

VISVESVARAYA TECHNOLOGICAL UNIVERSITY
Jnana Sangama, Belagavi-590018, Karnataka



A PROJECT REPORT
ON

“Prediction of Chronic Kidney Disease Using Machine Learning”

*Submitted in partial fulfillment of requirements for the award of degree of
Bachelor of Engineering
in*

COMPUTER SCIENCE & ENGINEERING

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An Autonomous Institute

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CERTIFICATE

This is to certify that the project work, entitled “**PREDICTION OF CHRONIC KIDNEY DISEASE USING MACHINE LEARNING**” is a bonafide work carried out by

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in partial fulfillment for the award of degree of Bachelor of Engineering in Computer Science & Engineering of the Visvesvaraya Technological University, Belagavi during the academic year 2021-22. It is certified that all the corrections/suggestions indicated for Internal Assessment have been incorporated in the Report. The project report has been approved as it satisfies the academic requirements.

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DECLARATION

We, hereby declare that the entire work titled “**PREDICTION OF CHRONIC KIDNEY DISEASE USING MACHINE LEARNING**” embodied in this project report has been carried out by us during the 8th semester of BE degree at MVJCE, Bangalore under the esteemed guidance of **Mrs. S. Suguna**, Assistant. Prof, Dept. of CSE, MVJCE affiliated to Visvesvaraya Technological University, Belagavi. The work embodied in this dissertation work is original and it has not been submitted in part or full for any other degree in any University.

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ABSTRACT

One of the biggest health problems in the world is chronic kidney disease (CKD), which can lead to renal failure, cardiovascular disease (CVD), and early death. For patients who progress to end-stage kidney disease, insufficient availability to kidney replacement therapy is a problem in many regions of the world. According to research, 1 in 10 people may have CKD, however it is more prevalent in adults than children, where it affects 1 in 50 people and 1 in 2 persons over the age of 75. Millions of people worldwide die each year from CKD, which affects 10% of the population and typically has no symptoms due to aberrant test findings. Data classification can be used to identify CKD, but the system might not be able to accurately estimate the proportion of CKD patients. The innovative aspect of this study is the method used to diagnose chronic renal disease. The goal of this project is to compare the accuracy-based performance of different machine learning algorithms employing Linear Classifier, K-Nearest Neighbor (KNN), Support Vector Machine (SVM), Naive Bayes, Random Forest, and XGBoost.

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CHAPTER 1

INTRODUCTION

1.1 Project Overview

Chronic kidney disease (CKD) is a world's major health issue, with adverse outcomes of kidney failure, cardiovascular disease (CVD), and premature death. In many parts of the world, poor access to kidney replacement therapy is an issue for patients who advance to end-stage kidney disease. Research suggests that 1 in 10 of the population may have CKD, but it is common in adults, being present in 1 in 50 people, for those aged over 75, CKD is present in 1 out of 2 people. Of the world's population, 10% are affected by CKD and millions die each year, as in most cases it doesn't cause any symptoms due to the abnormality in the test results. CKD can be detected by classifying the data but the system may fail to give an accurate percentage of the people having CKD. The novelty of this study lies in the diagnosis system to detect chronic kidney disease. This project's aim is to differentiate the performance of various machine learning algorithms that are primarily based on their accuracy by using Linear Classifier, K-Nearest Neighbor (KNN), Support Vector Machine (SVM), Naïve Bayes, Random Forest, and XG Boost.

1.2 Purpose of Project

The sole purpose of this project is to detect CKD accurately based on features of the CKD dataset. In this KNN, Linear classifier, Random Forest, SVM, XG Boost, and Naïve Bayes Classifier algorithms are used. For each classifier, results were computed based on the accuracy, precision, recall, and F1 score.

1.3 Scope of Project

More than 1 in 7 adults in the United States may have chronic kidney disease (CKD), with higher rates in low-income and racial/ethnic minority groups. And most people with CKD don't know they have it. Healthy People 2030 focuses on preventing, diagnosing, and treating CKD. People with CKD are more likely to have heart disease and stroke and to die early. Managing risk factors like diabetes and high blood pressure can help prevent or delay CKD. Strategies to make sure more people with CKD are diagnosed early can help people get the treatment they need. Recommended tests can help identify people with CKD to make sure they get treatments and education that may help prevent or delay kidney failure.

and end-stage kidney disease (ESKD). In addition, strategies to make sure more people with ESKD get kidney transplants can increase survival rates and improve quality of life.

1.4 Problem Statement

Chronic kidney disease (CKD) is the presence of lasting functional abnormalities in the renal structure and function that can worsen over time. Kidneys are vital for the whole body to work properly. Therefore, when a person has CKD, they may also have problems with the functioning of the rest of their body. Some of the common complications of CKD include anemia, bone disease, heart disease, high potassium, high calcium, and fluid accumulation.

