

The Ethical Implications of AI in Education: Privacy, Bias, and Accountability

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Abstract

As artificial intelligence (AI) technologies become more prevalent in education, their influence on privacy, fairness, and responsibility is under increasing scrutiny. This paper investigates the ethical implications of AI in educational settings, specifically examining concerns about student privacy, potential biases embedded in AI systems, and the accountability structures necessary to manage AI responsibly. The paper advocates for proactive policies and practices that prioritize ethical standards, thereby ensuring AI serves as a tool for equity and improved learning outcomes rather than one that exacerbates existing inequalities.

Keywords: Artificial Intelligence, Education, Ethics, Privacy, Bias, Accountability, AI Governance

1. Introduction

Introduction of Artificial Intelligence (AI) in education has become widespread on the international level, and it is expected that its implementation will lead to qualitative changes most of the sectors connected with education [1]. In recent years Institution has turned to Artificial Intelligence technologies to promote effective processes that support Independent Learning Environment and Administrative Solutions, and offer timely data feedback on student success and activity levels. These applications therefore hold the promise to meet individual learning requirements, increase effectiveness and aid in the decision making for learning professionals such as teachers and policy makers [2]. Personalized Learning: AI systems are designed to process terabytes of information about students' activities and learning profiles as well as their achievements. This permits educational agencies to provide custom material and flexible assessment, suitable to all learner's abilities and weaknesses [3]. For instance, it can control the tempo of delivery, suggest relevant learning materials and find out deficiencies that may exist in the student's understanding and thus create a unique learning trajectory that would help in improving performance and interest by the students. Administrative Efficiency: Notable here is that AI goodies offers ease in teaching and learning activities such as grading, attending taking and even time tabling's among others. Technology-based interventions, such as like automated grading systems allow instructors to quickly grade assignments so they can use their valuable time in more teacher-student direct instruction learning opportunities[4] . In the same way, the tools in attendance and behavior help execute daily routine more systematically, thus making the management of the learning process more effective and less challenging. Data-Driven Insights: Through big data and particularly data mining technique of predictive analytics, AI also helps educational institutions to make useful decisions. Through data on students' performance and attendance AI systems can be programmed to alert education institutions regarding students who are at risk of poor performance or dropout. This is useful to educators and administrators to be able to identify students who need help at an early stage so as to be given the needed help in order to enhance the students retention and success rates [5]. Other areas where the use of predictive models might serve a purpose involve adapting counselors in academics such that they can assist student's in course or career paths that may suit them best [7].

However, while these advancements are promising they actively enter novel and pressing ethical concerns [8]. AI implementation for student benefit will be at the cost of students' privacy as their data will be collected, stored, and analyzed. The nature of the questions include; who has access to this data, how the data is stored and for how long, and issues to do with the clarity of consent processes. Faculty, staff and students' personal information needs to be protected from data breaches and misuse; failure to do so could greatly endanger student privacy and erode trust [9]. Also, AI systems contain biases; in this case, they tend to magnify the biases already present in any given society. AI models are generally trained on several previous datasets and thus can sometimes replicate prejudices founded in the data set and thus is damaging to minorities. For instance, AI-aided predictive analysis may assign students from some backgrounds as at-risk regardless of their learning abilities or potentiality affecting their learning education. Some of these biases could be for the long-term and hence a child will get locked out of the system and the cycle of discrimination in the system is repeated. Finally, there is the use of accountability measures in adopting AI in education in order to make sure that the process will be responsible and humane. As more AI technologies influence decisions affecting students' education, adoption of the legal frameworks to regulate the manifold decision making processes involving students' education experiences to ensure that the institutions, developers, and other stakeholders are responsible for these decisions. AI needs to be transparent and explainable, educators, students and parents, must be able to comprehend why the results containing AI were generated.

Lack of clear accountability structures causes AI tools to be implemented with little supervision opening up the possibility that they can violate student rights or limit access to education [10].

