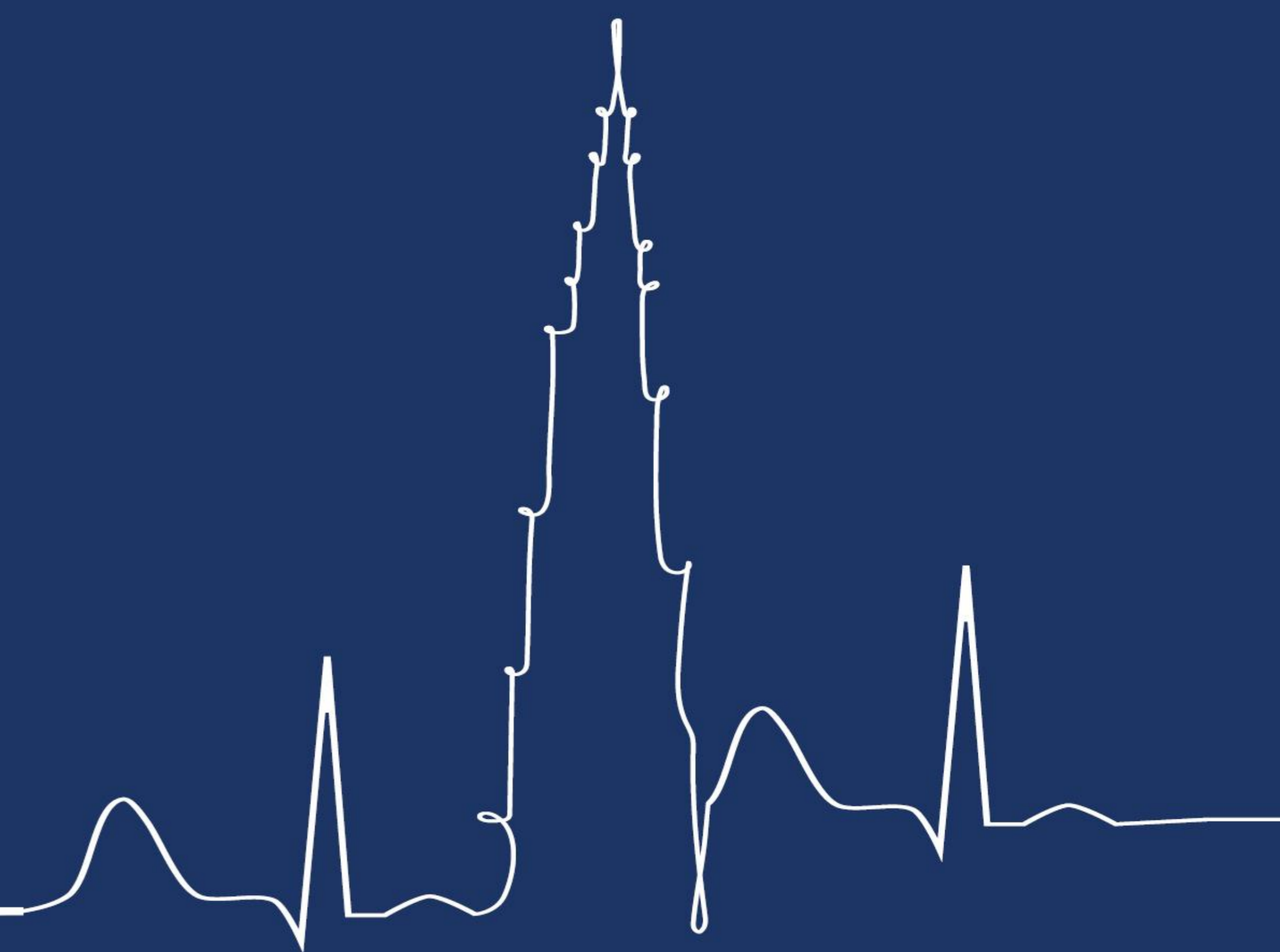




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ePOSTER



Artificial Intelligence in the Diagnosis and Prognosis of Arrhythmias: Revolutionizing Patient Care

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INTRODUCTION

Arrhythmias are among the most common and clinically significant cardiac conditions, contributing substantially to morbidity and mortality worldwide. These disorders, which include atrial fibrillation (AF), ventricular tachycardia (VT), and ventricular fibrillation (VF), pose significant diagnostic and prognostic challenges. Standard diagnostic methods, including electrocardiograms (ECGs) and electrophysiological studies, have been integral to detecting and managing arrhythmias. However, these traditional approaches often struggle to identify subtle, high-risk patterns and accurately predict long-term outcomes, particularly in patients with complex clinical presentations.

