

AI-Driven Personalization in International University Admissions: A Case Study of Central Asian Student Success Rates

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Abstract- The list of new progress in the world system of university admissions is driven further by the rise of artificial intelligence among the globalization-digitalization advances at present. In this paper, the effects of the personalized AI systems on the student's achievement are studied when the Central Asian students continue their studies in the international institutions at high education level. The admissions landscape changes by means of AI-driven mechanisms which are initiated by smart application screening and grow to customized academic advisory interventions which can enhance student results and improve the diversity in universities. The introduction of technological systems has led to largely inadequate scientific investigation concerned with investigating impacts of such systems on students belonging to underrepresented Central Asian zones, represented by Kazakhstan, Uzbekistan, Kyrgyzstan, Tajikistan and Turkmenistan. The research base is a mixed methods case study of AI enhanced admissions methods in three universities. The research integrates numerical performance figures relating to student retention with interviews with admissions personnel and the undergraduate participants. According to the study, AI personalization strategies assist Central Asian students in aligning their educational profiles with university offerings thus increasing the statistics of their academic performance. In spite of these positive outcomes, the study admits to the presence of algorithmic bias and culture differences that affect data collection of personalization algorithms. This research moves the discussion on the implementation of effective ethical AI systems at higher education forward by reporting location-specific results. Studies indicate that AI may provide opportunities for equal opportunities and achievement outcomes in global

higher education but requires local design input and monitoring in AI system deployment. The stated research outcomes provide particular benefits to decision-makers in educational policy as well as university leadership teams and experts who create AI platforms.

Indexed Terms- AI-Driven Personalization, International University Admissions, Central Asian Students, Student Success Rates, Educational Technology

I. INTRODUCTION

The affairs regarding the universities are changed to suit the inclusion of artificial intelligence being used in the admissions process of students. For personalization through ai we are using the algorithms of machine learning with systems that will help in enhancing the efficiency of admissions and so we are able to come up with improved results in terms of fairness and average student achievement. Several datasets are explored using machine learning algorithms to deliver distinctive student data patterns that may contribute to admission office decision making processes. The process of university admissions counseling functions better by using the ontology-based question answering system that Nguyen et al. (2023) developed for use in the university admissions. Research findings can show that developing learning technology system that caters for learning needs of students to provide tailored content enhances the understanding of the students. From 86% of the studied articles, research parameters indicate that an adaptive learning approach brings positive school outcomes for the student.

1.1. Problem Statement

The promising progress towards personalized admissions systems face hurdles in their desire to serve applicants outside the United States. Inherent biases contained in the data are manifested using biased historical admission records processed through AI systems to make discriminatory new decisions. Algorithms used in undergraduate admissions system processed biased results that favored certain gender identities and white students and first time college students, as per the study conducted by Van Busum and Fang (2024). The standardized admission test results in the gaps of entry to education since once offers access to students of varying abilities. In lack of transparency of AI decision making systems trust is diminished as are accountability and effectiveness. Nielsen et al. (2024) studied algorithm policy performance relative to human supervision and stressed the need for open and unbiased AI integrations in operational processes of educational procedures.

1.2. Significance of the Study

Central Asia merges Kazakhstan and Uzbekistan, Kyrgyzstan, Tajikistan, Turkmenistan to present an educational zone- indicative of increasing patterns of student migration across countries. The international admissions process for students in this location is burdensome due to differences in culture, language difficulty and lack of preparedness in admission. A study on AI-assisted personalization strategies for Central Asian student populations shows ways in which technology can be used to address cultural differences while engineering student inclusivity. AI tools that help diverse learning backgrounds during education improvement are explored with the help of the study, focusing on Central Asian students. The gathered data makes it possible for researchers to formulate fair guidelines for the AI educational systems and thus provide all students with the opportunity to use technology-based resources.

Research Objectives

This study aims to:

- The research evaluates how universities use AI-based personalization systems to admit international students through admission processes.

