



Integrating Artificial Intelligence In Language Acquisition: A Comparative Study Of Adaptive Learning Platforms In Indonesian Efl Classrooms

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Abstract: This study investigates the implementation and effectiveness of artificial intelligence-driven adaptive learning platforms in English as a Foreign Language (EFL) classrooms across urban and rural Indonesian educational settings. Through a mixed-methods approach involving 247 students and 12 teachers from eight institutions, this research comparatively analyzes three prominent AI language learning platforms: LinguaLearn AI, AdaptEd English, and AI Tutor Pro. The findings reveal significant differences in learning outcomes, with AI-enhanced platforms demonstrating a 27% improvement in language proficiency compared to traditional methods. However, implementation challenges including digital literacy barriers, internet connectivity issues, and cultural adaptation requirements were identified, particularly in rural contexts. The study proposes a contextually responsive framework for integrating AI language learning technologies in developing EFL environments, contributing to the growing discourse on educational technology in linguistically diverse settings. This research highlights the potential of AI to personalize language acquisition while emphasizing the necessity of culturally and infrastructurally appropriate implementation strategies.

Keywords: Artificial Intelligence, Adaptive Learning, EFL, Indonesian Education, Language Acquisition, Educational Technology

1. INTRODUCTION

The integration of Artificial Intelligence (AI) in educational contexts represents one of the most significant paradigm shifts in contemporary teaching and learning approaches. For English as a Foreign Language (EFL) instruction, AI-powered adaptive learning systems offer unprecedented opportunities to personalize learning experiences, identify individual learner needs, and provide targeted instruction (Yang & Warschauer, 2021). This potential is particularly relevant in linguistically diverse countries like Indonesia, where English language proficiency varies significantly across regions and socioeconomic boundaries (Lauder, 2022).

Indonesia's position as Southeast Asia's largest economy with a population exceeding 270 million presents unique challenges and opportunities for educational technology implementation. While urban centers increasingly embrace digital learning, rural areas often contend with infrastructure limitations and varying levels of technological readiness (Tanveer et al., 2023). The country's commitment to improving English proficiency as articulated in the national curriculum reforms creates an ideal context for examining how AI-enhanced language learning platforms perform across diverse educational environments.

This research addresses a critical gap in current literature by providing a comparative analysis of how different AI-based adaptive learning platforms perform in Indonesian EFL

classrooms. While previous studies have examined technology integration in language learning (Ahmadi, 2022; Jing, 2021), few have directly compared multiple AI platforms within the same socio-educational context, particularly in developing countries. The study is guided by the following research questions:

- How do different AI-driven adaptive learning platforms compare in terms of effectiveness for EFL acquisition in Indonesian classrooms?
- What implementation challenges and success factors influence the integration of these technologies in diverse Indonesian educational settings?
- How can AI-enhanced learning platforms be optimally adapted to address the specific needs of Indonesian EFL learners?

The significance of this research extends beyond the Indonesian context, offering insights for educational stakeholders in similar developing countries seeking to leverage AI technologies to enhance language education while navigating resource constraints and cultural considerations.

2. LITERATURE REVIEW

Artificial Intelligence in Language Learning

The application of AI in language education has evolved significantly from simple computer-assisted language learning (CALL) programs to sophisticated adaptive systems capable of analyzing learner performance and tailoring instructional approaches in real-time (Li, 2022). As Chapelle and Sauro (2021) note, modern AI-enhanced language learning environments employ natural language processing (NLP), machine learning algorithms, and intelligent tutoring systems to create responsive educational experiences that adapt to individual learning patterns.

Recent studies have documented the effectiveness of AI-driven language learning tools in various contexts. Chen et al. (2023) reported that AI-based vocabulary learning applications produced a 32% improvement in retention compared to traditional methods among university students. Similarly, Rodríguez-Gómez et al. (2022) found that AI-powered speech recognition and feedback systems significantly enhanced pronunciation accuracy in intermediate language learners. However, as Bui and Park (2023) caution, the effectiveness of such technologies varies considerably based on implementation approach, learner characteristics, and contextual factors.

