

Economic Impact Analysis: Cost-Benefit Assessment of AI-Driven Mental Health Support Systems in Public Education

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Abstract- *This article examines the financial implications and economic benefits of implementing AI-driven mental health support systems in public schools. As student mental health challenges continue to rise, traditional support mechanisms often fall short due to resource constraints and high costs. This study explores how AI technologies can bridge these gaps, offering early detection, counseling, and intervention capabilities that contribute to a more comprehensive mental health strategy. Key findings indicate that AI integration can substantially reduce costs related to crisis intervention. Also, AI-driven support enhances academic performance, contributing to higher graduation rates and improved long-term economic prospects. These systems also contribute to broader societal costs associated with untreated mental health issues, such as unemployment and healthcare burdens. By strategically investing in AI solutions, policymakers can ensure immediate educational benefits with an enduring societal and economic growth, effectively creating the need for regulatory frameworks that address data security and ethical concerns.*

Indexed Terms- *AI mental health, public education, financial implications, crisis intervention, academic performance, societal benefits, early intervention, data security.*

I. INTRODUCTION

In recent years, mental health challenges among students in public education have risen at an unprecedented rate, becoming a critical concern for educators, policymakers and families alike. According to the 2024 World Health Organization reports, globally one in seven individuals aged 10-19 experiences a mental disorder, contributing to 15% of the global burden of disease within this age group

(World Health Organization, 2024). Health systems worldwide have not sufficiently addressed the needs of those with mental disorders, leaving a large gap between the demand for treatment and the quality of care provided (World Health Organization 2022). These challenges have been further increased by societal stressors such as economic instability, social isolation, and the aftermath of the COVID-19 pandemic. Schools, often viewed as the primary settings for addressing student well-being, have struggled to provide adequate support, leaving gaps in intervention and treatment caused by limited funding (Heinrich et al 2023).

Traditional mental health support systems in public schools, while essential, often fail to meet the growing demand for services. Limited funding, a shortage of qualified mental health professionals, and fragmented service delivery models hinder the ability to offer timely and effective interventions. The financial burden of hiring and maintaining mental health staff, compounded by issues like lack of insurance coverage, inadequate community-based interventions, unequal access to evidence-based practices, stigma, workforce shortages, and geographical provider maldistribution, poses significant challenges for many underfunded U.S. districts. These systemic inefficiencies creates the need for innovative, scalable, and cost-effective solutions (Mongelli et al., 2020).

This article aims to evaluate the cost-benefit dynamics of implementing AI-driven mental health support systems in public education. By examining financial implications such as reduced crisis intervention costs, enhanced academic performance, and the long-term economic benefits of early mental health intervention, the research provides a comprehensive framework for assessing the feasibility and sustainability of such technologies.

The findings of this analysis hold significant implications for policymakers and educators striving to optimize resource allocation within public education. By identifying the economic and social benefits of AI-driven mental health interventions, this research supports informed decision-making and long-term planning. Furthermore, the study sheds light on how early mental health support can mitigate future costs, reduce dropout rates, and improve overall student outcomes. These insights not only address immediate challenges but also contribute to building a more resilient and inclusive educational system for the future.

II. LITERATURE REVIEW

- **AI in Mental Health**

Existing literature underscores the transformative potential of artificial intelligence (AI) in addressing mental health challenges. Olawale et al. (2024) emphasize AI's immense potential in mental health care, highlighting applications such as early detection of mental health disorders, personalized treatment plans, and AI-driven virtual therapists. Marr (2023) highlights that AI can predict which patients are likely to benefit from cognitive behavioral therapy (CBT), reducing the need for medication. This approach is significant because it can help avoid the life-limiting side effects of antidepressant and antipsychotic drugs, ultimately improving patient outcomes. Alhuwaydi (2024) similarly notes that predictive analytics, which uses historical data to create preventative interventions, aligns with the move toward personalized and preventive mental healthcare. In screening and diagnostics, AI technologies like machine learning and deep learning have proven effective in analyzing mental health data and identifying patterns associated with various conditions. Haque & Rubya (2023) explain that AI-powered mental health chatbots offer support, guidance, and resources through conversational interfaces, simulating human interactions to provide personalized care for issues like anxiety, depression, and stress, and sometimes connecting users with mental health professionals. These advancements demonstrate AI's utility in early diagnosis through predictive analytics, leveraging behavioral and speech patterns to identify risks. They emphasize AI's potential to complement human expertise, particularly

in areas with limited mental health professionals. However, ethical concerns, such as data privacy and bias in AI algorithms, remain critical issues that need addressing.