Together, the published research provides evidence that AI systems present great possibilities for improving the educational experiences of learners while suggesting that AI should be used in a more nuanced manner that promotes ethical practices. AI needs to be responsible for privacy, free of bias and must be accountable in education to ensure that there is trust, equality and fairness. This paper discusses these ethical effects and presents a conceptual model to help elucidate and address privacy perils, prejudice, and liability related to AI in education. By so doing, the implementation of the following guidelines should seek to promote fairness when use of the AI technology is applied to educational advancement as opposed to the prevalent social inequality

2. AI in Education: Current Applications

Artificial Intelligence (AI) has taken its place as a valuable tool in today's education system that provides revolutionary solutions to improve different aspects of how students learn and how teachers facilitate the learning process, as well as how institution heads administer educational institutions. Integrated AI affairs in education are aimed at handling a number of tasks, which is to individualize the education process as well as to optimize the executive tasks. Such uses hold a lot of potential, with potential for increased effectiveness, increased levels of student involvement, and better learning. However their strong focus on data and algorithm choice presents several ethical issues that range from data privacy, algorithm explainability and algorithms equality. This section outlines some of the major use of Artificial Intelligence in education and discusses some of the ethical issues that come with each of the use.

2.1 Personalized Learning and Student Data

One of the most popular use of AI is in individualization of learning. Due to the utilization of Artificial Intelligence systems, knowledge content can be selected and adapted according to one's ability, learner's rhythm, and deficiency. For example, the use of AI yields the option of following and learning a student's behavior, learning process, and response patterns; in order to provide tailored instructions [11]. Depending on which learning activity was effective and which learning activity was not, these systems could select learning activities in real-time to meet those learning needs. Examples are intelligent tutoring systems that change exercises such as quizzes and provide hints to students when they guess wrong as well as the method of self-paced learning which adapts quizzes for students based on their answers. However, the task of delivering such contextualised experiences means that extensive data about each student must first be gathered and processed. Such data may contain performance of the student, behavior data, engagement with learning materials, and other data of the student. The collection and retention of such a variety of data raise principal privacy issues. Consumers can be students and parents, and often they are not extensively aware of the data that are collected from them, how these data are going to be used, and who will be able to use it. Thirdly, there is danger in retaining information for long because important information collected in the course of carrying out business may attract some insecurity threats. This points to the need for protection of data and clear and voluntary compliance with data protection law for stewardship of students' data.

2.2 Predictive Analytics for Student Success

Another one is predictive analysis since AI uses machine learning algorithms to analyze historical and real-time data to forecast on learner performance. These models are commonly employed to look for students who may be at risk of poor academic performance, dropping out or whatever disadvantage. Such students are best identified early in the session so that educators and administrators can guide them to seek help in form of support; undergo counseling; or receive further additional instructions that would help him/her to succeed. Predictive analytics may also help to determine the lessons chosen as it also offers better course placements for the students, potential career paths in relation to the target, improve the proper alignment of education and results. As with most emerging technologies, there is great promise for predictive analytics to improve students' success while raising important questions about ethical considerations, such as bias and fairness. Most predictive models are built with the use of data collected in the past and it is no secret that such data usually has a bias that considers the socioeconomic status, ethnicity or otherwise. This means that if the training data is bias, the resulting algorithm will be bias and will predict that a particular demographic of student is at a higher risk than the other more than it actually is. Any classification that is misleading or unjust can be catastrophic to the students so classified as it may exclude them from many opportunities [12]. For example, if a model systematically claims that students coming from some specific background are likely to perform poorly in their studies, one may unconsciously adapt to that notion and make those students be treated differently, and in this way – the model may help to perpetuate that bias. To that effect, it is necessary to implement bias identification strategies in the predictive models and develop the methods for the verification

of the data, that is fed to the models, from prejudice. Moreover, there are procedures in how such models work and who is to blame for the resulting classification, so it is clear to educators, students, and parents on how these predictions are made and how to challenge unfair predictions.

2.3 Automated Grading and Assessment

Another application area highly dependent upon AI is automated grading, which is particularly vital in grading of standardized examinations, essays and short answer questions. Applying natural language processing algorithms, the AI systems can assess and give the feedback for the student responses, which can minimize the grading load of the educators etcetera. These systems are most effective in large-scale assessment where grading is done manually would be a herculean task. In addition, the use of computers in grading enables the student to get results immediately as they enable the students to correct their mistakes. However, similar to all beneficial technologies, problems remain, specifically in relation to assessing accuracy and equity considerations of automated grading. Such designs may also not well capture the various responses that students give due to uniqueness in their perceptions or variation in language in case of multiple languages[13]. It can also be seen that grading algorithms are biased to certain response forms or patterns, which may negatively impact creative responding. Kinds of biases that could be detrimental to students in particular circumstances include ones whose deviations from expected norms impact the abilities of students who do not fit a specific mold or whose communication patterns deviate because of linguistic or cultural differences. Due to such problems, it is prospective for educational institutions to pilot and fine-tune grading algorithms to ensure its effectiveness in evaluating arrays of responses. Even for multiple-choice questions which include inclusion of human overseer in the grading process will also reduce prejudices.

