NITTE MEENAKSHI INSTITUTE OF TECHNOLOGY

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APPROVED BY AICTE & GOVT.OF KARNATAKA



LA PROPOSAL Heart Disease Prediction Using Machine Learning Algorithms

Submitted in partial fulfilment of the requirement for the LA Component Introduction to Machine Learning Course (18CSE71) of 7th Semester

Bachelor of Engineering

in

Computer Science and Engineering

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ABSTRACT

Heart disease is one of the most significant causes of mortality in today's world. Prediction of cardiovascular disease is a critical challenge in the area of clinical data analysis, with the current epidemic scenario doctors need a support system for more accurate prediction of heart disease. Machine learning algorithm opens new door opportunities for precise prediction of heart diseases. Diagnosis and prediction of heart related diseases requires more precision, perfection and correctness because a little mistake can cause havoc or death of the person. To deal with this problem there is an essential need of prediction system for awareness about diseases. Machine learning is the branch of Artificial Intelligence, it provides prestigious support in predicting any kind of event which take training from natural events. Our objective is to calculate accuracy of various machine learning algorithms for predicting heart disease, algorithms used are k-nearest neighbor, decision tree, logistic regression, random forest, Naïve bayers and support vector machine (SVM). We are the existing dataset from the Cleveland database of UCI repository of heart disease patients.

INTRODUCTION

Machine learning is one of the most rapidly evolving domains of artificial intelligence. Machine learning algorithms can analyze huge data from various fields, one such important field is the medical field. Its primary focus is to design systems, allow them to learn and make predictions based on the experience. It trains machine learning algorithms using a training dataset to create a model. The model uses the new input data to predict heart disease. Using machine learning, it detects hidden patterns in the input dataset to build models. It makes accurate predictions for new datasets. The dataset is cleaned and missing values are filled. The model uses the new input data to predict heart disease and then tested for accuracy.

The machine learning methods that will be used to predict the heart disease are Logistic regression, Naïve bayers, Support vector machine (SVM), K-NN, decision tree and random forest.

DATASET

Characteristics of the dataset:

Data Set Characteristics	Multivariate
Attribute Characteristics	Categorical, Integer, Real
Associated Tasks	Classification
Number of Instances	303
Number of Attributes	75
Missing Values	Yes
Area	Life

This database contains 76 attributes, but all published experiments refer to using a subset of 14 of them. In particular, the Cleveland database is the only one that has been used by ML researchers to this date. The "goal" field refers to the presence of heart disease in the patient. It is integer valued from 0 (no presence) to 4. Experiments with the Cleveland database have concentrated on simply

attempting to distinguish presence (values 1,2,3,4) from absence (value 0). The names and social security numbers of the patients were recently removed from the database, replaced with dummy values.

One file has been "processed", that one containing the Cleveland database. All four unprocessed files also exist in this directory. To see Test Costs (donated by Peter Turney), please see the folder "Costs".

Attribute Information:

Only 14 attributes used:

- 1. #3 (age)
- 2. #4 (sex)
- 3. #9 (cp)
- 4. #10 (trestbps)
- 5. #12 (chol)
- 6. #16 (fbs)
- 7. #19 (restecg)
- 8. #32 (thalach)
- 9. #38 (exang)
- 10. #40 (oldpeak)
- 11. #41 (slope)
- 12. #44 (ca)
- 13. #51 (thal)
- 14. #58 (num) (the predicted attribute)

Complete attribute documentation:

- 1. id: patient identification number
- 2. ccf: social security number (I replaced this with a dummy value of 0)
- 3. age: age in years
- 4. sex: sex (1 = male; 0 = female)
- 5. painloc: chest pain location (1 = substernal; 0 = otherwise)
- 6. painexer (1 = provoked by exertion; 0 = otherwise)
- 7. relrest (1 = relieved after rest; 0 = otherwise)
- 8. pncaden (sum of 5, 6, and 7)
- 9. cp: chest pain type
 - Value 1: typical angina
 - Value 2: atypical angina
 - Value 3: non-anginal pain
 - Value 4: asymptomatic
- 10. trestbps: resting blood pressure (in mm Hg on admission to the hospital)
- 11. htn
- 12. chol: serum cholestoral in mg/dl
- 13. smoke: I believe this is 1 = yes; 0 = no (is or is not a smoker)
- 14. cigs (cigarettes per day)
- 15. years (number of years as a smoker)
- 16. fbs: (fasting blood sugar > 120 mg/dl) (1 = true; 0 = false)

- 17. dm (1 = history of diabetes; 0 = no such history)
- 18. family history of coronary artery disease (1 = yes; 0 = no)
- 19. restecg: resting electrocardiographic results
 - -- Value 0: normal
 - -- Value 1: having ST-T wave abnormality (T wave inversions and/or ST elevation or depression of > 0.05 mV)
 - -- Value 2: showing probable or definite left ventricular hypertrophy by Estes' criteria
- 20. ekgmo (month of exercise ECG reading)
- 21. ekgday(day of exercise ECG reading)
- 22. ekgyr (year of exercise ECG reading)
- 23. dig (digitalis used furing exercise ECG: 1 = yes; 0 = no)
- 24. prop (Beta blocker used during exercise ECG: 1 = yes; 0 = no)
- 25. nitr (nitrates used during exercise ECG: 1 = yes; 0 = no)
- 26. pro (calcium channel blocker used during exercise ECG: 1 = yes; 0 = no)
- 27. diuretic (diuretic used used during exercise ECG: 1 = yes; 0 = no)
- 28. proto: exercise protocol
 - 1 = Bruce
 - 2 = Kottus
 - 3 = McHenry
 - 4 =fast Balke
 - 5 = Balke
 - 6 = Noughton
 - 7 = bike 150 kpa min/min (Not sure if "kpa min/min" is what was written!)
 - 8 = bike 125 kpa min/min
 - 9 = bike 100 kpa min/min
 - 10 = bike 75 kpa min/min
 - 11 = bike 50 kpa min/min
 - 12 = arm ergometer
- 29. thaldur: duration of exercise test in minutes
- 30. thaltime: time when ST measure depression was noted
- 31. met: mets achieved
- 32. thalach: maximum heart rate achieved
- 33. thalrest: resting heart rate
- 34. tpeakbps: peak exercise blood pressure (first of 2 parts)
- 35. tpeakbpd: peak exercise blood pressure (second of 2 parts)
- 36. dummy
- 37. trestbpd: resting blood pressure
- 38. exang: exercise induced angina (1 = yes; 0 = no)
- 39. xhypo: (1 = yes; 0 = no)
- 40. oldpeak = ST depression induced by exercise relative to rest
- 41. slope: the slope of the peak exercise ST segment