1.5 Existing System

Machine learning refers to a computer program, which calculates and deduces the information related to the task and obtains the characteristics of the corresponding pattern. This technology can achieve accurate and economical diagnoses of diseases; hence, it might be a promising method for diagnosing CKD. It has become a new kind of medical tool with the development of information technology and has a broad application prospect because of the rapid development of electronic health records. In the medical field, machine learning has already been used to detect human body status, analyze the relevant factors of the disease and diagnose various diseases.

For example, the models built by machine learning algorithms were used to diagnose heart disease, diabetes and retinopathy, acute kidney injury, cancer, and other diseases. In these models, algorithms based on regression, tree, probability, decision surface, and neural network were often effective. In the field of CKD diagnosis, Hodneland et al. utilized image registration to detect renal morphologic changes. Vasquez-Morales et al. established a classifier based on the neural network using large-scale CKD data, and the accuracy of the model on their test data was 95%. In addition, most of the previous studies utilized the CKD data set that was obtained from the UCI machine learning repository. Chen et al. used k-nearest neighbor (KNN), support vector machine (SVM), and soft independent modelling of class analogy to diagnose CKD, KNN and SVM achieved the highest accuracy of 95.2%. In addition, they used fuzzy rule-building expert system, fuzzy optimal associative memory and partial least squares discriminant analysis to diagnose CKD, and the range of accuracy in those models was 94.8%.

Current System do not used many algorithms for prediction purpose. Dataset used by current system is not having more parameter. Algorithms which are used by current system are not giving highest accuracy.

1.5.1 Disadvantages of Existing System

In the above model, the mean imputation is used to fill in the missing values and it depends on the diagnostic categories of the samples. As a result, existing method could not be used when the diagnostic results of the samples are unknown. In reality, patients might miss some measurements for various reasons before diagnosing. In addition, for missing values in categorical variables, data obtained using mean imputation might have a large deviation from the actual values. For example, for variables with only two categories, we set the categories to 0 and 1, but the mean of the variables might be between 0 and 1. Polat et al. developed an SVM based on feature selection technology, the proposed models reduced the computational cost through feature selection, and the range of accuracy in those models was from 94.75%-95.5%. J. Aljaaf et al. used novel multiple imputation to fill in the missing values, and then MLP neural network

(MLP) achieved an accuracy of 96.1%. The studies of, focus mainly on the establishment of models and achieve an ideal result. However, a complete process of filling in the missing values is not described in detail, and no feature selection technology is used to select predictors as well. Almansour et al. used SVM and neural network to diagnose CKD, and the accuracy of the models was 94.75% and 95.75%, respectively. In the models established by Gunarathne et al., zero was used to fill out the missing values and decision forest achieved the best performance with the accuracy was 94.1%. To summarize the previous CKD diagnostic models, we find that most of them suffering from either the method used to impute missing values has a limited application range or relatively low accuracy.

Therefore, in this work, we propose a methodology to extend application range of the CKD diagnostic models.

1.6 Proposed System

This article objects to predict Chronic Kidney Disease based on full features and important features of CKD dataset. For feature selection different technique has been applied: Feature selection. In this perception, six classifiers algorithm were applied viz. Linear Classifier, K-Nearest Neighbor (KNN), Support Vector Machine (SVM), Naïve Bayes, Random Forest, and XGBoost. For each classifier, the results were computed based on full features, Accuracy achieved by different classifier are: Linear Classifier is giving accuracy about 96%, KNN is achieving accuracy of 96%, SVM is giving accuracy of 95%, Naïve Bayes classifier is giving 96%, Random Forest and XGBoost is giving accuracy of 98%. Result is concluded by considering accuracy, precision, recall and F1-score. Using these machine learning algorithms, we are proposing this CKD system for classification.

1.6.1 Objectives

Chronic Kidney Disease is one of the most critical illnesses nowadays and proper diagnosis is required as soon as possible. Machine learning technique has become reliable for medical treatment. With the help of a machine learning classifier algorithms, the doctor can detect the disease on time. Chronic Kidney Disease dataset has been taken from the Kaggle repository. Six classifier algorithms have been applied in this research such as Linear Classifier, K-Nearest Neighbor (KNN), Support Vector Machine (SVM), Naïve Bayes, Random Forest, and XGBoost. Linear Classifier is giving accuracy about 96%, KNN is achieving accuracy of 93%, SVM is giving accuracy of 97%, Naïve Bayes classifier is giving 96%, Random Forest and XGBoost is giving accuracy of 96%. Result is concluded by considering accuracy, precision, recall and F1-score. Among these algorithms, XGBoost is giving more accuracy of 98%.

Summary

This chapter gives an overview of project, its scope, existing system and its disadvantages, proposed system and objectives of proposed system.

CHAPTER 2

LITERATURE STUDY

2.1 Overview

Literature study is carried out to get all the related information of current project, which is used to get an idea for the enhancement as well as changes that can be made to improve existing approaches. A literature study is done on various machine learning techniques. Following section describes about all the related papers which is used in current project.