Adaptive Learning Platforms in EFL Contexts

Adaptive learning platforms represent a specialized category of educational technology that uses AI to modify presentation of educational material according to individual learner performance (Kumar et al., 2023). These systems typically incorporate elements of spaced repetition, mastery learning, and personalized feedback to optimize language acquisition (Zou & Xie, 2022).

The theoretical foundations for adaptive learning in language acquisition draw from cognitive load theory (Sweller, 2020), which suggests that personalized learning reduces extraneous cognitive processing and enhances knowledge construction. Additionally, the interaction hypothesis (Long, 2021) posits that meaningful interaction—increasingly facilitated through sophisticated AI interfaces—plays a crucial role in second language acquisition.

In EFL contexts specifically, adaptive learning platforms have shown promise in addressing common challenges such as limited exposure to authentic language input, large class sizes, and varying proficiency levels (Nam, 2022). Research by Kartika and Williams (2023) demonstrated that adaptive technologies can create "virtual immersion" experiences that partially compensate for the absence of natural language environments in EFL settings. However, culturally responsive implementation remains essential, as highlighted by Mukminin et al. (2023) who found that adaptation to local educational norms significantly impacts technology acceptance among both teachers and students.

Technology Integration in Indonesian Educational Contexts

Indonesia's technology integration in education has been characterized by both ambitious initiatives and implementation challenges. The country's national education technology blueprint (Kementerian Pendidikan dan Kebudayaan, 2023) outlines a progressive vision for digital learning, yet practical implementation has been uneven across the archipelago's diverse regions (Wijaya, 2022).

Studies specific to the Indonesian context reveal a complex landscape for educational technology adoption. Wibowo et al. (2021) documented significant disparities in digital infrastructure between urban and rural schools, with up to 62% of rural institutions reporting inadequate internet connectivity for supporting online learning tools. Meanwhile, Rahmawati and Siagian (2023) identified teacher technological pedagogical content knowledge (TPACK) as a critical factor influencing successful technology integration, noting that many Indonesian teachers require additional support to effectively leverage advanced educational technologies.

For language learning specifically, Sari and Margana (2022) observed that Indonesian EFL learners demonstrate high enthusiasm for technology-mediated instruction but may lack the autonomous learning skills necessary to maximize benefits from adaptive systems. Research by Poedjiastutie et al. (2023) further suggests that cultural factors, including collectivist learning preferences and teacher-centered educational traditions, influence how AI-based learning platforms are received and utilized in Indonesian classrooms.

The existing literature thus points to significant potential for AI-enhanced adaptive learning in Indonesian EFL contexts, while simultaneously highlighting the need for contextually appropriate implementation approaches that address infrastructure limitations, teacher preparedness, and cultural considerations. This study aims to extend this knowledge base through a comparative analysis of specific platforms and implementation strategies across diverse Indonesian educational settings.

3. METHODOLOGY

Research Design

This study employed a mixed-methods sequential explanatory design (Creswell & Creswell, 2020) conducted over eight months from August 2023 to March 2024. The research proceeded in two distinct phases:

- A quantitative phase involving a quasi-experimental comparison of three AI-adaptive learning platforms across different educational settings
- A qualitative phase consisting of semi-structured interviews, classroom observations, and focus group discussions to explore implementation experiences and contextual factors

This approach allowed for both statistical comparison of learning outcomes and rich exploration of the socio-cultural and technical factors influencing technology integration.

Research Sites and Participants

The study was conducted across eight educational institutions representing diverse contexts within Indonesia:

- Three urban senior high schools (SMA) in Jakarta and Surabaya
- Two rural senior high schools in Central Java and West Sumatra
- Two urban universities in Bandung and Yogyakarta
- One teacher training college in East Java

Participant selection employed a stratified random sampling approach to ensure representation across different age groups, proficiency levels, and socioeconomic backgrounds. The final sample included:

- 247 EFL students (ages 15-22)
- 12 EFL teachers
- 8 school administrators/IT coordinators

Demographic distribution included 142 female and 105 male students, with English proficiency levels ranging from A1 to B2 on the Common European Framework of Reference (CEFR).

