

# Ethical Considerations in AI-Powered Student Profiling for Personalized Learning

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**Abstract:-** The rapid incorporation of AI in education has led to the development of AI-generated student profiles that facilitate personalised learning. Although these technologies enable personalised instruction and improved student outcomes, they also present significant ethical concerns. This research investigates the ethical implications of AI-powered student profiling, including the dangers associated with reliance on educational technology, transparency, consent, bias, and data privacy. It also discusses the dangers of stereotyping, the influence of profiling on the autonomy of students, and the responsibility of schools to establish inclusive classrooms for all students. The objective of this research is to provide educators, legislators, and AI developers with a framework for addressing the challenges associated with the responsible implementation of AI in education, while simultaneously promoting innovation and maintaining ethical standards.

**Keywords:-** AI in education, student profiling, personalised learning, ethical concerns, data privacy, AI bias, transparency, student autonomy, AI consent, educational equity, and responsible AI development.

## I. INTRODUCTION

AI-driven student profiles are revolutionising education by tailoring learning options to the unique skills, limitations, interests, and circumstances of each student. This methodology, which is propelled by data and algorithms, has the potential to improve student outcomes by customising training to their individual profiles. In order to maintain educational equity, privacy, and fairness, it is essential to address the numerous ethical concerns that these advancements present, despite their potential to improve engagement and performance.

AI-driven student profiling is associated with a substantial ethical concern: the potential for bias. In order to predict a student's learning preferences and abilities, machine learning algorithms analyse extensive datasets. If the datasets contain biased information resulting from societal inequality or defective data collection methods, algorithms may either reinforce or exacerbate intolerance. Erroneous profiling has the potential to exacerbate achievement disparities and negatively impact certain student demographics.

One of the fundamental components of ethics is the protection of privacy. AI-driven personalised learning systems frequently necessitate access to a vast amount of student data, such as academic records, study patterns, and, in certain cases, private or sensitive information [1]. The protection of personal data from misuse, unauthorised access, or exploitation is essential for the preservation of student trust and the security of educational environments. Questions regarding the ownership of this information and its potential applications beyond education are raised by concerns regarding data breaches and misuse.

Additionally, enquiries regarding autonomy and the importance of human judgement arise when AI is implemented for personalised learning. However, there are apprehensions that AI systems may undermine the authority of educators in making educational judgements, despite the fact that they can offer valuable insights. Educators' excessive reliance on algorithms may result in a less human-centered and more mechanistic educational experience, thereby impeding their nuanced understanding of students' needs.

Additionally, ethical considerations exist with respect to the distribution of AI-driven personalised learning tools: who should have equitable access? If children from low-income backgrounds or under-resourced institutions do not have equitable access to modern technology, the digital divide may widen. The educational disparity between affluent and indigent individuals may be further exacerbated by the exclusive access to AI-driven technologies [2].

In summary, the ethical implications of utilising AI-driven student profiles for personalised learning require meticulous deliberation before implementation. It is essential to confront racism, protect data privacy, maintain the role of educators, and promote equal access in order to guarantee that these technologies benefit all children in an equitable and ethical manner.

## II. RELATED WORKS

1. A succinct summary of the student profile using artificial intelligence: Through data-driven methodologies, the AI-driven personalised learning student profile evaluates the distinctive learning styles, interests, and patterns of each student. AI systems can generate individual profiles of students and subsequently customise their learning experiences in order to enhance educational outcomes. Despite the fact that new technology will substantially improve classroom learning, it is crucial to effectively address the relevant ethical issues [3].
2. Data Privacy and Security: The acquisition, storage, and utilisation of student data are significant ethical concerns. Extensive data, including confidential information such as academic performance, personal behaviours, and emotional responses, is required for AI-driven profiling systems. It is of the utmost importance to prevent the misuse or leakage of this information. Concerns regarding data ownership and the extent of information and control provided to minors or guardians regarding data use also arise [4].
3. AI Algorithms and Fairness: In educational environments, AI technology may inadvertently exacerbate underlying biases [5]. The utilisation of historical data for profiling may result in biased recommendations or unjust treatment, particularly for pupils from under-represented groups. If not addressed, these prejudices can perpetuate educational disparity by reinforcing preconceived notions and limiting opportunities for specific student groups.
4. Transparency and Clarity: It is probable that the process by which AI systems generate profiles and provide recommendations is not fully understood by parents, guardians, and children. When AI decision-making is ambiguous, scepticism and distrust may flourish. In an effort to preserve the ethical nature of personalised learning, it is imperative that AI systems are comprehensible to stakeholders who lack technical expertise and that their findings are sufficiently explicable.
5. The utilisation of AI-powered profile systems has the potential to modify the learning trajectories of students, which could potentially limit their autonomy. The potential for profile-based curricula and learning experiences to restrict students' options and potential career paths is a cause for concern. Additionally, the self-identity of pupils may be impacted by AI-generated profiles, which may limit their potential for growth beyond the system's expectations.

