

# Artificial Intelligence in Educational Management: A Systematic Literature Review on AI-Based Primary Curriculum Design

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**Abstract.** The integration of Artificial Intelligence (AI) into primary education curriculum governance remains constrained by conceptual ambiguity, ethical concerns, and digital inequalities, particularly in under-resourced education systems where technological readiness is limited. This study addresses these challenges by conducting a systematic literature review to explore how AI can be ethically, effectively, and contextually embedded into curriculum decision-making. Grounded in five theoretical frameworks—Data-Driven Decision Making, Adaptive Learning, AI-Based Decision Support Systems, Contextual Curriculum Design, and Technology Ethics in Education—the review synthesizes findings from peer-reviewed publications over the past decade. Results reveal that AI holds significant potential to strengthen curriculum planning through real-time assessment, personalized learning trajectories, and prescriptive analytics that enhance evidence-based decisions. Nevertheless, systemic barriers such as poor digital infrastructure, limited AI literacy among educators, and fragmented policy directions continue to hinder large-scale adoption and sustainability. To respond to these challenges, the study proposes an integrative conceptual model that repositions AI not merely as a technological tool but as an ethically grounded and contextually adaptive agent within curriculum governance. Such a model emphasizes that the meaningful and equitable application of AI requires strong cross-sectoral collaboration, coherent policy alignment, and sustained capacity-building initiatives. By advancing this perspective, the study underscores the importance of positioning AI as a catalyst for inclusive and transformative educational change, ensuring that technological innovation aligns with ethical imperatives and local contextual needs.

**Keywords:** Artificial Intelligence; Curriculum Governance; Data-Driven Decision Making; Educational Management; Primary Education

## 1. Introduction

The rapid advancement of Artificial Intelligence (AI) has fundamentally reshaped the global educational landscape, positioning AI as a core strategic driver in educational management rather than a peripheral administrative tool. In the current era of data-driven governance, AI enables personalized learning at scale through the automation of learning analytics, real-time academic performance monitoring, and adaptive curriculum design recommendations (Luckin & Holmes, 2016). This paradigm shift reflects a move away from rigid, standardized decision-making models towards algorithmically informed frameworks that are more responsive, context-sensitive, and learner-centered. AI integration has further improved the efficiency, transparency, and accountability of educational decision-making processes, particularly in areas such as curriculum development, assessment, and instructional planning (Liu et al., 2023; Ng, 2015; Rochaendi et al., 2024).

However, while significant advancements have been documented in high-resource educational systems, a critical gap remains in understanding and implementing AI-based

curriculum governance in developing contexts, such as Indonesia. Existing literature tends to focus on technical affordances and policy-level adoption in advanced economies, often neglecting the socio-technical challenges faced by under-resourced primary education systems. These include inadequate infrastructure, limited digital literacy among educators, the absence of AI-specific national curricular policies, and the ethical risks of dehumanizing the learning process (Ahmad et al., 2023; Sallai et al., 2024). Furthermore, many studies have not sufficiently addressed the interplay between AI-driven decision-making and the holistic development of learners, particularly at the foundational education level.

This article seeks to address the existing conceptual and contextual gaps by critically examining the role of artificial intelligence (AI) in supporting managerial decision-making within primary education. Specifically, it focuses on how AI can inform curriculum design that is not only adaptive and data-responsive but also ethically grounded and inclusive. Building upon the recommendations of Farhan et al. (2024) and van Noordt et al. (2025), this review proposes a synthesized framework that integrates AI capabilities with pedagogical intentionality, equity considerations, and long-term educational outcomes. This academic exploration is crucial because, without a comprehensive understanding, the adoption of AI risks exacerbating educational inequalities, introducing algorithmic biases, or marginalizing humanistic values in curriculum processes. Therefore, this article aims to provide strategic insights for policymakers, educators, and researchers to navigate both the opportunities and constraints of AI implementation in curriculum governance for primary education systems across diverse and evolving contexts (Ariyani et al., 2024).

### **1.1. Problem Statement**

The core issue is the disjunction between AI's transformative potential and systemic readiness—especially in primary education. As AI tools permeate sectors, their advancement necessitates redesigned curriculum frameworks. AI is reshaping learning modalities and the decision-making behind curriculum development, demanding more dynamic, data-driven, and adaptive approaches. Although widely acknowledged as enabling data-informed planning, early need identification, adaptive instruction, and predictive policy (Holmes et al., 2019), practical application remains aspirational in many contexts. (Damodaran & Kanwar, 2025) note persistent constraints—limited digital infrastructure, insufficient teacher data literacy, and incoherent regulatory frameworks—hindering meaningful adoption, particularly in developing countries. These impediments are also institutional: policy misalignment, leadership gaps, and human-capital shortfalls. Disparities in access, concerns about algorithmic bias, and the inadequate integration of AI competencies in teacher training (Al-Zahrani, 2024) exacerbate the risk that AI will be confined to superficial administrative tasks, missing its promise for curriculum reform and pedagogical innovation (OECD, 2024).

This study addresses these challenges via a comprehensive literature review on AI integration in primary-level curriculum design. It illustrates how AI can support curriculum development through data analytics, personalized pathways, predictive modeling of student outcomes, and continuous evaluation. Anchored in Indonesia's context, it seeks a conceptual model positioning AI as a transformative mechanism for equitable, inclusive, and ethically grounded curriculum governance. By examining enabling and constraining factors—highlighted by the (World Economic Forum., 2024) and other global frameworks—the study offers strategic insights for policymakers, educators, and researchers navigating AI integration in foundational education (Rochaendi, 2025).

### **1.2. Related Research**

The current body of literature on Artificial Intelligence (AI) in education has provided essential foundational insights, particularly regarding AI's role in supporting data-informed teaching and learning processes. Luckin & Holmes (2016), for example, conceptualize AI as both an intelligent tutoring system and a learning analytics tool, designed to facilitate personalized instruction by bridging the gap between pedagogical decisions and student performance data. While this contribution is significant, it remains predominantly situated within the instructional domain, offering limited exploration of AI's potential role in higher-order

educational governance, particularly in the areas of curriculum development and strategic educational planning.