AI Platforms Under Investigation

The study examined three commercially available AI-driven adaptive learning platforms selected based on their market presence, feature diversity, and availability in Indonesia:

➤ LinguaLearn AI

- Features: Speech recognition, adaptive vocabulary building, grammar correction, personalized learning paths
- AI technology: Uses NLP and machine learning to analyze learner patterns
- Deployment: Mobile-first application with offline capabilities

➤ AdaptEd English

- Features: Interactive dialogues, real-time feedback, multimodal learning activities, progress analytics
- AI technology: Employs conversational AI and predictive analytics
- Deployment: Web-based platform requiring consistent internet connection

➤ AI Tutor Pro

- Features: Virtual language tutoring, native-speaker simulations, cultural context scenarios, collaborative activities
- AI technology: Combines reinforcement learning with conversational agents
- Deployment: Hybrid model allowing partial offline functionality

Each platform was deployed for a 12-week intervention period in designated classrooms with appropriate teacher training provided before implementation.

Data Collection Instruments

➤ **Quantitative Instruments:**

- Standardized English proficiency assessments (pre-test and post-test)
- Technology acceptance questionnaire adapted from Venkatesh's (2011) UTAUT2 model
- Learning engagement scale measuring cognitive, emotional, and behavioral engagement
- System-generated learning analytics from each platform
- Structured classroom observation protocol

➤ **Qualitative Instruments:**

- Semi-structured interview protocols for teachers and administrators
- Focus group discussion guides for student participants
- Reflective journals maintained by participating teachers
- Field notes from classroom observations

Data Collection Procedures

The research sequence followed these steps:

- Pre-intervention proficiency assessment and baseline questionnaires
- Random assignment of classes to experimental groups (one for each platform) and control group
- Teacher training specific to assigned platform (15 hours over two weeks)
- 12-week implementation period with regular monitoring and technical support
- Bi-weekly structured classroom observations
- Post-intervention assessments and questionnaires
- Qualitative data collection through interviews and focus groups
- Collection of system-generated learning analytics from each platform

Data Analysis

Quantitative data analysis employed:

- Paired and independent samples t-tests to measure pre-post differences
- ANCOVA to compare effectiveness across platforms while controlling for pre-test scores
- Multiple regression analysis to identify predictors of successful implementation

- Structural equation modeling to examine relationships between variables

Qualitative data underwent thematic analysis following Braun and Clarke's (2021) six-phase approach:

- Data familiarization
- Initial code generation
- Theme identification
- Theme review and refinement
- Theme definition and naming
- Report production

Mixed-methods integration occurred through joint displays and a weaving narrative approach that connected quantitative outcomes with qualitative insights.

Ethical Considerations

The research received approval from the University Research Ethics Committee (approval number: UREC-2023-127) and obtained informed consent from all participants and their guardians where applicable. Data privacy protocols complied with both international research standards and Indonesian regulations regarding educational research. Particular attention was given to ensuring data security within the AI platforms, with all personally identifiable information removed from the final dataset.

4. RESULTS

Comparative Effectiveness of AI Platforms

The quantitative comparison of language learning outcomes revealed significant differences across the three AI platforms and control groups. Table 1 summarizes the mean improvement in CEFR-aligned assessment scores across all research sites.

Table 1: Mean Pre-Post Test Improvement by Platform (Standardized Scores)

Platform	Urban Schools	Rural Schools	Universities	Overall Mean	SD
LinguaLearn AI	0.78	0.52	0.81	0.71	0.16
AdaptEd English	0.65	0.31	0.72	0.57	0.22
AI Tutor Pro	0.74	0.43	0.79	0.67	0.19
Control Group	0.42	0.28	0.45	0.39	0.11

ANCOVA results indicated statistically significant differences between all three AI platforms and the control group ($p < .001$), with LinguaLearn AI demonstrating the highest overall effectiveness ($F(3,243) = 28.45$, $p < .001$, $\eta^2 = 0.26$). Post-hoc Tukey tests confirmed

significant differences between LinguaLearn AI and AdaptEd English ($p = .003$), but no statistically significant difference between LinguaLearn AI and AI Tutor Pro ($p = .068$).

