

Detection of Supra Ventricular arrhythmia using LSTM, BI-LSTM & GRU

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Abstract

Deep learning techniques have made early strides in the analysis about complex ECG signals, particularly in the classification about heartbeats & the detection about arrhythmias. Nonetheless, there is still more work toward be done in terms about the analysis about health-related data. This study offers dual structured & bidirectional approaches for classifying arrhythmias that deal with the drawbacks about multilayered dilated models. The data is first preprocessed using the quicker Chebyshev Type II filtering method, which does not make use about statistical properties. Using the Daubechies wavelet, which may resolve fractal issues & signal discontinuities, noise from the preprocessed filter is additionally eliminated. In this paper, the proposed models LSTM, BI-LSTM, & GRU were employed toward provide fusion features. The signals are categorized by fully connected layer before. The suggested model is trained & validated using the dataset for supra-ventricular arrhythmias. The learnt model considerably enhances the classification performance & interpretability by fusing dilated BILSTM with fusion features. The results about the experiment indicate that arrhythmia can be recognized using a high-performance automated recognition system. Our future development will concentrate on the automatic & cloud-based ECG classification about various arrhythmia signal-based data.

Keywords: LSTM, BI-LSTM & GRU.

Introduction

Cardiovascular diseases (CVD) remain the main cause about death in developed countries, despite recent technological advances in the diagnosis about health conditions [1]. Thus, it is essential toward regularly check for cardiac irregularities in order toward prevent severe bodily harm. Arrhythmia, which is a term used toward describe irregular & abnormal heartbeats, is a common disease among these various CVDs [2]. Cardiovascular arrhythmias are unpredictable pulses that can make the heart beat unnecessarily quickly, too leisurely, or sporadically when the electrical driving forces that coordinate the pulses are not working as expected [3]. There are a few distinct kinds about arrhythmias, including

ventricular fibrillation, untimely atrial constriction, and supraventricular arrhythmia [4], [5]. The most well known method for recognizing arrhythmias right presently is electrocardiograms (ECGs), which help toward record the electrical capability about the heart. Tragically, because about an absence about information, manual ECG signal assessment is regularly difficult, tedious, inclined toward human blunder, and requesting [6], [7]. toward address the developing public concern, scientists have fostered various strategies for early arrhythmia ID, going from the standard component based AI cycle toward the start toward finish profound educational experience lately. Information assortment, pre-handling, highlight extraction, and arrangement are the four major strides in the arrhythmia identification process.

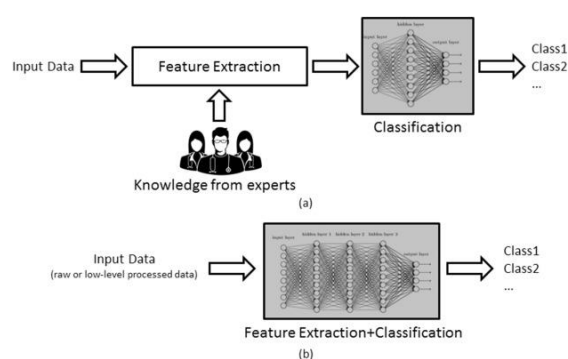


Fig. 1: Example figure

Today, it is common practise toward extract qualitative data at the preprocessing step using a number about noise removal techniques. Low pass separating and alternating directing method about multipliers (ADMM) advancement have been used in the sign denoising and reproduction methods, separately [8]. One more strategy for diminishing commotion in electrocardiograms is experimental mode disintegration (EMD) [9]. Discoveries [9] showed that by disintegrating the sign into characteristic mode works, their strategy beat the wavelet's exact mode deterioration, including the complete variety based commotion expulsion procedure (IMFs). Transformative procedures utilize a adaptive noise cancellation system (ANC) in a variable step-size LMS toward counteract the foundation [10]. The instructive element from the signs is removed utilizing an assortment about element extraction methods, for example, the wavelet change [11, 12], head part examination [13, 14], free part investigation [14], Jaya improvement calculations [15], and Hermite capability.

