

Enhanced Pre-Processing Based Cardiac Valve Block Detection Using Deep Learning Architectures

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Article History	Abstract
Received: 22 January 2022 Revised: 14 April 2022 Accepted: 19 May 2022	Nowadays, in non-infective diseases heart disease is marking more lethality rate. Many people in all over the world are suffering from the heart disease. Because of no prior knowledge in predicting the disease, many people are getting affected by heart disease. Most of the heart diseases are caused due to the smoking habit and imbalanced diet. So this paper proposes a novel method improved genetic neural network algorithm (IGNN) to measure blood pressure, cholesterol level, and blood sugar level and to identify the smoking habit, blood circulation rate in heart. Then on basis of the parameters measured, to preprocess the heart image to identify the block in heart valve using Standard Linear Discrimination Analysis (SLDA) this can transform the attributes with a Gaussian distribution. Keywords: heart disease, improved genetic neural network algorithm (IGNN), Standard Linear Discrimination Analysis (SLDA), Gaussian distribution
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1 Introduction:

The heart is one of the principal organs of the human body. It siphons blood box the veins of the circulatory framework. The circulatory framework is critical in light of the fact that it transports blood, oxygen and different materials to the various organs of the body [1]. Heart assumes the most critical part in circulatory framework. Heart illnesses or cardiovascular infections (CVD) are a class of sicknesses that include the heart and veins. Cardiovascular illness incorporates coronary conduit sicknesses (computer aided design) like angina and myocardial dead tissue (ordinarily known as a respiratory failure) [2]. Generally, blood pressure monitoring, blood test for identifying cholesterol and ECG were considered as the main choices for checking cardiovascular soundness of an individual. Deep learning calculations, specifically convolutional networks, have quickly turned into a procedure of decision for examining clinical pictures [3]. Brief outlines are given of studies per application

region: neuro, retinal, pneumonic, advanced pathology, bosom, and heart, abdominal, musculoskeletal [4].

2 Literature Review:

Diagnosis of cardio vascular disease is complicated and important task that need to be executed accurately and effectively. This section shows some of the survey based on cardio vascular disease. Author in [5] propose, Another method in light of factual hypothesis to identify heart status: typical versus unusual in uproarious climate. A choice in regards to the different heart states is spread the word (preparing) and obscure (test) highlights variety with the assistance of a classifier. Work in [6] presented neuro-fluffy coordinated framework for coronary illness. Work [7] investigated each boundary connected with cardiovascular harmless imaging modalities. The picture handling angles have been managed as per the need of the cardiologist. Creator in [8] propose the utilization of help vector bunching to recognize on the off chance that a patient presents a coronary illness or not. To diminish the quantity of passings, it is vital to think about advance notice signs, and knowing how to answer rapidly and appropriately when it happens. Work [9, 10] presents another technique for discovery, depiction and solo order of PCG sounds. First, the PCG signal was preprocessed and then for detection and delineation of the PCG genuine sounds incidence locations, two windows were slid on the pre processed PCG signal.

3 Research methodology:

Most of the heart diseases are caused due to the smoking habit and imbalanced diet. So this paper proposes a novel method improved genetic neural network algorithm (IGNN) to measure blood pressure, cholesterol level, and blood sugar level and to identify the smoking habit, blood circulation rate in heart. Then on basis of the parameters measured, to preprocess the heart image to identify the block in heart valve using Standard Linear Discrimination Analysis (SLDA) this can transform the attributes with a Gaussian distribution. The proposed architecture is as follows:

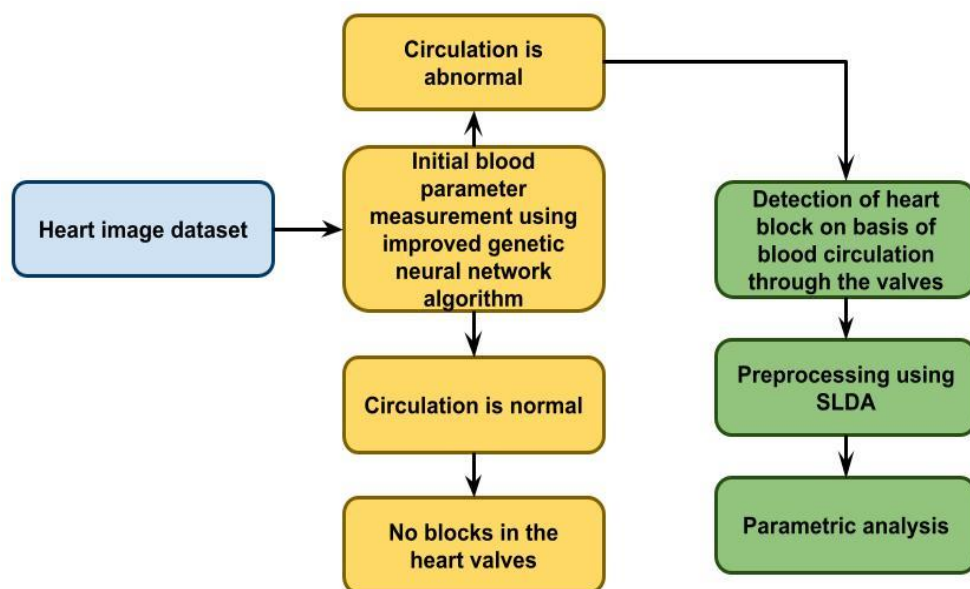


Fig-1 proposed architecture

3.1 Improved genetic neural network algorithm (IGNN):

IGNN design impersonates the educational experience of human cerebrum. The essential engineering of IGNN includes interconnected neurons, which are characterized in three particular classifications: input layer neurons, yield layer neurons and secret layer neurons as displayed in Fig. 1. The input data

are introduced through input layer neurons, and the reaction of the info information is presented at yield layer neurons. Neurons are associated by scalar capabilities known as weights that participate in the educational experience of organizations. In back proliferation algorithm, which is generally utilized in preparing of IGNNs, a progression of information and result information is introduced to the framework. Each secret layer neuron and result layer neuron process this information by duplicating its comparing loads, and utilizing an exchange capability. The S-molded sigmoidal bend, which is generally utilized as the exchange capability is given as eq. (1):

$$f(x) = 1/(1 + e^{-x}) \quad (1)$$

The learning of the organization is brought out through changing the loads by constant cycles and limiting the blunder between tentatively estimated reaction and IGNN model anticipated reaction. IGNNs, for example, the three-layer back proliferation network have been demonstrated to be widespread capability approximations. IGNNs have proactively been applied to tackle and foresee different issues in natural frameworks. The back propagation network is the most common administered IGNN learning model, which utilizes the gradient decent calculation to address the loads between interconnected neurons.

- a. Brain networks by and large comprise of five parts:
- b. A coordinated diagram known as the organization geography whose hubs address the neurodes (called handling components) and whose circular segment address associations.
- c. A state variable related with each neurode.
- d. A truly esteemed weight related with every association.
- e. A genuinely esteemed inclination related with each neurodes.

The condition of each neurode $f[\sum_{i=1}^n [w_{ix_i} - \beta]]$, being the exchange capability, b being the inclination of the neurode, w_i are the loads on the approaching associations, x_i are the conditions of the neurodes on the opposite finish of associations.

Numerically, a three-layer brain network with I input hubs, j stowed away hubs and k result hubs is communicated as eq. (2)

$$O_{pk} = f_1 \left[\sum_{j=1}^L \left[W_{jk} \cdot f_2 \left[\sum_{i=1}^n [W_{ij} \cdot x_{pi}] \right] \right] \right] \quad (2)$$

Where O_{pk} is the result from the k th hub of the result layer of the organization for the p th vector (data of interest); x_{pi} are the contributions to the organization for the p th vector (data of interest); W_{jk} is the association weight between j th hub of the secret layer and k th hub of the result layer; W_{ij} is the association weight between i th hub of the info layer and j th hub of stowed away layer; f_1 and f_2 are enactment capabilities.

They require the accompanying five parts:

1. An approach to coding answers for the issue on chromosomes.
2. An assessment capability, which returns a rating for every chromosome given to it.
3. An approach to instating the number of inhabitants in chromosomes.
4. Administrators that might be applied to guardians when they repeat to modify their hereditary structure; standard administrators are transformation and hybrid.
5. Boundary settings for the calculation, the administrators, etc.

3.2 Standard Straight Segregation Examination (SLDA):

The objective of the SLDA method is to project the first information framework onto a lower layered space. To accomplish this objective, three stages should have been performed. The initial step is to work out the detachability between various classes (for example the distance between the method for various classes), which is known as the between-class fluctuation or between-class lattice. The subsequent step is to work out the distance between the mean and the examples of each class, which is known as the inside class difference or inside class lattice. The third step is to build the lower layered space which expands the between-class difference and limits the withinclass change. This part will make sense of these three stages exhaustively, and afterward the full depiction of the SLDA calculation will be given.

To make sense of how the between-class fluctuation or the between-class framework (SB) can be determined, the accompanying presumptions are made. Given the first information lattice $X = \{x_1, x_2, \dots, x_N\}$, where x_i represents the i th test, example, or perception and N is the all out number of tests. Each example is addressed by M elements ($x_i \in \mathbb{R}^M$). At the end of the day, each example is addressed as a point in M -layered space. Expect the information network is parceled into $c = 3$ classes as follows, $X = [\omega_1, \omega_2, \omega_3]$ (step (A)). Each class has five examples (for example $n_1 = n_2 = n_3 = 5$), where n_i represents the quantity of tests of the i th class. The all out number of tests (N) is determined as eq. (3),

$$N = \sum_{i=1}^3 n_i \quad (3)$$

The means of discriminant investigation are as per the following:

1. Recognizable proof of factors that best separates among the different gatherings.
2. Utilization of distinguished factors to foster a condition or a capability for processing another variable or record that will stingily address the distinctions between different gatherings.
3. Utilization of discriminant capability to arrange future perceptions into any of the pre-characterized gatherings.