2.2 Related Papers

PAPER 1: Kidney stone recognition and extraction using Directional Emboss & SVM from Computed Tomography Images

“Akanksha Soni et al” The intention of this paper is to develop an automatic stone detection system from the Computed Tomography (CT) Image. The primary aim is to inspect the CT scan to recognize whether the kidneys are influenced by a stone. The algorithm used is the SVM (Support Vector Machine) to classify the stone. SVM is a linear model for classification and regression problems [1]. It can solve linear and non-linear problems and work well for many problems. This algorithm creates a line or hyperplane which separates the data into classes.

SVM is used for classifying two different classes of data in two particular shells and construct a hyperplane that can be valuable for regression. Emboss is a computer graphic technique in which each pixel of an image is replaced either by a highlight or a shadow, depending on light/dark boundaries on the original image. Low contrast areas are replaced by a gray background.

The proposed system helps to detect the stone in the kidney but fails to identify multiple stones in kidney. In the future scope it is suggested to develop the system which could detect multiple number of stones in the kidney with more accuracy [1]. Automatic identification is one of the trending in medical science which is helpful for saving human health and wealth as well.

PAPER 2: Kidney Stone Detection Using Neural Networks

“M.Akshaya al” Kidney may fail functioning due to the stone formation in the kidney. That is highly impractical to handle the large amount of data to detect manually. Kidneys remove wastes and extra fluid from our body. Kidneys also remove acid that is produced by the cells of our body and maintain a healthy balance of water, salts and minerals such as sodium, calcium, phosphorus and potassium in our body [2]. Kidney may fail functioning due to the stone formation in the kidney. That is highly impractical to handle the large amount of data to detect manually. By using AI based methods along with neural networks and feature extraction has shown the great potential using back propagation network was applied to detect the stone in the kidney.

The feature extraction is done by using principal component analysis and the image classification is done using back propagation network. Back propagation gives a perfect classification with greater accuracy.

The algorithm used is the Back Propagation Network (BPN) [2]. BPN is used to detect stone in the kidney. BPN gives a quick and precise classification and one of the better tools for identifying the stone in the kidney. This method also presents a segmentation method using FCM (Fuzzy C-Mean) clustering algorithm.

PAPER 3: Explainable Prediction of Chronic Renal Disease in the Colombian Population Using Neural Networks and Case-Based Reasoning

Gabriel R. Vasquez- Morales al” This paper presents a neural- network primarily based classifier that helps in predicting whether or not someone is in danger of developing Chronic excretory organ Disease (CKD). Nowadays computer science plays a very important role in health sector. Machine Learning Algorithms are widely utilized in malady prediction. Some Algorithms like Support Vector Machine (SVM), Random Forest, Neural Networks are accustomed to predict and classify patients’ polygenic disorder, heart condition, cancer et al. Neural Networks have currently been used as support tools for early malady detection. This paper presents a neural- network primarily based classifier that helps in predicting whether or not someone is in danger of developing Chronic excretory organ Disease (CKD). The model is trained with demographic knowledge and treatment info of 2 population teams.

Current analysis proposes the employment of dual system. Twin system refers Neural Network (NN)- Case primarily based Reasoning (CBR) that is applied and valid for the

reason of CKD predictions. Neural Networks are accustomed notice CKD and alternative effects associated with urogenital apparatus.

PAPER 4: A Comprehensive Unsupervised Frame work for Chronic Kidney Disease Prediction

“L. Antony et al” Early detection helps to provide proper treatment and care to patients. In many hospitals there is a shortage of nephrologists so patients have to wait for a longer to get diagnosed. CKD decreases the kidney function over time as indicated by glomerular filtration rate (GFR) and markers of kidney damage. People with CKD are dying at early age. It is crucial to detect CKD at an early stage. Early detection helps to provide proper treatment and care to patients. In many hospitals there is a shortage of nephrologists so patients have to wait for a longer to get diagnosed. Thus, with the help of AI technology this system classifies patients into CKD or Non-CKD which will help doctors to treat multiple patients faster.

Dataset is online data repository of the University of California Irvine (UCI) and contains data of 400 patients. It consists of 1 class attribute and 24 clinical attributes. The dataset consists of 250 CKD and 150 non-CKD cases. Dataset includes missing data as well.

Unsupervised learning extracts patterns from CKD related clinical data. Various feature selection mechanisms related to filter methods wrapper methods, embedded methods and unsupervised methods are implemented to identify important features. algorithms such as K-Means Clustering, Isolation Forest, DB-scan and Auto encoder are implemented on selected features. Evaluation metrics are generated to compare with existing models. In classifying CKD with Non-CKD, K-means clustering achieved the accuracy of 94%.

PAPER 5: Chronic Kidney disease prediction using Machine Learning

“Pankaj chittora al” Chronic kidney disease (CKD) means kidneys are damaged and are not filtering the blood way it should. if the person has suffered from the CKD means that waste is collected in the body. This disease is chronic because of the damage gradually over a long period. The causes of the CKD include diabetes, high blood pressure, heart disease, it also depends on age and gender. CKD does not give any symptoms till kidney is damaged badly. There are few diagnostic tests to check the condition of CKD: 1. estimated glomerular filtration rate (eGFR) 2. Urine test 3. Blood pressure. Algorithm used are Filter Method, Wrapper Method, and Embedded Method. The primary role of kidney is to filter the extra water and waste from blood to produce urine. Chronic kidney disease (CKD)

means kidneys are damaged and are not filtering the blood way it should. If the person has suffered from the CKD means that waste is collected in the body. This disease is chronic because of the damage gradually over a long period. The causes of the CKD include diabetes, high blood pressure, heart disease, it also depends on age and gender. CKD does not give any symptoms till kidney is damaged badly [11]. There are few diagnostic tests to check the condition of CKD: 1. estimated glomerular filtration rate (eGFR) 2. Urine test 3. Blood pressure.

