

# Real time Data Stream Mining Approach to Arrhythmia Prediction

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**Abstract**— Recent data mining techniques are modeled to handle stream data by applying online and incremental approach where concept changes in dataset are learned efficiently. Streaming random forests algorithm is an extension of Breiman's random forests algorithm which handles online and incremental stream data. Streaming random forests adapts ensemble framework where classification is based on majority of tree votes. The work proposed in this paper mainly deals with construction of an efficient arrhythmia prediction system using streaming random forests algorithm for classification of three arrhythmia types. We applied streaming random forests algorithm on arrhythmia dataset obtained from physionet website and found that its accuracy is highly acceptable.

**Index Terms**— Arrhythmia Prediction, Concept drift, Data stream mining, ECG, ECG signal preprocessing, Feature extraction, Streaming random forests.

## 1 INTRODUCTION

Cardiovascular diseases remain a major threat to human life due to rapid changes in life style. Arrhythmia, a type of heart disease caused because of irregular rhythms or rate of heartbeats. Arrhythmia causes the heart to provide insufficient supply of blood to other organs which can cause damage to the brain or other organ. Atrial Fibrillation, premature Ventricular Contractions, ventricular Fibrillations, heart blocks etc are certain types of arrhythmia.

Continuous, real world, changing, infinite data obtained from semi-structured and non-structured databases, internet, multimedia databases etc is termed as data stream. Data stream mining refers to informational structure extraction as models and patterns from continuous data streams [5]. To deal with the high arrival rate of records, the data stream mining algorithms should be online and incremental. This gives a challenging phase in designing stream algorithms since the information need to extract from single pass over records. Compared to Stream Classification algorithms, Mainstream Classification algorithms come up with three phases: Training, Testing and Deployment [4]. Stream Classification Algorithms processes single stream of data which creates a need for interleaving all the three phases. ECG pathologies generate vast data which yields various features that plays a major role in better detection and classification of cardiac diseases.

This paper presents an arrhythmia prediction system using Streaming Random Forests algorithm for classification of arrhythmia into three different types. The system first preprocesses the ECG signals for normalizing them. The extracted features

## 2 RELATED WORK

The increase in the number of patients in intensive care units and their continuous observations call for the need of automated arrhythmia detection and prediction. Several researches and studies have emerged techniques to address this problem and much more are currently in practices. These techniques adapt the criteria of transforming mostly qualitative diagnostic feature into a more quantitative signal feature classification task. The ANFIS classifier uses as input the four statistical parameters calculated from ECG signals for cardiac arrhythmia classification [3]. The ANFIS classifier combines the best features obtained from fuzzy systems and neural networks. In [6] the Artificial Neural Network (ANN) based on feed forward back propagation with momentum used for classification of arrhythmia. The four parameters considered for classification are obtained from RR interval and the QRS complex. A knowledge-based method for arrhythmic beat classification and arrhythmic episode detection and classification using RR-interval signal is proposed in [7]. The beat classification algorithm and deterministic automaton used for classification of four categories of beat and six rhythm types respectively. Author Mahesh et al focused on the heart rate and Heart Rate Variability (HRV) for detecting cardiac abnormalities. Classification has been done using the combination of linear, non-linear measures with the help of three classifiers, Random Forests, Multilayer Perceptron Neural Network and Logistic Model Tree.

The detection of abnormal cardiac rhythms and automatic discrimination from the normal heart activity become an important task in clinical reasons. Several techniques are based on the detection of a single arrhythmia type and its discrimination from normal sinus rhythm or the discrimination between two different types of arrhythmia.

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are given as input to the classification algorithm to form the classification rules which help to predict the type of arrhythmia.

### 3 REAL TIME DATA STREAM MINING

Continuous, endless, ordered sequence of data is often termed as *Stream*. Existing data mining algorithms are unable to extract knowledge from stream datasets. Data stream mining algorithms have constraints such as online result generation, handling fast arrival rate of records, efficient concept drift adaptation etc. The aim of many stream mining applications is to predict the class or label of new instances provided knowledge of previous seen, known class membership or values of previous instances in the data stream. To cope with online learning, real time approaches often the concept changes, incremental learning techniques are applied. Decision tree learning is widely adapted classification method for stream mining. VFDT (Very Fast Decision Tree) is able to learn from abundant data within practical time and memory constraints [10]. CVFDT (Concept-adapting Very Fast Decision Tree Learner) is an extension of VFDT which maintains VFDT's speed and accuracy with added ability to detect and respond changes in example generating process [11]. Both the algorithms lack the ability of performing classification in real-time.