A similar focus on classroom-level application is evident in the evaluation of the Eduten Playground platform implemented in Finland. Thrupp et al. (2023) report that AI technologies can effectively personalize mathematics instruction, leading to increased learner engagement and improved academic performance. However, despite these positive outcomes, the analysis—much like that of Luckin & Holmes (2016)—remains confined to the micro-level of teaching practice. Both studies stop short of addressing the systemic implications of AI, particularly its integration into curriculum governance structures and broader institutional decision-making processes. This narrow scope is also reflected in the findings of Lee et al. (2022), who examine the deployment of AI-Tutor systems in South Korea. Their research highlights the technological efficiencies gained in content delivery and formative assessment but does not extend to consider AI's strategic contributions to curriculum policymaking or educational management at the systemic level.

In contrast, more recent academic discourse has begun to adopt a wider perspective, positioning AI not only as a tool for instruction but as a key component in educational governance. Ejjami (2024) advocates for AI-driven curriculum models that transcend mere personalization, emphasizing the integration of equity considerations to ensure inclusive access for diverse learner populations. In line with this, Chen et al. (2020) emphasize that AI can streamline administrative processes and support real-time curriculum adaptation, aligning instructional content dynamically with continuous student feedback and learning analytics. Further advancing this discussion, Kreinsen & Achulz (2023) argue that ethical and pedagogical integrity in AI integration depends on enhancing teachers' digital literacy. They assert that professional development initiatives must be systematically designed to equip educators with the competencies necessary to critically engage with AI technologies, ensuring responsible implementation within both instructional and managerial contexts.

Building upon these evolving perspectives, the present study introduces a novel conceptual approach by repositioning AI not merely as an instructional enhancement, but as a strategic instrument for data-driven curriculum governance in primary education. Distinct from prior research that tends to focus on classroom-level applications or isolated technological functions, this study aims to construct a comprehensive framework that articulates AI's managerial role in curriculum design—particularly within the socio-technical and policy contexts of developing countries such as Indonesia. In alignment with Baidoo-Anu & Ansah (2023), who emphasize the potential of generative AI to create dynamic and responsive learning environments, and Ejjami (2024), who foregrounds the ethical dimensions of AI in curriculum models, this research contributes by integrating pedagogical, managerial, and ethical considerations into a unified model. The proposed framework seeks to support inclusive and adaptive curriculum governance, positioning AI as both a technological innovation and a transformative agent for educational equity in the digital era.

### 1.3. Research Objectives

This study seeks to undertake a comprehensive and systematic examination of the role of Artificial Intelligence (AI) as a strategic enabler of data-informed curriculum governance in primary education, with particular emphasis on under-resourced contexts such as Indonesia. The research is guided by three interrelated objectives: first, to identify and critically analyze both the enabling conditions and systemic constraints that influence the implementation of AI in curriculum planning at the foundational education level; second, to synthesize empirical findings and theoretical insights from existing literature that elucidate AI's capacity to support managerial decision-making in education; and third, to develop a conceptual framework for AI-based curriculum design that is pedagogically responsive, socially inclusive, and ethically grounded. Through the articulation of these objectives, this study aims to bridge the persistent divide between theoretical discourse and practical application, offering evidence-based guidance for policymakers, educational leaders, and curriculum designers operating within the complexities of emerging education systems. The central research question underpinning this inquiry is: *In what ways can AI be effectively integrated into primary education curriculum*

*governance to promote responsiveness, equity, and ethical accountability in decision-making processes?*

## **2. Theoretical Framework**

### **2.1. Data-Driven Decision Making (DDDM) in Educational Management**

This study grounds itself in Data-Driven Decision Making (DDDM), which holds that educational decisions should rely on accurate, systematic, context-sensitive data to ensure efficacy and equity. Within this paradigm, Artificial Intelligence (AI) acts as a transformative agent by automating data collection, analyzing learning trajectories, and producing prescriptive insights for curriculum policy (Prakash, 2024). Yet effective DDDM also requires the convergence of AI literacy, pedagogical discernment, and curricular flexibility (Southworth et al., 2023). In developing contexts such as Indonesia, infrastructural limitations and capacity gaps hinder schools from fully leveraging AI for data-informed curriculum governance.

Compounding these challenges, advanced AI models—such as game-theoretic fog computing—remain underutilized in digitally underprepared systems (Han et al., 2024). Successful DDDM depends not only on technological access but also on ethical and algorithmic literacy among educators and policymakers (Walter, 2024a). AI-based governance, therefore, must be embedded within inclusive, ethically grounded, and culturally responsive structures (Molina et al., 2024). Accordingly, this study treats DDDM as a normative lens intersecting principles of justice, human-centered design, and adaptive policymaking, recasting AI as a collaborative actor for advancing equity and pedagogical relevance in primary education systems.

### **2.2. Adaptive Learning and Personalization**

The Theory of Adaptive Learning and Personalization posits that effective education must be responsive to individual learners' cognitive profiles, capacities, and developmental trajectories. Within this framework, AI functions as a central enabler of personalized environments: through learning analytics, intelligent assessment systems, and machine learning, it supports continuous tracking of progress and preferences for data-informed, real-time adaptation (X. Chen, 2023; Fahimirad & Kotamjani, 2018; Meylani, 2024). Empirical studies show AI can tailor content, feedback, and difficulty to performance and motivational indicators (Ayodele et al., 2023; Miao et al., 2024; Miao & Cukurova, 2024), constructing individualized pathways that loosen standardized rigidity, though implementation remains uneven in under-resourced contexts with gaps in infrastructure, teacher readiness, and contextual customization.

Beyond technical and pedagogical design, AI-enabled personalization demands ethical and equity-centered scrutiny. Algorithmic personalization informed by biased data or opaque logic can reproduce inequities, risks heightened in primary education (Baker & Hawn, 2021; Holmes et al., 2019; Zacharis et al., 2024). Robust governance is therefore required to ensure data privacy, transparency, and inclusive design; equitable AI depends not only on technological sophistication but also on shared ethical standards, teacher empowerment, and sustained institutional capacity building (Holmes et al., 2022). This study thus frames adaptive learning as both pedagogical innovation and a socio-technical, ethical undertaking.

### **2.3. Theory of Artificial Intelligence-Based Decision Support Systems (AI-Based DSS)**

The theory of Artificial Intelligence-Based Decision Support Systems (AI-Based DSS) casts AI as a strategic actor that structures and synthesizes data for curriculum design and institutional planning. By integrating historical data with predictive analytics, these systems help leaders anticipate learning demands and generate prescriptive recommendations for proactive adjustments (Bengesi et al., 2024; Strielkowski et al., 2024). In primary education, AI thus moves beyond operational support toward foresight-oriented management, with tools like ClassCharts and generative models such as ChatGPT illustrating real-time diagnostics and scenario-based decision modeling (Tan et al., 2025). In low-resource contexts like Indonesia,

however, infrastructural constraints, limited technical expertise, and disjointed policy integration impede implementation.