Economic Models in Education

Economic evaluations of educational interventions are well-documented, often utilizing cost-effectiveness analyses to guide resource allocation. Levin and McEwan (2001) developed a framework to assess the costs of educational programs relative to their benefits, such as improved academic outcomes and reduced dropout rates. Expanding on this, Knapp and Wong (2020) provided a detailed update on economic evidence related to mental health, offering a lifespan perspective that analyzes costs and outcomes to highlight various critical issues. Nazari et al. (2023) found that web-based educational interventions improved mental health literacy but did not significantly impact stigma or help-seeking behaviors. They recommended using more rigorous methods, comprehensive and tailored interventions, and collaborative approaches to better address these issues among young people. Additionally, Daniele et al. (2023) noted that many interventions focused more on providing information through activities like lessons, rather than being participatory in nature.

Case Studies in Public Schools

The use of AI in public schools to address mental health needs is limited but growing. One prominent example is the implementation of AI-based monitoring systems in North Carolina schools. Tools like Gaggle analyze students' online activities to detect mental health risks. As of now, Gaggle is safeguarding over 440,000 students across 67 districts in the state. Data from the 2020–21 school year reveal a substantial number of student crises, with 40% of incidents flagged for suicide and self-harm. Notably, these interventions have saved 113 student lives, marking a 77% increase from the previous year. Moreover, there were over 23,000 student safety incidents that required district attention, as reported in 2022 by the North Carolina Department of Public Instruction. Gaggle monitors students' Google Drive and Outlook Email for issues related to self-harm, suicide, harassment, substance abuse, violence, nudity, and sexual content. Despite these achievements, there are ongoing concerns about the surveillance aspect and the

potential stigmatization of students who are flagged by the system (North State Journal, 2022; Gaggle 2022; gaggle, 2022).

Knowledge While existing literature provides valuable insights into AI-driven mental health interventions, significant gaps remain. Large-scale economic analyses of such interventions are limited, particularly concerning their cost-effectiveness and long-term impacts on academic outcomes and workforce readiness. A systematic review by Wolff et al. (2020) highlighted the scarcity of comprehensive cost-effectiveness studies dedicated to AI in healthcare, noting that many do not meet established quality criteria. Moreover, there is a lack of longitudinal studies assessing the sustainability and scalability of AI technologies in diverse educational contexts. While AI methodologies applied to longitudinal data have shown promise in bolstering accurate prognoses for psychiatric patients, their application within educational settings remains underexplored (Uwaoma et al, 2023). Addressing these gaps is critical for building a robust evidence base to guide policy and practice in integrating AI-driven mental health support systems in public education.

III. METHODOLOGY

- **Framework for Analysis**

The study employs a cost-benefit analysis framework to evaluate the financial and economic implications of implementing AI-driven mental health support systems in public schools. The framework categorizes costs into direct and indirect components. Direct costs include expenditures on AI technology acquisition, staff training, and initial implementation. Indirect costs cover ongoing system maintenance, software updates, and periodic performance evaluations. The framework considers the cost of data privacy compliance, which is crucial given the sensitive nature of student mental health information. The analysis also incorporates non-monetary factors such as the potential impact on student well-being and overall school culture. This holistic approach aims to capture both the tangible and intangible aspects of implementing AI technologies in educational contexts.

