

# Nursing Predictive Technologies for Timely Patient Drop Detection in Hospital Environments

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## Abstract

*Early warning of clinical worsening among hospitalized adults is important in enhancing patient outcomes as well as improving morbidity and mortality. Recent technological advances in the healthcare sector have allowed nurses to take advantage of predictive technologies that help in detecting minor physiological and behavioral changes before they become serious complications. In this scoping review, the scope of technological solutions used in clinical environments to assist nurses in predicting patient deterioration is investigated, such as early warning score systems, machine learning algorithms, wearable monitoring devices, and electronic health record-integrated alert systems. The review brings into context the effectiveness, issues of implementation and the possibilities of integrating these tools into normal nursing practice. The evidence shows that even though predictive technologies have the potential to improve clinical decision-making and prompt interventions, their successful use relies on usability, personnel training, and workflow process alignment. The review highlights the necessity of continuous studies to maximize these technologies and their safe, effective, and equitable applications in the care of the hospital.*

**Keywords:** Clinical deterioration, predictive nursing technologies, early warning systems, machine learning in healthcare, patient monitoring, hospitalized adults, nursing informatics, proactive patient care.

## 1.Introduction

The healthcare environment has experienced a radical change over the past decades, and patient safety has become a top priority that has led to innovation in clinical practice. One of the most important changes is the development of technologies aimed at forecasting clinical degradation in hospitalized patients before it turns into life-threatening. This is a broad study of the complex space of early warning mechanisms, assessment instruments, and predictive technologies that have transformed the manner in which medical practitioners, especially nurses, recognize and act on indications of patient degradation in clinical-surgical units.

The antecedent of systematic methods of identifying clinical deterioration can be dated back to the understanding that adverse events among hospitalized patients are hardly likely to occur without any warning. Instead, they are usually preceded by observable physiological alterations which once identified and understood properly can lead to timely interventions that in effect avoid devastating end results. This has spurred the creation of advanced scoring systems, with basic bedside assessment tools to advanced artificial intelligence programs built into electronic health records(1).

The development of these technologies is not just technological progress; it is a change of a paradigm in the philosophy of healthcare towards active and not passive care. The conventional models of health care tended to delay the intensive interventions in favor of observable signs of crisis. Modern methods, which are based on decades of research and clinical practice, note that the interval between the initial physiological alterations and irreversible worsening is crucial and operative. Such a paradigm shift has far-reaching consequences on patient outcomes, healthcare expenditure and the way of professional nursing practice.

The nurses play a special role in such ecosystem of early detecting and intervention. Being the medical practitioners, who have the most consistent interaction with patients in hospitals, nurses are the main source of the evaluation information and the first to notice the alarming changes. Their clinical judgement gained in practice and in schools is a precious element of most of the deterioration prediction systems. This dual role as both data gatherer and clinical interpreter positions nursing professionals at the forefront of patient safety initiatives.

Clinical deterioration is complex to predict because of the multifactorial characteristics of patient physiology and the variety in ways in which patients may undergo adverse events. Although vital signs are major components of assessment, they are but one aspect of patient condition. Modern technologies are gaining a greater interest in the use of neurological tests, indicators of renal functioning, laboratory results, or even a subjective measure of patient

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well-being or nurse anxiety. This multidimensional strategy represents increasing knowledge that successful early warning systems need to provide all the range of physiological and clinical signals that lead to deterioration.

The international nature of the issue of patient safety has encouraged cross-national cooperation in the creation and testing of these technologies. The evidence base of different early warning systems has been built by the research centers situated across continents, each of them has its own perspectives because of their healthcare settings, patient groups and resource base(2). This global aspect has not only added variety of approaches to the field, but also demonstrated the difficulties of developing universal solutions in a world where healthcare systems may be different and patient demographics may vary.

The technical complexity of the contemporary healthcare settings is both an opportunity and a challenge to the implementation of deterioration prediction systems. Electronic health records can gather data and analyze them in real-time when in former times it was impossible, and mobile technologies help to establish quick contact and reaction. Nevertheless, this technological feature also creates complexity in integration of systems, training of users, and also alert fatigue when poorly constructed systems raise too many false alarms.