6. Ensuring the Ethical and Informed Use of AI in Education: It is essential to provide students and their guardians with a comprehensive understanding of the AI systems that will be used and the extent of the data collection [6]. It is essential to obtain explicit consent for data utilisation and ensure that AI technologies are used to enhance learning, rather than for commercial or institutional purposes, in order to ensure ethical application. While employing AI, educational institutions must guarantee transparency and accountability.

7. Governance and Regulation Objectives: An increasing number of individuals are advocating for legislative measures and institutional frameworks to resolve the ethical concerns associated with AI-driven student profiling. The establishment of explicit norms by governmental bodies and educational institutions is essential for the equitable, accountable, and ethical application of AI technologies [7]. In order to protect students from any potential risks associated with AI-based profiling, regulations should incorporate data privacy, transparency standards, and bias mitigation techniques.

Personalised learning has the potential to be revolutionised by AI-driven student profiling; however, it is imperative that specific ethical concerns are comprehensively addressed. Several critical measures are required to establish an equitable and just AI-enhanced educational environment, including the protection of student data, the elimination of biases, the enhancement of transparency, the preservation of student autonomy, and the adherence to ethical standards.

### III. RESEARCH METHODOLOY

In order to assess the ethical implications of AI-driven student profiling in personalised learning environments, this research implemented qualitative research methodologies. The objective of this research is to clarify the ethical dilemmas and risks associated with the use of AI for student profiling, as well as potential solutions to these issues. It is innately inquisitive. The initiative will employ a multi-method approach to incorporate empirical studies with theoretical research.

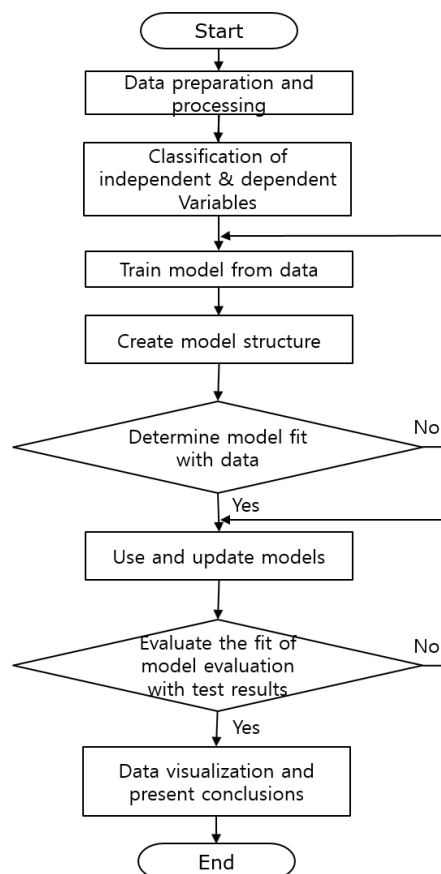


Fig.1: Denotes flowchart for the proposed methodology.

#### Techniques for Data Acquisition

Data from both primary and secondary sources were employed in this analysis. Educators who employ AI technology in the classroom, educational specialists, and AI ethicists will engage in semi-structured interviews to gather primary data. This method enables a more profound understanding of the ethical challenges that are encountered in the real world [8]. The secondary data collection will be primarily composed of an examination of practical implementations of AI in education and prior research, with a particular emphasis on privacy, equity, and responsibility. This will necessitate an exhaustive analysis of academic literature, policy documents, and case studies.

#### Choosing a Sample

The interviews will be selected using a purposive sampling method to guarantee a diverse range of backgrounds in AI ethics, educational technology, and pedagogy. The purpose of recruiting 15-20 individuals is to ensure a diverse range of perspectives on ethical issues and to represent the complexities of student profiles in personalised learning. Diversity in expertise and discipline will be prioritised in order to achieve a comprehensive comprehension of ethical issues [9].