Despite technical promise, AI-based DSS require ethical, epistemological, and political scrutiny. Taylor (2024) warns that equity and efficacy depend on transparent, fair algorithmic logic; without safeguards, systems may entrench inequities under the guise of objectivity. Bozkurt (2024) stresses ethical oversight and inclusive stakeholder engagement to prevent technocratic marginalization of pedagogical values. As Gabay & Funa (2025) argue, sustainable integration needs a holistic model—combining technological design, teacher training, participatory data governance, and culturally embedded value systems—reframing AI as a central pillar of ethically responsive, pedagogically aligned educational leadership.

#### **2.4. Contextual Curriculum Design**

Contextual curriculum design in primary education aligns content with learners' socio-cultural, economic, and technological realities, in contrast to standardized models. Tuomi (2018) emphasizes that contextual learning enables knowledge construction through real-world engagement, while Gligorea et al. (2023) assert that acquisition deepens when new concepts connect to prior experiences, local wisdom, and everyday social realities. This orientation is crucial in primary education, where foundational learning shapes cognitive, emotional, and social development.

Persistent barriers complicate AI-enabled contextualization: institutional rigidity, limited school autonomy, and gaps in teachers' digital literacy. Wang et al. (2025) stresses that success hinges on supportive policies that encourage local innovation, equitable infrastructure, and continuous professional development. Consistent with UNESCO's call for participatory curriculum development, involving teachers, communities, and policymakers fosters relevance, inclusion, and social responsiveness (Giannini, 2023).

This study therefore frames contextual design as a proactive strategy for inclusive, equitable, and culturally anchored ecosystems. AI is treated as an emergent contextual phenomenon reshaping access to knowledge, identification of diverse needs, and data-informed decisions (Ayodele et al., 2023). Empirical work such as Tan et al. (2025) shows AI-powered recommendation engines and predictive analytics personalizing content by learning styles, prior knowledge, and socio-cultural contexts, turning curriculum into a living, adaptive construct responsive to dynamic learner profiles and community realities.

Moreover, Funa et al. (2022) and Gabay & Funa (2025) underscore AI's role in advancing equity through differentiated strategies that address marginalization and disparities. AI-driven insights help design targeted interventions for underrepresented or disadvantaged groups, ensuring access aligned with lived experiences. This echoes Lyublinskaya & Du (2025), who argue that contextual curriculum must bridge national policy and classroom realities. Integrating AI into contextual curriculum design thus represents a strategic pedagogical shift toward long-term transformation that prioritizes inclusivity, cultural relevance, and ethical governance in primary education.

#### **2.5. Technology Ethics in Education**

The Theory of Technology Ethics in Education provides a normative basis for integrating Artificial Intelligence (AI) into curriculum governance, particularly in primary education where ethical concerns meet learner vulnerability. Central is safeguarding student rights—data privacy, non-discrimination, informed consent—so AI augments rather than replaces human agency. As Luckin & Holmes (2016) argue, AI should reinforce teachers' pedagogical roles and preserve the relational fabric of classrooms. Ethical deployment thus requires technical safeguards against algorithmic bias and systems designed for inclusion, transparency, and shared responsibility. AI is reframed as a co-intelligent partner in human-centered learning, emphasizing pedagogical co-agency over automation.

Yet implementation gaps persist. Tsai et al. (2019) and Holmes et al. (2019) warn that AI without ethical scaffolding can entrench discrimination, especially in unequal systems. While “*trustworthy AI*” frameworks promote fairness and accountability, practical adoption remains

limited. The Council of Europe (2024) advocates binding regulations grounded in democratic values, but many nations—particularly in the Global South—lack infrastructure for audits, oversight, and educator readiness. Farhan et al. (2024) and Ayodele et al. (2023) stress that without inclusive governance, AI risks control rather than empowerment. This study therefore centers technology ethics to uphold justice, participation, and learner dignity.

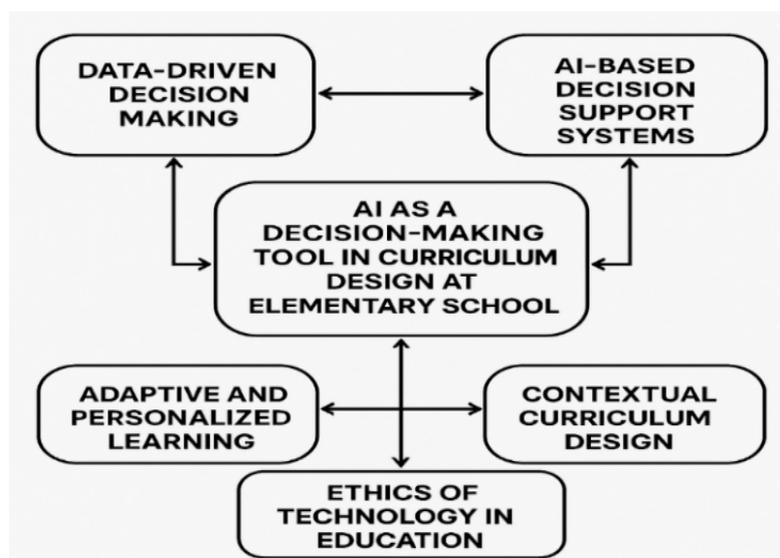
## 2.6. Conceptual Framework

The article's conceptual framework rests on five interrelated pillars that position AI as a multidimensional agent in curriculum governance, from optimizing instructional relevance via data analytics (L. Chen et al., 2020; Luckin & Holmes, 2016) to supporting ethical, contextualized decision-making (Ejjami, 2024; Kreinsen & Achulz, 2023). It reframes AI as a strategic, pedagogical, and ethical force for adaptive, learner-centered, socially just primary education ((Baidoo-Anu & Ansah, 2023). AI mediates between classroom complexity and policy through real-time insights, predictive analytics, and responsive recommendations aligned with primary learners' developmental and contextual needs (Tan et al., 2025). Yet translation into diverse schools—especially in Indonesia—remains limited by infrastructural inequities, policy fragmentation, and uneven AI literacy (Ayodele et al., 2023).