A. eGFR

eGFR value shows that how your kidney cleaning the blood. If your eGFR value is greater than 90, that means the kidney is normal. If eGFR value is less than 60, that means you have CKD.

B. Urine Test

The doctor also asks for urine test for kidney functionality because kidneys make urine. If the urine contains blood and protein, that means your kidney is not working properly.

C. Blood Pressure

Blood pressure range shows how your heart is pumping blood. If eGFR value reaches less than 15, that means the patient has end-stage kidney disease. At this point, there are only available treatments: (i) dialysis and (ii) kidney transplant. If dialysis is not possible, the doctor has only one solution, i.e., kidney transplantation.

Therefore, it is necessary to early recognition, monitoring and handling of the disease. since it does not show the symptoms at early stage it is essential to predict the striding of CKD with appropriate accuracy. The machine learning can also find data analysis and pattern detection. A variety of health dataset is present so machine learning algorithms are best fit to improve the accuracy of diagnosis prediction. As healthcare electronic dataset grows rapidly, machine learning algorithms are becoming more common in healthcare.

PAPER 6: An image pre-processing method for kidney stone segmentation in CT scan images

“Nilar Thein al” Automatic Kidney stone detection segmentation from CT images is challenging. The purpose of this system is to develop reader independent pre-processing

algorithm to assist the segmentation process [15]. Three algorithms based on the intensity, size and location are applied for unwanted regions removing such as soft-organ removing, bony skeleton removing and bed-mat removing. The digitalized transverse abdomen CT scans images from 30 patients with kidney stone cases were included in statistical analysis and validation.

The used algorithm takes image as an input recognizes soft-organ removing and bony skeleton removing and bed-mat removing and outputs the stone detected image as output.

The proposed system method for kidney stone segmentation will provide good support in detection of kidney stone. Intensity-based thresholding can produce the output image without any soft regions. Area-based thresholding can reduce many false positive and bedmat region. By using thresholding method based on the prior knowledge of the image, the proposed system is simple and easy to understand for the segmentation.

2.3 Literature Study

The most of the literature survey focuses on the techniques to detect chronic kidney disease using machine learning algorithms. However, they are not giving more accuracy. Literature survey made us think more logically in order to get more accuracy.

Summary

This chapter discusses about the literature study for the proposed system, related paper which gives idea of enhancements to the existing work. The related paper gives information about the existing systems, its advantages and disadvantages.

CHAPTER 3

SYSTEM REQUIREMENT

The system requirements specify features, components and behaviour of system which is to be developed. The following sections describe about functional, non-functional, performance related, features and behaviour of the solution. This includes the detailed description of the solution to be developed.

3.1 Resource Requirement

Jupyter Notebook is a non-profit organization created to "develop open-source software, open-standards, and services for interactive computing across dozens of programming languages". Spun-off from IPython in 2014 by Fernando Pérez, Project Jupyter supports execution environments in several dozen languages. Project Jupyter's name is a reference to the three core programming languages supported by Jupyter, which are Julia, Python and R, and also an homage to Galileo's notebooks recording the discovery of the moons of Jupiter. Project Jupyter has developed and supported the interactive computing products Jupyter Notebook, Jupyter Hub, and Jupyter Lab, the next-generation version of Jupyter Notebook. Jupyter Notebook (formerly IPython Notebooks) is a web-based interactive computational environment for creating Jupyter notebook documents. The "notebook" term can colloquially make reference to many different entities, mainly the Jupyter web application, Jupyter Python web server, or Jupyter document format depending on context. A Jupyter Notebook document is a JSON document, following a versioned schema, and containing an ordered list of input/output cells which can contain code, text (using Markdown), mathematics, plots and rich media, usually ending with the ".ipynb" extension.

3.2 Hardware Requirement

The following is the hardware requirements of the system for the proposed system:

Hardware Requirements:

Processor	: Intel(R) Core(TM) i3-8145U CPU @ 2.10GHz 2.30 GHz
RAM	: 256 MB (min)
Storage	: 20 GB
Standard Devices	: Keyboard, Monitor and Mouse

3.3 Software Requirement

The following is the software requirements of the system for the proposed system:

- OS : Windows
- Language : Python
- IDE/tool : Jupyter Notebook
- Dataset : <https://www.kaggle.com/datasets/mansoordaku/ckdisease>

CHAPTER 4

SYSTEM ANALYSIS

Analysis is nothing but finding solution to various problems. System analysis is defined as, process in which we get the information about the existing problems, requirements and to solve various problems related to system. Study of feasibility plays vital role in system analysis, which helps in providing goals for development and design.

