

## An Effectual Machine Learning Based Coronary Artery Disease Classification for Low Error Rates

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### Abstract:

Heart syndrome is the common and significant reason for death in the world nowadays. Estimation of cardiovascular infection is a critical experiment in the region of clinical data breakdown. Machine knowledge has been presented to be operative in support of making conclusions and estimations from the large number of records created by healthcare engineering. The prediction model is familiarized with diverse groupings of structures and several well-known classification practices. The proposed approach deals with the efficient machine learning model for the detection of the CAD which are having low validation and testing errors and achieves high true positive error rates. Our proposed model consists of hybridization of optimization processes using PSO and firefly nature-inspired and the classification is performed on the data using discriminant analysis. The proposed approach is achieving above 95% accuracy of the detection on different test samples to achieve high-performance classifications.

**Keywords:** Artificial Intelligence, Coronary artery diseases, Classification, instance selections

### 1. Introduction

It is very complex to recognize heart problems because of various risk aspects such as diabetes, high B.P, heavy weight, irregular pulse rate, and countless other factors. Different practices in data mining are used and various classifications have been engaged to find out the solutions to heart disease in humans. The severity of the illness is categorized on various procedures such as decision trees and many other hierarchical structures [1][2]. The environment of heart disease is difficult and hence, the infection must be handled sensibly. Not undertaking much heart or cause impulsive deaths. The perception of medical knowledge and data mining are recycled for determining various categories of metabolic disorders. Data mining with a grouping approach plays a significant part in the estimation of heart problems and data explorations. Most of the studies are done in predicting the accuracy of the specific functions related to heart syndrome. Various procedures have been recycled for understanding the heart-related data methods of data mining to diagnose efficiently and effectually [3][5].

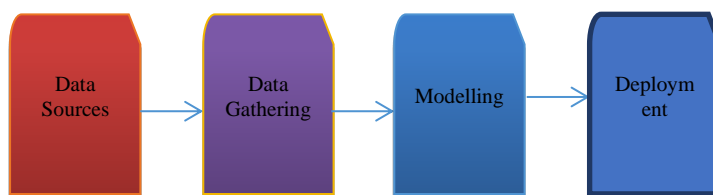


Fig 1: Data Mining Process [4]

The rest paper is divided into various phases. Section II shows the related work and review of the valid researches done. Section III shows the problem occurring in the current scenario. Section IV deals with the proposed workflow algorithm discussions. Section V shows the proposed result and discussion and the last section covers the conclusion and future scope of the proposed work.

### 2. Related Work

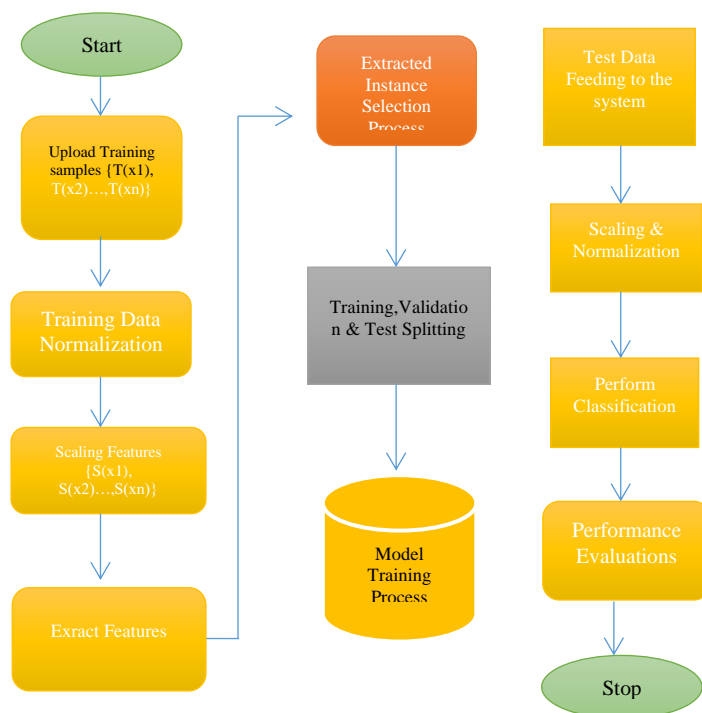
Heart problem these days is a very serious problem which needs better for the proper diagnosis. This section deals with the various research works which put light on the various previous researches done in the medical field. **Ilyaraja M, Meyyappan T** [6] worked on the predictive modeling to get the risk level heart disease patient. They have worked on the itemsets based on the support value. They have implemented their research in JAVA and achieved better accuracy. **Kaan Uyara et al.** [7] proposed an efficient genetic algorithm based on the fuzzy logic for heart problem diagnosis where they have achieved 97.78% accuracy using the Cleveland dataset based on heart disease. **Rashmi G Sabojiet al.** [8] proposed the diagnosis process of heart disease using Random Forest classification and they have achieved approx. 98% accuracy. **Mollet, Nico et al.** [9] proposed a HOG procedure that is more efficient in appropriate for the exploration and ordering of heart diseases than the traditional Histogram of Oriented Gradient. The overall arrangement has been evaluated through intensive tests using various classifiers.