As lenses on affordances and limits, the five pillars show: DDDM enables continuous feedback and diagnostics but lacks institutional mechanisms for sustainability (L. Chen et al., 2020; Liu et al., 2023); Adaptive and Personalized Learning promises tailored pathways, yet scalability is constrained by infrastructure and rigid mandates (Lee et al., 2022; Thrupp et al., 2023); AI-DSS extends to high-level decision-making and policy simulation (Zawacki-Richter et al., 2019), though analytic capacity is scarce ((Ayodele et al., 2023); Contextual Curriculum Design fosters social and cultural responsiveness (Baidoo-Anu & Ansah, 2023; Ejjami, 2024), but top-down standardization limits local adaptation (Gabay & Funa, 2025); Technology Ethics in Education sets boundaries for transparency, data justice, and privacy (Holmes et al., 2019; Kreinsen & Achulz, 2023), with persistent practice gaps (Gabay & Funa, 2025).

Collectively, these theories form an integrated framework situating AI at the intersection of innovation and accountability, pointing toward curriculum policies that are intelligent, responsive, equitable, and ethically grounded.

The conceptual visualization of this framework is illustrated as follows and can be observed in Figure 1, which systematically presents the key dimensions, theoretical underpinnings, and interconnections that form the foundation of the proposed mode: I



**Figure 1.** of Conceptual Framework: Artificial Intelligence in Educational Management: A Systematic Literature Review on AI-Based Primary Curriculum Design

### 3. Method

#### 3.1. Research Design

This study, titled “Artificial Intelligence in Educational Management: A Systematic Literature Review on AI-Based Primary Curriculum Design”, uses a qualitative, integrative review—exploratory and analytical—to synthesize theoretical, empirical, and conceptual developments on AI in primary curriculum governance. Following Xiao & Watson (2017), it applies transparent identification, categorization, and appraisal via predefined criteria, while a hermeneutic approach (Boell & Cecez-Kecmanovic, 2015) enables iterative interpretation and avoids single-method fragmentation.

Sources came from Scopus, Web of Science, and ScienceDirect, prioritizing the last decade with emphasis on the past five years (Ayodele et al., 2023; Baidoo-Anu & Ansah, 2023; Zawacki-Richter et al., 2019). Focusing on data-driven decision-making, learning analytics, adaptive learning, contextual curriculum design, and ethics, the review identifies key patterns and gaps through an iterative, transparent process, offering a foundation for ethical, context-sensitive, AI-informed curriculum.

#### 3.2. Participant

In this review, “participants” are scholarly sources—peer-reviewed articles, conference papers, institutional reports, and monographs—serving as units of analysis. Following Snyder (2019), the literature is treated as active data under a structured methodology with inclusion/exclusion criteria. The corpus focuses on publications from the past decade about AI in educational decision-making, emphasizing primary curriculum governance.

A thematic-analytical and hermeneutic approach guided iterative interpretation (Boell & Cecez-Kecmanovic, 2015), ensuring rigor and transparency per Booth et al. (2016). PRISMA guidelines structured identification, screening, and synthesis (Page et al., 2021), enabling synthesis and critical reflection. Table 1 summarizes the 52 reviewed documents by type, year, and thematic focus, analogous to participant profiling.

**Table 1.** Distribution of Reviewed Documents

Category	Sub-Category	Number of Documents (n=52)
Document Type	Journal Articles	38
	Conference Proceedings	7
	Institutional Reports	4
	Academic Monographs	3
Year of Publication	2015-2017	10
	2018-2020	17
	2021-2023	25
Main Themes	Data-Driven Decision Making (DDDM)	14
	Adaptive & Personalized Learning	12
	AI-Based Decision Support Systems	10
	Contextual Curriculum Design	9
	Technology Ethics in Education	7

#### 3.3. Data Collection

Data for this study were obtained through a structured literature review of sources from Scopus, Web of Science, ScienceDirect, and SpringerLink, limited to 2020–2025 except seminal works. Following Carrera-Rivera et al. (2022), stages included protocol development, inclusion/exclusion criteria, article selection, quality appraisal, and thematic synthesis. An integrative protocol consolidated conceptual and empirical literature. Keyword searches—“Artificial Intelligence in education,” “data-driven curriculum design,” “educational decision-making,” “learning analytics”—ensured thematic precision. The process adhered to traceability and auditability principles outlined by Booth et al. (2016) and, consistent with

Snyder (2019), exemplifies methodological rigor, scholarly depth, and ethical integrity in examining AI and curriculum governance.

### 3.4. Data Analysis

This study used an integrative synthesis aligned with a qualitative systematic literature review. Peer-reviewed sources were retrieved from Scopus, Web of Science, ScienceDirect, and SpringerLink using strategic keywords ("Artificial Intelligence in education," "data-driven curriculum design," "educational decision-making," "learning analytics"). The review followed a structured protocol of identification, screening, eligibility assessment, and critical appraisal.

NVivo 14 supported thematic coding to organize data, extract recurring patterns, and identify conceptual linkages. Following Mohammed & Ahmed (2021), the analysis integrated conceptual and empirical findings into a unified model of AI-based curriculum governance through iterative coding cycles. Consistent with Hart (2018), all steps were documented for replicability and scholarly integrity, mapping current knowledge, revealing research gaps, and proposing an ethical, adaptive, data-driven framework for primary education.

### 3.5. Validity and Reliability

This study ensured reliability and validity via a systematic literature review following a structured protocol and focusing—unlike surveys or interviews—on recent peer-reviewed publications from Scopus, Web of Science, and ScienceDirect, emphasizing conceptual rigor and empirical relevance to AI in educational management (Snyder, 2019). The review adhered to PRISMA for methodological consistency, replicability, and transparency (Page et al., 2021) and applied triangulation by integrating qualitative, quantitative, and mixed-method studies. NVivo 14 supported analysis for coding consistency, systematic theme identification, and bias minimization, enhancing inter-coder reliability and thematic saturation (Booth et al., 2016; Kraus et al., 2022), yielding a validated, empirically grounded framework for ethical, adaptive, data-driven curriculum governance in primary education.

## 4. Finding

### 4.1. Leveraging Prescriptive Analytics: The Strategic Role of AI in Curriculum Decision-Making for Primary Education

The findings highlight the strategic potential of Artificial Intelligence (AI) to strengthen data-informed curriculum governance in primary education by leveraging prescriptive analytics to monitor performance, detect learning patterns, and recommend timely pedagogical interventions. In doing so, AI shifts curriculum decision-making from a reactive, administratively driven model to a proactive, evidence-based process. Compared with traditional approaches that depend on periodic evaluations and static reports, AI enables educators and policymakers to address learning disparities in real time at both individual and classroom levels. A comparative analysis of selected studies (see Table 2) shows that AI-driven systems—such as Intelligent Tutoring Systems (Luckin & Holmes, 2016) and AI Dashboard Analytics (Ng, 2015)—are markedly more effective at identifying differentiated learning needs than conventional manual or routine reporting methods. Table 2 synthesizes the contrasts between traditional and AI-enabled approaches to curriculum decision-making.