- **Data Sources**

The analysis draws on diverse and verifiable datasets to provide a strong foundation for evaluation. Key data sources include statistical records on the costs and effectiveness of traditional mental health interventions, educational outcome metrics (such as graduation rates, academic performance trends, and behavioral indicators), and detailed financial reports on AI implementations in similar contexts. These datasets were obtained from peer-reviewed studies, government databases, and industry benchmarks, ensuring reliability and traceability. Comprehensive data on baseline mental health intervention costs and outcomes were extracted from institutional case studies and academic research. Also, longitudinal studies and district-level reports were analyzed to assess the impact of mental health programs on educational outcomes. Financial data on AI system deployment and maintenance were sourced from publications and industry standards. These integrated datasets form a cohesive picture of the economy, enabling a reliable comparison between traditional and AI-driven mental health interventions.

- **Analytical Tools**

The study uses a multi-faceted approach to economic modeling, incorporating advanced techniques like Net Present Value (NPV), Return on Investment (ROI), and Cost-Effectiveness Analysis (CEA) to evaluate the financial viability and broader impacts of AI systems. Net Present Value (NPV) is utilized to calculate the total expected financial benefits over time, adjusting for inflation and discount rates to present a realistic long-term value. Return on Investment (ROI) is used to assess the financial returns from the initial and ongoing investment in AI systems by comparing net benefits against implementation costs. Cost-Effectiveness Analysis (CEA) helps quantify how efficiently resources are utilized, providing insight into the cost per unit of benefit achieved, such as reductions in mental health crisis incidents or improvements in academic outcomes. Sensitivity Analysis is applied to test how changes in key variables, like system costs, student uptake rates, and intervention success rates, affect overall financial outcomes. This layered approach ensures that both immediate and long-term economic impacts are considered, offering a well-rounded understanding of

the potential value of AI-driven mental health interventions.

IV. FINDINGS AND DISCUSSION

A. Costs of Implementing AI Systems

The initial implementation of AI-driven mental health systems in public education requires significant investment. This includes expenditures on advanced software solutions capable of natural language processing and predictive analytics, alongside compatible hardware infrastructures. According to a report by Gartner (2023), Spending on AI software is projected to reach \$297.9 billion by 2027, with a compound annual growth rate (CAGR) of 19.1%. The market growth is expected to accelerate from 17.8% to 20.4% over the next five years. Additionally, the expenditure on generative AI software will increase significantly, growing from 8% of total AI software spending in 2023 to 35% by 2027. According to Solomon and Eucharia (2023), educational institutions must invest substantial financial resources to integrate AI applications, including high-performance computing systems and specialized software packages, which can be quite costly upfront. However, the introduction of chatbots has made a notable impact on the education sector. Recent surveys have shown a consistent rise in their use, with multiple studies highlighting how chatbots can enhance students' learning experiences and support their educational progress (Kamalov et al., 2023). David Nagel (2023) highlighted a report from P&S Intelligence forecasting that artificial intelligence will grow more than tenfold in the education sector over the next eight years. The report predicts that AI spending by schools will soar from \$2.13 billion in 2022 to \$25.77 billion by 2030. Furthermore, maintenance and training costs for school personnel to effectively utilize these systems are substantial. Integrating AI into education management involves a complex cost-benefit analysis with significant implications. While AI enhances personalized learning and improves administrative efficiency, measuring ROI requires considering both quantitative and qualitative factors to fully understand its value to educational institutions (Solomon and Eucharia, 2023).

