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E-mail :
editor.ijpast@gmail.com
editor@ijpast.in

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Examining the Accuracy of Machine Learning Disease Prediction Algorithms

Dr. V Venkata Ramana¹, Dr. V Lokeswara Reddy², Dr K Sreenivasa Rao³, Dr.M Ramanjeneya Reddy⁴

Abstract

Machine learning algorithms may take large, structured, unstructured, and complicated information and utilize a wide range of quantitative, probabilistic, and efficient methods to learn from the data and find relevant instances. In this way, data mining and machine learning aid in the diagnosis and treatment of a wide range of medical conditions. If the research employs a variety of effective machine learning algorithms, predictive analysis may help doctors better forecast the course of a disease and tailor therapy to the individual needs of each patient. We can now develop accurate and timely forecasts regarding the spread of illness using machine learning techniques. The purpose of this study is to evaluate and contrast many supervised machine learning methods for illness prediction using pathological data. Finally, we collect data on illness symptoms and show the findings to compare the performance of different machine learning methods.

Introduction

Each illness has its own unique set of symptoms, making early diagnosis and treatment challenging. Each symptom is given a certain weight in order to predict the illness's presence and aid in the diagnostic process, and the symptom with the highest influence on the condition is given the most attention. When it

comes to making choices about illness prediction, medical practitioners may benefit from data mining's supplemental and replacement information. In areas that typical statistical approaches have difficulty processing, our procedure obtains Hidden and undiscovered patterns, relationships, and information.

^{1,2,3,4}Professor,

Department of CSE, K.S.R.M College of Engineering(A), Kadapa

Accuracy and partiality in drawing conclusions are always what contribute to successful therapy. Data mining concepts are flexible enough to uncover hidden numbers, relationships within a database, and machine learning approaches to further analyze a patient's case based on the documented clinical information. Machine learning algorithms can forecast many different illnesses and help doctors create effective procedures to improve patients' health while minimizing healthcare costs, recuperation time, and the possibility of treatment errors. We are happy to see that machine learning algorithms are effective in illness prediction, and there are still numerous avenues to investigate. Machine learning is an enormous method for data analysis that incrementally improves with the help of learning algorithms and publicly available data.

In this research, we suggest a graphical user interface (GUI) program that makes use of machine learning methods for a disease prediction system, mines information on said illness's symptoms, and then detects said disease by contrasting the many methods now in use for doing so. Naive Bayes, Decision Trees, Support Vector Machines, and Random Forest are just some of the machine learning categorization techniques that may help you get the answers you need with each new input. Examine the data, both numerical and graphical, that reveals the universal

trend throughout all training programs. We also use a graphical user interface program to determine whether a patient will be afflicted with a disease. By immediately collecting suitable information and components identified with a different disease, distinguishing essential characteristics that could assist us with verifying the normal examples in selection of the model and parameters, the proposed framework aims to determine the issue of precise prediction of disease for a patient. Briefly, the presentation is broken down as follows: first, we discuss fundamental machine learning tactics for illness prediction; second, we analyze the research design, data, and methods used. Then, we get into the conclusions we draw at the very end.

Related Work

The proposed optimization hybrid approach method by Youngest Khourdifi [1] increased the predictive accuracy of medical data sets. These methods on comparison with the supervised procedures depend upon existing datasets of classification accuracy estimation which are utilized to assess the performance. KNN (99.65%) and RF (99.6%) are the precision score achieved by the proposed model using FCBF, PSO and ACO. H.Benjamin Fredrick David [2] proposed an algorithm which gave maximum precision with the classification carried of a typical and anomalous individual. From UCI machine learning repository heart disease dataset is used in his methodology for the assessment of the

performance of the algorithms. The conclusion is 8% exactness traced by the calculations using Random forest algorithm when contrasted with different calculations for coronary disease expectation. Meghan Shah [3] recommended data mining techniques for Heart Disease Prediction System. In the proposed strategy WEKA programming is utilized for programmed determination of sickness by providing characteristics of administrations in healthcare place. They used various methods like Support Vector Machine, Naive Bayesian, KNN, Association rule, ANN, and Decision Tree. They suggested Support Vector Machine is reasonable providing dense exactness as differentiated and various data mining procedures. A few investigations are led to clinical datasets utilizing numerous classifiers and features selection as suggested by M. Fatima [5]. The proposed method uses the classification of the heart disease dataset where good classification accuracy is traced. The hybrid effective algorithm is suggested by Malay [6] which is used to predict coronary disease. The popular clustering algorithm K means and ANN are used to extort anonymous information about coronary disease. An accuracy of 97% is given by the proposed methods. In this, a hybrid approach that includes merging various procedures like FCBF (Fast Correlation-Based Feature Selection) strategy to filter redundant features so as to improve the nature of heart disease classification.

