Effective heart disease prediction system using data mining techniques

Abstract: The health care industries collect huge amounts of data that contain some hidden information, which is useful for making effective decisions. For providing appropriate results and making effective decisions on data, some advanced data mining techniques are used. In this study, an effective heart disease prediction system (EHDPS) is developed using neural network for predicting the risk level of heart disease. The system uses 15 medical parameters such as age, sex, blood pressure, cholesterol, and obesity for prediction. The EHDPS predicts the likelihood of patients getting heart disease. It enables significant knowledge, eg, relationships between medical factors related to heart disease and patterns, to be established. We have employed the multilayer perceptron neural network with backpropagation as the training algorithm. The obtained results have illustrated that the designed diagnostic system can effectively predict the risk level of heart diseases.

Keywords: data mining, neural network, multilayer perceptron neural network, backpropagation, disease diagnosis

Introduction
Among various life-threatening diseases, heart disease has garnered a great deal of attention in medical research. The diagnosis of heart disease is a challenging task, which can offer automated prediction about the heart condition of patient so that further treatment can be made effective. The diagnosis of heart disease is usually based on signs, symptoms and physical examination of the patient. There are several factors that increase the risk of heart disease, such as smoking habit, body cholesterol level, family history of heart disease, obesity, high blood pressure, and lack of physical exercise.

A major challenge faced by health care organizations, such as hospitals and medical centers, is the provision of quality services at affordable costs. The quality service implies diagnosing patients properly and administering effective treatments. The available heart disease database consists of both numerical and categorical data. Before further processing, cleaning and filtering are applied on these records in order to filter the irrelevant data from the database. The proposed system can determine an exact hidden knowledge, ie, patterns and relationships associated with heart disease from a historical heart disease database. It can also answer the complex queries for diagnosing heart disease; therefore, it can be helpful to health care practitioners to make intelligent clinical decisions. Results showed that the proposed system has its unique potency in realizing the objectives of the defined mining goals.

Methods
The experiment was carried out on a publicly available database for heart disease. The dataset contains a total of 303 records that were divided into two sets, training set...
(40%) and testing set (60%). A data mining tool named Weka 3.6.11 was used for the experiment. Additionally, multilayer perceptron neural network (MLPNN) with backpropagation (BP) was used as the training algorithm.

**MLPNN**
MLPNN is one of the most significant models in artificial neural network. The MLPNN consists of one input layer, one or more hidden layers and one output layer. In MLPNN, the input nodes pass values to the first hidden layer, and then nodes of first hidden layer pass values to the second and so on till producing outputs as shown in Figure 1.

**BP network**
The BP algorithm has served as a useful methodology to train multilayer perceptron for a wide range of applications. The BP network calculates the difference between real and predicted values, which is circulated from output nodes backwards to nodes in previous layer. The BP learning algorithm can be divided into two phases, propagation and weight update.

First, this learning algorithm provides training data to the network and compares the actual and desired outputs. Then, it calculates the error in each neuron. Based on this, the algorithm calculates what output should be for each neuron and how much higher or lower output must be adjusted for desired output and finally adjusts the weights. The overall process is done to improve weights during processing.

**Results and discussion**
In order to predict the probability of patients having heart disease, a confusion matrix (Table 1) was created, where A denotes patients with heart disease, and B denotes patients with no heart disease.

![Figure 1 Multilayer perceptron neural network.](image)

### Table 1 A confusion matrix

|        | A (patients with heart disease) | B (patients with no heart disease) |
|--------|--------------------------------|----------------------------------|
| TP     | A (patients with heart disease) | TP                               |
| FN     | B (patients with no heart disease) | FN                              |

**Abbreviations:** TP, true positive; FN, false negative; FP, false positive; TN, true negative.

### Table 2 Description of 15 used parameters

| S no | Parameters | Parameter description | Values |
|------|------------|-----------------------|--------|
| 1    | age        | Age in years          | Continuous |
| 2    | sex        | Male or female        | 1 = male; 0 = female |
| 3    | thestbpx   | Resting blood pressure | Continuous value in mmHg |
| 4    | cp         | Chest pain type       | 1 = typical type 1; 2 = typical type angina; 3 = non-angina pain; 4 = asymptomatic |
| 5    | chol       | Serum cholesterol     | Continuous value in mm/dL |
| 6    | fbs        | Fasting blood sugar   | 1 ≥ 120 mg/dL; 0 ≤ 120 mg/dL |
| 7    | restecg    | Resting electrographic results | 0 = normal; 1 = having ST-T wave abnormal; 2 = left ventricular hypertrophy |
| 8    | thalach    | Maximum heart rate achieved | Continuous value |
| 9    | old peak   | ST depression induced by exercise relative to rest | Continuous value |
| 10   | exang      | Exercise induced angina | 0 = no; 1 = yes |
| 11   | ca         | Number of major vessels colored by fluoroscopy | 0–3 value |
| 12   | slope      | Slope of the peak exercise ST segment | 1 = unsloping; 2 = flat; 3 = downsloping |
| 13   | thal       | Defect type           | 3 = normal; 6 = fixed; 7 = reversible defect |
| 14   | obes       | Obesity               | 1 = yes; 0 = no |
| 15   | num        | Diagnosis of heart disease | 0%–50%; 1%–50% |

### Table 3 Results for neural network showing 100% accuracy

|        | A (patients with heart disease) | B (patients with no heart disease) |
|--------|--------------------------------|----------------------------------|
| TP     | A (patients with heart disease) | 109 (TP)                          |
| FN     | B (patients with no heart disease) | 0 (FN)                           |
| FP     | A (patients with heart disease) | 0 (FP)                            |
| TN     | B (patients with no heart disease) | 73 (TN)                          |

**Abbreviations:** TP, true positive; FN, false negative; FP, false positive; TN, true negative.
A confusion matrix contains information about real and predicted classifications done by a classification system. The data in the matrix are evaluated to know the performance of such systems.

The confusion matrix contains the following four entries:
- TP (true positive): The number of records classified as true while they were actually true.
- FP (false positive): The number of records classified as true while they were actually false.
- FN (false negative): The number of records classified as false while they were actually true.
- TN (true negative): The number of records classified as false while they were actually false.

The overall process of effective heart disease prediction system (EHDPS) is based on the following three steps:
1. Data collection
2. Data pre-processing and
3. The classification of data.

The data are collected from a standard dataset that contains 303 records. The 15 parameters, such as age, sex, chest pain type (CP), and cholesterol (chol), with some domain values associated with them, considered to predict the probability of heart disease are shown in Table 2.

The collected data were used to create a structured database system. The pre-processing was done by identifying the associated fields and removing all the duplications. After that, all the missing values were filled, and the data were coded according to the domain value.

After applying neural networks on training dataset, the results show that there are zero FN or FP entries (Table 3), suggesting that the system predicts heart disease with 100% accuracy. Figure 2 shows the actual work done by Weka 3.6.11 tool.

**Conclusion**

In this study, an EHDPS has been presented using data mining techniques. From ANN, an MLPNN together with BP algorithm is used to develop the system. The MLPNN model proves the better results and assists the domain experts and even the person related to the medical field to plan for a better and early diagnosis for the patient. This system performs realistically well even without retraining. Furthermore, the experimental results show that the system predicts heart disease with $\sim 100\%$ accuracy by using neural networks.

**Disclosure**

The authors report no conflicts of interest in this work.

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