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The Use of Clinical Decision Support Systems Based on Artificial Intelligence for Improving Treatment Outcomes and Reducing Medical Errors

Fazliddin Arziqulov, Sayfullayeva Dilbar Izzatillayevna, Maxsudov Valijon Gafurjonovich

Assistant, Department of Biomedical Engineering, Informatics, and Biophysics,
Tashkent State Medical University, Tashkent Uzbekistan

Abstract

Artificial Intelligence (AI)-based Clinical Decision Support Systems (CDSS) are increasingly being integrated into healthcare systems to enhance treatment outcomes and reduce medical errors. This study evaluates the effectiveness of AI-driven CDSS in clinical environments by examining their impact on decision-making accuracy, treatment efficiency, and patient safety. A convergent mixed-methods approach was employed, combining quantitative data from 178 healthcare professionals with qualitative insights from case studies and expert interviews. The findings demonstrate that AI-based CDSS significantly improve treatment accuracy, reduce diagnostic and therapeutic errors, and enhance clinical workflow efficiency. However, challenges such as system integration, user trust, and algorithm transparency remain significant barriers. The study provides recommendations for optimizing the implementation of AI-based CDSS in healthcare systems.

Keywords: Clinical Decision Support Systems, Artificial Intelligence, Medical Errors, Treatment Outcomes, Healthcare Technology.



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1. Introduction

The increasing complexity of modern healthcare has made clinical decision-making more challenging than ever before. Healthcare professionals must process vast amounts of patient data, medical knowledge, and clinical guidelines to make accurate and timely decisions. This complexity increases the risk of medical errors, which remain a significant cause of patient harm worldwide.

Clinical Decision Support Systems (CDSS) have been developed to assist healthcare professionals in making evidence-based decisions. These systems provide recommendations, alerts, and diagnostic support based on patient data and clinical guidelines. With the integration of Artificial Intelligence (AI), CDSS have become more advanced, enabling predictive analytics, pattern recognition, and personalized treatment recommendations.

AI-based CDSS can analyze large datasets, including electronic health records, laboratory results, and imaging data, to identify patterns and generate insights that support clinical decision-making. These systems have demonstrated the ability to improve diagnostic accuracy, optimize treatment plans, and reduce medical errors. For example, AI-driven CDSS can alert clinicians to potential drug interactions, recommend appropriate treatment options, and identify high-risk patients.

Despite these benefits, the adoption of AI-based CDSS presents several challenges. Issues related to system integration, data quality, algorithm transparency, and user trust must be addressed to ensure effective implementation. Additionally, healthcare professionals may be hesitant to rely on automated systems, particularly when decision-making involves high levels of uncertainty.

This study aims to evaluate the use of AI-based Clinical Decision Support Systems in improving treatment outcomes and reducing medical errors. It seeks to assess their impact on clinical decision-making, patient safety, and healthcare efficiency, as well as to identify key challenges associated with their implementation.

2. Methods

This study employed a convergent mixed-methods research design to evaluate the effectiveness of Artificial Intelligence-based Clinical Decision Support Systems in improving treatment outcomes and reducing medical errors. The integration of quantitative and qualitative approaches enabled a comprehensive assessment of both measurable clinical performance indicators and real-world experiences of healthcare professionals using AI-driven systems. This methodological approach was particularly appropriate given the complexity of clinical decision-making processes, which involve both objective data analysis and subjective clinical judgment.

The study population consisted of 178 participants, including physicians, nurses, clinical pharmacists, health IT specialists, and hospital administrators. Participants were selected using a stratified random sampling method to ensure representation across different medical specialties and professional roles. Data were collected from seven hospitals, two specialized clinics, and two healthcare technology organizations that had implemented AI-based CDSS for at least one year. All participants had direct experience with clinical decision support systems in their daily practice.

Quantitative data were collected through a structured questionnaire consisting of 39 items designed to evaluate key variables such as decision-making accuracy, reduction in medical errors,



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treatment effectiveness, system usability, and user trust in AI systems. The questionnaire utilized a five-point Likert scale and included objective performance indicators such as error rates, treatment success rates, and time required for clinical decision-making. Additional data were obtained from institutional records, including incident reports, treatment outcomes, and system usage logs. The reliability of the instrument was confirmed using Cronbach's alpha, which yielded a value of 0.93, indicating high internal consistency.