Proposed Work

Methodology

The fundamental commitments of this work are to extract the classified precision valuable for predicting various diseases, Although Machine learning models have been broadly contemplated and seen as extraordinarily fruitful, the disease prediction is a confounded issue and there are as yet numerous upgrades to be made and techniques to investigate. We assemble predictive models and look at them utilizing explicit execution measurements including Accuracy, Precision, Recall and F-score, recognizing the best ones that could be utilized for foreseeing different diseases. Programmed algorithms are used by machine learning to find out and upgrade the activities by breaking down info information to make forecasts inside a worthy choice. In concern with the recent information, these techniques will in general, build progressively exact expectations.

Dataset

Datasets from the effectively accessible repositories can be utilized for training the machine. Data pre-preparing or Data cleaning is one the significant angle to be completed before executing machine learning calculations for mining purposes. The dataset used in the proposed system contains 4988 rows and 133 columns. This dataset is downloaded from an open-source (<https://github.com/feat7/symptom-to-disease-prediction/tree/master/data/clean>). The training dataset has 4920 rows and 133 columns and the testing dataset has 68 row and 133 columns. Some of the attributes in the training dataset are shivering, skin rash, acidity, itching, obesity, mild fever, high fever, knee pain, puss-filled pimples, puffy face and eyes, blurred and distorted vision, phlegm, throat

irritation, redness of eyes, sinus pressure, runny nose, congestion, chest pain, weakness in limbs, fast heart rate, pain during bowel movements, pain in the anal region, bloody stool, red spots over body, belly pain, abnormal menstruation, diachronic patches, watering from eyes, increased appetite, polyuria, skin peeling, silver-like dusting, small dents in nails, inflammatory nails, blister, red soar around nose, yellow crust ooze, itching, headache, cough, high fever, fatigue, weight loss.

Architecture of the proposed system

In the proposed system, a relative study and performance evaluation between four different Machine learning classification algorithms is made using performance metrics such as accuracy, precision, recall and F-score. In this work, the various machine learning algorithms like used are Naïve Bayesian, Decision tree method, Random Forest method and Support Vector Machine method for calculating the accuracy in predicting disease. In machine learning, the most repeatedly used technique is classification. The disease prediction is carried out by the technique which is depicted in Figure1 which decided the investigation approach for developing a classification model necessary for the forecast of the patient disease. In order to make forecasts, a classifier should be prepared with the reports and then produce a classification model which is taken care of with another obscure report and the anticipation is made. The setup of this exploration incorporates the Performance Evaluation of the four algorithms used for classification.

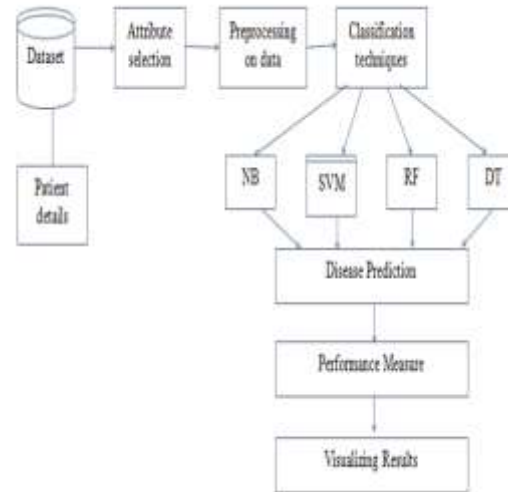


Figure1. Architecture of the proposed system

Data preprocessing

Data preprocessing is a very important innovate mining method particularly in health data. Data collection strategies square measure typically loosely controlled, leading to out of vary, inapplicable information mixtures, missing values etc. Data preprocessing cannot be done carefully screened for disease prediction which gives results that mislead the diagnosis. Data arrangement and sifting steps will took a lot of handling session. Data preprocessing incorporates cleaning, Instance choice, standardization, change, include extraction and determination. The result of information preprocessing is the last training set.

Classification techniques

The dataset is spitted into two parts in terms of percentage, 80% of the dataset is training dataset and the remaining 20% of the dataset is test dataset. In this work at the training phase, a classification model is built on the dataset which is trained by the four

classification methods like Naïve Bays, Decision Tree, Support Vector Machine, and Random Forest used in the training phase of the proposed method. Each of the four methods is depicted below.