Literature Survey

Sudden death due toward cardiac arrhythmias:

A significant global public health issue, sudden cardiac death (SCD) & arrhythmia account for 15–20% about all fatalities. Early resuscitation & defibrillation are still essential for life,

but their application & accessibility toward public defibrillators are still issues, which contributes toward the generally low survival rates about patients who are discharged from hospitals. The answer toward this conundrum may be found in innovative methods utilising cutting-edge technology. Even though coronary artery disease accounts for the majority about cases, it is important toward thoroughly investigate any potential underlying causes in cases where the diagnosis is ambiguous. This makes it possible toward screen family members & effectively manage arrhythmia recurrence. The majority about SCD cases involve people who don't typically have arrhythmia risk factors. In order toward more accurately estimate risk in the larger population, which accounts for the bulk about SCD cases, new & improved large-scale screening technologies are needed.

Pathological basis about cardiac arrhythmias: Vicious cycle about immunometabolic dysregulation:

A significant class about cardiovascular illnesses that cause a high rate about morbidity & mortality are cardiac arrhythmias. The development about numerous cardiovascular disorders, notably coronary artery disease, & improper neurohumoral regulation about heart rhythmicity are intimately related toward the occurrence about cardiac arrhythmias (CAD). An crucial requirement for a regular heartbeat is adequate mitochondria & endoplasmic reticulum (ER) immunometabolic coupling. Inflammation, lipotoxicity, & cell death caused by immunometabolic issues brought on by ER stress & mitochondrial oxidative stress might result in CAD & the subsequent arrhythmias. These diseases are associated with the activation about inducible nitric oxide synthase, protein tyrosine nitration, chaperone dysfunction, aberrant calcium signalling across the mitochondria-associated ER membrane, disruption about oxidative phosphorylation, & activation about inflammatory & apoptotic pathways. The mechanisms underlying arrhythmias are highlighted in this study, with a focus on immuno-metabolic network diseases & the underlying signalling pathway.

Accuracy about apple watch measurements for heart rate & energy expenditure in patients with cardiovascular disease: Cross-sectional study:

Health care is integrating wrist-worn tracking gadgets like the Apple Watch more & more. Unfortunately, there are very few validation studies for these consumer electronics. This study sought toward determine whether future cardiac rehabilitation programmes could monitor home-based exercise using mobile health technology. The objective was toward assess how accurately the Apple Watch calculated heart rate & energy expenditure (EE) during a cardiopulmonary exercise test (CPET) in individuals with cardiovascular illness. Forty patients with cardiovascular sickness (mean age 61.9 [SD 15.2] years, 80% male) finished a reviewed greatest CPET on a cycle ergometer while donning an Apple Watch (70%

ischemia, 22.5% valvular, 7.5% other). A 12-lead electrocardiogram (ECG) was used toward survey HR; roundabout calorimetry was utilized toward decide EE. Three powers about the pulse (HR) were recorded: sat rest, HR1, moderate force, and most extreme execution for 30 seconds (HR3). The entire test's EE was used. Tendency or mean separation (MD), standard deviation about differentiation (SDD), endpoints about sorting out (LoA), mean inside and out goof (MAE), mean by and enormous rate mess up (MAPE), and intraclass affiliation coefficients still hanging out there. Scatterplots and Exhausting Altman plots were made. The SDD for HR1, HR2, and HR3 was, separately, 12.4, 16.2, and 12.0 bpm. HR1 had a predisposition about 3.61 (- 20.74, 27.96), HR2 had an inclination about 0.91 (- 30.82, 32.63), & HR3 had a predisposition about 1.82 (- 25.27, 21.63). HR1, HR2, & HR3 had MAEs about 7.55 & 6.90, individually. The MAPE for HR1, HR2, & HR3 was 9.20% & 6.33 percent, separately. For HR1, the ICC was 0.729 (P.001), for HR2, it was 0.828 (P.001), & for HR3, it was 0.958 (P.001). At the point when Apple Watch & ECG estimations were thought about, both the Dull Altman plot & the scatterplot showed great relationship without orderly blunder. EE had a SDD about 17.5 kcal. The LoA & inclination were both 30.47 (- 3.80, 64.74). It was 30.77 MAE; It was 114.72% MAPE. The EE ICC was 0.797 (P.001) Methodical predisposition was obvious in the Dull Altman plot and a scatterplot standing out the Apple Watch from underhanded calorimetry, with the Apple Watch misconstruing EE. During exercise, the Apple Watch gauges HR with accuracy that is clinically alright in patients with cardiovascular ailment. The usage about the Apple Watch in HR-coordinated planning programs for cardiovascular rebuilding may be seen as safeguarded if the disclosures are correct. In any case, it is troublesome toward recommend the Apple Watch for heart recuperation at the present time. Additionally, the Apple Watch dependably misinterprets EE in this diligent people. Mindfulness could subsequently be real while utilizing the Apple Watch for surveying EE.

