

Assessment of Heart Rate and Risk Analysis of Cardiac Disorders using Supervised Learning

Presented by:

Sanatan Shrivastava: 2018KUCP1096

Yash Singh: 2018KUCP1088

Shaktiraj Daudra: 2018KUCP1092

Ajay Sharma: 2018KUCP1087

Rahul Singh Tanwar: 2018KUCP1043



Presented to:
Dr. Ankit Sharma
Faculty Coordinator,
Pattern Recognition (CST403)
Indian Institute of Information Technology
(IIIT), Kota.

Table of contents

01. Introduction

02. Dataset

03. ML Model, Classifier

04. Algorithms Used

05. Comparative Analysis

06. Conclusion

Problem Statement; Goals; Introduction to the topic.

Features, Dataset and some plots

Model to extract features for further classification

4 implemented algorithms, based on their utility.

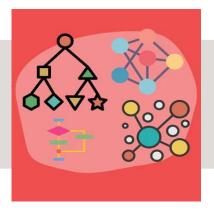
Algorithmic comparison based on their accuracy.

Conclusion of the project; further discussions.

Project goals









Goal 1

Picked Features like sex, age etc from the Dataset.

Goal 2

Pre-Process Features values.

Goal 3

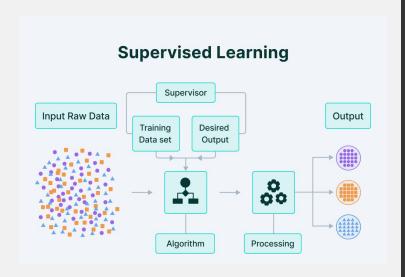
Perform algorithmic analysis using few selected algorithms

Goal 4

Predict values on the basis of algorithm with maximum accuracy; draw conclusion

Introduction

- ★ We have conducted an extensive study to find the solution to a classification problem of heart risk analysis and whether there is a probability of cardiac disorder.
- ★ We have used a variety of Machine Learning algorithms.
- ★ This is a classification problem, with input features as a variety of parameters, and the target variable as a binary variable, predicting whether heart disease is present or not.
- ★ The dataset used here is originally available on the University of California, Irvine (UCI) website.
- ★ Dataset used (Originally, at UCI) https://www.kaggle.com/ronitf/heart-disease-uci



Dataset

dataset.shape \Rightarrow (303, 14)

Features:

- age (AGE): age
- 2. **sex (SEX):** 1: male, 0: female
- 3. **cp:** chest pain type, 1: typical angina, 2: atypical angina, 3: non-anginal pain, 4: asymptomatic
- 4. **trestbps:** resting blood pressure
- 5. **chol:** serum cholesterol in mg/dl
- 6. **fbs:** fasting blood sugar > 120 mg/dl
- 7. **restecg:** resting electrocardiographic results (values 0,1,2)
- 8. thalach: maximum heart rate achieved
- 9. **exang:** exercise induced angina
- 10. **oldpeak:** oldpeak = ST depression induced by exercise relative to rest
- 11. **slope:** the slope of the peak exercise ST segment
- 12. **ca:** number of major vessels (0-3) colored by fluoroscopy
- 13. **thal:** thal: 3 = normal; 6 = fixed defect; 7 = reversible defect

Some snapshot of Dataset-

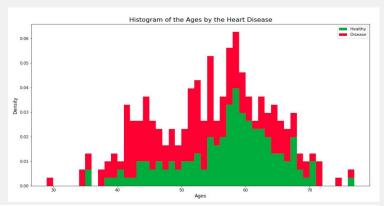
dataset.head(5)

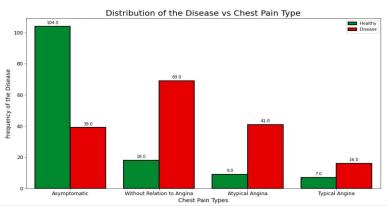
	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	thal	target
0	63	1	3	145	233	1	0	150	0	2.3	0	0	1	1
1	37	1	2	130	250	0	1	187	0	3.5	0	0	2	1
2	41	0	1	130	204	0	0	172	0	1.4	2	0	2	1
3	56	1	1	120	236	0	1	178	0	8.0	2	0	2	1
4	57	0	0	120	354	0	1	163	1	0.6	2	0	2	1

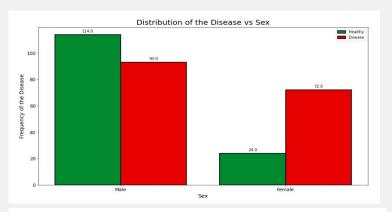
dataset["target"].unique() ⇒ array([1, 0])

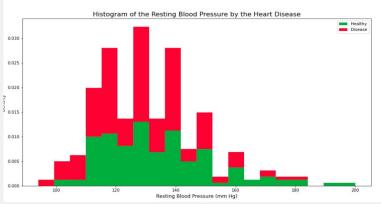
Clearly, this is a classification problem, with the target variable having values '0' and '1'

Parametric Analysis:



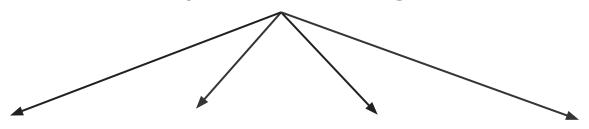






Algorithms used to process features:

Assessment of Heart Rate and Risk Analysis of Cardiac Disorders using Supervised Learning



K-Nearest Neighbor Support Vector Machine

Naive Bayes

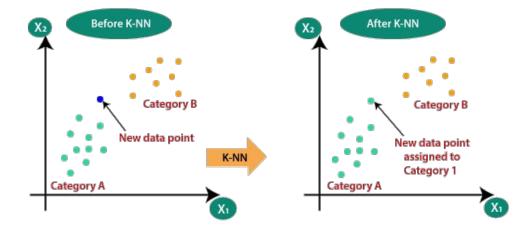
Random Forest

K Nearest Neighbor (KNN)

K-Nearest Neighbour is Machine Learning algorithm based on Supervised Learning technique.

