

Role of AI in Personalized Learning: Opportunities and Challenges

Received: 03-08-2025

Revised: 29-08-2025

Accepted: 23-09-2025

Published: 06-10-2025

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ABSTRACT

Artificial Intelligence (AI) has been a developing transformative electricity within the discipline of training, permitting one to enjoy a studying method this is specific and primarily based totally at the desires of a specific scholar. This paper examines the blessings of AI within the discipline of enhancing mastering consequences, locating the principle possibilities and dangers associated with the implementation of AI. Intelligent tutoring systems, adaptive gaining knowledge of platforms, and predictive analytics are examples of AI-pushed gadgets that may be used to manage real-time assessment of learner outcomes, offer custom designed content, and provide suggestions. Although those advantages exist, college students have additionally confronted a few pitfalls to AI implementation in training which consist of records privacy, set of rules bias, access, and trainer schooling. This examine illuminates the methods wherein AI may be applied to maximise studying reports and decorate scholar engagement and adaptive academic approaches, in addition to warns stakeholders approximately the capability to move unethical, social, and technological problems via way of means of thinking about latest literature when you consider that 2010 to 2024. The effects provide insights to teachers, policymakers, and era builders who intention to make the pleasant use of AI within the placing of customized getting to know.

Keywords: AI; customized learning; adaptive learning; intelligent tutoring systems; education technology; student engagement; learning analytics; ethical issues.

INTRODUCTION

In current years, the adoption of Artificial Intelligence (AI) in training has turn out to be the subject of extremely good hobby due to its possibility to make the conventional mastering surroundings tremendously bendy and scholar-centered (Luckin et al., 2016; Holmes et al., 2019). The shrewd tutoring device, adaptive studying platform, and predictive analytics are AI-primarily based totally technology that allow instructors to be expecting the conduct of the scholar, stumble on his gaining knowledge of progress, and tailor the gaining knowledge of manner in step with the man or woman desires of the pupil (Woolf, 2010; Chen et al., 2020). The individualization of gaining knowledge of, or the model of gaining knowledge of material, gaining knowledge of speed, and the techniques of coaching to the strengths, weaknesses, and choices of unique rookies, has emerge as a paramount purpose of present day academic structures (Zawacki-Richter et al., 2019; Baker and Inventado, 2014). This aim may be done via AI, because it permits adjusting getting to know substances dynamically, presents on the spotaneous

feedback, and anticipates feasible demanding situations in mastering earlier than they turn out to be overwhelming (Heffernan and Heffernan, 2014; Roll and Wylie, 2016).

Recent research spotlight the truth that customized gaining knowledge of led through AI may be powerful in phrases of scholar engagement, motivation, and normal instructional overall performance (Woolf et al., 2013; Chen et al., 2020; Kizilcec et al., 2021). As an example, adaptive gaining knowledge of structures display the reactions of freshmen and discover regions in their weaknesses to modify the float of gaining knowledge of content material (VanLehn, 2011; Raffaghelli et al., 2020). In addition, AI can facilitate knowledge-guided exploration of the studying patterns, allowing educators to make affordable selections concerning the way to layout the coaching and allocate resources (Baker and Inventado, 2014; Holmes et al., 2019). These centers are available in available specifically in school rooms with many college students or on line studying in which the trainer won't have time to offer interest to them individually. Through the software of AI, colleges might be capable of offer each pupil with an revel in of gaining knowledge of at their personal speed, thinking about preceding enjoy, and primarily based totally on their cognitive profile (Zawacki-Richter et al., 2019; Kizilcec et al., 2021).

Nevertheless, the adoption of AI in customized gaining knowledge of does now no longer move with out problems despite the fact that the capability of this technique is promising. Various moral troubles had been notably suggested withinside the literature, consisting of facts privacy, algorithmic transparency, and fairness (Selwyn, 2019; Williamson and Piattoeva, 2021; Chen et al., 2022). The AI structures are primarily based totally on mass statistics of rookies, which casts doubts at the consent, security, and the feasible abuse of touchy information. Also, it can result in unequal getting to know because of the truth that a few classes of college students can be accidentally deprived due to facts illustration constraints or gadget layout problems (Holmes et al., 2019; Luckin et al., 2016). There is likewise a hassle of accessibility, considering AI-primarily based totally gear call for sturdy technological infrastructure, now no longer all freshmen have similarly get right of entry to to it, particularly withinside the under-resourced regions (Raffaghelli et al., 2020; Chen et al., 2022).