**Kumar, Gandhi, et al. [10]** presented a computer-aided system for the recognition of heart disease by using IoT tools. They have used various IoT protocols for the real-time data and classify the diseases. Their analysis has worked on various significant parameters such as asymmetry, texture, analysis for regularization, and feature steps. The mined feature limitations are used to categorize the information. **Safdar, Saima, SaadZafar et al. [11]** proposed an evaluation for the arrangement to get the relationship of syndrome and non-disease through the classifiers. They collected information and perform data mining techniques for the handling of the data. They have performed a feature extraction process and classify the labels for the prediction using binarization. **Manogaran, R. Varatharajan et al. [12]** presented an innovative process for the recognition of heart infection. To sense the number of noises from records, the pre-processing phase is carried out by using filters. And consequently, the neuro-fuzzy methodology is executed to fill in the mysterious data. Fuzzy inference approaches are used to describe the information for the grouping of the infection. The process was estimated on a real-time dataset with efficient accuracy which is compared with the other same methods. **Ramalingam, Ayantan et al. [13]** performed analysis on several machines learning processes and classifiers. They have studied various current practices used for the categorizations and discussed the various difficulties used for the evaluations of their proposed method and implementation efficiently. **Kannan, Vasanthi, et al. [14]** presented their research on the predictive modeling using ROC curve using training and validation process. Their results are evaluated using true positive rates in terms of specificity and false negative which they have further compared with the previous studies.

### 3. Problem Statement

Classification and regularizations are certain tasks in the detection of coronary heart diseases. So it plays a significant role in the detection of disease data points in medical applications. Physical detection and examination of the diseased parts are a time-consuming and unpredictable task, and as the number of data points or values increases; the performance evaluation becomes very tough. However, it is comparatively time-consuming when the execution of the complex test cases becomes quite large and needs to be applied recurrently with different orientations. In many applied cases the simple thresholding is used but very less work is done in the optimization and classifications. The main focus of this research work is to enhance the accuracy and acceptance rate as well as to reduce the rejection rate of the detection. Lastly, the performance of the proposed system will be calculated to evaluate the performance and reliability of our proposed work.

### 4. Proposed Work



**Fig 2: Proposed Flow Diagram**

The proposed model is the following:-

#### 1. Dataset

The proposed work uses the well-known dataset named as Z-AlizadehSanithrough which the performance and classification are performed. This dataset involves 303 patient proceedings and 56 features or columns. , 216

individuals with CAD, and 88 individuals with the normal position. The structures used in the information gatherings are divided into several clusters i.e. demographic, indicator and investigation, ECG, and workshop and topographies. This dataset is generally used in the several latest types of research having a large number of characteristics as compared to various other CAD data through which we came to know that this dataset is more informatics. The motive for consuming this dataset is that it contains features that put heavy weights and having high correlations within the attributes of the datasets. [19].

## 2. Data Normalization & Scaling

After uploading the data, normalization of the data is a crucial part. The values in the data are alphanumeric which needs to be normalized and scale for the less variance and standard deviations. The scaling intends to standardize the information. The strings are converted into the numeric form so that the processing will be performed in the vector form instead of the alphanumeric which is the significant part of the implementation. If this is not properly done then the training set will achieve overfitting and underfitting of the model which is not good for our proposed work.