- -- Value 1: upsloping
- -- Value 2: flat
- -- Value 3: downsloping
- 42. rldv5: height at rest
- 43. rldv5e: height at peak exercise
- 44. ca: number of major vessels (0-3) colored by flourosopy
- 45. restckm: irrelevant
- 46. exerckm: irrelevant
- 47. restef: rest raidonuclid (sp?) ejection fraction
- 48. restwm: rest wall (sp?) motion abnormality
 - 0 = none
 - 1 = mild or moderate
 - 2 = moderate or severe
 - 3 = akinesis or dyskmem (sp?)
- 49. exeref: exercise radinalid (sp?) ejection fraction
- 50. exerwm: exercise wall (sp?) motion
- 51. thal: 3 = normal; 6 = fixed defect; 7 = reversable defect
- 52. thalsev: not used
- 53. thalpul: not used
- 54. earlobe: not used
- 55. cmo: month of cardiac cath (sp?) (perhaps "call")
- 56. cday: day of cardiac cath (sp?)
- 57. cyr: year of cardiac cath (sp?)
- 58. num: diagnosis of heart disease (angiographic disease status)
 - -- Value 0: < 50% diameter narrowing
 - -- Value 1: > 50% diameter narrowing

(in any major vessel: attributes 59 through 68 are vessels)

- 59. lmt
- 60. ladprox
- 61. laddist
- 62. diag
- 63. cxmain
- 64. ramus
- 65. om1
- 66. om2
- 67. rcaprox
- 68. rcadist
- 69. lvx1: not used
- 70. lvx2: not used
- 71. lyx3: not used

72. lvx4: not used

73. lyf: not used

74. cathef: not used

75. junk: not used

76. name: last name of patient (I replaced this with the dummy string "name").

The best 14 attributes are chosen for the Prediction.

Source:

Creators:

- 1. Hungarian Institute of Cardiology. Budapest: Andras Janosi, M.D.
- 2. University Hospital, Zurich, Switzerland: William Steinbrunn, M.D.
- 3. University Hospital, Basel, Switzerland: Matthias Pfisterer, M.D.
- 4. V.A. Medical Center, Long Beach and Cleveland Clinic Foundation: Robert Detrano, M.D., Ph.D.

Further EDA and pre-processing needs to be carried out on the dataset to understand and evaluate it.

Machine Learning Methods

- 1. Logistic Regression (Scikit-learn)
- 2. Naive Bayes (Scikit-learn)
- 3. Support Vector Machine (Linear) (Scikit-learn)
- 4. K-Nearest Neighbours (Scikit-learn)
- 5. Decision Tree (Scikit-learn)
- 6. Random Forest (Scikit-learn)
- 7. XGBoost (Scikit-learn)
- 8. Artificial Neural Network with 1 Hidden layer (Keras)

The end result would be to compare the accuracy of these models to see how correctly they have classified the diagnosis of heart disease.

Assessment

Accuracy of the algorithms depends on four values namely true positive(TP), false positive(FP), true negative(TN) and false negative(FN).

Accuracy= (FN+TP) / (TP+FP+TN+FN)

The numerical value of TP. FP. TN. FN defines as:

TP= Number of persons with heart diseases

TN= Number of persons with heart diseases and no heart

diseases

FP= Number of persons with no heart diseases

FN= Number of persons with no heart diseases and with heart diseases

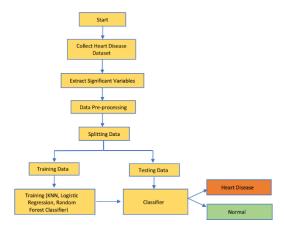
Presentation and Visualization

A graphical representation to compare the relationship between age ranges and cardiovascular disease.

Further Catplot can be used to visualize categorical values like chest pain type CP, sex, fasting blood sugar FBS and target .

Whereas distplot can be used to visualize distributive variables like age, chol – cholestol level, trestbps – resting blood sugar, thalach – maximm heart rate etc, oldpeak.

Design of the proposed Model



Roles

Exploratory Data Analysis EDA and pre-processing will be done by the team together Individual roles: The machine learning algorithms will be shared equally by the team members Samiksha Rajbhushan Ullal: Logistic Regression, Support Vector Machine, Decision Tree, XGBoost

Karan R: Naïve Bayes, K- Nearest Neighbour, Random Forest, ANN with 1 Hidden layer Comparison of accuracy metrics would be done together

Schedule

Date	Task to be completed
20/12/21	LA proposal
25/12/21	EDA and data preprocessing
05/01/22	Predicting using various machine learning models
07/01/22	Presentation
15/01/22	Report submission

Bibliography

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