4.1 Feasibility Analysis

Considering initial results, a deeper survey or feasibility study is done on existing results. “FEASIBILITY ANALYSIS” is nothing but meeting the requirement of system, to achieve desired goal, working on proposed system and getting detailed information on resources.

There are 8 steps to be considered for Feasibility Analysis:

- Develop a team for a project and assign leader for that team.
- Identify the strength of proposed work.
- Characterize and recognize qualities of proposed work.
- Identify cost of proposed work and show its performance.
- Show system performance.
- Choose best framework.
- Prepare report and submit to administrative

Three key features involve in “FEASIBILITY ANALYSIS” are:

- Financial Feasibility
- Technical Feasibility
- Public Feasibility

4.1.1 FINANCIAL

Financial feasibility is one of key factor of feasibility analysis, which is carried out to scrutinize economic cost of organization, limited fund is endowed by companies for development of system. Hence, the developed project cost is within budget, because of freely available resources and can be accessed by open-source.

4.1.2 TECHNICAL

Technical feasibility is entailed to scrutinize technical performance of system. System which is building should have less demand. Hence, this gives number high of demands asked by the clients. The proposed system is built in such way that there is no harm for the user as well the MR-SGSO executes very well. The project draws all the technical requirements from open source and resources are used which have higher execution time.

Summary

The objective of this chapter is to know the proposed system is feasible or not. The chapter mainly describes about the various keys of feasibility analysis i.e. financial, technical and public.

CHAPTER 5

SYSTEM DESIGN

Design is an innovative procedure; a successful system is developed through great designs. System design is a process of giving detailed information about the proposed work in a physical format. Different designs are built for development of system, which describes about features, components which are included and how client interact with system.

5.1 System Architecture

System architecture describes about each component of system:

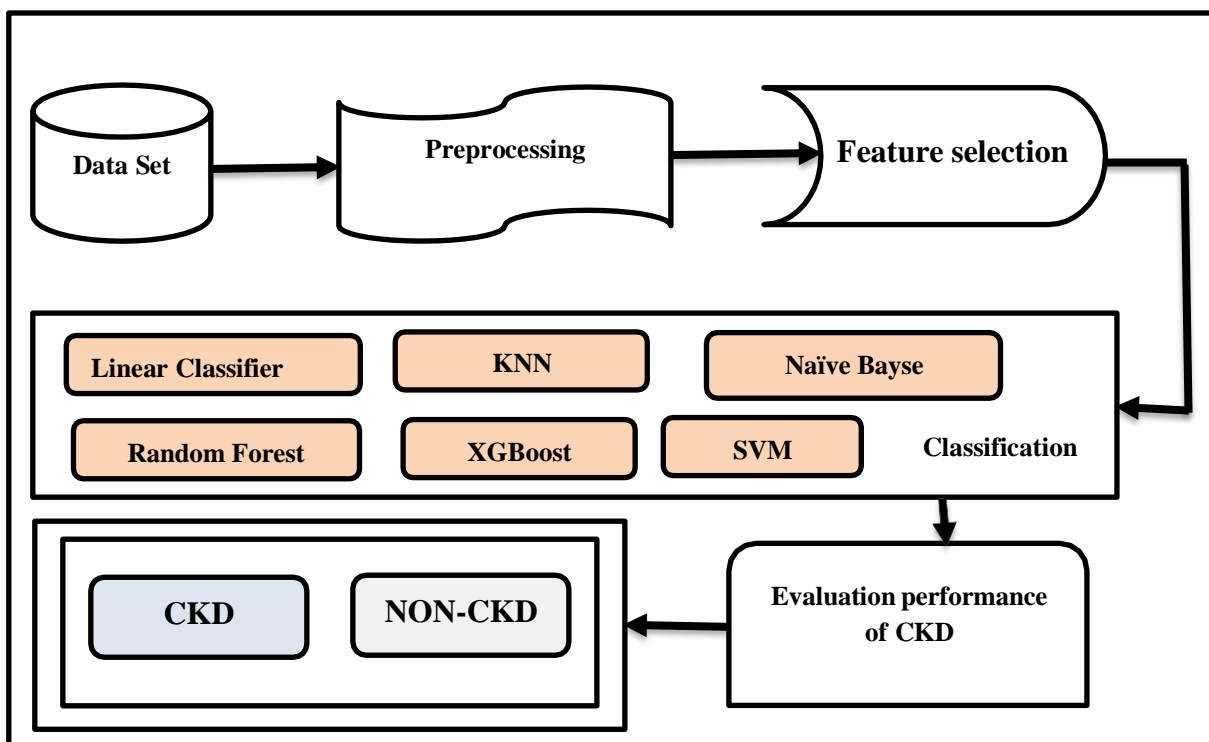


Fig 5.1: System Architecture

5.2 Use case Diagram

A **use case diagram** at its simplest is a representation of a user's interaction with the system that shows the relationship between the user and the different use cases in which the user is involved. A use case diagram can identify the different types of users of a system and the different use cases and will often be accompanied by other types of diagrams as well. These cases are represented by either circles or ellipses.

