

# Can AI Predict ICU Admission from Emergency Data?

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## Abstract

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## Introduction

Emergency departments (EDs) are the frontline of acute care, where rapid and accurate decision-making is critical for patient outcomes. One of the most significant challenges in this high-pressure environment is identifying which patients will require admission to an Intensive Care Unit (ICU). Overcrowding in EDs can lead to delays in treatment, increased risk of adverse events, and decreased quality of care [1]. Traditional triage systems, while valuable, often rely on subjective clinical judgment and can have variable accuracy. This has spurred the search for more objective, data-driven tools to support clinical decisions. Artificial intelligence (AI) and machine learning (ML) have emerged as powerful technologies with the potential to revolutionize this process by analyzing vast amounts of patient data to predict critical outcomes.

## The Role of Machine Learning in Predictive Triage

Machine learning models are designed to identify complex patterns in data that may not be apparent to human observers. In the context of the ED, these models can integrate a wide range of inputs—including vital signs, demographic information, chief complaints, and laboratory results—to generate a risk score for ICU admission. A systematic review published in *The American Journal of Emergency Medicine* analyzed the performance of various ML models in predicting the need for intensive care among ED patients. The review found that algorithms such as Gradient Boosting, Logistic Regression, Neural Networks, and Random Forests have demonstrated good performance in identifying critically ill patients [1]. These models showed promising results across different metrics, including sensitivity, specificity, and the Area Under

the Receiver Operating Characteristic Curve (AUROC), a key indicator of a model's discriminative ability.

However, the review also highlighted a crucial point: while the models perform well, there is a need for more high-quality, prospective research to validate these findings across different clinical settings and patient populations [1]. The performance of any given model can vary, and its real-world utility depends on rigorous testing and validation.

## **Developing and Validating AI-Powered Triage Models**

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A 2025 study published in *Scientific Reports* provides a concrete example of how these models are developed and validated. Researchers conducted a retrospective analysis of over 163,000 ED visits to create an ML model for predicting ICU admission [2]. They compared several algorithms, including logistic regression, random forest, and XGBoost, against the Canadian Triage and Acuity Scale (CTAS), a widely used manual triage system.

The results were compelling. The XGBoost model, a sophisticated gradient boosting algorithm, achieved superior predictive performance, with an AUROC of 0.917, compared to 0.882 for the CTAS system [2]. This indicates that the AI model was significantly better at distinguishing between patients who would need ICU care and those who would not.

### ***Leveraging Unstructured Data with Natural Language Processing***

A key innovation in this study was the integration of unstructured data, specifically the free-text chief complaints written by triage nurses. Using Natural Language Processing (NLP), the researchers converted these textual descriptions into numerical representations (semantic vectors) that the ML model could analyze. The analysis revealed that the chief complaint, along with the mode of arrival (e.g., ambulance vs. walk-in) and patient age, were among the most important predictors of ICU admission [2]. This demonstrates that AI can unlock valuable insights hidden in clinical notes, which are often underutilized in traditional risk-scoring systems.

## **The Future of AI in the Emergency Department**

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The evidence strongly suggests that AI can indeed predict ICU admission from emergency data with a high degree of accuracy, often outperforming traditional triage methods. By providing an objective and consistent risk assessment at the point of triage, these models can help clinicians prioritize care, allocate resources more effectively, and ultimately improve patient outcomes. They can act as a crucial decision support tool, augmenting the expertise of healthcare professionals without replacing it.

However, the journey from a validated model to a fully implemented clinical tool involves several important considerations. These include ensuring the model is fair and unbiased, integrating it seamlessly into existing hospital workflows, and conducting prospective studies to confirm its impact on clinical practice. As research continues and technology advances, AI-powered triage is poised to become an indispensable part of modern emergency medicine, helping to ensure that critically ill patients receive the timely care

they need.

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