2.4 Virtual Tutoring and Intelligent Tutoring Systems

Via helping students with tutors artificially implemented by AISs, virtual tutoring systems can offer students a one-on-one teaching model even when the class is not in session. These systems can respond to student inquiries, give directed explanation, and assist with problem solving and knowledge transference; some of which are reminiscent of human tutoring. Since virtual tutors can respond instantly, gaps for students who require extra help but cannot arrange for a human tutor or educator after school can be closed. However, Elliott points out that virtual tutoring entails a number of ethical concerns including dependence of the project on technological implementation and quality of interaction [14]. The AI tutors are helpful for assistance but as a result, it will be difficult for an AI to resolve queries that depict societies deep rooted reasoning and feel. This approach may also have implications on students' chances of interactions with authentic tutors – which are vital for holistic learning. It is critical to mention that when designs for AI tutoring, the invention has to augment instead of reducing human support for training. Overall, the projects involving the use of AI in education are numerous and beneficial but contain ethic issues that have to be discussed. Since the rise of the use personalization, big data, algorithmic evaluation, artificial intelligence, it is vital to tackle the matters of privacy, discrimination, equality and incapability of AI to make sound decisions. Through careful mitigation of the above challenges, it becomes possible to make AI tools to be of common utilise in learning, equitable, responsible and ethical.

3. Privacy Concerns in AI Applications

Most of the integration of AI in education requires the collection of data and this raises major issues of privacy. In the cases of adaptive learning, big data for analysis, and robotic teaching and counseling, AI interfaces seek student data and may comprise academic data, behavior record, and identity data. Given that the use of AI is continually deepening within learning institutions, the risks that data is susceptible to misuse or unauthorized access are also high; hence, privacy preservation should be of high value. In this section the author looks at the types of privacy threats and the existing privacy laws that exist to regulate the protection of student's privacy in using artificial intelligence applications including FERPA in United States and GDPR in European Union.

3.1 Data Collection and Consent

Another pressing privacy concern that arises together with using Artificial Intelligence in education is the huge amount and diversity of data that the learning system and institutions gather on students. In order to operate properly, AI systems use a variety of data inputs, including qualifications, scores, and not only behavioural data, which includes time on a task, engagement, and social interactions on learning platforms. While this data is useful for designing learning experiences, this data also contains PII that can potentially expose the most fragile aspects of a student's life, their family, and even their conduct. Often times learners or their caregivers do not have sufficient information on the extent of the information being

collected, and in what manner this information will be utilized, managed, and analyzed. This seemingly opaque collection practice questions if there is adequate and accurate informed consensus, if the students and parents fully comprehended the breadth of data consolidation and were offered sufficient control. The problem arises in cases where students are young, or in their ages they are legally incompetent and do not have the right to give informed consent, as caregivers can consent for students without understanding long-term privacy risks. To address this, there has to be clarity as to what information is being collected, for what purpose this information is being used and much more importantly, the rights that students and guardians have over collected information.

3.2 Surveillance and Data Security

The use of AI for monitoring and surveillance in education facility is on the rise as a safety measure, engagement meter, and as an administrative tool. It includes face recognition for attendance, a behavior tracking system for engagement in class, and others that make predictions of students who are likely to drop out. As these tools may enhance the management of the learning and teaching process and get valuable information about student's activity, they may lead to ethical issues that result in surveillance. Students may be given a feeling that they are being watched in some manner, which in some way could debilitate their liberty and cause them to lose trust in the administration. One more important issue which can be pointed out is the security of collected data. Schools and other education institutions do not possess the advanced protection means required to safeguard a significant number of students' personal records from hacking and cybertroubling. By obtaining and using this information without proper permission not only violates student rights to privacy but also opens the door to Identity theft or Reputation depreciation in future. To manage these risks, institutions must establish protocols that will protect client and consumer data, encrypt identified data and also conduct regular assessments to identify possible weaknesses in their systems. In addition, laws such as FERPA and GDPR may GUIDE institutions on how to handle and process consumer data securely and ethically, however, for them to serve the purpose, they have to be strictly implemented.