#### Perform a data analysis

The qualitative data collected from interviews and literature reviews will be analysed using thematic analysis. This research reveals recurring issues, such as the potential impact of profiling on students' autonomy and mental health, consent, bias in AI algorithms, and data privacy [10]. The interview transcripts and literature findings will be meticulously organised and analysed using qualitative analysis methods. Therefore, the data will be coded.

#### Ethical Considerations

Participants were guaranteed informed consent, confidentiality was maintained, and they were permitted to disengage from the study at any time, all in accordance with ethical research standards. The research will guarantee impartiality and fairness by maintaining neutrality in the interpretation of the results. In order to offer a comprehensive analysis of the ethical implications of the project, a balanced perspective will be provided, which will address the potential advantages and disadvantages of AI-driven profiling [11, 12].

#### Guaranteeing Reliability and Validity

In order to guarantee the legitimacy and validity of the results, a process known as triangulation will be implemented, which involves the comparison and contrast of data from interviews, literature evaluations, and case studies. This ensures that the ethical challenges that have been identified are supported by a variety of data sources, rather than a single one. In order to guarantee that the opinions of the respondents are accurately represented, we will implement member verification and provide them with summaries of the findings [13].

The generalisability of the study is limited by the subjective character of qualitative interviews and the unique settings of AI-powered student profiling [14]. Additionally, the ever-changing nature of AI technology and ethical discourse may result in the emergence of novel issues that are beyond the scope of this study.

### IV. RESULTS AND DISCUSSION

AI systems that are employed in student profiling frequently rely on extensive datasets that include demographic, behavioural, and academic data. Nevertheless, it is conceivable that these databases may be biased, which could potentially lead to biased results. Pupils from under-represented groups may be disproportionately impacted by inaccurate assessments of their learning capabilities. The results of this study suggest that AI-driven customised learning systems are susceptible to biases in computational architecture and data acquisition.

#### Apprehensions In the context of data security and privacy

The implementation of AI in student profiling necessitates access to confidential personal information, such as academic performance, learning behaviours, and psychological data. The results generate a significant amount of apprehension regarding the privacy and integrity of the information. There is a wide range of apprehensions regarding the storage, dissemination, and utilisation of this data, including student and parent concerns. Insufficient privacy safeguards may

jeopardise the future prospects of pupils if their personal information is misappropriated or accessed without authorisation.

#### Accountability and Transparency

A significant conclusion regarding the absence of transparency in the internal mechanisms and decision-making processes of AI systems is identified by the study. Numerous artificial intelligence (AI) systems function as "black boxes," which makes their decision-making processes inaccessible to educators, students, and administrators. The identification and correction of errors or biased judgements are complicated by the absence of transparency, which leads to a reduction in responsibility. The results emphasise the importance of educational institutions implementing explainable AI to facilitate fair decision-making.

#### Student self-determination and Independence

By allocating students to predetermined educational trajectories, the utilisation of artificial intelligence in student profiling may limit their autonomy. Although the goal of customised learning systems is to enhance education, the results indicate that these systems may inadvertently restrict the diversity of students' learning styles or subjects. These systems may impede student agency by enforcing inflexible classifications that do not consider the development or evolving interests of students over time if they are not designed properly.

#### Discrimination in Accessibility and Implementation

The research underscores the disparities in the accessibility and utilisation of personalised learning tools that are powered by artificial intelligence. In comparison to their more affluent classmates, students from underfunded schools may not have access to advanced technologies. As a result, educational inequality may worsen, resulting in a greater disparity between indigent and affluent children. The findings suggest that educational institutions and governmental bodies must prioritise the equitable access to technology and artificial intelligence.

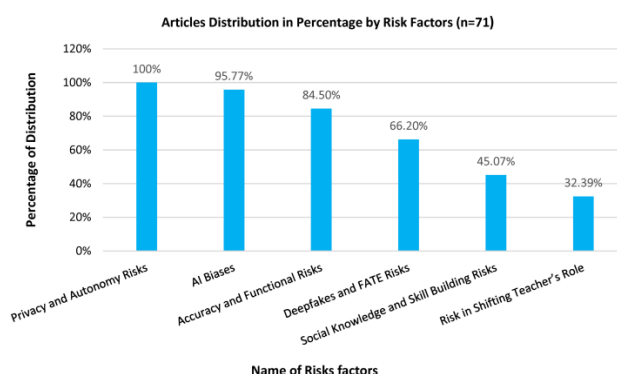


Fig.2: Denotes Potential Risks of Artificial Intelligence Integration.