**Table 2.** Comparative Features of Traditional vs. AI-Based Curriculum Decision-Making

Dimension	Traditional Approach	AI-Based Approach
Data Source	Periodic/manual evaluation reports	Real-time analytics and behavioral learning data
Response Time	Slow (monthly or term-based)	Immediate (daily or even instantaneous)

Intervention Accuracy	General and non-specific	Targeted and individualized pattern-based insights
Adaptive Capacity	Limited and standardized	Dynamic and context-sensitive to individual learner needs
Decision-Making Basis	Intuition or fixed procedural guidelines	Predictive data models and prescriptive analytics

As illustrated in Table 2, AI-based systems not only enhance the efficiency and precision of curriculum-related decisions but also open new avenues for inclusive, adaptive, and learner-centered instructional design grounded in authentic educational needs.

#### 4.2. Structural Barriers to AI Integration in Curriculum Planning: Challenges in Developing Country Contexts

Despite growing recognition of Artificial Intelligence (AI) in curriculum governance, implementation in developing countries such as Indonesia remains constrained by inadequate digital infrastructure, low educator data/AI literacy, and the absence of coherent national policies at the primary level. In contrast to high-income contexts where AI is embedded through institutionalized frameworks and robust technological ecosystems, resource-limited settings exhibit fragmented, underdeveloped enabling conditions. Evidence shows countries like Singapore and South Korea advancing via coordinated national strategies and substantial investment in digital ecosystems (Kim et al., 2021; Ng, 2015). Indonesia and similar systems face systemic obstacles that inhibit scalable AI adoption. Table 3 outlines enabling and constraining factors.

**Table 3.** Comparative Matrix: AI Implementation Readiness in Curriculum Planning

Dimension	High-Income Countries (e.g., South Korea, Singapore)	Developing Countries (e.g., Indonesia)
Digital Infrastructure	Advanced, equitable, and broadband-accessible	Uneven, limited, especially in rural areas
Educator AI/Data Literacy	Supported through continuous training and CPD	Low, with limited professional development
Policy Frameworks	Comprehensive, explicit AI integration strategies	Fragmented or absent for primary education
Institutional Support Systems	Strong inter-agency coordination and AI governance	Weak coordination across educational sectors
Financial Investment	High and sustained government funding	Dependent on donor support or pilot programs

This comparative analysis underscores that successful AI integration is not solely a matter of technological readiness but requires systemic alignment across infrastructure, human capacity, and policy. Without addressing these foundational barriers, attempts to implement AI in curriculum planning risk exacerbating educational inequalities rather than resolving them.

### 4.3. Enhancing Managerial Decision-Making through AI: Data-Driven Policy Simulations and Real-Time Curriculum Evaluation

A growing body of literature affirms that Artificial Intelligence (AI) enhances managerial decision-making in education by enabling data-driven policy simulations, real-time curriculum evaluation, and nuanced mapping of student learning needs. Unlike traditional practices reliant on delayed, aggregated, and often subjective reports, AI systems provide immediate predictive models, actionable insights, and dynamically generated scenarios. These capabilities help leaders anticipate disruptions, design differentiated strategies, and refine curriculum alignment using empirical evidence rather than intuition. Comparative implementations show AI tools—adaptive dashboard analytics, policy-simulation engines, algorithmic scenario planners—offer greater responsiveness, granularity, and decision reliability than conventional instruments. Table 4 outlines differences between traditional and AI-enhanced decision-making.

**Table 4.** Comparative Features: Traditional vs. AI-Enhanced Managerial Decision-Making in Education

Decision-Making Function	Traditional Approach	AI-Enhanced Approach
Policy Scenario Modeling	Rarely used, manual projections	Algorithmic simulations based on real-time datasets
Curriculum Evaluation	Periodic, summative, lagged feedback	Continuous, formative, real-time evaluation mechanisms
Learning Needs Mapping	Based on teacher observations	Based on learning analytics and behavioral pattern mining
Data Accessibility	Centralized, aggregated, delayed	Distributed, disaggregated, real-time
Decision Accuracy	Dependent on human interpretation	Augmented by machine-generated recommendations

This comparison highlights the epistemological shift from intuition-based to evidence-driven educational governance enabled by AI. As education systems become increasingly complex and data-saturated, the integration of AI into managerial frameworks offers not only technical efficiency but also the potential for more transparent, equitable, and adaptive curriculum planning.

### 4.4. AI as a Pedagogical and Social Agent: Advancing Personalized, Contextualized, and Differentiated Curriculum Design

AI functions not merely as a technological tool but as a pedagogical and sociocultural agent, reshaping curriculum toward personalized, contextualized, and differentiated designs. It engages learner diversity—cognitive profiles, socio-economic backgrounds, linguistic contexts, cultural identities—and adapts in real time, delivering personalized feedback, tailored content, and differentiated trajectories, surpassing uniform pacing and generalized delivery in conventional models.

Comparative studies show traditional curriculum design remains top-down and standardized, whereas AI-based frameworks use adaptive algorithms and localized data to craft individualized pathways and culturally relevant structures. Table 5 contrasts these approaches

across key educational dimensions, highlighting the pedagogical transformation enabled by AI-augmented models.

**Table 5.** Pedagogical Roles: Traditional Curriculum Design vs. AI-Augmented Curriculum Design

Pedagogical Dimension	Traditional Curriculum Design	AI-Augmented Curriculum Design
Instructional Personalization	Minimal; one-size-fits-all	High; individualized based on learner analytics
Cultural Contextualization	Limited; often generic or nationalized	Strong; localized content derived from contextual data
Differentiation Strategy	Based on teacher judgment	Algorithmically generated pathways based on performance
Learner Engagement	Passive consumption	Active interaction via adaptive platforms
Responsiveness to Diversity	Reactive and generalized	Proactive and targeted based on socio-demographic data

This comparative framework reinforces the argument that AI, when ethically and contextually embedded, enables a paradigm shift toward pedagogically just and socially inclusive curriculum design. It transforms educational practice from standardized delivery to dynamic, learner-responsive ecosystems—redefining how schools conceptualize equity, inclusion, and instructional relevance in the digital age.