## 3. Feature Extraction

This performs the feature vector extraction process using PCA which is one of the significant processes to find the eigenvalues and eigenvectors by using the covariance process. In the proposed work the feature engineering is used so the relevant feature vector is attained to identify the patterns using correlations which is transformed into a new vector representation. Most of the feature algorithms are having one main problem which is the computation time and execution processing complexity which is overcome by PCA and it will also reduce the nonlinearity among the data in the N-dimensional space. So in the proposed approach, the linear kernel PCA is used which will reduce the non-linearity and reduces the variance among the data points to extract the highly correlated features which are then transformed to the feature vector and are the significant information for the feature vector.

## 4. Feature Optimization

This section deals with the hybrid optimization process of two nature-inspired algorithms that perform the instance selection process which is the significant part of the implementation to select the relevant features to form the extracted feature vector which is the eigenvector arranged in the multi-dimension vector. The explanation of the algorithm is given below.

**a. Firefly-based modified instance selection**-This is inspired by the blinking fireflies. Various norms are implemented for these algorithms. They help solve the nonlinear, dynamic, anomaly problems. The main steps covered in the proposed work for the instance selections

- 1) Particles are fascinated by each other.
- 2) Magnetism is relative to the glow. Less optimistic is fascinated by the brighter particle or instance.
- 3) If the intensity is the same for both instances, then there will be a random movement.
- 4) New best solutions are produced using random walk.

There are various applications for Firefly algorithms

FA is applied in nonlinear problems, dynamic problems, feature selection, fault detection many more.

There are numerous advantages of FA over other optimization algorithms.

- 1) Automatic subdivision of the whole population into subgroups
- 2) Natural capability of dealing with multi-model optimization.
- 3) High randomness in the solutions.

Start

Determine the unbiased utility for the instance selection as an input.

such that:  $\{x_n\} = (x_{n\{1\}}, x_{n\{2\}}, \dots, x_{n\{d\}})$

Determine the instance as fireflies inhabitants such that  $\{p(i)\} \rightarrow (i = 1, 2, \dots, n)$ .

Estimate the intensity of the instance which is linked with  $f(x_n)$  i.e.  $\text{Int}\{\text{val}\} = f(x_n)$  where Int is the intensity value.

Define absorption capacity  $\gamma$

```

While (t(pn) < M(gn) i.e. max iterations
forxp = 1 : F(xn)
for: yp = 1 : F(xn)
    IF ({I{xp} > I{yp}})
Diverge A(x) i.e. fascination with dist ∈ Distan(xn{i})
Move F(xn) from xp → yp ;
EvaluateS(x{n}) and modifyInt{val}
end if
end for yp
end for xp
Perform best possible solutions and rank instances.
end While
Stop

```

**b. Particle Swarm based modified instance selection** -It is a metaheuristic process that cracks hard optimization complications based on complex computations. This procedure is encouraged by the communal flocking birds. It is a population-based searching process. Each instance is called a swarm particle. Each instance in the swarm has a rate of movement. They act as optimal function to control structure. It is effectively practical to the heart disease classification and showing the efficient outcome. In metaheuristic processes, the objective function is fed to minimize or maximize as per the requirement which is needed to be optimized. The proposed steps are given below.

```

For each occurrence as swarm input x{p} = 1, ..., S{p} do
Formulate the P{p} as position of the instance by distribution trajectory: xp(i) → U(L{b}, U{b})
Arrange the x{p} instance known location to its initial location: s{i} ← xp{i}
    If func(p{i}) < func(g):
modify the swarm best known loc: g{b} ← s{i}
Fixed the instance rate in a velocity vector: SV{i} ~ U(-|U{b} - L{b}|, |U{b} - L{b}|)
while a determined quantity does not occur to do:
    For each instance x{p} = 1, ..., S{p} do
        For each dimension d{s} = 1, ..., Nd{p} do
            Check random records: rnd{p}, r{g} ~ U{0,1}
Perform rate of speed updations of the each instance in the swarm:
SV(i, ds) ← ω (SV{i}, ds) + φ{p} rnd{p} (xp(i, ds) - x(i, ds)) + φ (g{b}) r (g{b}) (g{b} - x(i, ds))
            Modify the position of the instance in the vector: x{i} ← x{i} + LR (SV{i})
if f(x{i}) < f(p{i}) then
                Modify the instance best known locus: pi ← xpi
            iffunc(pi) < func(g) then
                Adjust the instance best-known pos: gb ← pi
        End For
    End For