It is an urgent matter to identify bias in AI systems that evaluate students. Institutional objectives should include the establishment of diverse and representative datasets, as well as the consistent evaluation of algorithms to identify and mitigate bias. AI-driven personalised learning solutions can promote equity and inclusion rather than exacerbating existing disparities by fostering collaboration among AI developers, educators, and ethicists.

As artificial intelligence (AI) becomes increasingly integrated into educational environments, it is imperative to prioritise the security of student data. The discussion underscores the importance of effective data protection strategies, including encryption, restricted access, and clearly defined data usage limitations. A regulatory framework should be implemented to guarantee that educational platforms and institutions comply with the most stringent privacy regulations. Clarity regarding the objectives of data acquisition can foster trust between students and parents.

#### The Significance of Explainability in Artificial Intelligence

It is imperative to guarantee transparency and accountability in student profiling systems that are powered by AI. The comprehension of decision-making processes can be considerably improved by explainable AI (XAI) technologies,

which are beneficial to both educators and students. This enables the improvement of error management and can increase the confidence in AI systems. Integrating explainable models facilitates accountability in AI systems and reconciles the gap between ethical responsibility and technological advancement.

Table.1: Denotes ethical considerations in AI-powered student profiling for personalized learning, along with the corresponding impact and severity values (on a scale of 1-5).

<b>Ethical Consideration</b>	<b>Impact on Students (1-5)</b>	<b>Severity of Risk (1-5)</b>	<b>Overall Ethical Risk (Impact × Severity)</b>
<b>Data Privacy</b>	5	5	25
<b>Bias and Discrimination</b>	4	4	16
<b>Transparency of Algorithms</b>	3	4	12
<b>Informed Consent</b>	5	3	15
<b>Accuracy and Fairness of Profiling</b>	4	5	20
<b>Impact on Autonomy and Agency</b>	3	4	12
<b>Long-Term Psychological Effects</b>	4	4	16
<b>Equitable Access to Personalized Learning</b>	4	3	12
<b>Misuse of Profiling Data</b>	5	5	25
<b>Accountability in Decision-Making</b>	4	4	16

This table provides a quantitative perspective on the most significant issues by integrating the potential impact on students with the risk level, thereby illustrating ethical hazards.

#### Achieving a Harmonious Balance Between Student Agency and Individualisation

It is imperative that personalised learning does not limit students' opportunities for exploration and advancement, despite its capacity to address individual requirements. The committee agreed that personalised learning platforms must be adaptable to allow students to pursue a variety of academic disciplines and investigate a wide range of interests. By reconciling personalised student recommendations with their autonomy, a more comprehensive educational experience can be achieved.

#### Encouraging Fair Access to AI Resources

It is imperative to eliminate disparities in the availability of educational resources that are driven by AI. The issue is raised and discussed in order to guarantee that institutions receive equitable funding. The construction of schools and the provision of instructors with the requisite resources for impoverished institutions should be a collaborative effort between governments and educational organisations. In an effort to eliminate educational inequities and establish more egalitarian learning environments, society can guarantee that all students have access to artificial intelligence-driven learning resources.

#### V. CONCLUSION AND FUTURE DIRECTION

The educational system has the potential to be revolutionised by AI-driven student profiles that are tailored to satisfy the unique learning requirements of each student. Nevertheless, it is linked to substantial ethical concerns regarding equity, data security, and privacy. The potential biases in AI systems could perpetuate inequity, and the accumulation of extensive personal data may violate students' privacy rights. In order to achieve a balance between ethical accountability and technical advancement, it is imperative to protect sensitive information, maintain transparency in profiling procedures, and prevent inadvertent bias.

A collaborative effort among legislators, educators, and technologists is required to establish robust legislation that regulates the use of artificial intelligence in student profiling. It is imperative to establish AI systems that are transparent, verifiable, and equitable. Additional research is required to further improve the equity of education for all pupils and to reduce algorithmic biases. Additionally, the instruction of digital literacy to both students and educators will improve their understanding of AI, thereby enabling its ethical and responsible integration into the classroom.

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