#### 4.5. Ethical Imperatives in AI Integration: Ensuring Fairness, Transparency, and Data Justice in Educational Governance

The ethical dimension is foundational to integrating AI in education amid concerns over algorithmic opacity, student data protection, equitable access, and systemic bias. As AI increasingly shapes curriculum decisions and personalization, governance must prioritize ethical accountability, inclusivity, and democratic participation; unlike conventional technologies, AI's autonomous interaction with complex datasets heightens design and implementation scrutiny.

Comparative analyses show uneven ethical preparedness: high-income nations such as Finland and Canada have institutionalized AI ethics through national digital charters and AI-specific educational guidelines, whereas many developing countries, including Indonesia, lack comprehensive legal instruments or oversight mechanisms. A comparative overview maps key ethical dimensions and their institutionalization levels.

**Table 6.** Ethical Dimensions in AI Integration: A Comparative Perspective

Ethical Dimension	High-Preparedness Systems (e.g., Finland, Canada)	Low-Preparedness Systems (e.g., Indonesia, Ghana)
Algorithmic Transparency	Legally mandated algorithm audits and open standards	Limited visibility; black-box models dominate

Data Privacy Protection	Strong child data protection and consent regulations	Weak or fragmented data policies in school systems
Equity of Access	National strategies for digital inclusion	Digital divides remain unresolved, especially in rural areas
Bias Mitigation Mechanisms	Institutional bias audit protocols	Absence of systemic monitoring frameworks
Participatory Governance	Stakeholder inclusion in AI policy formation	Top-down implementation with minimal educator involvement

This comparative analysis underscores the urgent need for ethically grounded AI governance, particularly in contexts where technological enthusiasm risks outpacing regulatory maturity. Without robust ethical frameworks, the promise of AI to enhance education may instead perpetuate existing inequities. Thus, inclusive, transparent, and participatory policy architectures are not optional, but essential prerequisites for building trust and legitimacy in AI-enhanced educational ecosystems.

**4.6. Toward an Ethical and Adaptive AI-Based Curriculum Governance: An Integrative Conceptual Framework**

This study proposes an integrative conceptual framework synthesizing five interlocking theoretical perspectives—Data-Driven Decision Making (DDDM), Adaptive and Personalized Learning, AI-Based Decision Support Systems (AI-DSS), Contextual Curriculum Design, and Technology Ethics in Education—to build a multidimensional model of curriculum governance grounded in ethical, adaptive, and inclusive principles. Positioned as mutually reinforcing, these logics enable AI to function as an instrument of educational transformation by integrating data analytics, pedagogical differentiation, managerial simulation, socio-cultural sensitivity, and normative accountability in primary education systems.

A comparative analysis of each pillar's strengths and orientations (see Table 6) shows their convergence provides robust guidance for AI adoption in diverse settings. Distinct from models isolating technological affordances or pedagogical aims, this holistic framework bridges epistemic innovation, ethical restraint, and local responsiveness within one governance logic. Table 7 offers a comparative overview.

**Table 7.** Comparative Overview of the Five Theoretical Pillars in the Conceptual Framework

Theoretical Pillar	Core Contribution	Practical Focus Area	Ethical Considerations Included
Data-Driven Decision Making	Enables evidence-based curriculum decisions	Learning analytics, gap identification	Partial (relies on data integrity)
Adaptive and Personalized Learning	Supports differentiated learning trajectories	Instructional adaptation, learner profiling	Equity through personalization
AI-Based Decision Support Systems	Assists policy simulation and scenario planning	Strategic governance,	Limited unless paired with ethics frameworks

		resource optimization	
Contextual Curriculum Design	Aligns curriculum with local socio-cultural needs	Content localization, cultural integration	Emphasizes relevance and learner dignity
Technology Ethics in Education	Safeguards against bias and misuse of AI	Data privacy, algorithmic fairness, inclusion	Central (guides responsible AI application)

This integrated framework contributes to a more nuanced and operationalizable understanding of AI in curriculum governance—one that resists reductionism and instead embraces the pedagogical, managerial, contextual, and ethical dimensions as equally vital. As such, it provides a comprehensive foundation for educational leaders, policymakers, and researchers seeking to navigate the socio-technical complexities of AI-driven transformation in primary education.

#### 4.7. Bridging the Gap: Addressing the Disjunction Between AI's Conceptual Potential and Institutional Readiness in Education

This study finds a persistent gap between AI's conceptual promise in educational governance and institutions' readiness to implement it. While AI offers personalization, optimized curriculum management, and policy responsiveness, many low- and middle-income countries remain ill-prepared. Constraints include fragmented regulation, underdeveloped infrastructure, weak inter-agency coordination, and limited professional development in AI competencies. Bridging this disjunction requires a cross-sectoral approach uniting education, technology, policy, and civil society in a cohesive governance ecosystem.

Comparative assessments (see Table 8) show successful AI integration hinges on systems-level alignment among legal, institutional, and human-resource capacities, not merely technological access. Table 8 contrasts high policy-operational coherence with fragmented adoption frameworks.

**Table 8.** Comparative Readiness for AI Implementation: Conceptual Potential vs. Institutional Capacity

Country/Context	Conceptual Embrace of AI	Institutional Readiness	Nature of the Gap	Collaborative Mechanisms in Place
Finland	High	High	Minimal gap; theory and practice aligned	Strong cross-sectoral policy integration
South Korea	High	Moderate–High	Addressed through adaptive digital strategies	Government–industry–education synergy
Indonesia	Moderate–High	Low	Wide gap; policy lags behind conceptual vision	Emerging multi-stakeholder pilot projects
Nigeria	Moderate	Low	Structural gaps in policy and infrastructure	Lacks sustained collaborative platforms

The findings suggest that realizing AI's transformative potential requires more than isolated technological innovation; it demands a systems-thinking paradigm that mobilizes cross-sector collaboration, aligns national digital education strategies with ethical frameworks, and institutionalizes AI literacy as a core component of teacher training. In doing so, educational ecosystems can transition from aspirational rhetoric to sustainable, equity-driven AI integration—ensuring that the power of technology is harnessed not selectively, but inclusively and responsibly.

## 5. Discussion

### 5.1. From Reactive Bureaucracies to Predictive Intelligence — Reframing Curriculum Governance through AI

AI—especially prescriptive analytics—has begun transforming curriculum decision-making in primary education by shifting governance from delayed, cyclical assessments to proactive intelligence via real-time data synthesis and predictive modeling (Holmes et al., 2019; Luckin & Holmes, 2016). Tools such as Intelligent Tutoring Systems and dashboard analytics support early detection of learning gaps and personalized interventions (Ng, 2015).