B. Immediate Economic Benefits

AI-based systems yield immediate financial advantages, primarily through the reduction of traditional mental health intervention costs. For instance, deploying AI for early detection of mental health risks limits the reliance on expensive crisis response teams. This system enables students to receive personalized mental health support and guidance while predicting potential mental health crises. AI-enabled mental health screenings in schools have reduced the need for emergency interventions, facilitated the diagnosis and treatment of disorders, and promoted workforce development and cultural diversity (Zhou Tian & Deng Yi, 2024; Kazandjian & Neylon, 2024). Recognizing mental health issues early can help prevent them from worsening and reduce the costs associated with intensive therapy. AI-integrated early intervention programs can decrease per-student mental health expenses. Yadav (2021) noted that AI could alert clinicians to factors such as hospitalization history, medication changes, and other conditions that might lead to a crisis. Additionally, AI can monitor social media for changes in mood or behavior to detect the onset of depression or suicidal thoughts. Predictive analytics use historical data to design preventative interventions, aligning with the shift toward personalized and preventative mental healthcare. In screening and diagnostics, AI technologies like machine learning and deep learning have proven effective in analyzing mental health data sets and predicting patterns of various mental health issues (Alhuwaydi 2024).

C. Long-Term Economic Benefits

The long-term economic advantages of integrating AI into mental health support systems are significant. Enhanced mental health outcomes contribute to improved academic performance, as students are better equipped to focus and succeed. Berger et al. (2022) observed that school-based programs aimed at improving students' mental health and well-being had lasting beneficial effects on both their physical and psychological health. Another study by Zajac et al. (2023) found that students who consistently received mental health support had higher graduation rates and a lower likelihood of dropping out. This improvement translates into increased lifetime earnings for these students. In 2019, untreated mental illness among approximately 429,407 Indiana residents resulted in

societal costs amounting to \$4.2 billion. This included \$3.3 billion in indirect costs, \$708.5 million in direct healthcare costs, and \$185.4 million in non-healthcare expenses (Taylor et al., 2023). Addressing mental health challenges early reduces societal costs by mitigating the negative impacts on unemployment and healthcare expenses, while untreated conditions hinder poverty reduction and economic growth (Le et al., 2021). A 2023 UNICEF report highlights the economic benefits of mental health and psychosocial support (MHPSS) interventions for children and adolescents in humanitarian emergencies. It found that failing to address these needs for 10–17-year-olds could result in a global loss of \$203 billion in potential lifetime earnings. The report also shows that MHPSS interventions improve mental health, psychosocial support, and learning outcomes, leading to individual and socio-economic benefits over a person's productive lifespan (UNICEF, 2023)..

D. Non-Monetary Benefits

Beyond financial considerations, AI-driven mental health systems offer critical non-monetary benefits. They significantly enhance accessibility to mental health resources, particularly in under-resourced schools. According to the National Center for Education Statistics' 2021-22 report on Crime, Violence, Discipline, and Safety in U.S. Public Schools, based on responses from nearly 2,700 P-12 school principals, highlights key issues affecting students and schools. Nearly 90% of schools increased social and emotional support for students, but 39% faced challenges due to a lack of licensed professionals and mental health funding. During this period, \$275.7 million of emergency school funds were allocated to mental health support for students and staff (Tamez-Robledo, 2024). AI systems bridge this gap by providing scalable and readily available resources. Additionally, improved mental health support ensures a positive school environment, reducing absenteeism and disciplinary actions. Focusing on well-being, building a sense of community, and incorporating evidence-based social and emotional learning (SEL) programs, schools can significantly improve the mental health and overall well-being of students and educators, leading to more vibrant communities (Frazier & Fosco, 2024).

V. CASE STUDIES

• AI-driven mental health chatbot

A case study by Oghenekaro and Okoro (2024) assesses the impact of an AI-driven mental health chatbot implemented in a public school district. The study uses two matrices: user engagement metrics and pre/post-intervention mental health scores. User engagement data, including chat frequency, session length, and resource usage, informs chatbot improvements and calculates the engagement rate as $(\text{Number of engaged users} / \text{Total number of users}) \times 100$. Mental health scores are measured using PHQ-9 and GAD-7 scales, categorizing depression and anxiety severity to gauge the chatbot's effectiveness in ensuring mental well-being. The mixed-methods approach combines quantitative assessments of mental health metrics with qualitative insights from user interviews. The findings reveal that the chatbot reduced the need for traditional crisis interventions, leading to cost savings and enhanced student satisfaction by offering a stigma-free platform for mental health support, demonstrating its effectiveness in improving mental well-being and guiding its development.