Naïve Bays Method

The most important aspects of machine learning are classification and prediction where the world brimming with AI and machine learning consciousness encompassing, nearly everything around. Naïve Bays is a basic yet surprisingly incredible algorithm for predictive examination. It is a classification strategy dependent on Bays the hypothesis with suspicion of freedom among predictors. It involves two sections which are Naïve and Bays, in straightforward terms In the Naïve Bayesian method, the classifier will accept when the nearness of a precise feature of a class is random with presence of another attribute. Regardless of whether these features rely upon one another or upon the presence of different features these properties autonomously add to the probability that is the reason named after seeing that Naïve.

For huge datasets especially we can use Naïve Bays model which is easy to build. In probability theory and measurements, Bays hypothesis which is then again known as the Bays law or the Bays rule depicts the likelihood of an event dependent on earlier information on the conditions that may be identified with the event. Bays theorem is a technique to build out the conditional probability. By considering the prior knowledge of a condition that relates to the event, the condition probability of the event occurrence can be calculated by using Bays theorem.

The conditional likelihood is the likelihood of an event happening given that it has some relationship to at least one different event. Bays hypothesis is marginally more nuanced more or less; it gives the real likelihood of an event. For the given information Naïve Bays method is defined as given a speculation H and the proof E, Bayesian theorem expresses the connection involving the likelihood of the theory before getting the proof P (H) and the likelihood of the speculation in the wake of getting the proof is given by

$$P(H|E) = \frac{P(E|H) \cdot P(H)}{P(E)}$$

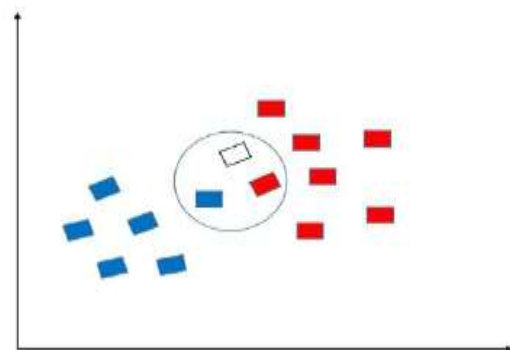


Figure2. Illustration of Naïve Bays method

This relates the probability of theory before cutting the proof which is P (H) to the likelihood of the speculation in the wake of getting the proof which is P (H|E) thus P (H) is known as the prior probability while P (H|E) is known as the posterior probability and the factor P (E|H) that narrate both is known as the likelihood ratio. Figure 2 shows the delineation of the Naïve Bays method. Now using this term Bays theorem specifies differently as the procedure

probability equals the prior probability times the likelihood ratio.

Support Vector Machine method

The various applications of SVM that are generally used with it are detecting face, normal text in hypertext classification, image classification and bio-information. The support vector machine is specific to supervised learning machine learning model learns from the past input data and makes future predictions as output. SVM is a method that looks at data and sorts into one of two categories. In the larger picture of the machine learning model and under supervised learning we can see that the support vector fits in under classification deciding what yes-and-no is and there is also a regression version but it is primarily used for classification. SVM exists in both linear and non-linear forms. There are two data sets like train dataset and test dataset which is involved in SVM.

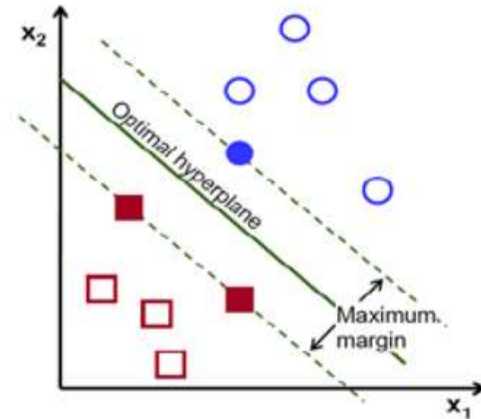
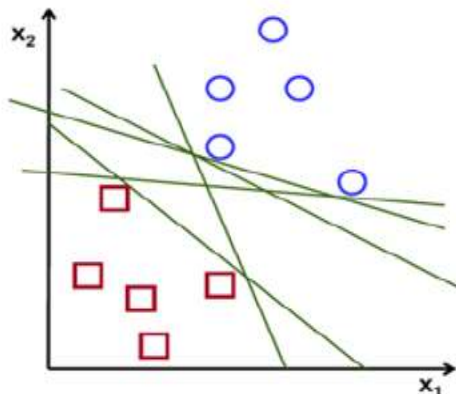


Figure3. Illustration of Support vector machine

In general, if two classes are there, it splits the two classes perfectly which is called linearly separable. On the other side, a set of lines can be used to split the dataset; among them one separating line is chosen as the best line. The ideal line is selected in such a way that the distance should be maximum to the closest purposes of the two classes in the training dataset. The separation between the support vector and the hyper plane ought to be beyond what many would consider possible and this is the place the support vectors are the outrageous focuses in the dataset.

