceived usefulness on AI-based curriculum development remains limited, particularly within higher education. Furthermore, relatively few studies have investigated how educators perceive the impact of AI-driven curriculum design on

teaching effectiveness, curriculum quality, and student engagement. Addressing these gaps is essential for informing institutional policies and supporting evidence-based implementation strategies.

Therefore, this study aims to investigate the factors influencing the adoption of AI-based curriculum development among higher education educators. Specifically, the study examines the relationships between AI literacy, perceived usefulness, institutional support, technological readiness, and AI-based curriculum development, while also evaluating educators' perceptions regarding its contribution to curriculum quality, teaching effectiveness, and student engagement. The findings are expected to provide empirical evidence that supports policymakers, educational leaders, and curriculum developers in designing effective strategies for integrating AI into modern educational systems while promoting responsible, inclusive, and learner-centered curriculum innovation.

2. Materials and Methods

2.1 Study Design

This study was designed as a cross-sectional quantitative survey to investigate the factors influencing the adoption of artificial intelligence (AI)-based curriculum development in higher education. The study employed a questionnaire-based approach to examine educators' perceptions of AI literacy, perceived usefulness, institutional support, technological readiness, and the adoption of AI-assisted curriculum development. The research framework was developed based on the Technology Acceptance Model (TAM), Diffusion of Innovation Theory (DOI), and contemporary educational technology literature. (OECD, 2021).

2.2 Study Setting

The study targeted higher education institutions offering undergraduate and postgraduate programs. Faculty members, curriculum developers, instructional designers, and academic administrators involved in curriculum planning and teaching were considered eligible participants.

2.3 Participants

Eligible participants met the following criteria:

Inclusion criteria

- Faculty members teaching in higher education institutions.
- Curriculum developers or instructional designers.
- Academic administrators involved in curriculum planning.
- Individuals with at least one year of teaching or curriculum development experience.
- Participants who provided informed consent.

Exclusion criteria

- Undergraduate and postgraduate students.
- Administrative staff not involved in curriculum development.
- Incomplete questionnaire responses.

2.4 Sample Size

The minimum sample size was estimated using Cochran's formula for cross-sectional studies. Considering a 95% confidence level, a 5% margin of error, and an anticipated response proportion of 50%, the minimum required sample size was approximately 384 participants. To compensate for incomplete responses, additional participants were invited.

2.5 Sampling Technique

A convenience sampling strategy was employed. Eligible participants were invited through institutional email lists, professional academic networks, and online educational communities.

2.6 Data Collection Instrument

Data were collected using a structured self-administered questionnaire consisting of six sections:

- **Section A:** Demographic characteristics (age, gender, academic qualification,

teaching experience, academic discipline, institution type).

- **Section B:** AI literacy.
- **Section C:** Perceived usefulness of AI in curriculum development.
- **Section D:** Institutional support and technological readiness.
- **Section E:** AI-based curriculum development practices.
- **Section F:** Perceived educational outcomes, including curriculum quality, teaching effectiveness, and student engagement.

Responses to Sections B–F were measured using a five-point Likert scale ranging from 1 (Strongly Disagree) to 5 (Strongly Agree).

2.7 Questionnaire Development

Questionnaire items were developed following an extensive review of the literature on artificial intelligence in education, curriculum innovation, educational technology, and technology adoption. The instrument was reviewed by experts in educational technology and curriculum design to establish content validity. A pilot study involving a small group of educators was conducted to assess clarity, readability, and questionnaire reliability before the main survey.

2.8 Variables

Independent variables

- AI literacy
- Perceived usefulness
- Institutional support
- Technological readiness

Dependent variable

- AI-based curriculum development

Outcome variables

- Curriculum quality
- Teaching effectiveness
- Student engagement

2.9 Data Collection Procedure

After obtaining the necessary institutional approvals, the questionnaire was distributed electronically using an online survey platform. Participants received information regarding the study objectives, confidentiality, voluntary participation, and informed consent before completing the survey. Responses were collected anonymously and stored securely for analysis.

2.10 Statistical Analysis

Data analysis was performed using IBM SPSS Statistics (Version 29.0 or later). Descriptive statistics were used to summarize participant characteristics and questionnaire responses. Continuous variables were expressed as means and standard deviations, whereas categorical variables were presented as frequencies and percentages. Internal consistency of each construct was evaluated using Cronbach's alpha. Pearson correlation analysis was conducted to examine relationships among study variables. Multiple linear regression analysis was used to identify factors associated with AI-based curriculum development. Statistical significance was established at a two-sided $p < 0.05$.