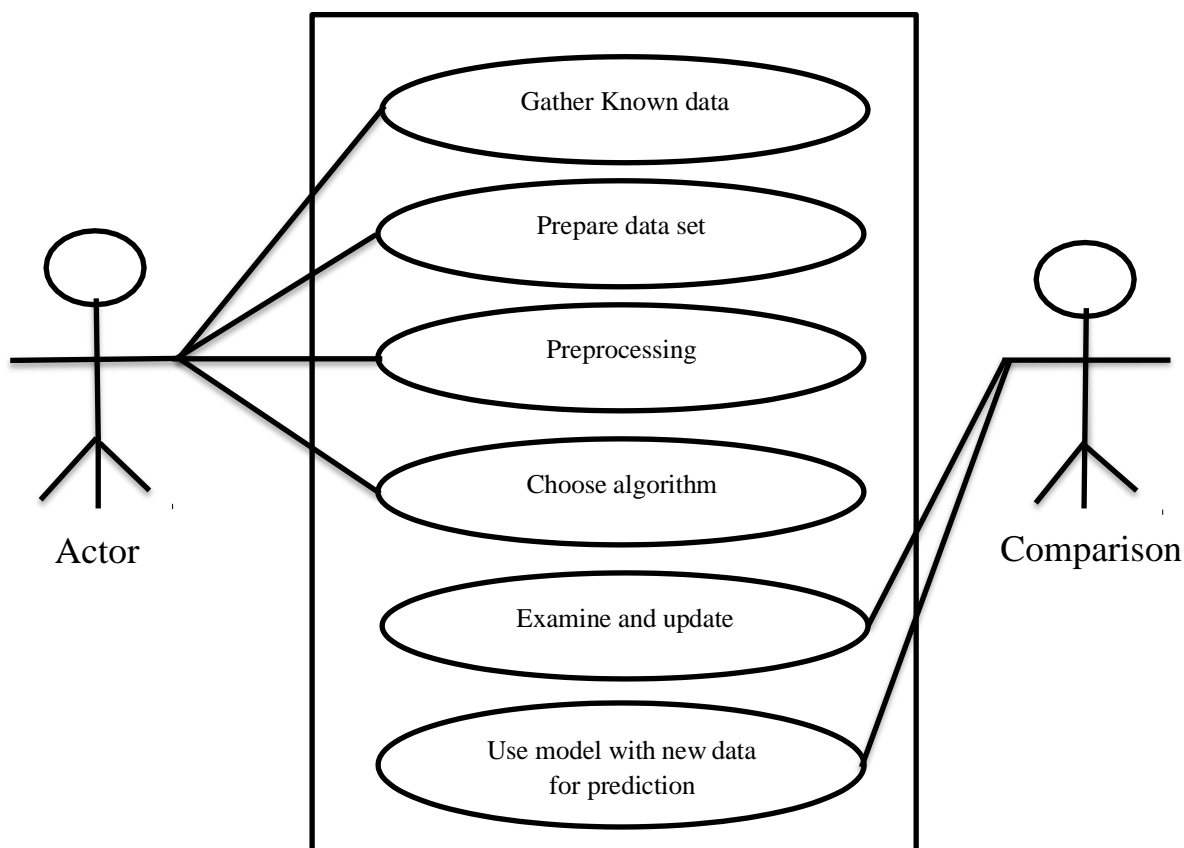


Fig 5.2: Use-case Diagram

5.3 Class Diagram

Class diagrams are generally used for conceptual modeling of static view of a software application, and for modeling translating models into programming code in a detailed manner. At time of developing or construction software systems, a class diagram is widely used. They are also used for data modeling. It is used to show classes, relationships among them, interface, association, etc. Class in a class diagram simply is a blueprint of an object. It simply describes and explains different type of objects in system, and different types of relationships that exist between them.