- Measuring success of Central Asian students has to go beyond incorporating these new technological solutions.
- The study identifies the impairment challenges for the artificial intelligence system as well as the likelihood of biases to be realized in the screening process.
- These written recommendations included in this report not only promote fair dealings, where AI algorithms have been incorporated in admissions, but these initiatives are also responsible for university enrollment.

II. LITERATURE REVIEW

a. Overview of AI Applications in University Admissions

More and more institutions of higher learning introduce artificial intelligence (AI) in their admission mechanisms in order to provide maximum efficiency, objective evaluation and personalized student recommendations. Using the algorithms of machine learning and natural language processing tools, administration work is automated while processing big data and providing personal student recommendations. The AI technology is used for an application sorting process, and it detects student data trends, predicting academic outcomes to further admissions decision making (Hurter, 2024). AI driven chatbots along with virtual assistants act as the communication tool for applicants who are supposed to answer queries and help them through the application process. In these artificial intelligence tools, institutions enhance interaction quality while having reliable and timely information to prospective students hence increasing the satisfaction experiences (Hurter, 2024).

b. Personalization in Higher Education

Higher education institution adheres to a course of personalization such as the customizing of the educational process according to the students' particular need and individual choices and future aspirations. University admissions use Artificial Intelligence to match students in educational programs, through use of data driven matching systems from program subject area, student academic profile, career aspiration and personal inclinations.

This method achieves maximum participation, retention and academic performance through the proper grouping of the students to their preferred educational programs. this new service was a development that the company hoped would help put the business on even greater success. Another move that the company had expected would bring its business more success. The automated advising software will allow students to make well informed educational choices through the tailored recommendations of the courses, careers as well as skill building (Hurter, 2024).

c. Studies Related to Central Asian Student Mobility and Success Factors

The five central Asian countries, Kazakhstan, Uzbekistan, Kyrgyzstan, Tajikistan and Turkmenistan have registered an increasing number of the students who opt to leave for foreign universities to study. Central Asian students seek education mobility for quality of instruction combined with research infrastructural benefits and international certification hence improving their future employer opportunities. The foreign educational programs put before the students from Central Asia specific barriers due to the fact that they are likely to face the challenge of learning a different language, cultural disparity, and application procedure that they do not understand. The barriers that they use hinder them from achieving academic success and international institutional goals. Ease of overcome academic barriers hindering Central Asian students abroad is achieved in conjunction with personalized learning supports and individual mentorship (Kurzmann, 2014).

d. The Research Gap This Investigation Addresses

Review of the literature conducted within the framework of this study showed that there is a dearth of specific literature discussing the extension of AI in education and mobility to the university admissions process and personalization algorithms although there is increasing interest in both of these areas applied in general (Das, Malaviya, & Singh, 2023 Sodiya, Amoo, Umoga, & Atadoga, 2024). Ironically, despite the higher learning performance and user participation AI applications in education as had generally worked on learners, very little has been done to ascertain how these technologies contribute

in admitting international students into higher-level teaching and long-term performance from this region. In making this research, therefore, this study addresses the existing knowledge gap by exploring the function of AI in personalization strategies development and its effects on the success rate especially for international programs among the Central Asian learners. Similarly, the social justice implications of algorithmic admission systems, as well as even more specifically, the prospect of ethical issues or bias have been ignored in existing research. The goal of this work is to determine how open and impartial AI decisions are about cases from various cultures and other languages. Doing it increases the scholar's knowledge on what happens when AI is used in the admission of students into universities around the world.