Random Forest method

Random forest algorithm which is a popular supervised machine learning technique used in regression and classification. It is called random forest because the forest has trees and a tree in the machine learning world means a decision tree. The random forest it constructs is based on an ensemble of decision trees using the bagging method. This bagging method combines various learning methods

to increase the overall results. The random forest is built on various decision trees and combines all to form accurate and stable predictions.

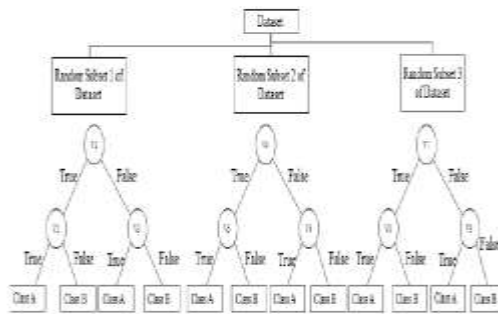


Figure3. Illustration of Random forest method

Random forests are a blend of tree indicators utilizing a decision tree to such an extent that each tree relies upon the estimations of an irregular vector examined autonomously and with a comparative conveyance for all trees in the forest. The theory mistake of a forest of tree classifiers relies on the idea of the individual trees in the forest and the association between's them. They are continuously enthusiastic regarding commotion. It is a supervised classification algorithm utilized for the desire and it is considered as the ideal because of its enormous number of trees in the forest giving improved precision than decision trees. Normally, the trees are trained autonomously and the desires for the trees are merged through averaging. Random forest algorithm can utilize both for classification and the regression dependent to the difficult space.

Initially, the k features are removed from absolute m features. Figure 4 shows the illustration of Random forest method. In the following phase, every tree haphazardly chooses k features so as to discover the

root node by utilizing the best split loom. The following phase on the disease dataset includes computing the daughter nodes utilizing a similar best split methodology. Likewise, the tree is shaped from the top hub, that is the root and until all the leaf hubs are produced from the features. A random forest is created by using the tree formed by randomly which is being used for making disease prediction.

Decision Tree Method

A decision tree is simple and used as one of the machine learning implementations of classification methods, which are represented in the form of a hierarchical structure. The decision tree algorithms can handle both numerical and categorical which is a supervised learning algorithm. In a decision tree, the data items are being mapped based on certain predictive conclusions. The classification in the decision tree categorizes the input data into an outcome. To clearly specify the accuracy of the algorithm is directly depended upon the features it selects for training model. Figure 5 shows the illustration of Decision tree method. The nodes of the decision tree contain different levels where the highest node is represented as the root node. Nodes which are having at least one child, all internal nodes signify tests on input factors or properties. Contingent upon the test result, the classification algorithm branches towards the fitting child node where the procedure of test and fanning rehashes until it arrives at the leaf hub. The leaf or terminal hubs compares with choice results.

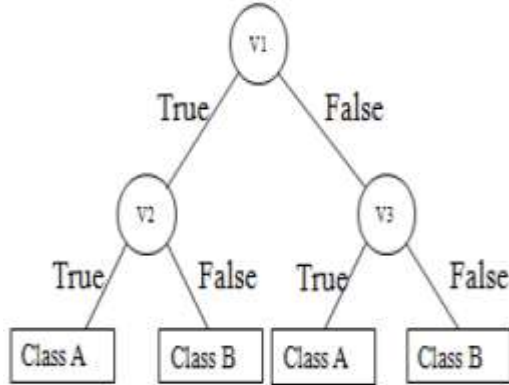


Figure5. Illustration of Decision tree method
Evaluating classification algorithms
performance metrics

In this paper, the various performance metrics used to assess and compare classification algorithms are Accuracy, Precision, Recall and F-score. These are measured on the basis of a two by two matrix called Confusion matrix. This confusion matrix holds the models predicted values to the actual class values. All the measures are focused on the values present in the four quadrants of the Confusion matrix.