4. Bias in AI Systems: Implications for Equity

With adoption of AI in learner management for purposes of grading, admission and student evaluation among others, society's prejudices find their way into these systems and are even reinforced. The central drawback of most AI algorithms is that the former is developed with an intention to learn from history which itself perhaps harbors bias. Training data contain biases that limit the opportunities for marginalized or underrepresented student populations if not expunged, threatening Education by AI applications. In this section, the author describes how bias within AI impacts learning for learners and how the latter can be solved.

4.1 Algorithmic Bias in Educational Assessment

In this area, there are high chances which reveal that AI bias obtains its footing among the most significant domains. For example, automated grading systems are designed to use machine learning to grade assignments and tests; however, these very algorithms may reproduce certain discursive practices and reward particular language use, or, conversely, punish the deviation from such patterns. In addition, there are biases that are instantiated in students' assessment for admission or placement since such models learn from past data that may be prejudiced by past discrimination. For example, if certain statistics show that a school has always discriminated kids based on race, gender or social class, an AI system will hold similar prejudices, giving less chances to the underprivileged children. It may be seen that introduction of unbiased AI based methods can lead to middle bias, and this is evident from case studies showing that AI systems can be prejudiced. The automated grading system has in some occasions penalized student from the concerned minority or even labeled them as underachievers despite excellent scores. To avoid such biases there is a need to ensure diverse and inclusive training sample, sampling for auditing the inputs as well as developing fairness metrics to track performance across the included demography. In particular, overcoming reliance on remedies based solely on the ideas of artificial intelligence can be facilitated by the inclusion of human control in this process, which will allow us to eliminate biases that the artificial intelligence system is unable to distinguish.

4.2 Implicit Bias in Student Performance Analytics

For early intervention, another efficiency of predictive analytics is targeting students most likely to regress and the best approaches to use with them. However, the criteria by which these determinations are made are often informed by predictive algorithms which may encapsulate preconceptions, hence limiting some cohorts of students. For example, predictive algorithms that include a factor of socioeconomic status or demography will label students from poor or black families as more risky. Despite the good intentions of creating these classifications for provision of differentiated support, the kind of

classification used can lead to stigmatization of students and differentiation of attention or resource provisions on the basis of these classifications. The implications of such biases are plentiful. Learners who are misidentified as vulnerable may be expected less academically by teachers, excluded from honors or AP classes or receive special attention which alters their demeanor or grades. Moreover, these biases can become self-reinforcing: the at-risk status derived from the preposterous predictions would mean that students so labeled receive differential treatment that influences their performance as they confirm the categorization. Overcoming such biases, thus, calls for fair and equitable approach to building and deploying predictive models. For instance, institutions should use approaches like bias sensing and equity validation so as to determine how such models affect the students. Where such instrumental uses of sensitive attributes are not prohibited (for instance where one has to use race, gender, level of income etc in model construction), extra caution must be taken to ensure that their use does not lead to disparate treatment of members of one group as compared to others. Also, when we refer to the 'human-in-the-loop' approach, appropriate to AI applications in education and counseling, educators and counselors should always review the data in which AI-based decisions are grounded, which means that the problem of reliance on biased prediction will definitely arise.

5. Accountability in AI-Driven Education

Any AI to be used in educational systems has to be fully accountable to avoid the misuse of the technology which will lead to loss of trust from the general public. Accountability means that some individuals or institutions **MUST** be held responsible for the growth, deployment, as well as supervision of AI uses in education. At the same time, one must realise that without proper accountability mechanisms, AI may decide something, or affect an outcome, in a completely opaque and potentially negative way for any student or other stakeholder. This section explores the fundamental pieces required to integrate accountability into AI systems in education: Openness, interpretability and sensible management frameworks.

5.1 Transparency and Explain ability

There is a strong focus on the need to make AI involved in education as transparent as possible as well as making them easily explainable to stakeholders. Transparency is about the educational institution or technology's provider's understanding of how an AI system works, what data it employs, and the rationale for its actions. In other words, explain ability entails offering clear rationales for decision making power by an AI. For example, if an AI system assigns a student a poor grade of 'at risk' or puts a student in a particular learning stream, the users, whether students, parents, and teachers, should be able to understand how and why the AI system arrived at such decisions. This paper asserts that explain ability must occur so that stakeholders can ask for reasons, demand change, and demand further understanding of decisions made by AI systems. It also helps institutions do an outlook of the system and correct any errors or bias that might be there. For example, if a grading AI gives certain grades that are not normal, educators should be able to know why this AI has graded like that, so that they can counter check the algorithms, with a view of correcting the wrong results generated by those algorithms. secondly, explain ability enhances accountability through traceability to make sure that the AI system is not only serving its educational purpose but is also not causing harm where it is not wanted. Interpretable machine learning models that provide information on feature inputs that affect outputs, and AI tools developed for educators and administrators can reduce opacity of AI systems in learning environments.

