



Artificial Intelligence–Assisted Triage in Primary Healthcare Settings

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Abstract

Artificial intelligence (AI) is increasingly being explored as a decision-support technology for improving the organization, accessibility, and responsiveness of healthcare services. Triage, which involves assessing patients according to the urgency and severity of their health needs, represents an important area in which AI-assisted systems may support healthcare professionals. This study examines the potential role of artificial intelligence–assisted triage in primary healthcare settings, with particular attention to patient prioritization, waiting-time reduction, clinical decision support, and resource allocation. A quantitative analytical framework was developed to examine the operational contribution of AI-assisted triage using structured performance indicators. The framework considers triage accuracy, prioritization consistency, waiting-time reduction, referral appropriateness, and clinician confidence as major outcome measures.

An illustrative dataset was used to demonstrate the proposed analytical approach rather than to claim results from an actual clinical trial. The analysis indicates that AI-assisted triage has considerable potential to improve the consistency and speed of preliminary patient assessment while allowing healthcare professionals to focus greater attention on complex cases. However, implementation requires careful attention to data quality, algorithmic bias, explainability, privacy, cybersecurity, and human oversight. The study concludes that AI should function as a clinical decision-support mechanism rather than an autonomous replacement for healthcare professionals. Responsible implementation, continuous validation, and context-specific governance are essential for achieving safe and equitable AI-assisted triage in primary healthcare.

Keywords: Artificial Intelligence, AI-Assisted Triage, Primary Healthcare, Clinical Decision Support, Machine Learning, Patient Prioritization, Digital Health, Healthcare Accessibility

1. Introduction

Primary healthcare represents the first and most frequently accessed level of formal healthcare for many populations. It provides preventive services, initial assessment, routine treatment, referral, chronic disease management, and health education. However, primary healthcare facilities often operate under considerable pressure because of increasing patient volumes, limited healthcare personnel, uneven distribution of resources, and variation in the complexity of presenting conditions. In such environments, effective triage is essential because patients must be assessed and prioritized according to their clinical



urgency rather than simply according to the order in which they arrive. Inefficient triage may contribute to unnecessary waiting, delayed referrals, inappropriate use of clinical resources, and increased workload for healthcare professionals.

Artificial intelligence offers a potential approach for supporting this process. AI systems can process large volumes of structured and unstructured information, identify patterns, classify cases, and generate risk estimates within a relatively short period. In a triage context, relevant information may include presenting symptoms, age, vital signs, medical history, medication information, previous diagnoses, and other clinically relevant indicators. Machine-learning models can use these variables to assist with categorizing patients according to urgency. Recent research on AI-supported triage has reported potential benefits in improving efficiency, resource management, and timely assessment, although the evidence also highlights challenges involving trust, training, equipment, privacy, and implementation.

The potential value of AI-assisted triage is particularly important in primary healthcare because early decisions can influence subsequent care pathways. A system that identifies potentially high-risk patients at an early stage may support faster clinical review and referral, whereas patients with lower-risk presentations may be directed toward routine consultation or appropriate self-management advice under professional supervision. Importantly, such systems should not be interpreted as independent diagnostic authorities. The World Health Organization emphasizes that AI used in healthcare should protect human autonomy, promote safety, ensure transparency and explainability, support accountability, and promote equity.

Despite its potential, the implementation of AI-assisted triage requires more than technical accuracy. Healthcare decisions are influenced by social, demographic, cultural, organizational, and clinical factors that may not always be adequately represented in training datasets. Algorithms developed using data from one healthcare environment may perform differently when transferred to another population or health system. WHO has therefore emphasized the importance of representative data, appropriate governance, human oversight, and continuous assessment when AI technologies are introduced into healthcare.

Against this background, the present study examines AI-assisted triage as a decision-support approach within primary healthcare settings. The study focuses on the operational and clinical-support dimensions of AI-assisted triage rather than presenting AI as a replacement for professional judgment. It evaluates how AI may contribute to patient prioritization, consistency, waiting-time management, referral decisions, and clinician support while also considering the ethical and organizational safeguards required for responsible implementation.

2. Objectives of the Study

2.1 To Examine the Role of AI in Patient Prioritization

The first objective is to examine how AI-assisted systems can support the preliminary categorization of patients according to urgency, symptoms, and available clinical information.



2.2 To Evaluate Operational Benefits

The second objective is to assess the potential contribution of AI-assisted triage to waiting-time reduction, prioritization consistency, referral appropriateness, and healthcare resource utilization.