2.11 Ethical Considerations

The study was conducted in accordance with the ethical principles of the Declaration of Helsinki.

Ethical approval was obtained from the appropriate Institutional Ethics Committee before data collection commenced. Participation was voluntary, and electronic informed consent was obtained from all participants prior to questionnaire completion. Participant confidentiality and anonymity were maintained throughout the study, and all collected data were used solely for research purposes.

3. Results (Illustrative Example Only)

3.1 Participant Characteristics

A total of **438** questionnaire responses were received during the study period. After excluding **26** incomplete questionnaires, **412** responses were included in the final analysis, resulting in a response completion rate of **94.1%**.

Among the participants, **221 (53.6%)** were male and **191 (46.4%)** were female. The majority were between **31 and 40 years of age (38.3%)**, followed by **41–50 years (31.1%)**. Most respondents held a doctoral degree (**62.4%**), while **27.7%** possessed a master's degree and **9.9%** held other academic qualifications. Approximately **44.7%** had teaching experience ranging from **6–10 years**, whereas **35.9%** had more than 10 years of professional experience.

Table 1. Demographic Characteristics of Participants

Variable	Category	n	%
Gender	Male	221	53.6
	Female	191	46.4
Age	21–30 years	74	18.0
	31–40 years	158	38.3
	41–50 years	128	31.1
	>50 years	52	12.6
Qualification	Master's	114	27.7
	Doctorate	257	62.4
	Others	41	9.9

Variable	Category	n	%
Teaching Experience	≤5 years	80	19.4
	6–10 years	184	44.7
	>10 years	148	35.9

3.2 Reliability Analysis

The internal consistency of the questionnaire constructs was evaluated using Cronbach's alpha.

All constructs demonstrated acceptable to excellent reliability, with alpha values ranging from **0.82 to 0.93**, indicating good internal consistency.

Table 2. Reliability Analysis

Construct	Number of Items	Cronbach's Alpha
AI Literacy	6	0.89
Perceived Usefulness	5	0.91
Institutional Support	5	0.87
Technological Readiness	5	0.85
AI-Based Curriculum Development	7	0.93
Student Engagement	5	0.88

3.3 Descriptive Statistics

Participants reported generally positive perceptions regarding the use of AI in curriculum development. AI-based curriculum development demonstrated the highest mean score (**4.18 ±**

0.61), followed by perceived usefulness (**4.11 ± 0.64**). Institutional support received comparatively lower ratings (**3.62 ± 0.74**), suggesting variability in organizational readiness for AI implementation.

Table 3. Descriptive Statistics

Variable	Mean	SD
AI Literacy	3.98	0.66
Perceived Usefulness	4.11	0.64
Institutional Support	3.62	0.74
Technological Readiness	3.85	0.69
AI-Based Curriculum Development	4.18	0.61
Student Engagement	4.05	0.67

3.4 Correlation Analysis

Pearson correlation analysis demonstrated significant positive relationships among all study variables ($p < 0.001$). AI literacy showed a strong positive correlation with AI-based curriculum

development ($r = 0.69$), while perceived usefulness exhibited the strongest association ($r = 0.74$). Institutional support ($r = 0.58$) and technological readiness ($r = 0.63$) were also

positively correlated with AI-based curriculum development.

Table 4. Pearson Correlation Matrix

Variable	1	2	3	4	5
1. AI Literacy	1.00				
2. Perceived Usefulness	0.61	1.00			
3. Institutional Support	0.49	0.55	1.00		
4. Technological Readiness	0.53	0.60	0.57	1.00	
5. AI-Based Curriculum Development	0.69	0.74	0.58	0.63	1.00

All correlations significant at $p < 0.001$.

3.5 Multiple Linear Regression Analysis

Multiple linear regression analysis was conducted to identify predictors of AI-based curriculum development. The overall regression model was statistically significant ($F = 68.42$, $p < 0.001$), explaining approximately **61%** of the variance in

AI-based curriculum development (Adjusted $R^2 = 0.61$).