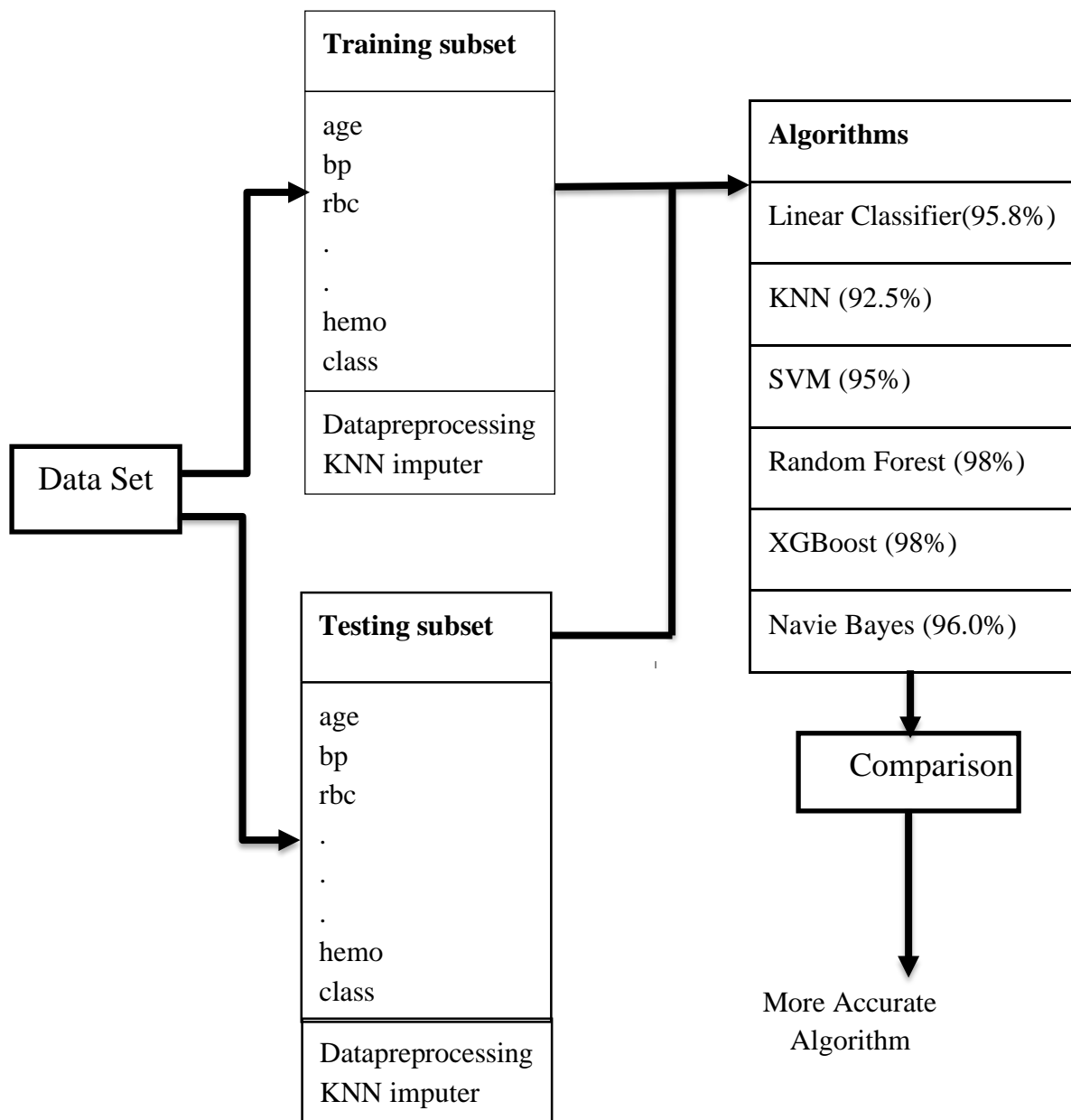


Fig 5.3: Class diagram

Summary

This chapter mainly concentrates on input design, output design, system architecture and various designs which is involved in proposed system.

CHAPTER 6

IMPLEMENTATION

It is main part of project where project is implemented. Implementation must be clearly defined, planned carefully and systematically otherwise it causes confusion and leads to generation of problems. Following tasks involved in stage of implementation:

- Planning carefully
- System Investigation.
- Investigating different types of systems and constraints.
- Selecting the most accurate and correct language for developing the application.
- Understanding and correctly evaluating several changeover methods.
- Rightful decision is made about the selection of the platform.

6.1 Language used for implementation

Implementation is an important step where execution of project is done. Hence language used for implementation should be adaptable, error free, user friendly and etc. Most of the time because of some mistake project get ruined or spoiled due to inappropriate programming language.

Python is widely considered as the preferred language for ML (Machine Learning). Few simple reasons are:

It has an excellent collection of in-built libraries: Python claims a huge number of in-built libraries for data mining, data manipulation, and machine learning. For example, NumPy, this is used for scientific calculation. Scikit-learn have tools for data mining and analysis which optimizes the Python's brilliant machine learning usability. Panda is another package which provides developers with high- performance structures and data analysis tools and helps them to reduce the project implementation time. Similarly, SciPy is used for advanced computation, and Pybrain is used for machine learning.

Moderate learning Curve: Python Programming language is very accessible and easy to learn and use. It focuses on code readability. It is versatile and well- structured language.

It is a general-purpose programming language: Well, Python is a good choice if the project requirements are more than just information such as developing a functional website.

Easy to integrate: Python programming language incorporates better than “R” in business environments. It is easy to integrate Python with other lower-level languages such as C, C++, or Java. Similarly, the Python based-stack is easy to incorporate with data scientist’s work, which allows it to bring efficiently into production.

High productivity: Syntax in Python is extremely readable and easy to understand similar to other programming languages, whereas R programming language has different syntax. The readability syntax in Python programming language confirms high productivity of development teams.

Easy to create prototypes: It is already stated that Python is easy to learn and fast to develop. It requires less coding, which means that you can create prototypes and test your concepts quickly and easily in Python as compared to other programming languages. Developing prototypes in Python not only saves developers time, but it also decreases your company’s costs.