The economic consequences of clinical deterioration are so far reaching that the immediate expenses of intensive interventions are not considered very high. Patients that undergo avoidable worsening commonly have lengthy hospital stays, intensive care unit admissions and complicated procedures that would otherwise have been avoided by earlier identification and management. In healthcare system terms, the investments in effective early warning technologies are not only patient safety projects; they are also cost-beneficial approaches to using resources more efficiently and increasing the overall efficiency of the system.

Deterioration prediction technologies have been developed and refined using quality improvement methodologies. Repeated testing of systems, their results, performance assessment, and the refinements of methods have resulted in more and more sophisticated and efficient tools. This is an iterative improvement process that acknowledges that there is no one universal technology or system that is universally applicable across all patient populations and clinical settings, and that constant changes and improvements are required.

Incorporation of deterioration prediction technologies into already established clinical workflow is also another major implementation issue that is not limited to technicalities. To become adopted successfully, user experience, integration of the workflow, and change in organizational culture must be given due consideration. Technologies that do not fit into current clinical processes or cause new load without obvious benefit are frequently unable to be implemented in the long term despite their perceived efficacy in theory(3).

In prospect, the area is rapidly developing as artificial intelligence, machine learning and predictive analytics offer new and more advanced methods of early detection. These technological advances, however, have to be considered in a balance with the primary role of clinical judgment, patient centered care, and the human aspects of healthcare that cannot be substituted as inseparable elements of effective, safe patient care. The true effectiveness of any deterioration prediction system is not its technical complexity but its capacity to make clinical decisions about potential deteriorations in a patient more effective and to focus on the human factors that constitute compassionate effective healthcare.

## **2.Methods**

The methodological approaches to systematic exploration of clinical deterioration technologies must be rigorous enough to adequately represent the complexity and variety of tools available and offer a valuable contribution to clinical practice. This comprehensive scoping review process is a well-designed method of tracing the landscape of early warning systems and assessment technologies among nursing professionals when working with hospitalized adult patients. The methodological framework used relies on the recommendations of the Joanna Briggs Institute, which guarantees that the systematic evidence synthesis is developed in accordance with internationally accepted recommendations.

The choice of a scoping review methodology as opposed to alternative systematic reviews techniques reflects the exploratory nature of this study and the necessity to map the available technologies in detail as opposed to assessing the efficacy of particular interventions(4). In this way, it recognises that the area of clinical deterioration prediction has a wide range of tools, systems, and methods which would not necessarily be readily comparable by means of conventional meta-analytic methods. Rather, the scoping review approach allows a wide overview of existing evidence and pinpoints patterns, gaps and future research and development opportunities.

The process of developing the research question relied on the Population, Concept, Context (PCC) model and targeted the group of adults hospitalized and kept in clinical-surgical wards as the population of interest. The concept of study includes all technologies, systems and tools that are oriented to predicting or detecting early signs of clinical deterioration that are within the scope of nursing practice. The setting of the context is hospital, where general medical and surgical wards are explicitly discussed, where continuous monitoring may be restricted as opposed to intensive care units(5).

**TABLE 1** Research Methodology Framework

| Component                          | Details                                   | Rationale   |
|------------------------------------|---|---|
| <b>Study Design</b>                | Scoping Review (JBI Framework)            | Enables comprehensive mapping of diverse technologies rather than effectiveness evaluation  |
| <b>Research Question Framework</b> | PCC (Population-Concept-Context)          | Population: Adults in clinical-surgical wards; Concept: Deterioration prediction technologies; Context: Hospital nursing practice |
| <b>Database Selection</b>          | MEDLINE, CINAHL, Scopus, Web of Science   | Comprehensive coverage of nursing and medical literature  |
| <b>Search Strategy</b>             | Two-phase approach: Broad then refined    | Phase 1: Clinical Deterioration + Inpatients; Phase 2: Specific technology terms  |
| <b>Gray Literature</b>             | Clinical manuals, institutional protocols | Recognition that significant knowledge exists outside peer-reviewed channels  |
| <b>Screening Process</b>           | Independent dual review                   | Minimizes selection bias and ensures comprehensive capture  |
| <b>Data Extraction</b>             | Standardized forms                        | Captures study characteristics, technology details, variables, and outcomes   |
| <b>Quality Assessment</b>          | Level of evidence classification          | Oxford Centre Evidence-Based Medicine hierarchy   |
| <b>Analysis Method</b>             | Mixed quantitative/qualitative            | Descriptive statistics plus thematic analysis   |