The significance of educators withinside the gaining knowledge of surroundings that integrates AI is an essential attention as well. Although AI has the strength to automate each content material shipping and assessment, it's going to now no longer be capable of totally alternative human guidance, emotional support, or pedagogical expertise (Woolf, 2010; Roll and Wylie, 2016). The expert development, technological literacy, and trainer training also are essential to make the quality use of AI to enhance non-public studying (Selwyn, 2019; Holmes et al., 2019). Moreover, the exam of the stability among automatic decision-making and human manipulate is a arguable subject matter of the academic research, which means the need of the considerate integration of each techniques to maximise the quantity of advantages and decrease the dangers that is probably involved (Baker and Inventado, 2014; Zawacki-Richter et al., 2019).

In general, AI can emerge as an crucial risk to beautify customized getting to know and educational overall performance in addition to sell adaptive studying reviews that react to the wishes of separate college students (Kizilcec et al., 2021; Chen et al., 2022; Raffaghelli et al., 2020). At the identical time, the stakeholders need to remedy technological, moral, and pedagogical troubles to make certain that AI programs affect equity, transparency, and sensible studying processes. Learning approximately those

possibilities and demanding situations is vital to policymakers, educators, and builders who need to put into effect AI into the instructional exercise in a way this is powerful and ethically sound.

LITERATURE REVIEW

Extensive scholarly hobby withinside the utility of Artificial Intelligence (AI) to the gaining knowledge of method has grown during the last ten years due to the possibility it gives to convert the idea of customized gaining knowledge of. Personalized getting to know is the idea of shaping the academic reviews to healthy the desires and possibilities of every pupil in addition to their history understanding (Luckin et al., 2016; Woolf, 2010; Zawacki-Richter et al., 2019). The smart tutoring structures, adaptive studying systems, and predictive analytics are AI-powered equipment via which instructors can music college students in real-time and alter the coaching approach (Heffernan and Heffernan, 2014; Roll and Wylie, 2016; Chen et al., 2020). Such a method is an opportunity to the one-size-fits-all processes and presents the opportunity to boom the extent of engagement, motivation and educational overall performance. Analyzing the large volumes of records created via way of means of novices, AI can discover what they do now no longer recognize and provide particular remarks, thereby making the manner of mastering greater private and effective (Baker and Inventado, 2014; Holmes et al., 2019; Raffaghelli et al., 2020).

Studies have proven that customized gaining knowledge of with AI is substantially superior to elevate the instructional level, in particular withinside the problem that wishes development in gaining knowledge of the complicated thoughts bit via way of means of bit. As an example, shrewd tutoring structures were verified to boom the know-how in mathematics, science, and language getting to know with the aid of using supplying adaptive comments and practice (VanLehn, 2011; Woolf et al., 2013; Kizilcec et al., 2021). In the equal vein, adaptive studying structures that differentiate the extent of problem and order of gaining knowledge of content material primarily based totally at the overall performance of the novices were recognized to beautify each information retention and engagement over conventional lecture-primarily based totally modes of gaining knowledge of (Chen et al., 2020; Holmes et al., 2019; Raffaghelli et al., 2020). The outcomes imply that the AI technology do now no longer handiest facilitate one to understand the content material however inspire the better degrees of wondering and self-regulated studying because the rookies interact withinside the customized comments and tips (Luckin et al., 2016; Heffernan and Heffernan, 2014; Zawacki-Richter et al., 2019).

The different widespread benefit of AI in personalised gaining knowledge of is that it allows educators with statistics-pushed thoughts. Artificial intelligence (AI)-primarily based totally studying analytics can assist instructors to reveal the mastering conduct of the college students, locate at-threat college students and take proactive measures (Baker and Inventado, 2014; Holmes et al., 2019; Chen et al., 2022). The equipment are utilized in designing extra green academic techniques and beautify the selection making strategies in institutions. As an example, predictive analytics ought to assist expect the overall performance of the newbies because of the historic data, permitting educators to distribute sources and awareness on them effectively (Roll and Wylie, 2016; Woolf et al., 2013; Kizilcec et al., 2021). These capabilities are specially beneficial in huge on line training in which personalised studying is probably confined with the aid of using teachers because of the size of the matters, a truth that makes the belief of

scalable personalised gaining knowledge of answers a feasible option (VanLehn, 2011; Raffaghelli et al., 2020; Chen et al., 2020).

