

Heart Disease Prediction

Abstract:

Heart related diseases or Cardiovascular Diseases (CVDs) are the main reason for a huge number of death in the world over the last few decades and has emerged as the most life-threatening disease, not only in India but in the whole world. So, there is a need of reliable, accurate and feasible system to diagnose such diseases in time for proper treatment. Machine Learning algorithms and techniques have been applied to various medical datasets to automate the analysis of large and complex data. Many researchers, in recent times, have been using several machine learning techniques to help the health care industry and the professionals in the diagnosis of heart related diseases. This paper presents a survey of various models based on such algorithms and techniques and analyze their performance. Models based on supervised learning algorithms such as Support Vector Machines (SVM), K-Nearest Neighbour (KNN), Naïve Bayes, Decision Trees (DT), Random Forest (RF) and ensemble models are found very popular among the researchers.

Existing System:

Heart diseases have emerged as one of the most prominent cause of death all around the world. According to World Health Organisation, heart related diseases are responsible for the taking 17.7 million lives every year, 31% of all global deaths. In India too, heart related diseases have become the leading cause of mortality [1]. Heart diseases have killed 1.7 million Indians in 2016, according to the 2016 Global Burden of Disease Report, released on September 15, 2017. Heart related diseases increase the spending on health care and also reduce the productivity of an individual. Estimates made by the World Health Organisation (WHO), suggest that India have lost up to \$237 billion, from 2005-2015, due to heart related or Cardiovascular diseases . Thus, feasible and accurate prediction of heart related diseases is very important.

Medical organisations, all around the world, collect data on various health related issues. These data can be exploited using various machine learning techniques to gain useful insights. But the data collected is very massive and, many a times, this data can be very noisy. These datasets, which are too overwhelming for human minds to comprehend, can be easily explored using various

machine learning techniques. Thus, these algorithms have become very useful, in recent times, to predict the presence or absence of heart related diseases accurately.

Proposed System:

Dimensionality Reduction involves selecting a mathematical representation such that one can relate the majority of, but not all, the variance within the given data, thereby including only most significant information. The data considered for a task or a problem, may consist of a lot of attributes or dimensions, but not all of these attributes may equally influence the output. A large number of attributes, or features, may affect the computational complexity and may even lead to overfitting which leads to poor results. Thus, Dimensionality Reduction is a very important step considered while building any model. Dimensionality Reduction is generally achieved by two methods - Feature Extraction and Feature Selection.

Modules:

A. Feature Extraction

B. Feature Selection

C. Naïve Bayes

D. Support Vector Machine

E. K – Nearest Neighbour

F. Decision Tree

G. Random Forest

A. Feature Extraction

In this, a new set of features is derived from the original feature set. Feature extraction involves a transformation of the features. This transformation is often not reversible as few, or maybe many, useful information is lost in the process. In [3] and [4] Principal Component Analysis (PCA) is used for feature extraction. Principal Component Analysis is a popularly used linear transformation algorithm. In the feature space, it finds the directions that maximize variance and finds directions that are mutually orthogonal. It is a global algorithm that gives the best reconstruction.

B. Feature Selection

In this, a subset of original feature set is selected. In [5], key features are selected by CFS (Correlation based Feature Selection) Subset Evaluation combined with Best First Search method to reduce dimensionality. In [6] chi-square statistics test is used to select the most significant features.

c. Naïve Bayes

Naive Bayes is a simple but an effective classification technique which is based on the Bayes Theorem. It assumes independence among predictors, i.e., the attributes or features should be not correlated to one another or should not, in anyway, be related to each other. Even if there is dependency, still all these features or attributes independently contribute to the probability and that is why it is called Naïve.

In [7], Naive Bayes has achieved an accuracy of 84.1584% with the 10 most significant features which are selected using SVM-RFE (Recursive Feature Elimination) and gain ratio algorithms whereas in [8], Naive Bayes has achieved an accuracy of 83.49% when all 13 attributes of the Cleveland dataset [25] are used.