3.3 Performance Analysis:

The parameter to be obtained is blood pressure, cholesterol level, blood sugar level measurement, blood circulation rate in heart, heart beat rate using improved genetic neural network algorithm (IGNN) for the analysis of block in heart valve.

Table-1 Parameter comparison for various heart image dataset

Parameters	VD_IM	PCG_GS	EP_CVBD_DL
Accuracy	91	93	95
Precision	81	83	85
Recall	77	79	82
F1_Score	55	59	62
RMSE	51	52	56
MAP	41	43	44

The above table-1 shows comparative analysis between proposed and existing techniques in terms of accuracy, precision, recall, F_1 score, RMSE, MAP. Here the analysis has been carried out based on number of epochs. Accuracy calculation is done by the generalprediction capability of projected DL method. For calculating F-score, number of images processed are EEG signal for both existing and proposed technique. The F-score reveals each feature ability to discriminate independently from other features. For the first feature, a score is generated, and for the second feature, a different score is obtained. However, it saysnothing about how the two elements work together. Here, calculating the F-

score using exploitation has determined the prediction performance. It is created by looking at the harmonic component of recall and precision. If the calculated score is 1, it is considered excellent, whereas a score of 0 indicates poor performance. The actual negative rate is not taken into consideration by F-measures. The accuracy of a class is calculated by dividing the total items classified as belonging to positive class by number of true positives. Probability that a classification function will produce a true positive rate when present. It is also known by the acronym TP amount. In this context, recall is described as ratio of total number of components that genuinely fall into a positive class to several true positives. How well a method can recognise Positive samples is calculated by recall. Recall increases as more positive samples are determined. When training regression or time series models, RMSE is one of the most widely used metrics to gauge how accurately our forecasting model predicts values compared to real or observed values. MSE squared root is used to calculate RMSE. The RMSE calculates the change in each pixel as a result of processing.

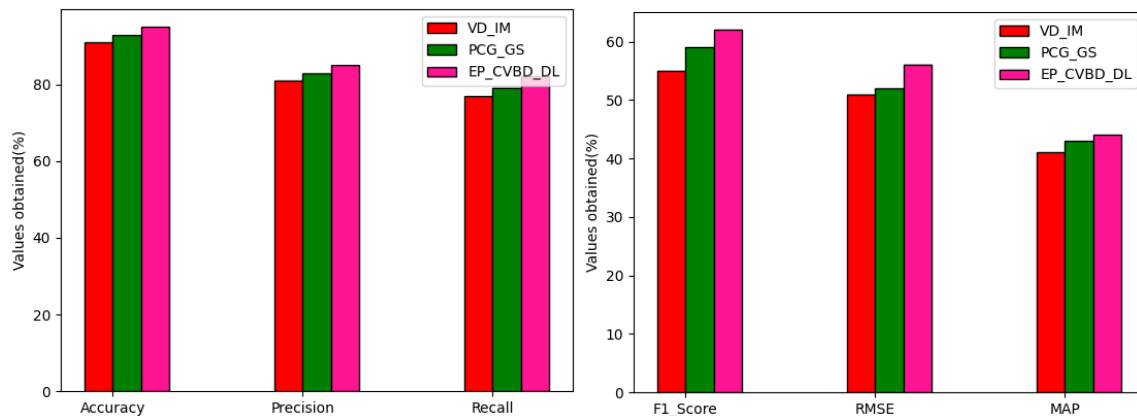


Figure 2: Comparison of proposed and existing methods

From above figure 2 shows comparative analysis between proposed and existing technique. the proposed technique attained accuracy of 95%, precision of 85%, recall of 82%, F-1 score of 62%, RMSE of 56%, MAP of 44%.

4 Conclusion:

This research propose novel method improved genetic neural network algorithm (IGNN) to measure blood pressure, cholesterol level, and blood sugar level and to identify the smoking habit, blood circulation rate in heart. identify the block in heart valve using Standard Linear Discrimination Analysis (SLDA) this can transform the attributes with a Gaussian distribution. The experimental analysis has been carried out in terms of accuracy, precision, recall, F_1 score, RMSE, MAP. the proposed technique attained accuracy of 95%, precision of 85%, recall of 82%, F-1 score of 62%, RMSE of 56%, MAP of 44%.

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