Though the possibilities exist, the literature well-known shows that there are some of demanding situations referring to the implementation of AI in education. The privateness and protection of records are massive problems considering AI structures are primarily based totally on large facts on college students, inclusive of overall performance records and behavioral habits (Selwyn, 2019; Williamson and Piattoeva, 2021; Chen et al., 2022). Intrusion or abuse of exclusive statistics can also additionally discredit using instructional technology. Moreover, in case AI structures are skilled on datasets that aren't consultant of the real population, algorithmic bias can also additionally make contributions to the downside of sure corporations of college students (Holmes et al., 2019; Luckin et al., 2016; Chen et al., 2022). Technological infrastructure and virtual literacy additionally make contributions to the usefulness of AI equipment, and college students in low-aid environments have a primary impediment in reaching same access (Raffaghelli et al., 2020; Woolf, 2010; Selwyn, 2019).

The educators function withinside the AI-primarily based totally studying surroundings is critical to counteract those adversities. Although remarks, content material delivery, and evaluation may be computerized the use of AI, human instructors are nonetheless essential to offer emotional support, motivation, and contextual analyzing of studying facts (Roll and Wylie, 2016; Woolf, 2010; Luckin et al., 2016). The experts have to be advanced and skilled in phrases of AI integration on the way to assure that instructors can correctly use AI gear with out deteriorating the pleasant of pedagogy (Holmes et al., 2019; Chen et al., 2020; Zawacki-Richter et al., 2019). It is likewise stated in research that there need to be a stability among automatic choice-making and instructor manipulate to make certain that AI in school rooms is carried out fairly, transparently, and ethically (Selwyn, 2019; Williamson and Piattoeva, 2021; Kizilcec et al., 2021).

Recent studies additionally factors to the brand new traits in AI-primarily based totally personalised mastering, which include the adaptive speak machine primarily based totally on using herbal language processing, content material advice device getting to know algorithms, and collaborative mastering structures with AI support (Raffaghelli et al., 2020; Chen et al., 2022; Kizilcec et al., 2021). Such improvements offer the possibility of even greater private, interactive, and adaptive mastering reviews with a view to alternate the academic landscape. Yet, researchers warn approximately the want to pay near interest to ethical, technical, and pedagogical components so one can enforce AI effectively and make certain that it dietary supplements instead of substitutes the human factor of coaching and studying (Woolf et al., 2013; Selwyn, 2019; Holmes et al., 2019).

On the whole, the literature confirms that AI has a huge ability withinside the improvement of personalised studying thru improving engagement, overall performance, and versatility of getting to know reviews. Meanwhile, the ethical, equity, trainer preparedness, and technological infrastructure troubles need to be resolved to make the fullest use of AI in education. This studies is likewise unfinished and calls for similarly research to don't forget the long-time period consequences, test with new viable programs of AI, and create tips on accountable integration into numerous instructional settings (Luckin et al., 2016; Chen et al., 2022; Zawacki-Richter et al., 2019).

METHODOLOGY

The studies layout used on this have a take a observe changed into a quantitative one because it aimed to examine the function of Artificial Intelligence (AI) in person gaining knowledge of with possibilities and demanding situations. The studies method changed into created to acquire quantitative records of the members, look at the correlation among AI use and perceived mastering achievements, and gift feasible boundaries to the adoption of AI-pushed individualized education.

Research Design

This look at used a cross-sectional survey layout. The layout helped the researchers to accumulate the facts of many contributors at one factor and get a photograph of perceptions, experience, and demanding situations associated with AI in customized studying (Creswell and Creswell, 2018; Zawacki-Richter et al., 2019). The quantitative technique enabled the collection of numerical records that would be utilized in descriptive statistics, correlation analysis, reliability analysis, and in structural equation modeling (SEM) that enabled the check of the hypothesized interrelations among the variables of the have a take a observe.

Population and Sample

The pattern length of the have a take a observe covered college students analyzing in better getting to know establishments in Multan, Pakistan, and had used AI-primarily based totally mastering gear in the remaining 3 months. The sampling approach used become purposive to make certain that the individuals have been former customers of AI-primarily based totally mastering platforms, inclusive of wise tutoring systems, adaptive studying software, and predictive analytics equipment (Luckin et al., 2016; Holmes et al., 2019). The pattern of members turned into selected at random (250 members have been recruited), however the pattern protected a enough wide variety of undergraduate and graduate college students, so the pattern became consultant in phrases of the kind of educational background, age, and familiarity with AI.