The development of database selection and search strategy was thoroughly consulted with the help of information specialists and systematic review methodologists in order to cover the relevant literature optimally. The search strategy was sensitivity and specificity balanced as it used wide initial searches to encompass the entire area of the available literature and the further refinement steps were based on the need to narrow down on the most relevant and high quality evidence. The gray literature sources, such as clinical practice manuals and institutional protocols, acknowledge the fact that much knowledge about the field is found beyond the mainstream peer-reviewed publication channels.

Several reviewers operating independently were used in the screening and selection to reduce selection bias and to ensure that all the relevant studies were captured. Any disagreements between reviewers were addressed by discussion and consultation with senior investigators in an effort to remain transparent and consistent during the selection process. The inclusion criteria were intentionally inclusive to capture the range of potential technologies available whilst the exclusion criteria were aimed at eliminating those studies that dealt with extremely specialized patient populations or clinical settings that would restrict external generalization to typical medical-surgical ward settings.

The process of data extraction consisted of standardized forms prepared specifically to investigate this problem, and the focus of which was not only the general properties of the study but also the detailed data on the technologies under study, their multiple variables, implementation procedures, and reported results. This all-embrative method of data extraction allows studying the peculiarities of technology in detail and finding the similar tendencies in various systems and methods(6).

The quality assessment of scoping reviews is not the same as that used in scoping studies and efficiency reviews because it emphasizes relevancy and contribution to the overall mapping effort as opposed to risk of bias assessment. Nevertheless, some consideration was given to study design features and level of evidence to highlight

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the context under which findings should be interpreted and in which a higher quality of evidence might be required to inform clinical decision-making.

**TABLE 2** Inclusion and Exclusion Criteria

| Criteria Type             | Inclusion  | Exclusion  |
|---------------------------|--|--|
| <b>Study Types</b>        | Original studies, literature reviews, monographs, theses, dissertations, editorials  | None based on study type   |
| <b>Publication Period</b> | 1997 onwards (first EWS development)   | Studies before 1997  |
| <b>Language</b>           | No restrictions  | None   |
| <b>Patient Population</b> | Adults hospitalized in clinical-surgical wards                                       | Pediatric patients, specific clinical conditions (pregnancy, sepsis, cancer, psychiatric disorders, palliative care, COVID-19) |
| <b>Clinical Setting</b>   | General medical-surgical wards   | Pre-hospital care, emergency departments, post-ICU discharge   |
| <b>Technology Focus</b>   | Systems, scores, programs for early warning of clinical deterioration in nursing use | Technologies not applicable to nursing practice  |
| <b>Geographic Scope</b>   | Global   | None   |

The method of analysis utilized quantitative and qualitative methods in this case but as a descriptive statistic was utilized to characterise the body of evidence and as a qualitative thematic analysis to identify common themes and trends across various technologies and researches. This analytical method combined with mixed-methods can provide an opportunity to explore the breadth and depth of available evidence and provide meaningful insights on future research directions and clinical practice.

Geographic and temporal analysis of distribution gives a significant background on development and evolution of the clinical deterioration technologies. The presence of research predominantly belonging to particular geographic areas could be indicative of disparities in priorities within the healthcare system, research infrastructures, or patient safety efforts that might affect the creation and acceptance of the technologies.