```

### 5. Discriminant Analysis and Classifying using k-fold cross-validation

In the proposed approach we have performed statistical analysis to evaluate the linear grouping of features that describes or splits two or more objects based on the classification process. It will act as a linear classifier which reduces the non-linearity in the training of the model and is used to perform high true positive based predictive modeling. It is also used to analyze the variance in the data because if the data perform high variance then it will increase the deviations in the data which will result in improper classifications. In our proposed approach discriminant analysis is performed on the independent variables in the continuous form and the categorical dependent variable as prediction labels.

### 5. Proposed Algorithm

**Step 1:** Input Records such that  $t\{s\} = t\{s1\}, t\{s2\} \dots t\{n\}$  as data and execute the framing of the data to process information efficiently.

**Step 2:** Standardize & data scaling

For  $x=1$  to  $\text{len}(t\{s\})$

$S_D = \text{StdScaling} \{ t\{s\} \}$  to reduce the variances among data points.

EndFor

Where  $t$  is the total training samples

**Step 3:** Implement the extraction of the features & perform covariance data in the vector form

For  $p=1$  to  $S_D$

$Cv(p) = \text{COV}(S_D)$

EndFor

**Step 4:** Eigenvalues and vectors extractions for the transformation  $T\{x\} = X \times W$  for the information for new space vector generation.

Where  $VE(v)$  where  $VE = \{VE_1, VE_2 \dots VE_N\}$  is processed vector.

**Step 5:** Perform instance selections using hybrid optimization procedures for the selection of relevant instances using the feature reduction method

While  $(t(pn) < M(gn))$  i.e. max iterations

IF  $(\{I\{xp\} > I\{yp\}\})$

Diverge  $A(x)$  i.e. fascination with  $\text{dist} \in \text{Distnace}(x_n \{i\})$

Move  $F(x_n)$  from  $x_p \rightarrow y_p$  ;

Evaluate  $S(x\{n\})$  and modify  $\text{Int}\{val\}$

Perform best possible solutions and rank instances.

end While

For each occurrence as swarm input  $x\{p\} = 1, \dots, S\{p\}$  do

Formulate the  $P\{p\}$  as position of the instance by distribution trajectory:  $x_p(i) \rightarrow U(L\{b\}, U\{b\})$

Arrange the  $x\{p\}$  instance known location to its initial location:  $s\{i\} \leftarrow x_p\{i\}$

If  $\text{func}(p\{i\}) < \text{func}\{g\}$ :

while a determined quantity does not occur to do:

For each instance  $x\{p\} = 1, \dots, S\{p\}$  do

For each dimension  $d\{s\} = 1, \dots, Nd\{p\}$  do

Check random records:  $\text{rnd}\{p\}, r\{g\} \sim U\{0,1\}$

Perform rate of speed updations of the each instance in the swarm:

$SV(i, ds) \leftarrow \omega (SV\{i\}, ds) + \varphi\{p\} \text{rnd}\{p\} (x_p(i, ds) - x(i, ds)) + \varphi(g\{b\}) r(g\{b\}) (g\{b\} - x(i, ds))$

Modify the position of the instance in the vector:  $x\{i\} \leftarrow x\{i\} + LR (SV\{i\})$

```

if f(x{i}) < f(p{i}) then
    Modify the instance best known locus: pi ← xpi
    if func(pi) < func(g) then
        Adjust the instance best-known pos: gb ← pi
    Endif
endif

End For

EndFor
    
```

**Step 6:** Prepare data for training and testing phase

$N_D = \{T_x(N_D)\}$

70% training data and 30% for the test data.