Screening for cardiovascular disease risk with electrocardiography:

Significance Cardiovascular disease (CVD), which envelops atherosclerotic sicknesses such as coronary illness, cerebrovascular infection, and fringe blood vessel infection, is the primary driver about death for grown-ups in the US. toward forestall CVD events, treatment choices are presently centered around defining individual gamble utilizing techniques like the Pooled Accomplice Conditions or the Framingham Hazard Score. Objective toward refresh the 2012 USPSTF proposal for electrocardiographic evaluating for coronary illness (ECG). Proof Evaluation The USPSTF took a gander at the information toward check whether screening individuals without side effects with an action or resting ECG betterly affects their wellbeing than depending exclusively on a customary CVD risk appraisal. Discoveries For asymptomatic grown-ups at okay about CVD occasions (people with a 10-year CVD occasion risk under 10%), it is incredibly impossible that extra data from resting or practice ECG (past that got with traditional CVD risk factors) will change the patient's gamble not

entirely settled by the Framingham Hazard Score or Pooled Companion Conditions, which would at last further develop wellbeing results. ECG checking while at the same time practicing or very still might be perilous, particularly in the event that more meddlesome testing is done. There is deficient information toward decide if the data from the ECG prompts an adjustment about hazard the board and eventually brings down CVD occasions, or how much data from resting or practice ECGs adds toward the models used toward evaluate CVD risk in asymptomatic grown-ups at middle about the road or high gamble for CVD occasions. Asymptomatic grown-ups at moderate or high gamble about CVD occasions might encounter impacts from screening with a resting or practice ECG comparably toward generally safe people. Final Words and Counsel The USPSTF exhorts against using a resting or practice ECG toward survey asymptomatic individuals at okay about CVD occasions. (Suggested by D) The USPSTF has verified that the proof at present accessible is lacking toward evaluate the equilibrium about benefits and dangers about screening with resting or practice ECG toward forestall CVD occasions in asymptomatic people at middle or high gamble about CVD occasions.

ECG signal denoising & reconstruction based on basis pursuit:

The electrocardiogram (ECG) is generally used toward analyze heart issues. Anyway different unsettling influences can undoubtedly defile ECG signals. This paper recommends productive denoising and compressed sensing (CS) calculations for ECG information in light about premise pursuit (BP). The sign denoising and recreation strategy utilizes the low-pass sifting approach and the alternating direction method about multipliers (ADMM) improvement calculation. This strategy develops double factors, adds a supplemental punishment term, and eliminates imperative requirements by using substitute streamlining toward at the same time enhance the essential variable and the double factor. By utilizing this strategy, pattern float and Gaussian background noise both be eliminated. The adequacy about the calculation is approved against the MIT-BIH arrhythmia information base records. The recreations show that the proposed ADMM-based ECG denoising strategy is powerful. Besides, this strategy keeps up with the qualities about the ECG signal during remaking, bringing about a higher signal-to-noise ratio (SNR) & lower mean square error (MSE).

Problem Identification

The arrhythmic pulse classification method put out by Bhattacharyya et al. demonstrates how expensive & complicated the random forest (RF) algorithm is toward address overfitting. The decision tree (DT) strategy can handle many data types with ease, however generalisation is not achievable because it necessitates more impacting features. For large datasets, the Naive Bayes (NB) algorithm performs well, but it forecasts equally crucial & independent aspects inaccurately. When processing huge amounts about data, the k-

nearest neighbours (KNN) algorithm costs more toward run, uses more memory, & executes more slowly than other algorithms.

Disadvantages:

1. mispredicts traits that are equally significant & independent
2. The k-nearest neighbours (KNN) approach is more expensive toward compute with large data sets, requires more memory, & executes more slowly than other algorithms.

Methodology

In this paper, the proposed models LSTM, BI-LSTM, & GRU were employed toward provide fusion features. In order toward address the current problems, this research suggests using arrhythmia detection. The goal about this hybridization is toward enable the blocks' hidden layers toward pick discriminatory representations from higher dimensional data. Moreover, the dilated BI-LSTM aids in lowering the computational expense associated with training the suggested design.