Table1. Confusion matrix

Class-Actual values	Class-Predicted values	
	P	N
P	TP	FN
N	FP	TN

In the above table 1 the terminology is given as

- Observation positive is given by P
- Observation negative is given by N

Positive observation and negatively predicted is given by TP

Accuracy

Accuracy or success rate is defined as the proportion of correctly classified test instances.

$$\text{Accuracy} = \frac{TP + TN}{TP + TN + FP + FN}$$

Precision

Precision or positive predicted value is the proportion of the total number of correctly classified positives with the total number of predicted positive instances.

$$\text{Precision} = \frac{TP}{TP + FP}$$

Recall

Recall or True positive rate is characterized as the proportion of the absolute number of accurately grouped positive cases to the all out number of positive cases. High Recall demonstrates effectively perceived (for example few FN).

Results and Discussions

Figure 6 shows the output with a GUI for a disease dataset. From the dataset the input symptoms are chosen as blackheads, puffy_face_and_eyes, puss_filled_pimples and irritability. The predicted disease for the given symptoms by all the classification algorithms like Support vector machine, Random Forest, Naïve Bays and Decision tree is Acne.



Figure 6: Disease predictor-Acne

Figure 7 shows the results for the attributes like Accuracy, Precision, Recall and F-score for the classification algorithms Support vector machine, Random Forest, Naïve Bays and Decision tree.

The results obtained after running the code are compared and analyzed. Decision Tree algorithm gives an accuracy of 0.9527, Random Forest gives an accuracy of 0.9504, Naive Bayesian algorithm gives an accuracy of 0.9693 and Support vector machine algorithm gives an accuracy of 0.9504 .

From the above results it is clear that Naïve Bays algorithm gives the highest accuracy with the given training and testing datasets when compared to other algorithms like Support vector machine, Random Forest and Decision tree.

```

Training set : (4920, 133)
Testing set : (331, 133)
-----DECISION TREE-----
ACCURACY OF DECISION TREE ALGORITHM IS: 0.9527436477372293
RECALL 0.9795585785462744
PRECISION 0.9759001161440187
F1 SCORE 0.9742369361909198
=====RANDOM FOREST=====
ACCURACY OF RANDOM FOREST ALGORITHM IS : 0.9504805065138826
RECALL 0.986775419702249
PRECISION 0.9651567944250871
F1 SCORE 0.969393526233399
-----NAIVE BAYES-----
ACCURACY OF NAIVE BAYES ALGORITHM IS: 0.969300632105023
RECALL 0.9890983000739096
PRECISION 0.9814169570267132
F1 SCORE 0.9828081603797404
=====SUPPORT VECTOR MACHINE=====
ACCURACY OF SVM ALGORITHM IS: 0.9504805065138826
RECALL 0.986775419702249
PRECISION 0.9651567944250871
F1 SCORE 0.969393526233399
    
```

Figure 7: Calculations of the metrics for the disease Acne

Figure 8 compares the values of the obtained attributes from all the supervised learning methods.

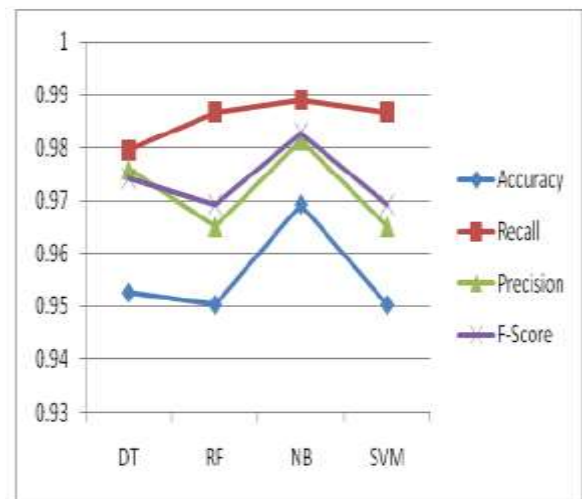


Figure 8: Graphical representation of comparative analysis

Conclusion and Future work

The primary objective of this article is to use Machine Learning techniques to better accurately predict the occurrence of the disease. Decision Tree, Random Forest, Support Vector Machine, and Naive Bays were some of the methods we used. These algorithms take symptoms into account as input variables, allowing for an evaluation of Accuracy, Precision, Recall, and F-score on a disease dataset. The Naive Bays algorithm provides the greatest accuracy in illness prediction among all the compared algorithms. Because this new condition may emerge at some point in the future, the system may be improved further. Adding additional symptoms to the data set allows for the prediction of these emerging illnesses. The GUI is made more user-friendly by using more potent machine learning supervised algorithms, which allows for the provision of more specific information about the condition to patients, and the inclusion of a component prescribing medication to the patient in the event of an emergency.

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