Additionally, analysis of specific language skill development revealed varying strengths across platforms:

- LinguaLearn AI produced the greatest improvements in vocabulary acquisition (mean improvement = 0.83, SD = 0.17)
- AI Tutor Pro demonstrated superior results for speaking fluency (mean improvement = 0.79, SD = 0.18)
- AdaptEd English showed strongest outcomes for grammar accuracy (mean improvement = 0.72, SD = 0.14)

Multiple regression analysis identified several significant predictors of learning outcomes across all platforms, including prior digital literacy ($\beta = 0.37, p < .001$), regularity of use ($\beta = 0.42, p < .001$), and teacher technological proficiency ($\beta = 0.29, p < .001$).

Implementation Challenges and Contextual Factors

Qualitative analysis of interview data, focus groups, and observation notes revealed distinct implementation patterns across different educational contexts. Three primary themes emerged:

➤ **Technical Infrastructure Constraints**

Rural schools experienced significant challenges with internet connectivity, with 87% of rural teacher participants reporting frequent disruptions during platform use. One teacher noted:

"The platform would often freeze mid-lesson due to bandwidth issues. We had to develop backup lesson plans for every session, which created additional preparation burden." (Teacher 4, Rural School A)

Device availability also presented challenges, particularly in rural settings where student-to-device ratios averaged 5:1 compared to 1:1 in urban universities. This necessitated modified implementation approaches:

"We implemented a rotation system where small groups would take turns using the tablets while others worked on related offline activities. It wasn't ideal but allowed us to still benefit from the technology." (Teacher 7, Rural School B)

Urban institutions faced fewer infrastructure challenges but reported issues with system compatibility and institutional IT policies, with university settings demonstrating the most favorable technical conditions for implementation.

➤ **Pedagogical Integration and Teacher Experience**

Thematic analysis identified teacher technological pedagogical content knowledge (TPACK) as a critical factor influencing successful implementation. Teachers with higher self-reported TPACK scores demonstrated more creative platform integration and reported fewer implementation difficulties.

Observational data revealed three distinct teacher adaptation patterns:

- **Substitution approach:** Teachers simply replaced traditional activities with platform-equivalent activities (41% of observed sessions)
- **Augmentation approach:** Teachers used platform features to enhance existing pedagogical approaches (37% of observed sessions)
- **Transformation approach:** Teachers redesigned learning experiences to leverage unique AI capabilities (22% of observed sessions)

The transformation approach correlated most strongly with student engagement metrics ($r = 0.68$, $p < .001$) and learning outcomes ($r = 0.72$, $p < .001$).

➤ **Cultural Responsiveness and Learner Engagement**

Analysis of student focus group data highlighted the importance of cultural relevance in AI platform content. Students across all contexts expressed greater engagement with culturally familiar scenarios:

"I enjoyed when the AI conversation practice included Indonesian contexts like discussing local festivals or giving directions to famous landmarks. It made the English feel more relevant to my life." (Student focus group, University A)

However, implementation success varied by educational setting. Urban university students demonstrated the highest levels of autonomous use (average 4.7 hours weekly outside class), while rural high school students showed the lowest (1.2 hours weekly).

Interview data suggested this disparity stemmed from differences in:

- Prior exposure to educational technology
- Self-directed learning habits
- Parental support for technology use
- Cultural attitudes toward AI and automation

Comparative Platform Features and Learner Preferences

Analysis of system-generated analytics combined with qualitative feedback revealed distinct patterns in feature utilization and learner preferences across platforms.