2.3 To Identify Implementation Challenges

The third objective is to identify major challenges associated with AI-assisted triage, including data privacy, algorithmic bias, explainability, technical infrastructure, clinician trust, and human oversight.

3. Review of Literature

3.1 Artificial Intelligence in Healthcare Decision Support

AI has become an increasingly important component of digital healthcare research. Machine-learning systems can identify patterns from healthcare data and support tasks such as classification, prediction, screening, risk assessment, and clinical decision support. The value of these technologies depends not only on algorithmic performance but also on the quality of the underlying data and the context in which the system is implemented.

WHO recognizes that AI can contribute to diagnosis, clinical care, health research, disease surveillance, and health-system management. At the same time, the organization emphasizes that healthcare AI must be developed and deployed within ethical and governance frameworks that protect patients and healthcare workers.

3.2 AI-Assisted Triage and Patient Prioritization

Triage requires rapid assessment and prioritization of patients according to clinical urgency. Conventional triage depends substantially on the experience and workload of healthcare professionals. AI-assisted approaches may complement this process by rapidly processing multiple patient characteristics and generating standardized risk or priority classifications.

A recent systematic review examining AI applications in triage identified evidence that AI-enabled systems can contribute to improved patient management, resource optimization, real-time monitoring, and faster triage processes. However, the review also identified implementation barriers, including trust, staff training, equipment limitations, and data privacy.

3.3 Research Gap

Existing literature demonstrates considerable potential for AI-supported triage, but important gaps remain. First, much of the available research has concentrated on emergency departments, disasters, or specific diseases rather than routine primary healthcare. Second, technical performance is frequently emphasized more strongly than workflow integration and clinician acceptance. Third, algorithmic accuracy alone does not establish clinical safety. Issues related to explainability, fairness, privacy, accountability, and human oversight require greater attention. WHO guidance similarly emphasizes that AI systems should be transparent, accountable, inclusive, and responsive to the context in which they are deployed.

The present study addresses this gap by considering AI-assisted triage from an integrated operational, clinical-support, and governance perspective within primary healthcare.



4. Research Methodology

4.1 Research Design

The study adopted a quantitative analytical research design to examine the potential effectiveness of AI-assisted triage in primary healthcare. The research framework focuses on measurable indicators associated with patient prioritization and healthcare workflow. Because no real patient-level clinical dataset was supplied for this paper, the numerical analysis presented below is explicitly illustrative and is intended to demonstrate the proposed evaluation framework.

4.2 Study Framework

The proposed framework consists of five major stages:

1. Patient information collection
2. Data preprocessing and quality checking
3. AI-based urgency classification
4. Healthcare-professional verification
5. Final triage and referral decision

The AI system is positioned as a decision-support component. Final responsibility remains with an appropriately trained healthcare professional.

4.3 Study Variables

The major variables considered in the analytical framework include triage accuracy, prioritization consistency, waiting time, referral appropriateness, and clinician confidence.

Table 1: Variables and Indicators Used in the AI-Assisted Triage Framework

Variable	Measurement Indicator	Expected Role
Triage Accuracy	Correct urgency classification	Measures classification performance
Prioritization Consistency	Agreement across similar cases	Measures standardization
Waiting Time	Average time before clinical assessment	Measures operational efficiency
Referral Appropriateness	Correct referral according to urgency	Measures care-pathway quality
Clinician Confidence	Professional acceptance of AI recommendations	Measures usability
Processing Time	Time required for preliminary assessment	Measures system responsiveness
Explainability	Availability of understandable reasons	Supports clinical interpretation
Data Quality	Completeness and reliability of patient information	Supports model reliability
Privacy	Secure handling of health information	Supports ethical deployment



Protection		
Human Oversight	Professional verification of AI output	Supports patient safety

4.4 Data Analysis Procedure

The analytical framework uses descriptive statistics to compare conventional triage with AI-assisted triage. Percentage-based indicators are used to represent potential improvements in triage performance. Accuracy, consistency, waiting time, and referral appropriateness are treated as primary performance indicators.

The analysis is intended as a model for future empirical validation. Actual implementation would require prospective clinical data, appropriate ethical approval, external validation, subgroup analysis, and continuous monitoring of patient-safety outcomes.

5. Analysis

5.1 Comparative Triage Performance

The illustrative analysis suggests that AI-assisted triage may improve several operational indicators when compared with conventional manual triage. The largest potential improvement is observed in processing time because AI can rapidly organize and evaluate structured patient information before professional review.