**Step 7:** Perform Discriminant analysis classification

$L\{m\} = \{FitTransform[TR(N_D)]\}$

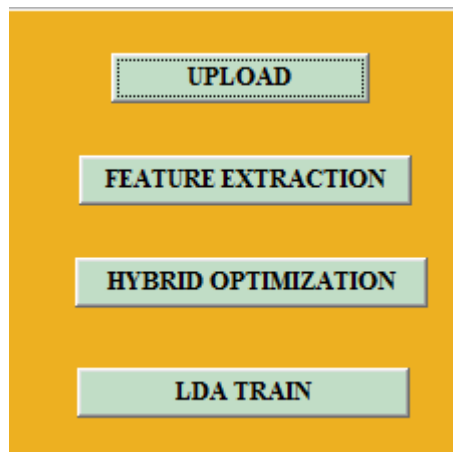
Where fit\_transform generate the configuration of the training model and TR(x) is the training samples  
**Step 8:** Upload Test data such that  $TD_S = \{TD_{S1}, TD_{S2}, TD_{S3}, TD_{S4} \dots TD_{SN}\}$

**Step 9:** Training model loading and Implement classification on  $T_{SN}$ .

**Step 10:** Evaluate Performance Evaluations and Repeat Step 5 to 9 until all configurations get completed.

### 6. Result And Discussions

This section covers the implementation part of the proposed model which is implemented in the MATLAB environment. No external library is used in the proposed system. A detailed explanation is given below.



**Fig 3: User Panel**

Fig 3 shows the proposed panel shows the user interface consists of the graphical user interface tools which deal with the user interactions with the machines. The panel consists of the static texts, pushbuttons through which the user clicks on the pushbuttons and get the output which perform specific functions.



**A ROBUST SUPERVISED SYSTEM USING MACHINE LEARNING TECHNIQUES:  
CLASSIFICATION AND PROGNOSIS IN HEALTH DOMAIN**

53	Male	28.3878	0	1	1	0	0	Y	110	80	N	15.6	N
67	Female	28.3987	0	1	0	0	0	Y	140	80	N	13.9	N
54	Male	28.0773	0	0	1	0	0	N	100	100	N	13.5	mid
66	Female	28.8386	0	1	0	0	0	Y	100	80	N	12.1	Severe
50	Female	27.1652	0	1	0	0	0	Y	110	80	Y	13.2	Severe
50	Male	24.4898	0	0	1	0	0	N	118	70	N	15.6	N
55	Male	28.3848	0	0	0	1	0	Y	110	80	Y	14.1	mid
72	Male	28.1224	1	0	1	0	0	Y	130	70	N	16.1	mid
58	Female	31.6158	0	0	0	0	0	Y	90	50	N	11.6	N
60	Male	24.5675	1	0	0	0	0	N	130	70	N	13.9	N
58	Male	28.5731	0	1	0	1	0	Y	170	70	N	14.5	mid
80	Female	28.6215	0	1	0	0	0	Y	140	100	N	10	mid
70	Female	30.7004	1	1	0	0	0	Y	140	84	Y	12.3	Moderate
67	Female	28.5643	1	1	0	0	0	Y	150	74	N	14.3	mid
66	Female	28.2227	1	1	0	0	0	Y	130	80	N	12.9	mid
59	Male	28.0437	1	0	0	0	0	Y	120	70	N	13.3	Moderate
41	Male	23.8887	0	0	1	0	0	Y	130	80	Y	11.4	Severe
68	Female	22.7815	0	0	0	0	0	N	115	70	N	11	N

Fig 4: Training Panel

Fig 4 shows the training panel which consists of the list boxes in which the uploaded data is displayed from excel and the characteristics and nature of the data taken. It shows that the individual properties on which the processing will be performed. The data shown on the panel is limited but the back data which is processed is of a total of 56 attributes for the one individual which will give more insights into the data.

<b>ENTROPY</b>	<b>VARIANCE</b>	<b>STD</b>	<b>MEAN</b>
1.9728	1.0027	1.0013	0.032528

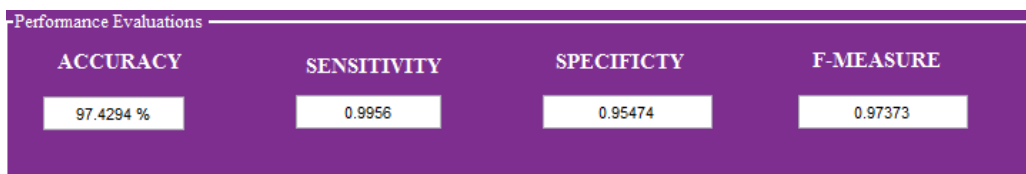
Fig 5: Extracted Features

Fig 5 shows the extracted features from the feature vector. From fig 6 the processing of the data can be controlled and it will give proper information of the feature values with the other data values. The entropy is considered as the significant parameter which shows the disorder among the data points. Variance, the standard deviation will tell us the measure of dispersion and spreading of the data, and mean shows us the tendency of data points to be centered for the groupings. It gives us important information about the total population concerning the observed values.