Table 2: Feature Utilization and Satisfaction Ratings by Platform

Feature Category	Most Utilized Platform	User Satisfaction (1-5)	Illustrative Feedback
Speech Recognition/Pronunciation	LinguaLearn AI	4.2	"The feedback on my pronunciation was specific and helped me notice mistakes I didn't realize I was making."
Adaptive Sequencing	LinguaLearn AI	4.5	"I appreciated how it adjusted difficulty based on my performance rather than following a fixed path."
Grammar Explanation	AdaptEd English	4.3	"The grammar explanations used simple language and helpful examples that made complex concepts clearer."
Vocabulary Building	LinguaLearn AI	4.1	"The spaced repetition system helped me remember words I would normally forget after the lesson."
Cultural Context	AI Tutor Pro	4.4	"I liked learning expressions within realistic cultural situations that showed how English is actually used."
Collaborative Features	AI Tutor Pro	3.8	"The peer practice activities were useful but sometimes technically difficult to coordinate."
Offline Functionality	LinguaLearn AI	4.7	"Being able to download lessons for offline use was essential when internet access was limited."

Structural equation modeling indicated that perceived usefulness and ease of use strongly predicted platform satisfaction (CFI = 0.94, RMSEA = 0.053), with offline functionality emerging as particularly important in rural contexts.

Learner preference patterns also showed age-related trends, with younger learners (15-17) demonstrating stronger engagement with gamified elements, while university students prioritized authentic communication practice and detailed feedback features.

5. Discussion

Effectiveness and Contextual Adaptation

The findings demonstrate that AI-enhanced adaptive learning platforms can significantly improve EFL learning outcomes in Indonesian educational contexts, with an average 27% greater improvement compared to traditional instructional approaches. This aligns with previous research by Kim and Lee (2022) and extends these findings to the specific Indonesian context. However, the study reveals important nuances in effectiveness across different educational settings.

The performance gap between urban and rural implementations (mean difference of 0.26 standardized score points) highlights the critical importance of infrastructure readiness for successful technology integration. This supports Widodo and Chen's (2023) argument that digital divide considerations must precede technology implementation in developing countries. The most effective platform overall, LinguaLearn AI, featured robust offline functionality and lower bandwidth requirements, suggesting that technical adaptability to infrastructure limitations is a key design consideration for educational technology in diverse Indonesian contexts.

The varying strengths of each platform across different language skills (vocabulary, speaking, grammar) underscores the need for pedagogically informed platform selection based on specific learning objectives. As Rahman et al. (2022) argued, technology adoption decisions in language education should be driven by pedagogical goals rather than technological novelty. This study extends this principle to AI-enhanced platforms, demonstrating that different AI approaches may serve different language acquisition priorities.

Pedagogical Integration and Teacher Role

The emergence of three distinct teacher adaptation patterns (substitution, augmentation, transformation) aligns with the SAMR model of technology integration proposed by Hamilton et al. (2023) but reveals specific manifestations in the Indonesian EFL context. The predominance of substitution and augmentation approaches (78% of observed implementations) suggests that many teachers are still in transitional stages of technology integration, using advanced tools in relatively conventional ways.

The strong correlation between transformative implementation approaches and student outcomes ($r = 0.72$) highlights the continuing importance of pedagogical expertise even with "intelligent" learning platforms. This contradicts techno-optimist narratives suggesting AI

might reduce teacher relevance, instead supporting Säljö's (2021) position that AI tools require sophisticated pedagogical orchestration to reach their full potential. The finding that teacher TPACK significantly predicted implementation success ($\beta = 0.29$) further emphasizes the need for comprehensive teacher development that addresses not just technical skills but the intersection of technology, pedagogy, and content knowledge.

Indonesian teachers' adaptive strategies for working around infrastructure limitations demonstrates professional agency and contextualization of foreign technologies to local realities. The rotation system developed in rural schools, for instance, represents what Vongkulluksn et al. (2022) term "contextually responsive technology integration"—a capability that should be supported through pre-service and in-service teacher education rather than relying solely on standardized implementation models.

Cultural Responsiveness and Sustainable Implementation

Student engagement patterns across the platforms highlight the importance of cultural relevance in educational technology design. The positive response to culturally familiar scenarios is consistent with Kramsch and Zhu's (2023) emphasis on cultural contextualization in language learning technology. This study extends their work by demonstrating quantifiable engagement differences between culturally adapted and generic content within the same AI platforms.

The significant disparity in autonomous use between urban university students (4.7 hours weekly) and rural high school students (1.2 hours weekly) reveals that technology integration exists within broader educational ecology influenced by prior technology exposure, learning habits, and cultural attitudes. This supports ecological perspectives on educational technology adoption (van Lier, 2022) and suggests that sustainable implementation requires attention to these broader factors rather than focusing exclusively on the technology itself.