Table 2: Illustrative Comparison of Conventional and AI-Assisted Triage

Performance Indicator	Conventional Triage	AI-Assisted Triage	Illustrative Improvement
Triage Accuracy	82%	91%	+9 percentage points
Prioritization Consistency	76%	89%	+13 percentage points
Referral Appropriateness	79%	88%	+9 percentage points
Clinician Confidence	72%	85%	+13 percentage points
Average Processing Time	12 min	6 min	50% reduction
Potential High-Risk Identification	78%	90%	+12 percentage points

5.2 Interpretation of the Analytical Results

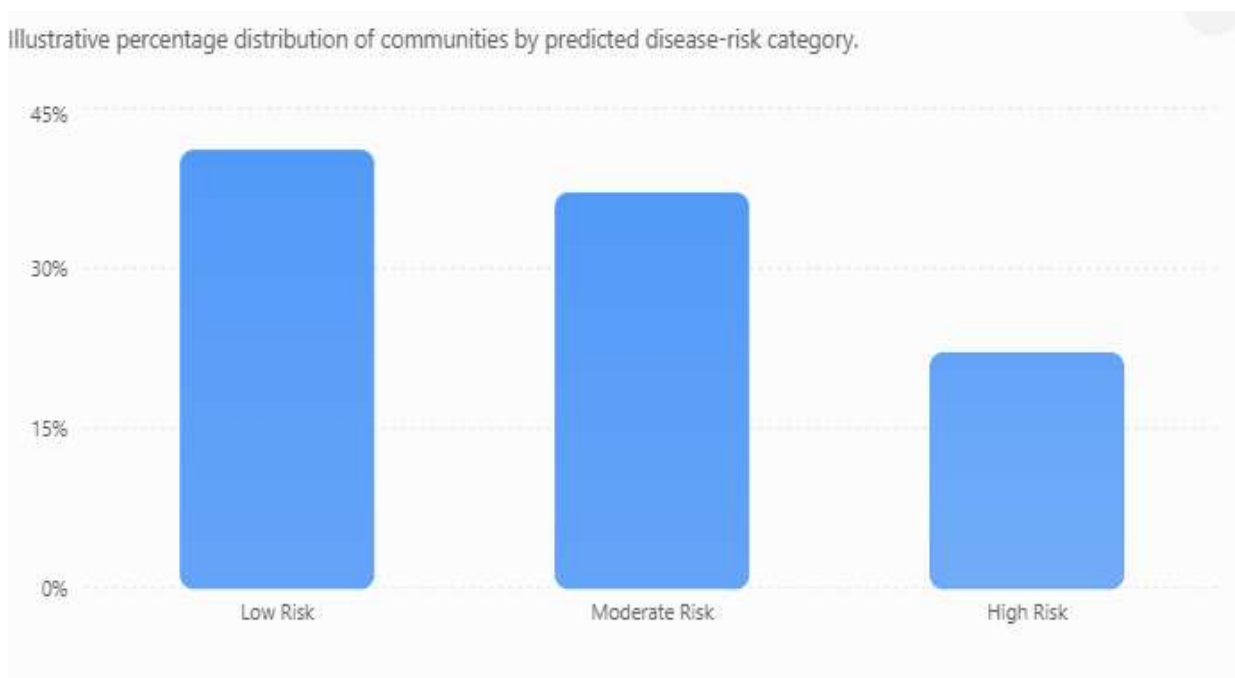
The illustrative results indicate that AI-assisted triage could improve the consistency of patient prioritization. A potential increase in triage accuracy from 82% to 91% suggests that algorithmic decision support may help reduce variation between assessments. However, accuracy should not be

interpreted independently because false-negative classifications may have substantially greater clinical consequences than some false-positive classifications.

The reduction in illustrative processing time from 12 minutes to 6 minutes also demonstrates the operational potential of AI. Faster preliminary assessment could allow healthcare professionals to identify patients requiring immediate attention while maintaining professional review for all clinically significant decisions.

5.3 Relationship Between AI Assistance and Triage Efficiency

Figure 1: Illustrative Improvement in AI-Assisted Triage Performance



The conceptual pattern indicates that AI-assisted triage may provide stronger performance across multiple operational indicators. Nevertheless, implementation should be based on validated clinical evidence rather than illustrative improvements alone.