D.Support Vector Machine

Support Vector Machine is an extremely popular supervised machine learning technique(having a pre-defined target variable) which can be used as a classifier as well as a predictor. For classification, it finds a hyper-plane in the feature space that differentiates between the classes. An SVM model represents the training data points as points in the feature space, mapped in such a way that points belonging to separate classes are segregated by a margin as wide as possible. The test data points are then mapped into that same space and are classified based on which side of the margin they fall.

E. K – Nearest Neighbour

In 1951, Hodges et al. introduced a nonparametric technique for pattern classification which is popularly known the K-Nearest Neighbour rule[13]. K-Nearest Neighbour technique is one of the most elementary but very effective classification techniques. It makes no assumptions about the data and is generally be used for classification tasks when there is very less or no prior knowledge about the data distribution. This algorithm involves finding the k nearest data points in the training set to the data point for which a target value is unavailable and assigning the average value of the found data points to it.

In KNN gives an accuracy of 83.16% when the value of k is equal to 9 while using 10-cross validation technique. In KNN with Ant Colony Optimization performs better than other techniques with an accuracy of 70.26% and the error rates is 0.526.Ridhi Saini et al. have obtained a efficiency of 87.5% [15], which is very good.

F. Decision Tree

Decision tree is a of supervised learning algorithm.This technique is mostly used in classification problems. It performs effortlessly withcontinuous and categorical attributes. This algorithm dividesthe population into two or more similar sets based on the most significantpredictors.Decision Treealgorithm, first calculates the entropy of each and every attribute. Then the dataset is split with the help of thevariables or predictors with maximum

information gain or minimum entropy. These two steps are performed recursively with the remaining attributes.

In [10] decision tree has the worst performance with an accuracy of 77.55% but when decision tree is used with boosting technique it performs better with an accuracy of 82.17%. In [9] decision tree performs very poorly with a correctly classified instance percentage of 42.8954% whereas in [16] also uses the same dataset but used the J48 algorithm for implementing Decision Trees and the accuracy thus obtained is 67.7% which is less but still an improvement on the former. Renu Chauhan et al. have obtained an accuracy of 71.43% [17]. M.A. Jabbar et al. have used alternating decision trees with principle component analysis to obtain an accuracy 92.2% [18]. Kamran Farooq et al. have achieved the best results on using decision tree-based classifier combined with forward selection which achieves a weighted accuracy of 78.4604% [19].

G. Random Forest

Random Forest is also a popularly supervised machine learning algorithm. This technique can be used for both regression and classification tasks but generally performs better in classification tasks. As the name suggests, Random Forest technique considers multiple decision trees before giving an output. So, it is basically an ensemble of decision trees. This technique is based on the belief that more number of trees would converge to the right decision. For classification, it uses a voting system and then decides the class whereas in regression it takes the mean of all the outputs of each of the decision trees. It works well with large datasets with high dimensionality.

REQUIREMENT ANALYSIS

The project involved analyzing the design of few applications so as to make the application more users friendly. To do so, it was really

important to keep the navigations from one screen to the other well ordered and at the same time reducing the amount of typing the user needs to do. In order to make the application more accessible, the browser version had to be chosen so that it is compatible with most of the Browsers.

REQUIREMENT SPECIFICATION

Functional Requirements

- Graphical User interface with the User.

Software Requirements

For developing the application the following are the Software Requirements:

1. Python
2. Django
3. Mysql
4. Wampserver

Operating Systems supported

1. Windows 7
2. Windows XP
3. Windows 8

Technologies and Languages used to Develop

1. Python

Debugger and Emulator

- Any Browser (Particularly Chrome)

Hardware Requirements

For developing the application the following are the Hardware Requirements:

- Processor: Pentium IV or higher
- RAM: 256 MB
- Space on Hard Disk: minimum 512MB