EARLY HEART DISEASE PREDICTION USING DATA MINING TECHNIQUES

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ABSTRACT

The successful application of data mining in highly visible fields like e-business, marketing and retail has led to its application in other industries and sectors. Among these sectors just discovering is healthcare. The Healthcare industry is generally “information rich”, but unfortunately not all the data are mined which is required for discovering hidden patterns & effective decision making. Discovery of hidden patterns and relationships often goes unexploited. Advanced data mining modeling techniques can help remedy this situation.

This research paper intends to use data mining Classification Modeling Techniques, namely, Decision Trees, Naïve Bayes and Neural Network, along with weighted association Apriori algorithm and MAFIA algorithm in Heart Disease Prediction. Using medical profiles such as age, sex, blood pressure and blood sugar it can predict the likelihood of patients getting heart disease.

1. INTRODUCTION

Data mining is the process of finding previously unknown patterns and trends in databases and using that information to build predictive models. Data mining combines statistical analysis, machine learning and database technology to extract hidden patterns and relationships from large databases.

The World Health Statistics 2012 report enlightens the fact that one in three adults worldwide has raised blood pressure – a condition that causes around half of all deaths from stroke and heart disease. Heart disease, also known as cardiovascular disease (CVD), encloses a number of conditions that influence the heart – not just heart attacks. Heart disease was the major cause of casualties in the different countries including India. Heart disease kills one person every 34 seconds in the United States. Coronary heart disease, Cardiomyopathy and Cardiovascular disease are some categories of heart diseases. The term “cardiovascular disease” includes a wide range of conditions that affect the heart and the blood vessels and the manner in which blood is pumped and circulated through the body. Diagnosis is complicated and important task that needs to be executed accurately and efficiently. The diagnosis is often made, based on doctor’s experience & knowledge. This leads to unwanted results & excessive medical costs of treatments provided to patients. Therefore, an automatic medical diagnosis system would be exceedingly beneficial. Our work attempts to present the detailed study about the different data mining techniques which can be deployed in these automated systems.
2. METHODOLOGY

This paper exhibits the analysis of various data mining techniques which can be helpful for medical analysts or practitioners for accurate heart disease diagnosis. Due to resource constraints and the nature of the paper itself, the main methodology used for this paper was through the survey of journals and publications in the fields of medicine, computer science and engineering.

3. RESEARCH FINDINGS

3.1 Data Mining in the Heart Disease Prediction.

Different supervised machine learning algorithms i.e. Naïve Bayes, Neural Network, along with weighted association Apriori algorithm, Decision algorithm have been used for analyzing the dataset in [1]. The data mining tool Weka 3.6.6 is used for experiment. Weka is a collection of machine learning algorithms for data mining tasks. The algorithms can either be applied directly to a dataset or called from your own Java code. Weka contains tools for data pre-processing, classification, regression, clustering, association rules, and visualization. It is also well-suited for developing new machine learning schemes.

3.1.1 Decision Tree is a popular classifier which is simple and easy to implement. There is no requirement of domain knowledge or parameter setting and can high dimensional data can be handled. It produces results which are easier to read and interpret. The drill through feature to access detailed patients’ profiles is only available in Decision Trees.

3.1.2 Naïve Bayes is a statistical classifier which assumes no dependency between attributes. This classifier algorithm uses conditional independence, means it assumes that an attribute value on a given class is independent of the values of other attributes. The advantage of using naïve bayes is that one can work with the Naïve Bayes model without using any Bayesian methods.

3.1.3 Neural Networks

An artificial neural network (ANN), often just called a "neural network" (NN), is a mathematical model or computational model based on biological neural network. In other words, it is an emulation of biological neural system [9]. In feed-forward neural networks the neurons of the first layer forward their output to the neurons of the second layer, in a unidirectional fashion, which explains that the neurons are not received from the reverse direction. There is connection between each layer & weights are assigned to each connection. The primary function of neurons of input layer is to divide input $x_i$ into neurons in hidden layer. Neuron of hidden layer adds input signal $x_i$ with weights $w_{ji}$ of respective connections from input layer. The output $Y_j$ is function of

$$Y_j = f \left( \Sigma w_{ji} x_i \right)$$

where $f$ is a simple threshold function such as sigmoid or hyperbolic tangent function.
4. DATA SOURCE

Using medical profiles such as age, sex, blood pressure and blood sugar it can predict the likelihood of patients getting a heart disease. It enables significant knowledge, e.g. patterns, relationships between medical factors related to heart disease. The publicly available heart disease database is used which can be used for determining various heart diseases.