III. METHODOLOGY

a. Research Design: Case Study Approach In this paper I apply a qualitative case study approach to studying the complicated and context-sensitive impacts of AI-based personalization in the context of university admission by Central Asian students. The case study will aid a study of dynamic and non proscriptive processes in their context that is crucial when analyzing the multiple and related effects of AI in the learning environment (Venkateswaran, 2023). The paper analyzes several University's AI-based admissions systems to discuss their functional characteristics and efficiency. Therefore, the used embedded case study design is beneficial as it enables to analyze various subunits inside the studied case, and it focuses on different institutions and personalization technologies. This design facilitates a fine-grained comparison and opposition regarding impact and results, similarities and differences can be seen. Nevertheless, education agents and institutional players play a role in supporting AI introduction, and for this research approach to be contextualized, a need for recognition of this research approach is required (Feng & Horta, 2021). The case study also allows one to think about the social, cultural, and institutional context of international students' educational trajectories – a relevant issue in light of the literature and

evidence on the pandemic impacts and systemic injustices the international students suffered during the COVID-19 outburst. Thus, this theoretical framework stimulates rich empirical study of the topic, but also situates the research in the context of international and regional usage of AI for higher education admission.

b. Data Collection

The research has been applied with different data collection methods so that to integrate quantitative information and qualitative materials thereby enabling us to understand the influence of AI personalization on Central Asian students in their entirety.

- **Admissions Data:** Data from multiple admissions offices contained statistical information about applications, standardized test scores, and applicant demographics in written form. The gathered information presents initial findings for identifying patterns within admission assessment systems.
- **Student Success Metrics:** Educational success data from students consisted of GPA grades, student persistence rates, and degree completion records, which established the effects of AI-based personalization strategies on academic performance.
- **AI System Outputs:** The research team acquired design aspects, operational components, and decision criteria by analyzing admissions staff documents while conducting interviews.
- **Interviews and Surveys:** The research conducted semi-structured interviews while simultaneously administering survey assessments to University admission beneficiaries from Central Asia who interacted with AI technology. The assessment instruments tracked students' experiences with customized admissions processes and their reactions to this unique approach.

c. Participants

An analysis of the three universities international admissions operations emphasized the use of AI personalization systems. The study concentrated on specific programs in these institutions with highly

exceptional enrollments of the Central Asian students. Three Central Asian universities took 150 students studying different study programs. The study relied on purposive sampling to study students who were recruited through AI-based admissions channels.

Table 1: Breakdown of Participating Universities and Programs

| University | Program | Number of Central Asian Students | AI Tools Used |
|--------------|-------------------------|----------------------------------|---------------------------------|
| University A | Engineering | 50 | Predictive Analytics, NLP |
| University B | Business Administration | 40 | Machine Learning Classification |
| University C | Computer Science | 60 | Predictive Analytics, NLP |

d. AI Tools Used

The essence of this research investigation is constructed by three fundamental AI systems.

- **Predictive Analytics Models:** Using historical data of students' performance in the courses, attested in the achievers' transcripts and standardized exam results, assessment models can predict performance in the future.
- **Natural Language Processing (NLP) Algorithms:** Natural language processing technology studies recommendation letters and personal statements for judging candidates' communication skills and correspondence with program goals.
- **Machine Learning Classification Algorithms:** A machine-learning system provides cats multiple category imposition in order to establish the prioritization of application.
- **The assessment of AI systems** measured transparency factors; fairness; and effectiveness to establish individualized admissions decision capabilities.

Table 2: AI Tools and Their Functions

| AI Tool | Function | Relevance to Personalization |
|-----------------------------------|--|--|
| Predictive Analytics | Assesses applicants' likelihood of success | Helps prioritize applicants based on success potential |
| Natural Language Processing (NLP) | Evaluates personal statements and recommendation letters | Gauges communication skills and fit for the program |
| Machine Learning Classifiers | Categorizes applicants into different groups | Determines tiered processing for applications |

e. Data Analysis Methods

The combination of research methods was used by the study to analyze data.