Similarly, Alsayed et al. (2024) described a case study on AI-driven chatbots that have emerged as effective tools for managing mental health issues, especially among students in educational settings. A systematic review of studies published between 2019 and 2023 highlighted their significant impact on addressing psychological disorders such as anxiety and depression by providing immediate and accessible support. By screening 115 articles and analyzing 13 in-depth, researchers found that AI chatbots reduces mental health symptoms but also enhance student satisfaction and engagement by offering a stigma-free platform for assistance. These chatbots are increasingly used in universities and schools for mental health management and academic advising, effectively decreasing the need for resource-intensive interventions and improving student well-being. However, current research limitations, such as restricted database scopes and keyword-based searches, creates broader studies using comprehensive databases like Scopus and WoS. These studies will help understand the long-term behavioral impacts and scalability of chatbot use. Overall, these findings

underscore the transformative potential of AI chatbots in enhancing mental health outcomes and resource efficiency in educational environments.

- Comparative analysis of schools with and without AI systems

A study by Thomas et al. (2023) highlights the significant potential of AI-assisted tutoring to improve learning outcomes, particularly among students from low-income backgrounds. They introduced a hybrid human-AI tutoring model and conducted a three-study quasi-experiment across three urban and low-income middle schools: 125 students in a Pennsylvania school, 385 students (50% Latinx) in a California school, and 75 students (100% Black) in a Pennsylvania charter school, comparing learning analytics of students engaged in human-AI tutoring with those using only traditional math software. The findings indicate that the hybrid approach positively influences student proficiency and engagement, especially benefiting lower-achieving students. Annual costs per student for the hybrid model are approximately \$700, making it a cost-effective strategy for resource utilization. Despite engagement issues, the study emphasizes the transformative potential of AI systems in enhancing academic outcomes and optimizing resources in educational settings.

VI. CHALLENGES AND LIMITATIONS

- Implementation Challenges

To implement AI-driven mental health support systems in public education faces several hurdles. Funding constraints are a significant barrier, as public schools often operate under limited budgets. Heinrich et al. (2023) insufficient funding obstructs the development of sustainable mental health services programs. Allocating resources for AI infrastructure and training competes with other educational priorities, creating challenges in securing necessary investments (Onesi-Ozigun et al, 2024).

Data privacy is another critical issue, as mental health data is highly sensitive. Federal laws like the Family Educational Rights and Privacy Act (FERPA) and the Health Insurance Portability and Accountability Act (HIPAA) require stringent safeguards to prevent data misuse or breaches (American Academy of Pediatrics, 2024). Lan (2023) highlights the growing public

concerns about AI ethics in education, particularly regarding the security of student personal information, and emphasizes the ethical risks posed by AI technology, alongside recommendations that educational institutions, governments, and AI stakeholders collaborate to develop an efficient and enhanced framework for protecting data.

Resistance from educators and parents also presents challenges. Educators may lack confidence in AI tools due to insufficient training or skepticism about their efficacy. Parents might question the appropriateness of using AI for their children's mental health, fearing over-reliance on algorithms over human professionals. A survey by Barna Group. (2024) found that a third of parents expressed concerns about privacy and data security in AI applications for children.

- Ethical Considerations

The use of AI in sensitive mental health interventions raises ethical dilemmas, particularly due to the risk of algorithmic bias, which can perpetuate inequalities if the AI models are trained on unrepresentative data. For example, Norori (2021) demonstrated that biased training datasets can result in lower diagnostic accuracy for underrepresented demographic groups (Norori et al, 2021). Murikah et al. (2024) stress the importance of transparent AI development and regular audits to identify and correct biases, with auditors playing a crucial role, and recommend governance, pre-deployment risk assessment, ongoing performance monitoring, and policies ensuring trust and collaboration for equitable outcomes. Another ethical consideration is the potential for over-diagnosis or misdiagnosis. AI systems, while powerful, are not infallible and may flag students unnecessarily, leading to stigma or undue anxiety. Balancing the benefits of early detection with the risks of false positives is a critical ethical challenge (Yong et al., 2022; Jansen et al., 2024).