Conceptually, AI promotes adaptive, learner-centered governance over standardized models common in the global South (Zacharis et al., 2024). Yet adoption in developing countries is constrained by infrastructure gaps, low educator data literacy, and weak AI policies (Ahmad et al., 2023; Sallai et al., 2024), so pilots thrive while large-scale implementation lags (Sperling et al., 2023).

Realizing AI's promise requires embedding prescriptive analytics within ethical, transparent policy frameworks that enable systemic capacity building (Farhan et al., 2024; van Noordt et al., 2025). Key concerns include algorithmic fairness, data privacy, and teacher agency (Sallai et al., 2024), alongside alignment with local pedagogical cultures to prevent techno-centric reductionism (Ejjami, 2024; Oluyemisi, 2023).

In sum, AI is a strategic tool for data-informed curriculum reform whose success depends on cross-sector collaboration, ethical governance, and infrastructure equity. Without these, AI risks remaining an exclusive solution in a fragmented system, so leaders should position it to enhance pedagogical responsiveness, learner diversity, and educational justice.

### 5.2. Structural Barriers to Equitable AI Integration in Curriculum Planning

Despite AI's theoretical promise in curriculum governance (Holmes et al., 2022; Luckin & Holmes, 2016), significant structural barriers persist in developing contexts like Indonesia: insufficient digital infrastructure, fragmented policies, and limited educator capacity. These conditions echo concerns that digitalization outpaces institutional readiness and widens inequities (Ahmad et al., 2023).

By contrast, South Korea and Finland have advanced through coordinated policies, infrastructure investment, and sustained professional development (Lee et al., 2022; Ng, 2015). Indonesia's absence of a centralized AI-in-education policy yields isolated pilots with weak scalability (Estrellado & Miranda, 2023). Without a national strategy rooted in local contexts, integration remains fragmented and reactive (Evans et al., 2021; UNESCO., 2023), constraining culturally responsive, pedagogically aligned AI-enabled curriculum development, especially in underserved regions (Nazaretsky et al., 2022).

Addressing these gaps aligns with calls for multi-dimensional transformation: expanding infrastructure, fostering cross-sector collaboration, and providing AI-focused teacher training that integrates technical, pedagogical, and ethical competencies (Farhan et al., 2024). Policy should shift from scattered digital initiatives to comprehensive AI governance frameworks with clear objectives, ethical standards, and scalable pathways, while contextualizing AI to local languages, cultures, and educational realities (Bengesi et al., 2024).

Unless these constraints are tackled through participatory, coherent policy design, AI integration will remain aspirational; governments should reconceptualize AI as a public good, embedded in national education agendas and co-owned by policymakers, educators, and communities.

### **5.3. AI-Supported Managerial Decision-Making — From Data Collection to Strategic Foresight in Curriculum Governance**

This study highlights AI's transformative role in managerial decision-making for curriculum governance, shifting from static data to dynamic, real-time responsiveness; AI-driven dashboards, predictive analytics, and simulation models help leaders monitor implementation, map needs, and test policy scenarios (Baker & Hawn, 2021; Holmes et al., 2022).

Unlike traditional processes dependent on delayed evaluations and annual reports, AI supplies predictive insights and real-time diagnostics for adaptive adjustments (Luckin & Holmes, 2016; Ng, 2015), elevating leaders from reactive administrators to data-informed strategists with greater precision and pedagogical awareness (Capraro et al., 2024; Hancock et al., 2024; Tuomi, 2018).

Effective use nonetheless hinges on institutional culture, digital readiness, and leadership capacity; without data literacy or ethical safeguards, AI may be underutilized or misapplied, risking technocratic determinism (Ahmad et al., 2023; Sallai et al., 2024).

To mitigate these risks, Farhan et al. (2024) urge rethinking educational leadership toward AI-informed competencies—ethical data interpretation, critical analysis, participatory decision-making. AI should support human leadership, not replace it. For Indonesia and similar contexts, this entails infrastructure development, ethical data governance, and leadership training emphasizing accountability, cultural sensitivity, and adaptability (Baidoo-Anu & Ansah, 2023).

In summary, AI expands managerial decision-making capacity, but its impact depends on reconceptualized leadership; without systemic investment in ethical, human, and institutional capabilities, AI's potential remains automation rather than meaningful pedagogical advancement.

### **5.4. AI as a Pedagogical and Social Agent — Personalization, Contextualization, and Inclusive Differentiation in Curriculum Design**

AI is not merely a computational tool but a pedagogical and sociocultural agent, reshaping curriculum design into a responsive, inclusive process. As Luckin & Holmes (2016) note, AI-driven systems act as co-intelligent platforms mediating learners' cognitive development and curricular structures. They enable personalized learning by adapting content in real time to performance and behavior (Funa et al., 2022; Gabay & Funa, 2025).

Beyond personalization, AI strengthens curricular contextualization by embedding local narratives, languages, and cultural values (Ejjami, 2024; Oluyemisi, 2023). In multicultural societies like Indonesia, it helps curricula reflect diverse identities (Thrupp et al., 2023; Walter, 2024b).

AI also supports differentiated instruction by detecting micro-patterns in engagement and frustration (Sedrakyan et al., 2018), adjusting not only what is taught, but how and for whom (Jiali et al., 2024; Nazaretsky et al., 2022; Wang et al., 2025), fostering inclusive excellence.

However, realizing this potential requires ethical, intentional design. Absent data justice, cultural sensitivity, and algorithmic transparency, AI can reinforce inequities (Al-Zahrani, 2024; Sallai et al., 2024). Participatory design with teachers, students, and communities is vital so AI becomes a relational partner, not a depersonalized optimizer (Gligorea et al., 2023).

In sum, AI's role in curriculum governance should shift from mechanistic enhancement to ethically embedded, human-centered learning innovation, demanding a paradigmatic rethinking of learning design in the era of intelligent systems.

### **5.5. Ethical Governance as a Foundational Imperative in AI-Based Curriculum Reform**

The meaningful integration of AI in curriculum governance requires a comprehensive ethical framework addressing datafication, automation, and algorithmic decision-making. Holmes et al. (2022) emphasize transparency, fairness, and explainability to avoid black-box conditions that alienate educators and erode accountability (Rutti-Joy et al., 2023).