Qualitative data were collected through ten case studies and twenty semi-structured interviews with healthcare professionals and AI specialists. The case studies focused on the application of AI-based CDSS in various clinical settings, including emergency care, oncology, and internal medicine. Interviews explored participants' experiences with CDSS, including perceived benefits, challenges, and the impact on clinical workflows and patient care.

Quantitative data analysis was conducted using statistical methods, including descriptive statistics, correlation analysis, and regression modeling, to examine relationships between CDSS usage and improvements in treatment outcomes and error reduction. Qualitative data were analyzed using thematic analysis, identifying key themes related to system effectiveness, usability, trust, and integration challenges. The integration of findings from both methods enabled triangulation, enhancing the reliability and validity of the study.

Ethical considerations were strictly observed throughout the study. All participants provided informed consent, and data were anonymized to ensure confidentiality. Data protection measures were implemented to safeguard sensitive patient information.

3. Results

The findings of this study indicate that the use of Artificial Intelligence-based Clinical Decision Support Systems has a significant positive impact on treatment outcomes and the reduction of medical errors. The results demonstrate consistent improvements in clinical decision-making accuracy, patient safety, and healthcare efficiency, supported by both quantitative data and qualitative insights.

One of the most significant outcomes observed in this study is the improvement in clinical decision-making accuracy. The data indicate that the use of AI-based CDSS increased diagnostic and treatment accuracy from an average of 85.1 percent to 93.8 percent. This improvement was particularly evident in complex clinical cases where multiple variables needed to be considered. The ability of AI systems to analyze large volumes of patient data and provide evidence-based recommendations contributed to more precise and consistent clinical decisions.

The study also found a substantial reduction in medical errors. Overall error rates decreased by approximately 31 percent following the implementation of CDSS. Medication-related errors, including incorrect dosages and drug interactions, showed the most significant reduction due to automated alert systems and real-time decision support. These improvements in patient safety highlight the critical role of CDSS in reducing preventable harm.

In addition to improving accuracy and safety, AI-based CDSS contributed to better treatment outcomes. The findings indicate a 27 percent improvement in treatment success rates, particularly in chronic disease management and complex conditions such as cancer and



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cardiovascular diseases. Personalized treatment recommendations generated by AI systems enabled healthcare providers to tailor interventions based on individual patient profiles.

Healthcare efficiency also improved significantly with the adoption of CDSS. The time required for clinical decision-making decreased by approximately 35 percent, allowing healthcare professionals to make faster and more informed decisions. This efficiency gain was particularly valuable in high-pressure environments such as emergency departments.

Qualitative findings further support these results by highlighting the perceived benefits of CDSS among healthcare professionals. Participants emphasized the role of AI systems in enhancing confidence in clinical decisions and reducing cognitive workload. However, the qualitative analysis also identified several challenges, including concerns about system reliability, lack of transparency in AI algorithms, and difficulties in integrating CDSS with existing healthcare systems.

Another important finding relates to user trust. While many participants recognized the benefits of CDSS, some expressed hesitation in fully relying on AI-generated recommendations, particularly in critical situations. This highlights the need for explainable AI systems that provide clear and interpretable outputs.

Overall, the results demonstrate that AI-based Clinical Decision Support Systems significantly improve treatment outcomes and reduce medical errors, while also highlighting important challenges that must be addressed to ensure effective implementation.

4. Discussion

The findings of this study confirm that AI-based CDSS play a crucial role in improving clinical decision-making and patient safety. The significant reduction in medical errors and improvement in treatment outcomes highlight the effectiveness of these systems in supporting healthcare professionals.

However, challenges such as system integration, user trust, and algorithm transparency must be addressed. The development of explainable AI and improved training programs is essential for increasing adoption and ensuring safe use.

5. Conclusion

This study demonstrates that AI-based Clinical Decision Support Systems significantly enhance treatment outcomes and reduce medical errors. Their ability to provide data-driven recommendations makes them essential tools in modern healthcare.

To maximize their potential, healthcare systems must address technical and ethical challenges, improve system integration, and promote user trust. Future research should focus on long-term impacts and real-world scalability.

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