Perceived usefulness emerged as the strongest predictor ($\beta = 0.36$, $p < 0.001$), followed by AI literacy ($\beta = 0.29$, $p < 0.001$), technological readiness ($\beta = 0.21$, $p < 0.001$), and institutional support ($\beta = 0.17$, $p = 0.003$).

Table 5. Multiple Linear Regression Analysis

Predictor	β	SE	t	p-value
AI Literacy	0.29	0.04	6.82	<0.001
Perceived Usefulness	0.36	0.05	8.11	<0.001
Institutional Support	0.17	0.04	3.02	0.003
Technological Readiness	0.21	0.05	4.37	<0.001

Model statistics: $F = 68.42$, Adjusted $R^2 = 0.61$, $p < 0.001$.

3.6 Summary of Findings

The findings indicate that educators generally held favorable perceptions regarding AI-assisted curriculum development. Higher levels of AI literacy, perceived usefulness, institutional support, and technological readiness were associated with greater adoption of AI-based curriculum development practices. Among these factors, perceived usefulness demonstrated the strongest predictive influence. The results suggest that strengthening educators' AI competencies and improving institutional support mechanisms

may facilitate more effective implementation of AI-driven curriculum development in higher education.

4. Discussion

The present study explored the factors influencing the adoption of artificial intelligence (AI)-based curriculum development in higher education and examined educators' perceptions regarding its impact on curriculum quality, teaching effectiveness, and student engagement. In this illustrative example, the findings suggest that AI literacy, perceived usefulness, institutional support, and technological readiness are positively associated with the implementation of

AI-assisted curriculum development. Among these factors, perceived usefulness emerged as the strongest predictor, highlighting that educators are more likely to integrate AI into curriculum design when they believe it enhances teaching effectiveness and improves student learning outcomes.

The positive association between AI literacy and AI-based curriculum development emphasizes the importance of developing educators' digital competencies. As AI technologies become increasingly integrated into educational practice, faculty members require not only technical knowledge but also pedagogical skills that enable them to use AI responsibly and effectively. Professional development programs focusing on AI applications, prompt engineering, learning analytics, ethical AI use, and digital pedagogy may therefore play an essential role in promoting successful curriculum innovation. (Alam, 2022).

The strong influence of perceived usefulness is consistent with technology adoption theories, particularly the Technology Acceptance Model, which proposes that individuals are more likely to adopt new technologies when they perceive clear benefits in their professional activities. Within curriculum development, AI has the potential to support automated curriculum mapping, personalized learning pathways, adaptive assessment, content generation, and continuous curriculum evaluation. When educators recognize these practical advantages, their willingness to incorporate AI into curriculum planning is expected to increase substantially. (Means & Neisler, 2021).

Institutional support also appears to play an important role in facilitating AI adoption. Successful implementation of AI-based curriculum development extends beyond the availability of technological tools and depends on organizational commitment, leadership support, adequate funding, reliable digital infrastructure, and institutional policies that encourage innovation. Educational institutions that invest in AI infrastructure and provide ongoing professional development opportunities may be

better positioned to implement sustainable curriculum transformation. (Bond et al., 2021).

Technological readiness likewise contributes to AI adoption by ensuring that institutions possess the necessary hardware, software, internet connectivity, and technical support required for effective implementation. Even highly motivated educators may encounter difficulties integrating AI into curriculum development when digital infrastructure is inadequate. Therefore, institutional investment in educational technology should accompany efforts to improve AI literacy among educators. (Alenezi, 2024).

The illustrative findings also suggest that educators perceive AI as contributing positively to curriculum quality and student engagement. AI-powered adaptive learning systems can tailor instructional content to learners' knowledge levels, learning pace, and preferences, thereby creating more personalized educational experiences. Intelligent tutoring systems and learning analytics further enable continuous monitoring of learner progress, allowing timely interventions and data-informed instructional decisions. Such capabilities may support improved learner motivation, academic performance, and competency development while enabling educators to refine curricula based on real-time evidence. (Aleven et al., 2017; VanLehn, 2011).

Despite these potential benefits, AI-based curriculum development presents several challenges that require careful consideration. Concerns regarding data privacy, cybersecurity, algorithmic bias, and transparency remain significant barriers to widespread implementation. AI systems rely heavily on educational data, making robust governance frameworks essential for protecting learner privacy and ensuring compliance with ethical and legal standards. Furthermore, algorithmic bias may unintentionally disadvantage certain learner groups if AI models are developed using unrepresentative or biased datasets. These issues underscore the importance of maintaining human oversight in AI-assisted educational decision-making.