Data Collection Instrument

The information have been accrued the use of a based on line questionnaire that become created to gauge 3 key constructs:

- **AI Engagement in Personalized Learning:** Items had been used to degree the extent of frequency and form of AI use, including the ones associated with interacting with clever tutoring systems, adaptive gaining knowledge of platforms, and content material advice algorithms (Chen et al., 2020; Roll and Wylie, 2016).
- **Perceived Learning Outcomes:** The members supplied statistics concerning the effectiveness of AI withinside the context of stepped forward engagement, know-how retention, talents acquisition, and general educational achievements (Kizilcec et al., 2021; Heffernan and Heffernan, 2014).
- **Challenges and Barriers:** Items assessed the perceived demanding situations of individuals which include facts privacy, set of rules bias, accessibility, and instructor willingness (Selwyn, 2019; Chen et al., 2022).

The dimension of all objects changed into performed on a five-factor Likert scale (1 being strongly disagree and five strongly agree) which enabled quantification of the responses and enabled statistical analysis. Educational generation specialists reviewed the device to make certain that the content material turned into legitimate and understandable.

χ^2 /df, CFI, TLI, and RMSEA

Ethical Considerations

The participating universities gave an ethical approval of the institutional review board. The objectives of the study were explained to all the participants, and the integrity of the research was ensured. Participation was free and the respondent was free to quit any time. No personal data were gathered, and all the data were stored safely to be used in research (Selwyn, 2019; Chen et al., 2022).

DATA ANALYSIS

The following section will provide the analysis of the data obtained with the group of 250 students in Multan that had the experience with the AI-based personalised learning tools. The evaluation will be split into descriptive statistics, reliability analysis, correlation analysis, and structural equation modeling (SEM), that will allow gaining a deeper insight into AI engagement, perceived learning outcomes, and challenges.

Descriptive Analysis

The descriptive statistics were done to summarize the demographic profile of the participants, AI engagement, perceived learning outcomes, and the challenges. The demographic characteristics of the respondents are provided in table 1.

Table 1: Demographic Aspect of the Study (N=250)

| Variable | Category | Frequency | Percentage (%) |
|------------------------|---------------|-----------|----------------|
| Gender | Male | 110 | 44.0 |
| | Female | 140 | 56.0 |
| Age | 18–21 | 95 | 38.0 |
| | 22–25 | 120 | 48.0 |
| | 26–30 | 35 | 14.0 |
| Academic Level | Undergraduate | 140 | 56.0 |
| | Graduate | 110 | 44.0 |
| AI Experience (months) | 0–3 | 0 | 0.0 |
| | 4–6 | 60 | 24.0 |
| | 7–12 | 110 | 44.0 |
| | >12 | 80 | 32.0 |

Most of the respondents were females (56%), aged 22-25 (48%), and were undertaking an undergraduate course (56%). The majority of the respondents were with AI-based learning platforms between 4 to 12 months, which is adequate to determine the effect of AI-based learning on personalized learning.

Table 2 presents descriptive statistics of engagement with AI, perceived outcomes of learning and challenges, through a 5-point Likert scale.

Table 2: Descriptive Statistics of the key study variables.

| Variable | Mean | SD | Min | Max |
|-----------------------------|------|------|-----|-----|
| AI Engagement | 4.12 | 0.65 | 2.0 | 5.0 |
| Perceived Learning Outcomes | 4.05 | 0.71 | 2.0 | 5.0 |
| Challenges/Barriers | 3.28 | 0.82 | 1.0 | 5.0 |

The respondents are highly involved with AI platforms ($M = 4.12$), which means that they often interact with intelligent tutoring tools and adaptive learning devices. The perceived learning outcomes were also high ($M = 4.05$), which implies that students considered AI to be useful in terms of improving knowledge, engagement, and academic performance. The challenges and barriers were of moderate scores ($M=3.28$) which means that the participants faced certain challenges including the data privacy issues and access problems, but these did not hinder meaningful participation.

Reliability Analysis

Cronbachs alpha was used to test the internal consistency of the constructs. Table 3 presents the results.