The integration of evidence from different study designs and methodological approaches presents both opportunities and challenges for synthesis and interpretation. Although this diversity helps in the enrichment of knowledge of the field, it also demands a keen consideration of the levels of evidence variation and possible limitations that are accrued with the different approaches to research. The analytical framework used considers these complexities without necessarily aiming to give meaningful insights capable of making a difference in clinical practice and policy formulation.

### 3.Results

The adoption of technological clinical deterioration prediction systems has produced considerable data on its effect on patient outcomes, the performance of the healthcare system, and the effectiveness of clinical practice. This evidence provides a complicated terrain of achievements, problems, and constant prospects of improvement that all combine to help us comprehend how these technologies can be effectively used to provide better patient safety and clinical care quality(7).

Mortality decreasing is the final aim of the majority of deterioration prediction technologies, but the data on their effects on the total hospital mortality is a subtle phenomenon that should be interpreted with care. Although it has been shown in several studies that certain outcome measures have improved, e.g. lower rates of cardiac arrest, or lower rates of unexpected admissions to intensive care units, the implications of these improvements on statistically significant decreases in overall hospital mortality have been even more elusive. This seeming paradox probably indicates a complex, multifactorial aspect of hospital mortality, and difficulty achieving a definite outcome improvement attributable to individual interventions in complex healthcare settings.

The rate of activation of rapid response teams has continued to record high gains after the introduction of deterioration prediction technologies thereby indicating that the technologies are effective and effective in identifying patients who would otherwise have not been detected by conventional methods of assessment. The

higher activation rates are not only a victory of finding at-risk patients but also a drawback in the activation rates in resources and the possible alert fatigue of response teams. The sensitivity and specificity in such systems are still under consideration with implementers trying to strike the trade-off between not receiving the actual at-risk patients and flooding the response systems with false alarms.

**TABLE 3** Primary Clinical Outcomes Across Technologies

| Technology Type               | Primary Outcomes Measured              | Positive Impact                              | Mixed/Neutral                           | Challenges Identified           |
|-------------------------------|--|--|---|---------------------------------|
| <b>NEWS/Modified NEWS</b>     | Death, ICU transfer, cardiac arrest    | ✓ Reduced cardiac arrests (15-30% reduction) | ~ Hospital mortality (variable results) | Alert fatigue, false positives  |
| <b>MEWS Variants</b>          | RRT activation, ICU admission, death   | ✓ Increased early RRT calls (40-60%)         | ~ Length of stay (inconsistent)         | Staff training requirements     |
| <b>AI-Enhanced Systems</b>    | 6-hour prediction accuracy, escalation | ✓ Earlier identification (2-4 hours sooner)  | ~ Implementation complexity             | Technology integration barriers |
| <b>Nursing-Specific Tools</b> | Clinical deterioration recognition     | ✓ Improved assessment accuracy               | ~ Subjective variability                | Standardization challenges      |
| <b>Composite Scores</b>       | Multiple adverse events                | ✓ Comprehensive risk assessment              | ~ Resource utilization increase         | Cost-effectiveness unclear      |

Notably, the patterns of ICU admissions have shown significant improvement since the introduction of early warning systems and it has been shown in many studies that the number of patients who are admitted to a higher level of care and subsequently suffer cardiac arrest or other disastrous incident has been on the rise. Although such prior transfers could have prevented negative consequences, they also present significant issues on how to use resources and economic aspects of higher ICU use. It is difficult to tell the difference between justified early interventions that can avert worse outcomes and possibly unwarranted transfers that drain scarce resources despite a lack of commensurate benefits.

Prevention of cardiac arrest has become one of the most regularly reported advantages of deterioration prediction systems, with many studies indicating that cardiac arrests within the wards reduced significantly after the implementation of the system. Avoidance of cardiac arrest is not only an evident patient safety outcome, but also a cost-effective outcome because cardiac arrests in the hospital are expensive and frequently unfavorable. The fact that this finding is being consistent with the various systems and settings offers solid support to the usefulness of systematic methods of detecting deterioration.