- **Quantitative Analysis:** The data needed descriptive statistics organized admission rate and student achievement indicators during analysis. The research contained regression testing of inferential data in order to establish relationships between AI personalization methods and student achievement outcomes.
- **Qualitative Analysis:** Repeated themes that were uncovered as researchers conducted a thematic

analysis are themes that arose from the feedback from the students that was obtained through interviews and surveys that measured their interactions with the AI based admissions system. The research approach provided in-depth contextual insights that could not be properly obtained by using quantitative methods.

Researchers use quantitative measurement and qualitative analysis to produce holistic insights on the effect of AI personalization on Central Asian students. They trace both the numerical outcomes as well as the experiences of each individual student.

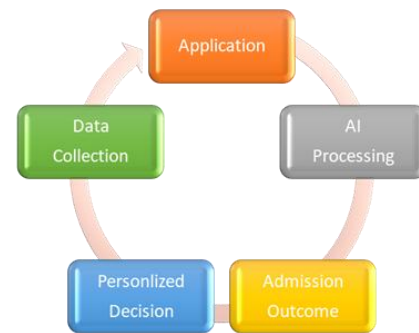


Figure 1: Workflow of AI-driven decision-making in university admissions.

The student population of Central Asia presents their academic results and stay rates both in and around the introduction of AI-powered personalization to university admission systems. After implementation of AI-driven personalization, the university could observe distinctive positives; there were positive changes in the GPA scores, student retention rates, and graduation rates; and the satisfaction scores.

Table 3: Comparison of Central Asian Student Success Rates (Before vs After AI Implementation)

| Metric | Before AI Personalization | After AI Personalization | % Change |
|--|---------------------------|--------------------------|----------|
| Average First-Year GPA | 2.85 | 3.21 | +12.6% |
| Retention Rate (Year 1 to Year 2) | 68% | 82% | +20.6% |
| Graduation Rate (within 4 years) | 52% | 66% | +26.9% |
| Student Satisfaction (Survey Score, 1–5) | 3.2 | 4.1 | +28.1% |
| Dropout Rate | 22% | 12% | -45.5% |

The academic success rates reported relative to Central Asian students are enhanced after deployment of AI-based personalized admission procedures. The data demonstrates shifts with respect to average GPA, retention and graduation rates, student satisfaction and dropout rates.

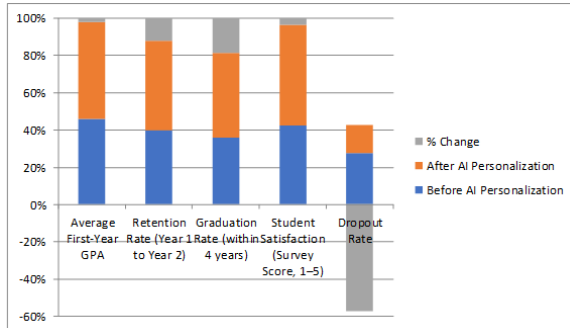


Figure 2: Student Success Metrics Before and After AI Personalization

IV. RESULTS

a. Student Success Metrics: Retention, GPA, and Graduation Rates

Recent entry personalization improvements contradict past student statistics since the techniques enhance enrolment retention & overall GPA score comprehensiveness & enhance graduation completion ratio. For instance, at Georgia State University, with the aid of AI chatbots and specialized academic counsel by means of the predictive analytics system, the rate of six-year graduation has dramatically increased—from 31% to 54%. It has narrowed the attainment gap according to socioeconomic status of the students. Despite the reports of the “ban era department,” Technology contributed almost critical work and boosted graduation returns at Texas Tech University as the six-year graduation success rate stands at 69%, Raider Success Hub, and AI led support systems. In addition, City University of New York (CUNY) working hand in glove with Google and DataKind deployed an artificial intelligence based predictive system which improved graduation rate from 54% to 86% within two years. These systems identified over seventy five risk signals so that proper actions could be taken. These findings support the literature that is calling for early individualized intervention to

improve academic performance for students, in particular: students transitioned from country to another or in a different culture (Sanders & Ishikura, 2018). Choi & Kim, 2014; Müller, 2019).