5.2 Ethical AI Governance in Education

In order to maintain basic ethical values of AI in educational processes, institutions should establish the governance structures that would regulate AI designing and using. Ethical AI management can include defining rules and frameworks within which the AI system will operate, this can also involve measures that have to be taken in order to ensure that all forms of AI applications run in accordance with standard norms of ethical practices as well as in compliance with legal requirements. There are various aspects to efficient governance structure; these are; engagement of stakeholders, legal requirements/regs and Audit. Another important aspect of governance is to establish committees or ethics squads for analysing the applications of artificial intelligence in educational organisations. These boards can weigh probability of risk, check algorithm for bias and consult on the most appropriate use of it. Besides, they can negotiate with the developers of the technology as well as mediate between the technology and the school community in order to guarantee that every AI implementation upholds the institutional norms and objectives of the learning institution. To avoid being biased the AI systems should have multiple stakeholders, for example, students, parents, lecturers, and external ethicists involved in decision making. These are seen to benefit in the sense that they enable the finding of concerns that may not be subscribed from a technical viewpoint or lens such as how students feel or are affected by AI-based assessments. Besides, ethical AI governance frameworks should follow current rules and act in accordance with the requirements of the applicable legislation, including FERPA and GDPR, which is crucial to protect the rights of students and privacy.

6. Ethical Guidelines and Recommendations

For AI in education to meet the ethical benchmark, institution and technology provider ought to adhere to the following seven principles of ethical guidelines. These guidelines should include protection of confidentiality, non-discriminatory language and authorising responsibility for work produced. Below are key recommendations to help guide the responsible use of AI in educational environments:

- **Implement Strong Privacy Protections and Transparent Consent Processes:** Institutions' educational data must minimize data collection to retain privacy, encrypt data and adhere to privacy acts. The consent procedures must be clear to indicate to the students and the guardians, the data that is being collected, why it is being collected, and how it is going to be utilized. This is by ensuring they make it easier to opt out of sharing the data and respecting the rights of students and parents to choose exactly what data will be shared.
- **Develop Bias-Detection Protocols and Regularly Audit AI Algorithms for Equity:** Schools should periodically analyze their AI driven algorithms for bias that may prejudice one particular group of students over the others. Facial recognition technologies along with bias-detection protocols such as fairness metrics and representative training datasets can help mitigate discriminative outcomes. They should be fully disclosed; results and recommendations ought to be publicly disclosed to prevent the building of a culture of mistrust regarding the functional efficacy of AI systems as well as maintaining this equitable recognition.
- **Establish Accountability Frameworks and Ethical AI Governance Boards:** Policing of AI applications can also be maintained through frameworks of accountability created specially within the educational setting. The ethical AI governance boards should be composed of stakeholders who are supposed to supervise AI functions as well as respond to new ethical concerns and modify the guidelines in the process. These boards should also allow for a constant process of improvement to reflect results from the feedback from students, educators and parents. As a whole, all these suggestions form a starting point for ethical AI use in education that can help to build up practice protecting educational aims and result from negative influence while using AI technologies.

7. Conclusion

Future of artificial intelligence in education is quite promising and has to do with personalized learning, using data analysis to anticipate student needs, use of intelligent administrative systems and intelligent approaches to test, assessment, and evaluation. Nevertheless, these advantages cannot be obtained without a number of ethical imperative obligations. To guarantee that the AI applications function for the public benefit of promoting equality, inclusion, and transparency in education, privacy risks, biases and accountability need to be found and mitigated. As this paper shows, there is a significant lack of discussion as to what exactly is ethical within the setting of AI in education, how practices should be conducted, and what principles should underpin AI governance. By affording priority to privacy protections, algo sensing and accountability, educational institutions can harness the AI to facilitate students a fair and ethical learning environment and opportunities. Overall, it is crucial to remember that with the proper levels of monitoring and compliance with proper ethical requirements, AI can become an instrument in the formation of better education for every learner.

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