The temporal dynamics of outcome changes give valuable clues regarding the processes by which deterioration prediction technologies can obtain their advantages. Most systems show the strongest effects in the hours of the immediate postimplementation, and some benefits may decay over time as clinical teams get used to new alert patterns and the population of patients may shift. Such temporal change implies the need to keep the system optimized and educate the staff to ensure effectiveness in the long term.

Length of hospital stay is another key outcome measure whose findings have yielded a mixed outcome in various studies and implementations. There are systems in which overall length of stay is also reduced, potentially due to the earlier detection and management of complications that could otherwise increase the length of hospitalization. The other implementations have however had found a neutral or a slightly longer length of stay which may perhaps indicate a more conservative management or a higher rate of vigilance resulting in longer periods of observation of borderline cases.

The consequences of the change in the use of healthcare resources are not confined to the actual clinical care but also to the laboratory tests, the usage of diagnostic imaging, and request consultations. The more frequent blood testing has been caused by some deterioration prediction systems that use laboratory values whereas more neutral effects have been found by those prediction systems that use clinical observations as the primary basis. It is essential to understand these patterns of utilization of resources by healthcare systems that may consider the implementation of these technologies.

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Patient and family satisfaction outcomes are another potential outcome of deterioration prediction technology impact, but they are less commonly measured. There is little evidence available that proactive methods of identifying and managing possible complications are well-received in families in general, especially when the methods are well-explained and are part of a larger communication concerning patient care plans. Yet, the possibilities of escalating anxiety based on the more regular monitoring or activation of alerts, should be thought through in terms of designing and implementing the system.

**TABLE 4** Quantitative Impact Measurements

| Outcome Category    | Pre-Implementation Baseline | Post-Implementation Results | Statistical Significance     | Study Quality Level |
|---------------------|-----------------------------|-----------------------------|------------------------------|---------------------|
| Cardiac Arrest Rate | 2.1-4.3 per 1000 admissions | 1.4-2.8 per 1000 admissions | $p < 0.05$ in 78% of studies | Moderate to High    |
| RRT Activation      | 8-15 per 1000 patient days  | 18-35 per 1000 patient days | $p < 0.001$ consistently     | High                |
| ICU Transfer Rate   | 12-25 per 1000 admissions   | 15-32 per 1000 admissions   | $p < 0.05$ in 65% of studies | Moderate            |
| Hospital Mortality  | 1.8-3.2% overall            | 1.6-2.9% overall            | $p > 0.05$ in 60% of studies | Variable            |
| Length of Stay      | 4.2-8.7 days average        | 4.1-9.2 days average        | Mixed results                | Low to Moderate     |

Improving the satisfaction and acceptance of healthcare providers are the key determinants of the long-term success of deterioration prediction technologies. Systems with clear, operational instruction with reduced false alarms are more likely to be accepted by the clinical staff. In contrast, alerts systems that do not have a strong relationship with patient outcomes might over time have reduced compliance because staff become susceptible to alert fatigue or lose trust in the reliability of the system(9).

The educational and training results linked to the introduction of deterioration prediction technology are not given enough attention despite the possibility of its significance in the long-term success. The systems that incorporate effective training schemes and continuous education elements might obtain improved results compared to the ones that depend mainly on the implementation of the technologies without the adequate development of the knowledge. The chance to improve clinical evaluation capabilities and pattern identification potential corresponds to a valuable prospect of benefit that goes beyond the particular technology being introduced.

## 4. Conclusion

The overall review of the technologies applied to the prediction of clinical deterioration among hospitalized adults can be discussed as a highly inventive domain of research that has become more sophisticated and has enormous opportunities to be further developed. The development of these technologies into the rudimentary, manual-calculated scoring systems into sophisticated, artificial intelligence-written algorithms reflects not only a technological development but an overhaul of the approach to patient safety and clinical risk management taken by healthcare systems.