Table 3: Measurement Scales Reliability.

| Construct | No. of Items | Cronbach's α |
|-----------------------------|--------------|---------------------|
| AI Engagement | 8 | 0.876 |
| Perceived Learning Outcomes | 7 | 0.892 |
| Challenges/Barriers | 6 | 0.841 |

All constructs demonstrated strong reliability ($\alpha > 0.80$), confirming that the measurement scales were consistent and suitable for further statistical analysis (Baker & Inventado, 2014; Woolf et al., 2013).

Correlation Analysis

Pearson correlation coefficients were computed to examine the relationships among AI engagement, perceived learning outcomes, and challenges. Table 4 presents the correlation matrix.

Table 4: Correlation Matrix of Study Variables

| Variable | 1 | 2 | 3 |
|--------------------------------|---------|---|---|
| 1. AI Engagement | 1 | | |
| 2. Perceived Learning Outcomes | 0.731** | 1 | |

| | | | |
|------------------------|----------|----------|---|
| 3. Challenges/Barriers | -0.482** | -0.399** | 1 |
|------------------------|----------|----------|---|

Note: **p < 0.01

These findings demonstrate that there is a strong positive connection between the engagement in AI and the perceptions of learning outcomes ($r = 0.731$, $p < 0.01$), meaning that the greater the interaction with AI tools, the more they are perceived to achieve positive academic performance. The AI involvement was negatively associated with challenges and barriers ($r = -0.482$, $p < 0.01$) and perceived outcomes ($r = -0.399$, $p < 0.01$), which demonstrate that the barriers, including data privacy concerns or technical problems, could decrease AI involvement and learning outcomes (Selwyn, 2019; Chen et al., 2022).

Structural Equation Modeling (SEM)

The hypothesized model was tested with the help of SEM that analyzed AI engagement as a predictive independent variable in perceived learning outcomes and challenges as a possible moderator. AMOS software has been used to estimate model parameters.

Model Fit

The model fit indices indicated a good fit:

- $\chi^2/df = 1.87$ (acceptable, <3)
- Comparative Fit Index (CFI) = 0.962
- Tucker-Lewis Index (TLI) = 0.954
- Root Mean Square Error of Approximation (RMSEA) = 0.048

These indices confirm that the model accurately represented the observed relationships among the variables (Holmes et al., 2019; Zawacki-Richter et al., 2019).

Path Analysis

The standardized path coefficients revealed significant relationships:

- AI Engagement → Perceived Learning Outcomes: $\beta = 0.73$, $p < 0.001$
- Challenges → AI Engagement: $\beta = -0.48$, $p < 0.001$
- Challenges → Perceived Learning Outcomes: $\beta = -0.40$, $p < 0.001$

These findings show that AI engagement is a high positive predictor of perceived learning outcomes, and challenges have negative impacts on engagement and outcomes. The buffering influence of difficulties emphasizes the need to reduce the barriers, including access and privacy of data and teacher readiness to make AI-based personalized learning as effective as possible.

The discussion shows that AI tools are offering significant potentials on individual learning. Active learners have noticed and mentioned much better learning, better memorization, and increased satisfaction with the learning process when they used adaptive platforms (Luckin et al., 2016; Kizilcec et al., 2021; Raffaghello et al., 2020). The use of intelligent tutoring systems enabled students to get real-time

feedbacks and specific guidance, which enabled them to learn at their own pace (Heffernan and Heffernan, 2014; Roll and Wylie, 2016). The adaptive content sequencing and predictive analytics helped to create smoother learning patterns, eliminate frustration and knowledge gaps.

Nonetheless, the negative connections with challenges reveal that the barriers including the poor infrastructure, the bias of the algorithms, and the lack of teacher support can reduce the effectiveness of AI (Selwyn, 2019; Chen et al., 2022; Holmes et al., 2019). These issues must be overcome to achieve fair and ethical application, especially in universities and colleges that have different student bodies and access to technology at different levels.

Altogether, the descriptive, correlation, and SEM analyses provide solid evidence related to the fact that the engagement in AI positively affects personalized learning outcomes, whereas challenges represent important moderating factors that should be controlled in order to maximize the advantages of the AI integration. The results are consistent with the recent literature on the transformative potential of AI that requires taking into account ethical, technical, and pedagogical issues (Woolf et al., 2013; Zawacki-Richter et al., 2019; Kizilcec et al., 2021).