Risks include automated bias, surveillance, and unjust classifications (Sallai et al., 2024). AI can reinforce structural inequalities, especially in primary education where early experiences shape lifelong development (Brazil et al., 2025; Du Boulay, 2023).

Mitigation demands a multidimensional approach combining ethical literacy, institutional safeguards, and participatory oversight. Taylor (2024) calls for national regulations with algorithmic audits and redress mechanisms (Chiu & Chai, 2020; Hancock et al., 2024; Miao & Cukurova, 2024; Walter, 2024a; Xu, 2025), while Evans et al. (2021) advocate empowering teachers to critically engage with AI outputs.

Teachers should act as co-designers so human values guide automated decisions, and policy should shift from top-down mandates to collaborative digital governance with students, teachers, parents, and civil society (Figaredo & Stoyanovich, 2023; Sperling et al., 2023). Without participatory frameworks, AI risks privileging efficiency and surveillance over learning diversity and dignity (Tao & Diaz-Prez, 2025).

In conclusion, ethical governance is foundational for just, pedagogically meaningful AI integration; ethical AI must go beyond compliance by embedding care, transparency, and justice into educational innovation.

### **5.6. Toward an Integrative Framework for AI-Based Curriculum Governance**

The sixth major finding proposes an integrative framework of five interrelated perspectives—Data-Driven Decision Making (DDDM), Adaptive and Personalized Learning, AI-Based Decision Support Systems (AI-DSS), Contextual Curriculum Design, and Technology Ethics in Education (Langeveldt, 2024; Liu et al., 2023; Owan et al., 2023). Addressing fragmented AI implementation in primary education, especially in under-resourced contexts like Indonesia's, it balances technological, pedagogical, managerial, contextual, and ethical dimensions. Aligned with UNESCO (2021), the framework synthesizes knowledge, surfaces gaps and directions, and repositions AI as an epistemic agent transforming curricula into responsive, data-informed, ethically grounded systems (Forum on Information & Democracy, 2024). Each pillar links evidence and theory: DDDM with real-time interventions (Prakash, 2024); Adaptive Learning with intelligent tutoring systems (L. Chen et al., 2020; Fahimirad & Kotamjani, 2018); AI-DSS with managerial decision support (Strielkowski et al., 2024); Contextual Curriculum with socio-cultural adaptation (Ayodele et al., 2023); and Technology Ethics with fairness, privacy, and democratic accountability (Holmes et al., 2019; Hutson & Ceballos, 2023; OECD, 2023; Pitrella et al., 2024; Schleicher, 2024). Practically, it guides policymakers and leaders to embed AI while preserving pedagogical relevance and ethical responsibility (Cheah et al., 2025; Haetami, 2025; Pawar, 2023), countering reductionism and offering a theoretical consolidation and practical roadmap for bridging algorithmic precision with humanistic education (Figaredo & Stoyanovich, 2023).

### **5.6. Bridging the Conceptual-Practical Gap: Toward Cross-Sectoral Collaboration in AI-Enabled Curriculum Governance**

The seventh finding underscores a gap between AI's conceptual potential in curriculum governance and institutions' practical readiness, especially in under-resourced systems. Despite promise for real-time analytics and data-driven design (Funa et al., 2022; Gabay & Funa, 2025), implementation remains fragmented, notably in Indonesia where absent national AI frameworks and limited digital readiness hinder scalability (Baker & Hawn, 2021; Fahimirad & Kotamjani, 2018; Kreinsen & Achulz, 2023).

Global and local analyses show success depends not only on technology but on coordinated efforts across leaders, infrastructure providers, teacher-training bodies, and regulators (Estrellado & Miranda, 2023; Molina et al., 2024). Disjointed regional adoption signals the need

for integrated, participatory governance; strategic alignment must go beyond regulation to enable inter-ministerial and cross-sector collaboration (Akintola et al., 2024; Damodaran & Kanwar, 2025).

To bridge this gap, the study proposes a multi-stakeholder collaboration framework engaging local policymakers, AI researchers, ed-tech startups, and civil society. Centered on inclusivity, transparency, and capacity building, it addresses the systemic neglect of education in national AI strategies (Schiff, 2022). Only sustained co-design among educators, technologists, and policymakers can translate AI's conceptual promise into equitable, context-sensitive curriculum reforms.

## 6. Conclusion

This study concludes that Artificial Intelligence (AI) holds transformative potential in primary education curriculum governance by enabling data-driven, adaptive, and contextually responsive decision-making. Through the synthesis of five key theoretical pillars—Data-Driven Decision Making, Adaptive Learning, AI-Based Decision Support Systems, Contextual Curriculum Design, and Technology Ethics in Education—this research provides a conceptual framework that repositions AI as a strategic and ethical co-actor in curriculum development.

The findings emphasize that realizing AI's potential requires more than technological adoption; it demands systemic readiness, ethical governance, and socio-cultural alignment, especially in contexts like Indonesia where infrastructure and policy integration remain limited. Therefore, this study offers a practical roadmap for policymakers, educators, and technologists to collaboratively implement AI in ways that prioritize equity, inclusivity, and long-term educational improvement.

## Limitation

This study—"Artificial Intelligence in Educational Management: A Systematic Literature Review on AI-Based Primary Curriculum Design"—has several limitations. First, exclusive reliance on secondary sources precludes field data, reducing contextual depth and external validity. Second, the primary-education focus limits generalizability to other levels. Third, methodological, epistemological, and regional heterogeneity across sources complicates thematic coherence and comparisons. Finally, because AI and policy evolve rapidly, conclusions are time-bound; without periodic updates, their applicability may erode as technologies, regulations, and pedagogies change over time significantly.

## Recommendation

Future research should address these limitations by employing mixed-method or qualitative fieldwork (e.g., interviews, focus groups, case studies) to generate empirical evidence, validate frameworks, and contextualize findings. Studies must extend beyond primary to secondary and tertiary levels to enhance generalizability. Standardized comparative criteria through meta-analysis or meta-synthesis can improve methodological consistency. Given AI's rapid evolution, longitudinal studies are essential to sustain relevance and strengthen inclusive, adaptive, and ethical governance in education.

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## Conflict of Interest

The Author(s) declare(s) that there is no conflict of interest.

## Declaration of Generative AI-assisted Technologies

The author acknowledges ChatGPT's contribution (≈15%) for drafting, refinement, and editing. All conceptual analysis, synthesis, interpretation, and revisions were independently completed. The author assumes full responsibility for originality and integrity.

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