Normal
Normal
Normal
Normal
Normal
Normal
Normal
Normal
CAD
CAD
CAD
CAD
CAD
CAD
CAD
CAD
CAD
CAD

Fig 6: Classified Outcomes

Fig 6 shows the classified outcome as a result of which it can be distinguished that how many individuals are having CAD and how many are in a normal state. It is classified using discriminant analysis where the trained data model is loaded which is processed on the unknown data which is considered as the test data from our side. In the proposed model the 70% of the total training data and 30% of the test data are taken to check the validity and performance of the proposed system.



**Fig 7: Performance Analysis**

Fig 7 shows the performance evaluation of the proposed work. It can be noticed that the proposed supervised learning model is achieving high performance in terms of true positive and negative rates. It shows that the recognition accuracy is approx. 98% which is the desired outcome and also the sensitivity and specificity is increasing which increases the true positive and true negative rate of the proposed model. Also, the F-measure should be high which indicates that the proposed approach is retrieving efficient relevant information based on the training model which shows increases the precision and recall of the information retrieval process. The performance is evaluated using the following equations.

$$P(x) = tp(x) \div (tp(x) + fp(x))$$

$$R(x) = tp(x) \div (tp(x) + fn(x))$$

$$Sp(x) = tn(x) \div (tn(x) + fp(x))$$

$$Sn(x) = tp(x) \div (tp(x) + fn(x))$$

$$A(x) = tp(x) + fp(x) \div (tp(x) + fn(x) + fp(x) + tn(x))$$

$$Fm(x) = 2 \times \left( \frac{P(x) \times R(x)}{P(x) + R(x)} \right)$$

Where P(x) and R(x) is the evaluated precision and recall of the proposed model. Sp(x) and Sn(x) is the sensitivity and specificity of the proposed model. A(x) and Fm(x) is the evaluated accuracy and F-measure of the proposed model.

**Table 1: Accuracy Performance**

Test No.	Accuracy (%)
1	97.429
2	97.319
3	98.185
4	98.401
5	96.297
6	96.310
7	96.739
8	97.185
9	98.006
10	97.071

**Table 2: Sensitivity Performance**

Test No.	Sensitivity
1	0.972
2	0.979
3	0.985
4	0.981
5	0.962
6	0.968
7	0.969
8	0.974
9	0.986
10	0.978

**Table 3: Specificity Performance**

Test No.	Specificity
1	0.954
2	0.961



3	0.958
4	0.955
5	0.957
6	0.960
7	0.951
8	0.958
9	0.959
10	0.961

**Table 4: F-Measure Performance**

Test No.	F-Measure
1	0.973
2	0.956
3	0.978
4	0.981
5	0.976
6	0.976
7	0.962
8	0.978
9	0.974
10	0.968

Table 1, 2, 3, 4 shows the performance analysis on different test samples and can check the variation in different test samples. It can be seen that there are not that many variations after applying different samples of the test which shows that our proposed approach is having the highly precise evaluation of the system to achieve high accuracy in terms of true positive and negative rate and low classification error rates.

## 7. Conclusion & Future Scope

The health care data should be monitored timely and accurately which will help doctors to diagnose patients efficiently. Still, the accuracies and precisions on the data are low and not upto the mark. So an efficient precise model is required which can give full and better insights among the data to gather information and characteristics about the patient's health. This paper put light on the robust predictive modeling using feature extraction and instance selection hybridization which gives betterment in the evaluation of the proposed model using discriminant analysis. It can be noticed from the proposed work performance that the developed predictive modeling is achieving satisfactory results with low false positive and negative rates which are the desired outputs and will help doctors with the effectual diagnosis of the CAD or normal state of the patient.

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