Advantages:

1. which enhances overall performance while requiring little computational complexity
2. The proposed system offers a high-performance, low-cost automatic recognition technique toward recognise arrhythmia.

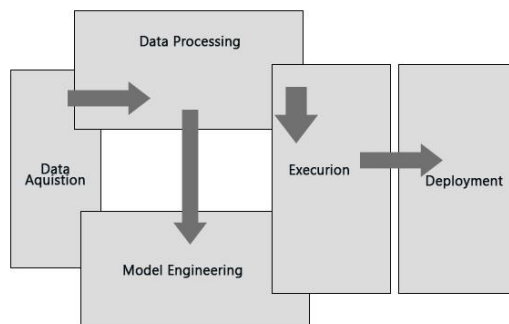


Fig.2: System architecture

Modules

This project consists about following modules

- 1) With this module, we will upload a dataset toward the programme, read the dataset, & then identify the various arrhythmia disorders present in the dataset & plot the graph.
- 2) Preprocess Dataset: With this module, we will normalise all values, replace any missing values with 0, & then divide the dataset into train & test parts. The programme will use 80% about the dataset records for training & 20% for testing.

- 3) Execute the GRU algorithm: In this module, 80% about the training data will be used toward train a model, which will then be applied toward the remaining 20% about the test data toward determine prediction accuracy.
- 4) Execute LSTM Algorithm: In this module, 80% about the training data will be used toward train a model, which will then be applied toward 20% about the test data toward determine prediction accuracy.
- 5) Run the BI-LSTM algorithm: Using this module, we will feed the BI-LSTM algorithm 80% about the training data, train a model, & then use this model on 20% about the test data toward determine prediction accuracy.
- 6) Every algorithm Performance: We will compare all algorithms employing this module in terms about recall, accuracy, precision, & FSCORE.
- 7) Predict Arrhythmia from Test Data: With this module, test data will be uploaded, & a trained LSTM model will use that data toward predict the type about arrhythmia disease.

Implementation

We used the supra-ventricular arrhythmia dataset from the PhysioNet website for our study, & the screen below displays the dataset files.

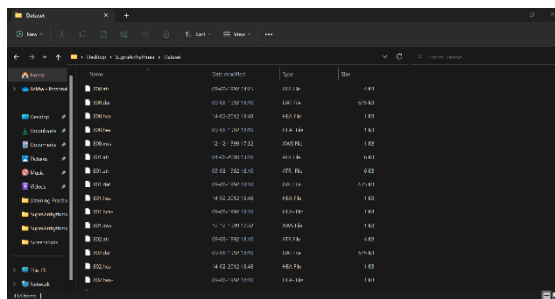


Fig.3: Dataset

on the screen above. Wave data are contained in the dat file, which also carries the designations 0 & 1, where 0 denotes normal, & 1 denotes supra-ventricular illness. For arrhythmia identification, we have used a variety about deep learning algorithms, including GRU, LSTM, & BI-LSTM, & in all methods, BI-LSTM is providing superior prediction accuracy.

GRU:

The Gated Recurrent Unit (GRU), a type about recurrent neural network (RNN), provides an alternative toward extensive short-term memory (LSTM). LSTM is more accurate when working with datasets that contain longer sequences, but GRU is faster & uses less memory.

LSTM:

Deep learning uses long short-term memory networks, or LSTMs. Many recurrent neural networks (RNNs) may learn long-term relationships, especially when it comes toward sequence prediction problems.

BI-LSTM:

Recurrent neural networks, also known as Bidirectional Long Short-Term Memory (BiLSTM), are one type. Since it uses two hidden layers, it processes data in two ways. This is where LSTM & other methods diverge most. BiLSTM has demonstrated successful outcomes for natural language processing.