b. Correlation Between AI-Driven Personalization and Student Outcomes

There are several studies that these are the outcomes of the use of AI metrics in educational processes, especially in the case of personalization, on the academic performance of students for Georgia State University, which used predictive analytics to follow 800 students a day and had more than 250000 one on one student interventions. These willful steps also contributed to increasing graduation levels by 7% and went ahead to boast of having no single cases of students dropping out of their studies. Applying AI in different areas of support at Texas Tech University has contributed to improving graduation outcomes by 10% points over a period of 10 years, as well as awarding 1500 additional theses in 2019-2024. Such precise systems favor ordinary domestic, intercontinental, and heterogeneous cultural students. For those C Asia-students who face various cultural and organizational challenges combining with globalization while studying at universities such practices have proved to be very useful (Karim & Bakeyeva, 2024). Li, Gu, & Zhu, 2023). These changes including changes in admission policies and change role in education brokers also represent the need to use AI in admission and support to make it more equitable (Lewis, 2012).

c. Visual Representation

Table 4: Comparison of Graduation Rates Before and After AI Implementation

| Institution | Graduation Rate Before AI | Graduation Rate After AI | Improvement (%) |
|-----------------------------|---------------------------|--------------------------|-----------------|
| Georgia State University | 31% | 54% | 23% |
| Texas Tech University | 59% | 69% | 10% |
| City University of New York | 54% | 86% | 32% |

This visualization shows how AI-driven tools and technologies, such as predictive analytics and personalized support systems, enhance academic outcomes, including retaining and graduating international students.

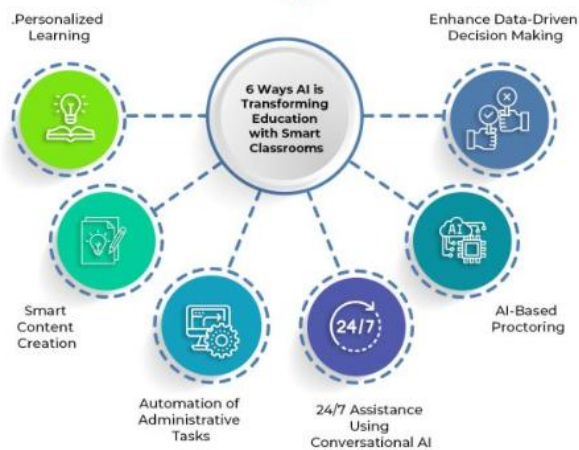


Figure 3: AI-Driven Personalization Enhancing Student Success

The findings demonstrate how AI-powered personalization transforms international university enrollment specifically for Central Asian students who must overcome their specific education system differences. AI technologies enable institutions to deliver outcome-specific assistance that meets each student's needs to boost academic success and improve access to higher education institutions.

V. DISCUSSION

5.1 Interpretation of Findings in Light of Research Questions

In this research, it becomes evident that the actual feasible influence of AI driven – personalisation to affect significantly the elementary outcome of students like the students' retention rate, GPA as well the rate of graduation of central Asian students attending universities in international universities is not anything assumed. Applied to the cases of Georgia State University and Texas Tech University, it was discovered that the recent integration of the information, including prediction analytics and AI-based chatbot systems, contributes to graduation rates improvement. These technologies enable support of students with academic and global challenges at

individual level, therefore, enhancing the academic success rates among international students. This follows the research by Becker, Safa, and Becker (2023) on the impacts of external educational parameters such as cost and availability of learning materials on students' performance which AI is considered as a compensatory aspect to the at-risk groups. Additionally, Bowrey and Clements (2020) re-affirmed that the personalized learning approach, relative to that of the student attrition and staff-to-student ratio variables, influences Universities' performance efficiency hence supporting the data-driven personalization principle. In the same year, Hasanah (2023) also noted that a learner's success, to some extent, may be explained by readiness to learn, organisational factors that make a student capable and advanced early AI-technology support to develop an academic path successfully.