Why to use it?

- K-NN algorithm assumes the similarity between the new data and available data and put the new data point into the category that is most similar to the available categories.
- 2. K-NN algorithm stores all the available data and classifies a new data point based on the similarity. This means when new data appears then it can be easily classified into a well suite category by using K- NN algorithm.



67.21%

Accuracy achieved by K-Nearest Neighbor (KNN) Algorithm.

Model Fitting:

```
from sklearn.neighbors import
KNeighborsClassifier
knn = KNeighborsClassifier(n_neighbors=7)
knn.fit(X_train,Y_train)

Y_pred_knn=knn.predict(X_test)
score_knn = round(accuracy_score(Y_pred_knn,Y_test)*100,2)

print("The accuracy score achieved using KNN is:
"+str(score knn)+" %")
```

Splitting of data:

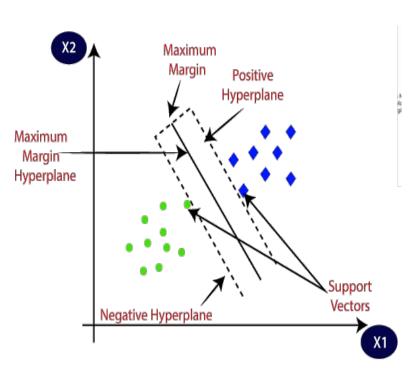
```
from sklearn.model_selection import
train_test_split

predictors = dataset.drop("target",axis=1)
target = dataset["target"]

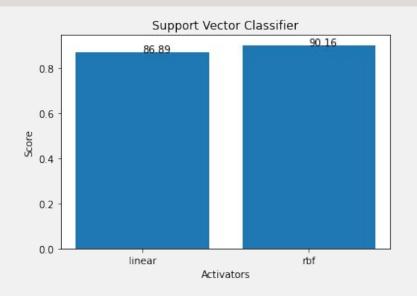
X_train, X_test, Y_train, Y_test =
train_test_split(predictors, target, test_si
ze=0.20, random_state=0)
```

Support Vector Machine (SVM)

- Used primarily for classification problems.
- Creates a decision boundary that can segregate n-dimensional space into classes.
- Power to reduce the *curse of dimensionality*.
- Hyperplane is chosen in a way to maximize the margin.
- The data points or vectors that are the closest to the hyperplane are known as Support Vectors.



SVM Results



```
kernels = ['linear', 'rbf']
svc = []
for i in range(len(kernels)):
    SVclassifier = SVC(kernel = kernels[i])
    SVclassifier.fit(X_train, Y_train)
    svc.append(SVclassifier.score(X_test, Y_test))
```

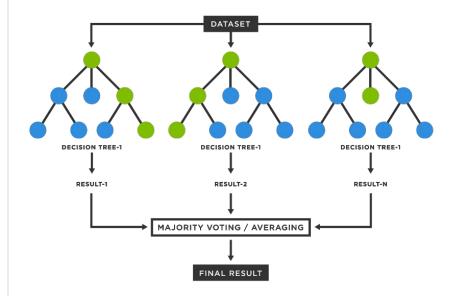
86.89%
Accuracy achieved by Linear Function

90.16%

Accuracy achieved by 'rbf' - Radial Basis Function

Random Forest Algorithm

- In terms of accuracy, it is unrivalled among contemporary algorithms.
- On huge databases, it performs well. Handling 10,000 inputs at a time.
- As the forest grows, it generates an internal unbiased estimate of the generalisation error.
- It offers a method for guessing missing data that works well and retains accuracy even when a high percentage of the data is missing.
- It includes techniques for balancing error in uneven data sets with a class population.
- The forests that are created can be preserved and used on other data in the future.
- Unlabeled data can be used to develop unsupervised clustering, data views, and outlier identification skills.



Code for Random Forest Algorithm

```
max accuracy = 0
for x in range (2000):
    rf =
RandomForestClassifier(random statex)
    rf.fit(X train, Y train)
    Y pred rf = rf.predict(X test)
    current accuracy =
round(accuracy score(Y pred rf,Y test*100,2)
    if (current accuracy) max accuracy):
        max accuracy = current accuracy
        best x = x
rf =
RandomForestClassifier(random statebest x)
rf.fit(X train, Y train)
Y pred rf = rf.predict(X test)
```

95.08%

Accuracy achieved by Random Forest.

Naive Bayes Algorithm

- Naive Bayes is a method to predict the probability of output based on various attributes.
- This algorithm is mostly used in classification and with problems having multiple classes.

Why to use it? Advantages are:

- It is easy and fast to predict the class of the test data set.
- It also performs well in multi-class prediction.
- When assumption of independence holds, a Naive Bayes classifier performs

better compared to other models like logistic regression and less training data is needed.

85.25%

Accuracy achieved by Naives Bayes Algorithm.

$$P(A|B) = \frac{P(B|A) P(A)}{P(B)}$$

Model fitting

```
from sklearn.naive_bayes import GaussianNB
nb = GaussianNB()
nb.fit(X_train,Y_train)
Y_pred_nb = nb.predict(X_test)
```

Conclusion

Algorithm	Accuracy				
K-Nearest Neighbor	67.21%				
SVM	90.16%				
Naive Bayes	85.25%				
Random Forest	95.08%				

Thanks!

We'll be glad to answer any questions.

We are thankful for your guidance and your understanding.