The prevalence of studies based on developed healthcare systems in the United States and the United Kingdom makes critical implications regarding the validity and transferability of these technologies worldwide. Although these healthcare systems have produced worthy of knowledge and new ideas, the geographic and economic concentration of research has created a concern of relevance and feasibility of these strategies in resource constrained environments or healthcare systems of varying organizational design and priorities.

The ongoing struggle of having to translate the gains on intermediate outcome measurements into absolute decreases in mortality within the hospital indicates that the linkage between forewarn systems and the final patient outcomes is more multifaceted than thought. This complexity can be indicative of the multifactorial character of hospital mortality, the effect of factors other than the timely identification of worsening, and the necessity of system-wide changes that go beyond the deployment of prediction technologies.

Indeed, the inclusion of nursing clinical judgment as a formal variable in deterioration prediction technologies is a major validation of nursing expertise and professional knowledge. Nonetheless, this integration also asks significant questions concerning how to systematically capture, quantify and standardize clinical intuition that has long been viewed as being mostly subjective and experiential. Future research avenues could also consider

elaborate methods of integrating professional judgment and yet upholding objectivity and reproducibility that defines an efficient early warning system.

The technological trend of more advanced artificial intelligence and machine learning applications in the medical domain indicates that the next generation of deterioration predictors will probably be more capable of detecting more subtle patterns and associations that a human eye cannot notice. Nonetheless, this technical complexity should be weighed against the issues of interpretability, transparency, and the extent to which clinicians need to be familiar with and have confidence in the systems to which they are solicited to appeal in making patient care decisions.

Considerations of implementation science have become more prevalent in the definition of the success or failure of technologies in predicting deterioration. The understanding that even the most effective technologies may not deliver the intended results unless they are carefully incorporated into the current workflows or insufficiently backed by organizational systems imply that any further developmental process should equally consider the issues of implementation and the technological complexity.

The economic consequences of large-scale implementation of deterioration prediction technologies are not fully assessed as they should have been so far in the current literature. The intuitive power of preventing adverse events by intervening early can be overwhelming, but the reality of whether these systems are cost-effective, or not lies in complicated interactions between the implementation costs, changes in the use of the resources and the improvement of the outcomes, which can only be grasped by means of a sophisticated economic analysis.

The possibility of these technologies becoming a cause of health inequities is a subject of future research. Assuming some groups of patients are less prone to the usefulness of early warning mechanisms because of their physiological presentation pattern differences, baseline health status, or interaction with the healthcare systems, the broad use of such technologies might contribute to the further decrease in the disparities in healthcare outcomes.

The application of quality improvement methodology is the potential way to the next level of improving the efficiency and sustainability of deterioration prediction technologies. A combination of technology implementation and systematical quality improvement measures can possibly bring more positive outcomes than those where the implementation is mainly based on the use of technology without any particular focus on the continuous improvement and system optimization.

Patient and family engagement in the process of recognizing deterioration and early warning systems is a very under-researched field that has much room to be innovative. Technologies addressing patient self-assessment, family assessments, or other types of interaction with the patient may have a higher sensitivity or sooner warning than technologies that are based solely on professional assessment.

The implications of deterioration forecasting technologies on professional development and education are that the field of nursing and medical education is able to improve its curricular in order to equip future healthcare practitioners with the skills to use the systems effectively. The development of technological competency coupled with the development of better clinical assessment skills can help train professionals that can become proficient users of early warning systems and retain the clinical judgment abilities that are still necessary to provide safe care to the patients.

The global aspect of patient safety implies that there are potential collaborative research and development opportunities that can thus speed up the innovation process and at the same time make sure that the technological progress can be translated to various healthcare settings. This kind of collaboration could resolve the existing geographic concentration of research, as well as innovation relevant to the healthcare needs of the world.

Considerations of research methodology raise the possibility of more advanced study designs that are more capable of measuring the complexity and system-level effects of deterioration prediction technologies. Conventional methods of clinical trials might not be sufficient to test interventions that impact whole systems of health care and various domains of outcomes jointly, and new research methods may be necessary to better measure these multifaceted interventions.

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### **Conflicts of interest**

The authors have no conflicts of interest to declare

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