The emergence of artificial intelligence (AI) technologies, specifically machine learning (ML) and deep learning (DL), has revolutionized the field of cardiology. AI models, trained on vast datasets, possess the ability to uncover intricate patterns in data, offering unprecedented precision in diagnosis and risk stratification. This study investigates the current role of AI in diagnosing and prognosticating arrhythmias, emphasizing its potential to transform patient care and improve outcomes.

AIM

The primary objective of this study is to critically evaluate the role of artificial intelligence in the diagnosis and prognosis of arrhythmias. Specifically, the study aims to:

- Assess the diagnostic accuracy of AI models in detecting various arrhythmias, including AF and VT.
- Explore the prognostic capabilities of AI in predicting arrhythmia-related outcomes, such as sudden cardiac death and recurrent arrhythmias.
- Identify limitations, challenges, and future directions for integrating AI into clinical cardiology.

METHODS

This review involved a systematic analysis of peer-reviewed articles published between 2018 and 2023. Databases such as PubMed, SCOPUS, and Web of Science were searched using keywords such as "artificial intelligence," "arrhythmias," "machine learning," and "ECG analysis."

Inclusion criteria were as follows:

- Studies focusing on the application of AI techniques in arrhythmia diagnosis and prognosis.
- Use of ML, DL, or neural networks applied to ECGs, electrocardiographic imaging, or clinical datasets.
- Studies with rigorous methodological design and clinically relevant outcomes.

Exclusion criteria included studies with small sample sizes, lack of validation cohorts, or those not directly related to arrhythmias. Data from selected studies were synthesized to evaluate the diagnostic and prognostic performance of AI models.

RESULTS

Diagnostic Capabilities of AI

Deep learning models trained on large ECG datasets demonstrated superior diagnostic accuracy in detecting arrhythmias, particularly atrial fibrillation (AF). Several studies reported sensitivities and specificities exceeding 90%, significantly outperforming traditional manual and rule-based diagnostic systems. Real-time detection capabilities of AI algorithms have shown promise in identifying arrhythmias during routine clinical monitoring, with potential applications in wearable devices and remote patient care.

Prognostic Applications of AI

AI models have been instrumental in predicting arrhythmia-related outcomes. For example, machine learning algorithms trained on ECG and clinical data achieved up to 85% accuracy in predicting the onset of ventricular tachycardia (VT) and sudden cardiac death. Risk stratification models developed for heart failure patients identified those at high risk for

arrhythmias, enabling timely interventions and personalized therapeutic strategies.

Detection of Subtle Markers

One of the key advantages of AI is its ability to identify subtle predictors of arrhythmogenic risk. AI models have successfully detected QT prolongation, T-wave alternans, and abnormalities in heart rate variability—markers often overlooked in manual analysis. These findings underscore the potential of AI to enhance early diagnosis and preventative care.

CONCLUSIONS

Artificial intelligence has demonstrated remarkable potential to improve the diagnosis and prognosis of arrhythmias. By providing accurate and real-time detection, risk stratification, and personalized treatment insights, AI represents a paradigm shift in cardiology.

However, significant challenges remain. The development of robust AI models requires access to large, diverse datasets, and the interpretability of these algorithms must be ensured for clinical adoption. Furthermore, ethical concerns related to data privacy and algorithmic bias must be addressed.

Future efforts should focus on refining AI models, validating their utility across diverse populations, and integrating them into hybrid diagnostic systems that combine AI insights with traditional clinical expertise. The continued evolution of AI technologies promises to transform the management of arrhythmic disorders, heralding a new era of precision medicine in cardiology.

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