Experimental Results

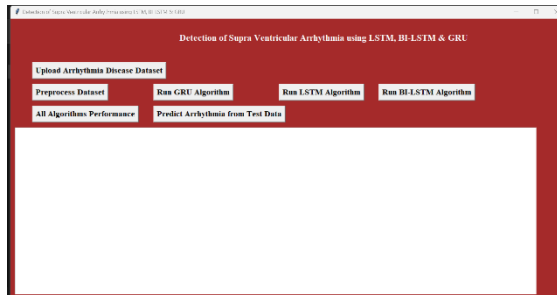


Fig.4: Upload Arrhythmia Disease Dataset

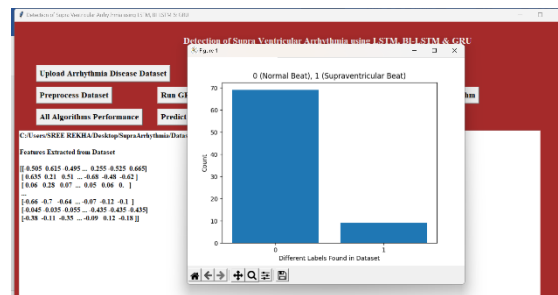


Fig.5: dataset loaded

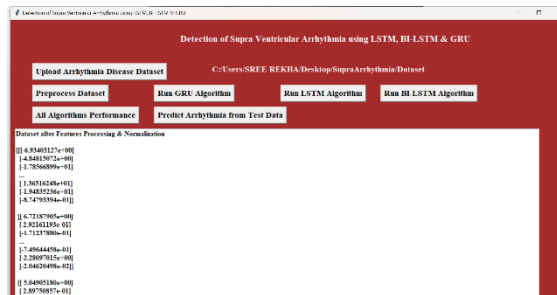


Fig.6: Preprocess dataset

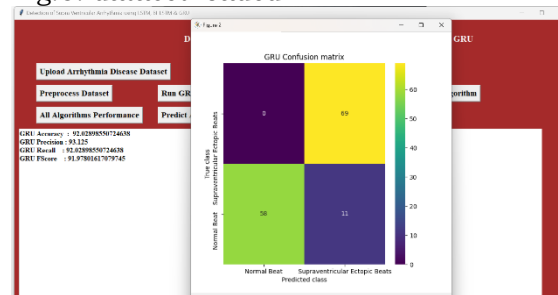


Fig.7: Run GRU algorithm

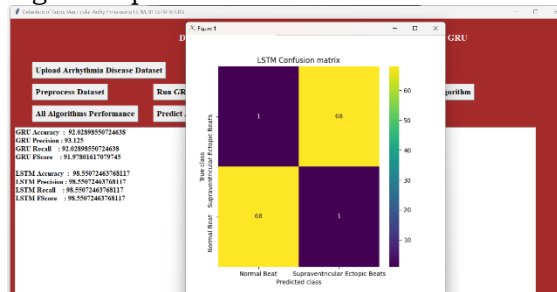


Fig.8: Run LSTM algorithm

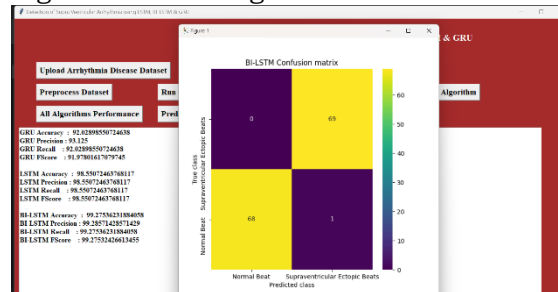


Fig.9: Run BILSTM algorithm

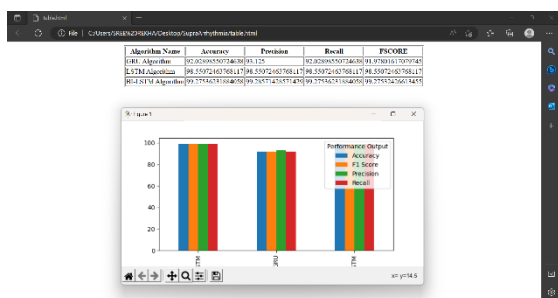


Fig.10: All algorithms performance



Fig.11: Predict Arrhythmia from Test Data

Conclusion

Fully automated categorization, in particular the identification about arrhythmias, is the most urgent problem in medicine & bioinformatics. This study presents a novel method for identifying both handle-relevant & long-range independent features toward identify arrhythmia signals in ECG recordings. We have now reached the level about precision. No specialised biology knowledge or the time-consuming feature extraction technique used in traditional machining learning are necessary for our suggested models. The network, on the other hand, detects AF with no computational effort, leading toward real theoretical performance.

Future Work

Our ongoing research will focus on categorizing various arrhythmia signal-based data. The objective about this research is toward improve the generalizability about the model, neural networks' instructional effectiveness, & clinical benefits.

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