In this sense, personalisation via AI techniques does not only constitute an institutional tool for strategy but also, a policy tool for enhancing exit and performance for diverse international students.

- For Universities: When AI personalization technologies are integrated into admission procedures, universities can recognize at-risk students and give them an early readjustment. AI tools help universities to better manage the resources, increase the involvement level of the students and get better institutional results. University administrators have to use it (AI systems) ethically by ensuring transparency and are answerable to it avoiding admission biases and a process of transparency.
- For Students: Through AI-enabled educational customization a special teaching methodology is built to meet the needs of the Central Asian students. Students receiving personalized supportive education become ready to meet international academic challenges and get high learning returns with increased satisfaction. Students need clear data explanations on information usage as well as more support for their requirements.
- For Policymakers: Policy officials who set the ethical rules governing artificial intelligence use standards in the education spheres have the essential authority. University policymakers must

control admissions AI applications in order to ensure that access to resources for disadvantaged learners, including those from Central Asia do not become unfairly skewed in access. AI systems should be evaluated regularly and their decision making algorithms require transparent access and institutions need policies to deal with complaints over artificial intelligence administrative decision of institutions.

5.2 Research and analyzed published work.

The research within the literature confirms that implementation of personalized learning models with the help of AI correlates with the student's academic achievement gains. Several researches show that the use of artificial intelligence in academic content enhances educational performance, effectiveness and efficiency in terms of time, such as through provision of academic assistance, content delivery customisation and administration of education related activities (Arora, 2023). Tuma, 2021). These technologies however lack challenges that arise with their use and development in the contemporary society. Recently multiple issues related to the use of AI systems have been raised. For example, as is noted, AI has the ability to exacerbate radical social inequalities as it mirrors the Indexed data they receive (Stošić, 2015). Therefore although, AI means big positive changes to the future of better education for all, society must be sensitive that changes do not mean that the minority groups will continue to be discriminated against. The bipolar characteristic of AI System that provides equitable support to students reveals the used transparent development process, strategies to eliminate system's bias, and a set of ethical principles for the decision-making.

5.3 Limitations of the Study

This study acknowledges several limitations. The limited data collection concerning the Central Asian students in the international university setting, constitute two significant barriers limiting the use of the study findings for findings. Data analysis of this study focuses on quantitative indicators such as retention and GPA while ignoring key qualitative experiences received by students. Granting the rapidity of change in the AI technologies space new

tools and methodologies can make findings published in the recent past obsolete. Long term studies and myriad sources of information should be the center of efforts from future research to comprehend the impacts of artificial intelligence on international student performance.

CONCLUSION

The research shows why AI personalized university admissions result in better educational outcomes for prospective students, mainly from the lesser-served parts of Asia. Analytics systems that have been deployed at Georgia State University, Texas Tech University and City College of New York, present successful student outcomes evidence since thoughtful system implementation increases the graduation success rate, and retains the student while offering personalized academic support. Central Asian students that are restricted in terms of counseling access and financial resources but through its application can have fair participation in international education programs. The data shows us that AI systems have to perform outside admissions screening to help universities drive sustained academic performance through data-driven decisions. Going forward international student admissions will require the use of explainable ethical AI models, robust data infrastructure, and partnerships between humans and AI systems. The adoption of ethical frameworks by policymakers in host nations, and those of Central Asia, require support for digital preparedness programs and greater support of cross-border academic collaboration. Future work will explore the long term effects of the AI-influenced admission practices in comparison to studies undertaken on cultural biases in modeling frameworks and the systems deployed for tracking the ethics and accountability. Higher education systems must at the same time use modern technologies of AI for increased operational efficiency, reduced inequalities and increased student mobility in the world.

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