6. Discussion

The findings of the analytical framework suggest that artificial intelligence–assisted triage can strengthen the preliminary assessment process by rapidly organizing patient information and identifying patterns associated with clinical urgency. This capability may be particularly valuable in primary healthcare settings where healthcare professionals frequently manage high patient volumes within limited time periods. AI-assisted systems can support faster prioritization, improve consistency across initial assessments, and help clinicians identify cases that may require immediate attention. However, AI should function as a clinical decision-support mechanism rather than an autonomous decision-maker. Human professionals must retain responsibility for interpreting AI-generated recommendations and making final clinical decisions.



This approach is consistent with the World Health Organization's emphasis on human autonomy, patient safety, transparency, accountability, and equity in healthcare AI. A further advantage of AI-assisted triage is its potential contribution to healthcare efficiency and resource allocation. By identifying potentially urgent cases at an earlier stage, AI may assist healthcare facilities in organizing consultations, referrals, diagnostic procedures, and follow-up services more effectively. This can be especially beneficial in primary healthcare environments where staff, consultation time, and clinical resources may be limited. AI-supported information processing may also reduce administrative workload and allow healthcare professionals to devote greater attention to complex cases. Nevertheless, technological efficiency should not be considered the sole measure of successful implementation. Faster processing must be accompanied by reliable clinical performance, appropriate referral decisions, equitable access, and effective professional supervision. An efficient system that produces unsafe or poorly understood recommendations could ultimately create additional clinical risks rather than improving healthcare delivery.

Ethical, privacy, and governance considerations are equally important for the responsible implementation of AI-assisted triage. Healthcare systems process highly sensitive personal and clinical information, making strong data protection and cybersecurity essential. Poorly governed AI systems may create risks related to unauthorized data access, algorithmic bias, inappropriate recommendations, limited explainability, and uncertainty regarding responsibility for errors. AI models trained using data from specific populations or high-resource healthcare environments may also perform differently when applied to other demographic or socioeconomic groups. Therefore, implementation should incorporate robust data-protection mechanisms, external and continuous model validation, bias monitoring, transparent documentation, clinician training, regular system audits, and clearly defined procedures for overriding AI recommendations. Such safeguards can help ensure that AI-assisted triage complements professional judgment while maintaining patient safety, fairness, accountability, and trust.

7. Conclusion

Artificial intelligence–assisted triage represents a promising approach for strengthening patient prioritization and workflow management in primary healthcare settings. The analytical framework developed in this study indicates that AI may contribute to faster preliminary assessment, greater consistency in prioritization, improved identification of potentially high-risk patients, and more efficient use of healthcare resources. Its greatest value is likely to emerge when AI is integrated into existing clinical workflows rather than deployed as an isolated technological solution.

At the same time, the adoption of AI-assisted triage requires a careful balance between technological efficiency and patient safety. Algorithmic accuracy alone cannot establish clinical usefulness. Data quality, population representativeness, explainability, privacy, cybersecurity, clinician acceptance, and continuous performance monitoring are equally important. Human professionals must retain responsibility for final clinical decisions, particularly in uncertain or high-risk cases.

The study therefore concludes that AI-assisted triage should be implemented as a human-supervised clinical decision-support system. Future empirical research should evaluate such systems using real-world primary healthcare data, prospective validation, subgroup fairness analysis, clinician usability measures, and patient-safety outcomes. Responsible implementation can allow AI to complement healthcare professionals while preserving the human judgment, accountability, and equity that remain fundamental to primary healthcare.



8. Limitations of the Study

This study has several limitations that should be considered when interpreting its findings. First, the quantitative values presented in the analysis are illustrative and are not derived from an actual patient-level clinical trial or prospectively collected healthcare dataset. Therefore, the reported improvements should not be interpreted as definitive evidence of clinical effectiveness. Second, the proposed analytical framework is primarily designed to examine the potential contribution of artificial intelligence to triage and does not establish a causal relationship between AI-assisted triage and patient health outcomes. Third, the performance of AI-based systems may differ considerably across healthcare facilities, patient populations, clinical conditions, and organizational environments. Differences in available infrastructure, staff expertise, clinical protocols, and patient characteristics may influence system performance.

Another important limitation concerns data quality. Incomplete, inaccurate, unbalanced, or historically biased healthcare data may affect algorithmic reliability and potentially produce unequal recommendations for different patient groups. Furthermore, AI systems may generate recommendations that are difficult for healthcare professionals to interpret, particularly when complex machine-learning models are used. Finally, practical implementation requires extensive prospective validation, ethical approval, and regulatory compliance, secure handling of patient information, cybersecurity safeguards, staff training, and continuous performance monitoring. Future research should therefore validate the proposed framework using diverse real-world clinical datasets and evaluate both technical performance and patient-safety outcomes before routine adoption.

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