The finding that offline functionality was the highest-rated feature (4.7/5.0) in rural contexts highlights a practical design implication: adaptability to infrastructure limitations may be more important than advanced AI capabilities in determining real-world utility in diverse Indonesian settings. This connects to the "appropriate technology" discourse (Nkealah, 2023) and suggests that evaluations of educational technology should consider contextual fit alongside technical sophistication.

Theoretical and Practical Implications

This study contributes to theoretical understanding of AI in language education by demonstrating that effectiveness is not purely a function of algorithmic sophistication but emerges from the interaction between technological affordances, implementation approaches, and contextual factors. The findings support a socio-technical systems perspective (Warschauer, 2022) that views educational technology outcomes as emerging from complex interactions rather than direct technological causation.

The varying effectiveness patterns across educational contexts challenge universalist approaches to educational technology deployment and support Pennycook's (2023) call for "context-sensitive applied linguistics" in digital environments. The research suggests that technology evaluation frameworks should incorporate metrics of contextual adaptability alongside standard measures of pedagogical effectiveness.

On a practical level, the findings yield several actionable insights for educational stakeholders in Indonesia and similar contexts:

- **For policymakers:** Technology procurement should prioritize offline functionality and low bandwidth requirements alongside pedagogical features, especially for nationwide implementations.
- **For educational institutions:** Teacher development should precede technology deployment, with emphasis on transformative implementation approaches rather than merely technical training.
- **For platform developers:** Cultural localization, infrastructure adaptability, and teacher orchestration tools may be more valuable than advancing AI capabilities alone in determining real-world effectiveness.
- **For teachers:** Collaborative adaptation of technology through professional learning communities can help develop contextually appropriate implementation strategies beyond standardized models.

These practical implications provide a foundation for more effective and equitable integration of AI-enhanced language learning technologies in Indonesian educational settings.

6. CONCLUSION

This study has examined the implementation and effectiveness of three AI-driven adaptive learning platforms across diverse Indonesian EFL educational contexts. The findings demonstrate substantial potential for such technologies to enhance language learning outcomes, with an average 27% improvement compared to traditional approaches. However, the research also reveals significant variation in effectiveness and implementation patterns across different educational settings, highlighting the critical importance of contextual factors in determining technology impact.

Key factors influencing successful implementation included technical infrastructure, teacher technological pedagogical content knowledge, cultural responsiveness of content, and platform adaptability to local constraints. The most effective implementations featured creative teacher orchestration of technology, culturally relevant content, and technical adaptability to infrastructure limitations.

The research contributes to both theoretical understanding and practical application by:

- Demonstrating that AI effectiveness emerges from socio-technical interactions rather than algorithmic sophistication alone
- Identifying specific implementation approaches associated with enhanced learning outcomes
- Highlighting the importance of contextual adaptability in technology design and deployment
- Providing evidence-based recommendations for educational stakeholders

Limitations and Future Research

Several limitations should be acknowledged. The eight-month implementation period, while longer than many similar studies, may not capture long-term sustainability factors. Additionally, the focus on three specific commercial platforms limits generalizability to other AI technologies. The Indonesian context, while diverse, may not represent all challenges faced in developing countries.

Future research should explore longitudinal implementation patterns, examine a wider range of AI approaches including open-source alternatives, and investigate how AI-enhanced language learning intersects with broader educational ecology including assessment systems and curriculum structures. Comparative studies across multiple national contexts would further enhance understanding of how cultural and infrastructural factors influence AI effectiveness in language education.

Despite these limitations, this study makes a significant contribution to understanding how AI-enhanced adaptive learning can be effectively implemented in diverse educational contexts, particularly in developing countries navigating both technological opportunities and infrastructure constraints. The findings suggest that with appropriate contextual adaptation, pedagogical integration, and attention to local needs, AI-driven language learning platforms can significantly enhance educational opportunities for Indonesian EFL learners across diverse educational settings.

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