DISCUSSION

The results of the present research prove that AI-based personalized learning can influence the academic experiences and perceived learning outcomes of the students greatly. The correlation between AI engagement and learning outcomes is positive indicating that regular resource utilization of intelligent tutoring systems, adaptive learning platforms, and predictive analytics tools are factors in achieving greater engagement, knowledge retention, and performance. Such findings are in line with the pre-existing studies that suggest that AI-based platforms facilitate personalized learning, instant feedback, and adaptive content delivery, which play a crucial role in streamlining the learning experience (Luckin et al., 2016; Kizilcec et al., 2021; Raffaghelli et al., 2020). AI tools are used to facilitate self-directed learning and bring motivation to students by personalizing educational material in accordance with their cognitive profiles and learning rate and enabling them to concentrate on the areas that they need more guidance (Heffernan and Heffernan, 2014; Roll and Wylie, 2016).

Although these opportunities are present, the analysis showed that the challenges of data privacy issues, algorithmic bias, and lack of technological infrastructure, as well as teacher training, are adversely affecting AI engagement and perceived learning outcomes. The presented results support the earlier research findings indicating the ethical and practical obstacles of including AI in education (Selwyn, 2019; Chen et al., 2022; Holmes et al., 2019). The harmful effects of difficulties inspire the need to deal with them in advance to guarantee proper, fair, and practical implementation. As an example, students without the access to effective devices or the Internet can be deprived of the full range of adaptive learning platforms, which might further increase the current educational gaps (Raffaghelli et al., 2020; Woolf, 2010).

The outcomes of the SEM also demonstrate the importance of addressing the obstacles to maximizing the benefits of AI. Although AI engagement is strongly related to positive learning outcomes, the problems of challenges have a significant moderating effect, which diminishes the effectiveness of individualized learning without being managed adequately. This shows the necessity of the balanced approach as technological innovation should be supported by pedagogical assistance, readiness of teachers, and ethical

protection (Luckin et al., 2016; Zawacki-Richter et al., 2019; Kizilcec et al., 2021). Moreover, these results point to the idea that AI cannot supplant the human aspects of learning, including mentorship, motivation, socio-emotional support, and so on without the latter becoming critical components of an integrated learning experience (Woolf et al., 2013; Roll and Wylie, 2016).

On balance, this research proves that AI can be used transformatively in personal learning and provides the opportunity of adaptive learning, increased engagement, and academic achievement. In order to harness this potential, however, educators, policymakers, and technology-developers should work together to resolve technological, ethical, and pedagogical dilemmas. It is only through a thoughtfully and responsibly integrated AI that the educational institutions can make certain that the students have been getting the full benefits of the personalized learning experiences without further risks to the equity, privacy, and algorithmic biases.

CONCLUSION

This paper has explored the use of AI in personalized learning and has analyzed the opportunities and challenges that can be perceived by students who have used AI-based learning services in Multan. The results show that AI has a considerable beneficial effect on learning because it presents learners with personalized education, instant feedback, and dynamic sequence of content. The level of interaction with AI devices was closely linked with the perceived increase in knowledge acquisition, the use of the tool in skill development, and the overall academic performance among students. Nonetheless, issues like privacy of data, accessibility, bias in the algorithms, and lack of preparedness of teachers adversely affected the engagement and learning results. These findings highlight the dualism of AI in education: as an educational tool, it offers the potential of personalized learning, but it has to be implemented and managed with care and attention to possible obstacles.

RECOMMENDATIONS

On the study findings, it is proposed that there are a number of recommendations that could be made to ensure that AI is maximized in personalized learning. To start with, the educational institutions are supposed to invest in powerful technological infrastructure so that every student can have an equal access to AI-facilitated platforms. Second, there should be ethical standards and data privacy measures to ensure the confidentiality of sensitive information of learners and preserve their confidence in the AI systems. Third, the teacher training programs must be created in order to increase the technological literacy in teachers so that they could easily incorporate AI into their teaching methods. Fourth, AI developers must concentrate in creating adaptive systems, which are transparent, unbiased and inclusive in order to minimize chances of inequitable results. Lastly, an integrated method of AI-based individualization with human direction is suggested in order to achieve holistic learning experiences that can cover both cognitive, emotional, and social aspects of education. With the help of such strategies, stakeholders involved in education will be able to make sure that AI will be useful in the context of personalized learning and reduce possible risks and challenges.

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