- Data Gaps

The effectiveness of economic models evaluating AI-driven mental health systems depends on the availability of comprehensive datasets. The integration of AI and big data can deliver numerous benefits for economic, scientific, and social progress; however, there is currently a lack of large-scale, longitudinal data capturing the costs, outcomes, and

societal impacts of such interventions. Addressing this gap is crucial to fully understand and harness the potential of these technologies for broader societal benefits (European Parliament, 2020). Furthermore, inconsistencies in data collection across school districts hinder meaningful comparisons and scalability assessments. Addressing these gaps requires coordinated efforts among policymakers, researchers, and educational institutions to develop standardized data collection and sharing protocols (Rodriguez, 2020).

VII. POLICY IMPLICATIONS

- **Integrating AI Mental Health Systems into Public Education**
Budgets
Policymakers must prioritize the inclusion of AI-driven mental health support systems in public education funding frameworks. A critical first step is the allocation of targeted grants to schools in underserved areas, ensuring equitable access to advanced mental health tools. According to the U.S. Department of Education (2023), federal funding for mental health services in schools increased to \$70million in 2024, however, most of these funds were directed towards traditional intervention programs. Allocating a portion of this funding specifically for AI-driven systems could enhance efficiency and scalability. Also, states should incentivize partnerships between schools and technology providers to reduce implementation costs (Bevan & DeWitt, 2024). For instance, the California Mental Health Services Oversight and Accountability Commission (2023) has initiated public-private collaborations to deploy AI tools in over 100 schools, demonstrating a scalable model for broader adoption.

Regulatory Frameworks for Data Security and Ethical AI
Use
The integration of AI in mental health support necessitates stringent regulatory frameworks to address data security and ethical considerations. Student data, particularly sensitive mental health information, must be protected under stringent guidelines. The Federal Trade Commission (FTC) and Department of Education should collaborate to update regulations like the Family Educational Rights and Privacy Act (FERPA) to explicitly include AI-driven data collection and processing. According to a 2023 report by the National Institute of Standards and

Technology (NIST), 78% of schools surveyed expressed concerns about the security of data managed by third-party AI providers. Establishing standards for encryption, data anonymization, and access control is critical.

Ethical considerations are equally essential, particularly in ensuring that AI systems operate transparently and without bias. Policymakers should mandate independent audits of AI algorithms to assess potential biases and their impact on different demographic groups. Despite organizations' efforts to use information responsibly, biases in technology processes can cause harmful impacts, posing significant challenges to building public trust in artificial intelligence (Schwartz et al., 2022). Regulatory bodies must ensure that AI systems adhere to principles of fairness, accountability, and explainability to gain the trust of educators, parents, and students.

CONCLUSION

AI-driven mental health support systems represent a transformative approach to addressing the rising mental health challenges in public schools. By enabling early identification and intervention, these technologies reduce the costs associated with traditional crisis responses and improve academic outcomes. Research has shown that early mental health support directly correlates with better long-term societal outcomes, including higher graduation rates and reduced societal burdens such as unemployment and healthcare expenses (Colizzi et al, 2022; Le et al., 2021). Furthermore, these systems enhance accessibility, ensuring that even resource-constrained schools can provide critical mental health services to students. Strategic investments in AI-driven mental health systems are not expenditures alone they possess long-term economic and societal investments. Policymakers, educators, and stakeholders must collaborate to integrate these systems into school budgets, emphasizing equitable access for underserved communities. A well-planned implementation can lead to a more resilient, mentally healthy student population, to enable long-term economic growth by preparing students for better workforce readiness and reducing future societal costs related to untreated mental health issues. The

integration of AI-driven solutions in schools is not without its challenges, but the potential benefits—both immediate and long-term—underscore the importance of adopting these technologies. As we face growing mental health challenges in education, embracing innovative solutions is imperative to safeguard student well-being and secure a prosperous future for society as a whole.

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