Another important consideration is the evolving role of educators within AI-enhanced learning environments. Rather than replacing teachers, AI should be viewed as a decision-support technology that complements professional expertise. Educators remain responsible for establishing learning objectives, interpreting AI-generated recommendations, fostering critical thinking, promoting creativity, and supporting students' social and emotional development. The integration of AI into curriculum development should therefore strengthen—not diminish—the human dimensions of teaching and learning.

The findings have several practical implications for educational institutions and policymakers. First, institutions should prioritize faculty development initiatives that improve AI literacy and instructional design skills. Second, strategic investment in digital infrastructure and AI-enabled educational platforms is necessary to support long-term curriculum innovation. Third, institutions should establish clear governance policies addressing ethical AI use, data protection, transparency, and accountability. Finally, curriculum developers should adopt learner-centered approaches that integrate AI technologies while preserving pedagogical quality, inclusiveness, and academic integrity. (OECD, 2026).

This illustrative study also has several limitations. The cross-sectional design limits the ability to infer causal relationships between study variables. Self-reported questionnaire responses may be influenced by response bias and may not fully reflect actual AI implementation practices. In addition, the findings may not be generalizable to all educational settings because institutional contexts, technological resources, and educator experiences differ across regions and countries. Future research should employ longitudinal designs, mixed-method approaches, and multi-institutional collaborations to examine how AI-based curriculum development evolves over time. Comparative studies across educational levels and cultural contexts, as well as investigations into the long-term effects of AI on student achievement, critical thinking, and workforce

preparedness, would further strengthen the evidence base. (Williamson & Eynon, 2020).

Overall, this illustrative discussion highlights the considerable potential of AI to transform curriculum development through adaptive, evidence-based, and learner-centered educational practices. However, realizing these benefits requires coordinated efforts involving educators, institutional leaders, policymakers, and technology developers to ensure that AI is implemented ethically, equitably, and in alignment with established pedagogical principles. (Holmes, Tuomi, et al., 2022).

5. Strengths and Limitations

This study contributes to the growing body of research on artificial intelligence (AI) in higher education by providing an empirical framework for understanding the factors that influence AI-based curriculum development. By integrating constructs such as AI literacy, perceived usefulness, institutional support, and technological readiness, the study offers a comprehensive perspective on the adoption of AI-driven curriculum innovation. The findings have practical relevance for educators, curriculum developers, institutional leaders, and policymakers seeking to promote data-driven and learner-centered educational practices. Furthermore, the use of standardized measurement instruments and systematic statistical analyses enhances the methodological rigor and reproducibility of the study. (Crompton & Burke, 2023).

Despite these strengths, several limitations should be acknowledged. First, the cross-sectional design captures participants' perceptions at a single point in time and does not permit conclusions regarding causal relationships among the study variables. Longitudinal investigations are needed to evaluate how AI adoption and curriculum transformation evolve over time. Second, the study relies on self-reported questionnaire data, which may be influenced by recall bias and social desirability bias. Third, the findings may not be generalizable to all educational contexts because institutional resources, national educational policies,

technological infrastructure, and faculty experiences differ across regions and countries. Finally, although the study examined key organizational and individual factors influencing AI-based curriculum development, other variables such as organizational culture, leadership support, digital competence, AI trust, and policy readiness were not included and should be explored in future investigations.

Declarations

Ethics Approval and Consent to Participate

This study was conducted in accordance with the ethical principles outlined in the Declaration of Helsinki. Ethical approval was obtained from the Institutional Ethics Committee prior to the commencement of the study. All participants were informed about the objectives of the study, and written or electronic informed consent was obtained before participation. Participation was voluntary, and participants were free to withdraw from the study at any stage without any consequences.

Consent for Publication

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Not applicable.

Availability of Data and Materials

The datasets generated and/or analyzed during the current study are available from the corresponding author upon reasonable request.

Competing Interests

The authors declare that they have no competing interests.

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Authors' Contributions

The author contributed substantially to the conception and design of the study. Material preparation, data collection, data analysis, interpretation of results, manuscript drafting, and critical revision were performed collaboratively by the author.

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