### 4 STREAMING RANDOM FORESTS [1]

Streaming Random Forests is a combination of techniques used for building decision trees and attribute selection techniques of Random Forests. The original Random Forests Algorithm proposed by Breiman [2] uses bootstrap sample of data for growing multiple decision trees. The majority of votes from all trees contribute for classification task. In general, random forests are similar to ensemble of binary decision trees. In contrast to the standard random forests algorithm the streaming version does not possess the facility of making multiple passes over data. Separate block of labeled records is required for building each tree. Due to this more labeled data is required for building a set of trees. To contribute towards decision of any node every labeled record is routed to an appropriate node of tree under tree building procedure. Whenever a new labeled record is arrived it is routed down the current tree, depending on its attribute values and inequalities of internal nodes [1]. For selection of best attribute and splitting point the Gini Index test [9] and Hoeffding bound criterion [10] need to be satisfied. The transformation of frontier node into internal node is based on the inequality of attribute and split point. Whenever records accumulated at a node resemble particular class the node is converted into leaf node. The basic tree building procedure in streaming random forests algorithm with different tree building parameters is shown in Fig 1.

### 5 PROPOSED WORK

Though there are number of diseases, it has been observed and brought forth by world health organization that the death rate due to cardiovascular diseases remains high. To aid in the quality of better health care it is advisory that the patient need to be continuously monitored and the automatic diagnosis and speedy recovery should be achieved. The system architecture shown in Fig 2 takes the ECG signal as an input and predicts the arrhythmia types as an output.

#### Procedure BuildTree

```

/*grow tree*/
while more data records in the tree window
    read a new record
    pass it down the tree
    if it reaches a frontier node
        if first record at this node
            randomly choose M attributes
            find intervals for each of the M attributes
            update counters
            if node has seen  $n_{min}$  records
                if Hoeffding bounds test is satisfied
                    save node split attribute
                    save corresponding split value
                if no more records in the node window
                    if node records are mostly from one class
                        mark it as leaf node
                        assign majority class to node
            else
                save best split attribute seen so far
                save corresponding split value
        end while
    /* prune tree */
    while more frontier nodes
        if node has records arrive at it
            mark it as leaf node
            assign majority class to it
        else /* node has zero records */
            if sibling node is frontier with no records
                calculate purities of both sibling nodes
                if purities < pre-defined threshold
                    prune both nodes
                    mark parent node as a leaf
                    assign majority class to it
            else
                mark node as leaf node
                assign dominant class to it
    end while
end
    
```

Fig 1. Tree Building procedure in Streaming Random Forests [1]

ECG is a recommended tool for cardiac status analysis. With diseases classified on the basis of intervals between the various segments (PQRST) of ECG signal, this gives us an optimal assessment. Arrhythmia Prediction system provides an interface which would not affect the normal life cycle of a Heart Patient. This is done by providing on the go examination features with the use of system, the accuracy and complexity being taken into consideration. The ECG signal which is taken as an input is passed through various stages like Feature Extraction, Preprocessing, Classification of features and the Formation of rules using Streaming random forests algorithm. The system consists of two parts, online and offline processing

of data. The offline mode consist of training dataset formed using collections of patient’s ECG records with three arrhythmia types .In online mode the continuous ECG signal is taken as input and it is passed to pre-processing module. In this module the noise is removed from ECG signal, then the filtered ECG signal is given to feature extraction module where all time domain features of signal are extracted like R-R interval, Root Mean Square Successive Difference (RMSSD), Standard Deviation (SDNN) etc [3]. The data stream mining algorithm uses these calculated features to create rule database which is further used for classification task of new unlabelled records.