Input Attributes:

1. age: age in years
2. sex: sex (1 = male; 0 = female)
3. cp: chest pain type
   Value 1: typical angina
   Value 2: atypical angina
   Value 3: non-anginal pain
   Value 4: asymptomatic
4. trestbps: resting blood pressure (in mm Hg on admission to the hospital)
5. chol: serum cholesterol in mg/dl
6. fbs: (fasting blood sugar > 120 mg/dl) (1 = true; 0 = false)
7. restecg: resting electrocardiographic results
   Value 0: normal
   Value 1: having ST-T wave abnormality (T wave inversions and/or ST elevation or depression of > 0.05 mV)
   Value 2: showing probable or definite left ventricular hypertrophy by Estes' criteria
8. thalach: maximum heart rate achieved
9. exang: exercise induced angina (1 = yes; 0 = no)
10. oldpeak = ST depression induced by exercise relative to rest
11. slope: the slope of the peak exercise ST segment
    Value 1: upsloping
    Value 2: flat
    Value 3: downsloping
12. ca: number of major vessels (0-3) colored by flouroscopy
13. thal: 3 = normal; 6 = fixed defect; 7 = reversible defect
14. num: diagnosis of heart disease (angiographic disease status)
   Value 0: < 50% diameter narrowing
   Value 1: > 50% diameter narrowing

Key Attributes

1. Patientid – Patient’s identification number

A total of 909 records were obtained from the Cleveland Heart Disease database. It has been observed during the analysis that Naive Bayes appears to be most effective as it has the highest percentage of correct predictions (86.53%) for patients with heart disease, followed by Neural Network (85.53%) and Decision Trees. Decision Trees, however, appears to be most effective in case of predicting patients with no heart disease, i.e. (89%) as compared to other two models.
In the research the number of attributes which were used for heart disease diagnosis were reduced. Earlier, 13 attributes were used for this prediction but this research work reduced the number of attributes to six only using Genetic Algorithm and Feature Subset Selection.

Genetic Algorithm [6] incorporates natural evolution methodology. The genetic search starts with zero attributes, and an initial population with randomly generated rules. Based on the idea of survival of the fittest, new population was constructed to match with fittest rules in the current population, as well as offspring of these rules. Offspring were generated by applying genetic operators; cross over and mutation. The process of generation continued until it evolved a population P where every rule in P satisfied the fitness threshold. With initial population of 20 instances, generation continued till the twentieth generation with cross over probability of 0.6 and mutation probability of 0.033. The genetic search resulted in six attributes out of thirteen attributes. After reduction of 13 attributes to 6 attributes, various classifiers are used on the dataset corresponding to these 6 attributes for heart disease prediction.

Performance analysis of these classifiers is shown in Table 4. It can be perceived from the table that Decision Tree has outperformed with highest accuracy and least mean absolute error.

<table>
<thead>
<tr>
<th>Techniques</th>
<th>Accuracy</th>
</tr>
</thead>
<tbody>
<tr>
<td>Naive Bayes</td>
<td>86.53%</td>
</tr>
<tr>
<td>Decision Tree</td>
<td>89%</td>
</tr>
<tr>
<td>ANN</td>
<td>85.53</td>
</tr>
</tbody>
</table>

5. ASSOCIATIVE CLASSIFICATION

Associative classification mining is a promising approach in data mining that utilizes the association rule discovery techniques to construct classification systems, also known as associative classifiers. Association rule mining is used to find associations or correlations among the item sets. It is an unsupervised learning where no class attribute is involved in finding the association rule. On the other hand, classification is a supervised learning where class attribute is involved in the construction of the classifier and is used to classify or predict the data unknown sample. Associative classification is a recent and rewarding technique which integrates association rule mining and classification to a model for prediction and achieves maximum accuracy. Associative classifiers are especially fit to applications where maximum accuracy is desired to a model for prediction. Various Techniques which can be used are apriori algorithm, éclat algorithm, FP-growth algorithm. Here we will be using Apriori Algorithm for discovering interesting relations in heart based diseases.
Apriori Algorithm:

Apriori is an algorithm for frequent item set mining and association rule learning over transactional databases. It proceeds by identifying the frequent individual items in the database and extending them to larger and larger item sets as long as those item sets appear sufficiently often in the database. A frequent itemset can be defined as a subset of frequent itemset i.e., if \{PQ\} is a frequent itemset, both \{P\} and \{Q\} should be a frequent itemset.

1. Iteratively discover frequent itemsets with cardinality from 1 to k (k-itemset).
2. Use the frequent itemsets produce association rules. Join Step: Ck is generated by joining Lk-1 with itself. Prune Step: Any (k-1)-itemset that is not frequent cannot be a subset of a frequent k-itemset. Initialize: K := 1, C1 = all the 1-item sets; read the database to count the support of C1 to determine L1. L1 := {frequent 1-item sets};

\[
k := 2;
\]
//represents the pass number//

While (Lk-1 ≠ \emptyset) do
begin
Ck := gen_candidate_itemsets with the given Lk-1 Prune (Ck) for all candidates in Ck do count the number of transactions of at least k length that are common in each item Ck
Lk := All candidates in Ck with minimum support;

k := k + 1;
end

Frequent Pattern mining using MAFIA

Mining frequent itemsets is an active area in data mining that aims at searching interesting relationships between items in databases[11]. It can be used to address a wide variety of problems such as discovering association rules, sequential patterns, correlations and much more. The proposed approach utilizes an efficient algorithm called MAFIA (Maximal Frequent Itemset Algorithm) which combines diverse old and new algorithmic ideas to form a practical algorithm. The proposed algorithm is employed for the extraction of association rules from the clustered dataset besides performing efficiently when the database consists of very long itemsets specifically.

Pseudo code for MAFIA :

MAFIA(C, MFI, Boolean IsHUT)
{
    name HUT = C.head C.tail;
    if HUT is in MFI
    stop generation of children and return
    Count all children, use PEP to trim the tail, and recorder by increasing support,
    For each item i in C, trimmed_tail
    {
        IsHUT = whether i is the first item in the tail
        newNode = C I
        MAFIA (newNode, MFI, IsHUT)
    }
    if (IsHUT and all extensions are frequent)
Stop search and go back up subtree
If (C is a leaf and C.head is not in MFI)
Add C.head to MFI

The cluster that contains data most relevant to heart attack is fed as input to MAFIA algorithm to mine the frequent patterns present in it. Then the significance weightage of each pattern is calculated using the approach described in the following subsection.

**Significance Weightage Calculation**

After mining the frequent patterns using MAFIA algorithm, the significance weightage of each pattern is calculated. It is calculated based on the weightage of each attribute present in the pattern and the frequency of each pattern.

\[
S_{Wi} = \sum_{i=1}^{n} W_{i}f_{i}
\]

Where \(W_{i}\) represents the weightage of each attribute and \(f_{i}\) denotes the frequency of each rule. Subsequently the patterns having significant weightage greater than a predefined threshold are chosen to aid the prediction of heart attack

\[
SFP = \{x : Sw (x) = \Phi\}
\]

Where SFP represents significant frequent patterns and \(\Phi\) represents the significant weightage. This SFP can be used in the design of heart attack prediction system.

**6. CONCLUSION**

In this paper the focus is on using different algorithms and combinations of several target attributes for effective heart attack prediction using data mining. Decision Tree has outperformed with 99.62% accuracy by using 15 attributes. Also the accuracy of the Decision Tree and Bayesian Classification further improves after applying genetic algorithm to reduce the actual data size to get the optimal subset of attribute sufficient for heart disease prediction.

Association classification technique apriori algorithm, was along with a new algorithm MAFIA was used .

Straight Apriori-based algorithms count all of the 2k subsets of each k-item set they discover, and thus do not scale for long item sets. They use “look a heads” to reduce the number of item sets to be counted. MAFIA is an improvement when the item sets in the database are very long.

**REFERENCES**