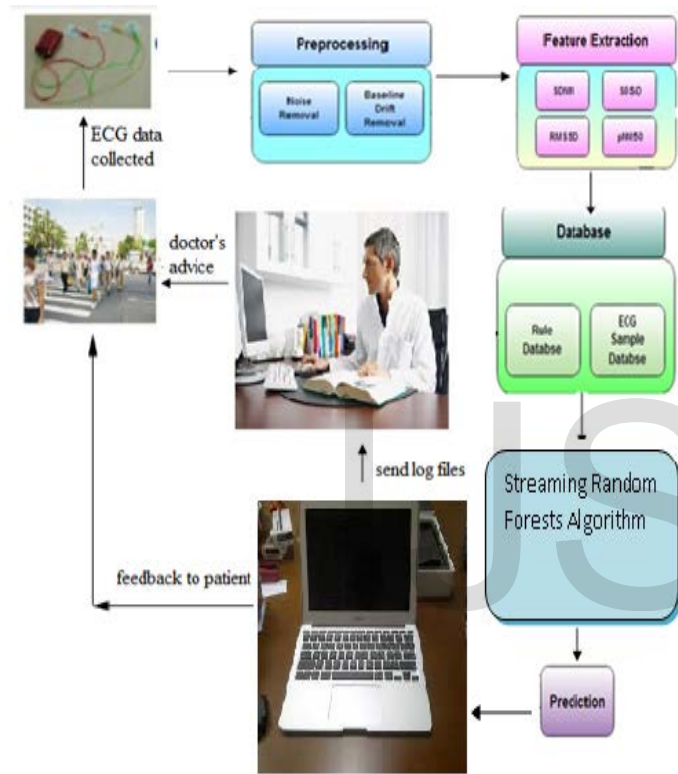


Fig 2. Proposed System Architecture for arrhythmia prediction

## 6 ARRHYTHMIA PREDICTION

The system is able to detect three types of arrhythmia namely PVC, AF and NSR. The ECG signals for training of classification algorithm were obtained from the MIT-BIH Atrial Fibrillation Database, MIT-BIH normal sinus rhythm database and MIT-BIH supraventricular Arrhythmia database. For testing purpose the MIT-BIH Arrhythmia database is used [12]. The range of four statistical parameters calculated as a result of training dataset is further used for classification of new unlabeled records. Figure 3 shows the schematic of the arrhythmia prediction performed.

Table 1 shows the testing result of streaming random forests on MIT-BIH arrhythmia dataset for AF, PVC and NSR diseases.

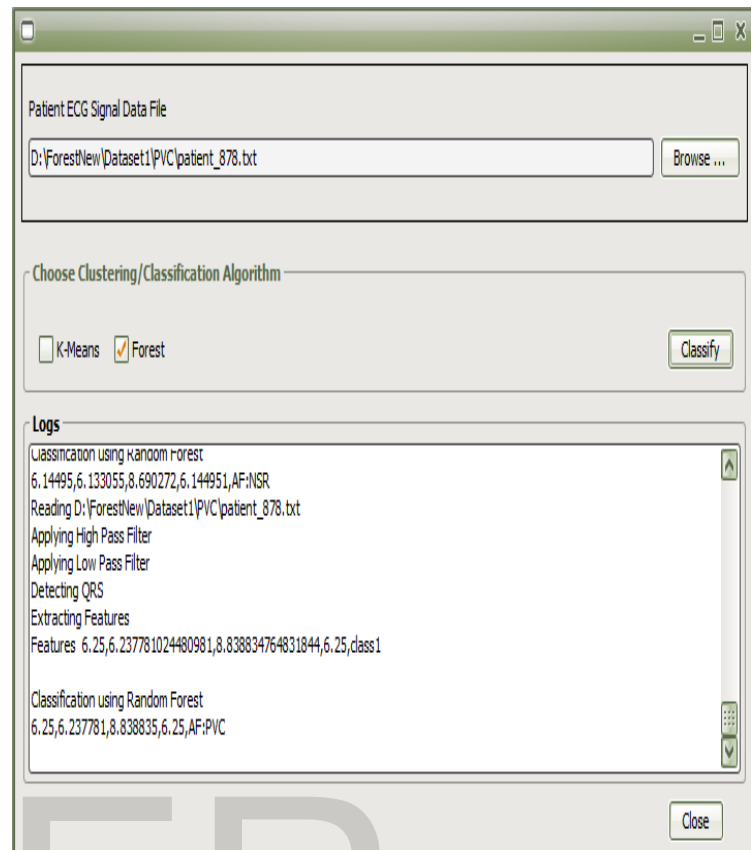


Fig 3. Arrhythmia Predicted as type PVC

TABLE 1  
 TESTING RESULTS ON MIT-BIH ARRHYTHMIA DATASET

Algorithm	Disease	No of Records for training	No of records for testing	Correctly classified	Accuracy
Random Forests	AF	10	5	4	80
			10	8	80
			15	11	73.33
Random Forests	NSR	10	5	5	100
			10	6	60
			15	11	73.33
Random Forests	PVC	10	5	3	60
			10	7	70
			15	13	86.66
				Average	76

## 7 CONCLUSION

Computerised detection and classification of ECG signals is gaining a worldwide attention due to quick diagnosis and preventive actions. The availability of a good classifier for classifying the arrhythmia types in a patient creates an opportunity for future development that implements the results of this research to work in the form of decision support system. The proposed and developed system performs analysis of various ECG signals and predicts the type of arrhythmia. The system helps an individual by evaluating his or her risk levels in terms of arrhythmia symptoms and will alert the patient as well as the physician from the problems one can suffer. The system is developed in such a way that it can handle all ECG signals whether normal or abnormal. The streaming random forests algorithm works on the extracted features efficiently for determining the type of arrhythmia present in a patient.

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