It is free and open source: Python programming language is available freely; this allows you to download Python for free, i.e., you can download its source code, make the modifications in it and then distribute it. The download comes with an extensive collection of libraries that supports you to carry out your tasks.

Object-oriented paradigm: Python programming language supports for both the Object-oriented and procedural programming models. Classes and objects in Object-oriented programming help us to model the real world while functions in procedural programming enable us to reuse the code. A class in Object-oriented programming encapsulates data and function together.

Portability: Python programming language is portable, i.e. code written in Python can be run on another platform; this is called Write Once Run Anywhere, i.e., WORA, but this is not possible with other languages like C++. When you have to run C++ code on other platforms, it requires making some modifications in the code to run. It is not advisable to use any system-dependent features.

6.2 Implementation platform

Platform plays an important role for development of software. A platform is a place where software is launched. Ubuntu platform is used for implementation of project. There are various analyses for choosing Ubuntu platform. It is an open-source operating system. It is used for many different purposes such as scrutiny, embedded system, security and etc. It also contains remote connection and option to restore, so that file can be restored if deleted. Ubuntu software is used for scalable processing, multi-tasking, Encrypted File System (EFS) and smart card support.

6.2.1 Linear Classifier

A linear classifier is a model that makes a decision to categories a set of data points to a discrete class based on a linear combination of its explanatory variables. As an example, combining details about a dog such as weight, height, color and other features would be used by a model to decide its species.

A linear classifier will draw a straight line (e.g., $y=mx+c$) to divide sets of points into 2 classes.

Formula:

$$Y = \beta_0 + \sum_{i=0}^n \beta_i \cdot X_i$$

6.2.2 K-Nearest Neighbor Algorithm

It is a supervised ML algorithm used to solve both regression and classification problems. It aims to locate all of the closest neighbors around a data point in order to figure out in which class it belongs. KNN is one of the simplest machine learning algorithms which is easy to implement.

It is a supervised machine learning algorithm that solves both classification and regression problems. This algorithm stores all the available data and classifies new data based on their similarity.

First select the number of neighbors. Then calculate Euclidean distance between the data points.

Euclidean distance is the distance between two points. Formula for Euclidean distance is given by:

$$\sqrt{(X_2 - X_1)^2 + (Y_2 - Y_1)^2}$$

6.2.3 Support Vector Machine

Support Vector Machine or SVM is one of the most popular Supervised Learning algorithms, which is used for Classification as well as Regression problems. However, primarily, it is used for classification problems in Machine Learning.

The goal of the SVM algorithm is to create the best line or decision boundary that can segregate n-dimensional space into classes so that we can easily put the new data point in the correct category in the future. This best decision boundary is called a hyperplane.

SVM chooses the extreme points/vectors that help in creating the hyperplane. These extreme cases are called support vectors, and hence algorithm is termed a Support Vector Machine. Consider the below diagram in which there are two different categories that are classified using a decision boundary or hyperplane.

Formula used for non-linear data:

$$\mathbf{Z} = \mathbf{X}^2 + \mathbf{Y}^2$$

6.2.4 Random Forest

Random Forest is an ensemble technique capable of performing both regression and classification tasks with the use of multiple decision trees and a technique called Bootstrap and Aggregation, commonly known as **bagging**. The basic idea behind this is to combine multiple decision trees in determining the final output rather than relying on.

industrial decision trees. Random Forest has multiple decision trees as base learning models. We randomly perform row sampling and feature sampling from the dataset forming sample datasets for every model. This part is called Bootstrap.

The random forest algorithm works by completing the following steps:

Step 1: The algorithm selects random samples from the dataset provided.

Step 2: The algorithm will create a decision tree for each sample selected. Then it will get a prediction result from each decision tree created.

Step 3: Voting will then be performed for every predicted result. For a classification problem, it will use mode, and for a regression problem, it will use mean.

Step 4: And finally, the algorithm will select the most voted prediction result as the final prediction.

6.2.5 Naïve Bayes Classifier

Naive Bayes classifiers are a collection of classification algorithms based on **Bayes' Theorem**. It is not a single algorithm but a family of algorithms where all of them share a common principle, i.e., every pair of features being classified is independent of each other.

Bayes' Theorem finds the probability of an event occurring given the probability of another event that has already occurred. Bayes' theorem is stated mathematically as the following equation:

$$P(y_i | X_1, X_2, \dots, X_n) = P(X_1 | y_i)P(X_2 | y_i) \dots P(X_n | y_i)P(y_i) / P(X_1, X_2, \dots, X_n)$$

where A and B are events and $P(B) \neq 0$.

- Basically, we are trying to find probability of event A, given the event B is true. Event B is also termed as **evidence**.
- $P(A)$ is the **priori** of A (the prior probability, i.e., Probability of event before evidence is seen). The evidence is an attribute value of an unknown instance (here, it is event B).
- $P(A|B)$ is a posteriori probability of B, i.e., probability of event after evidence is seen. Now, with regards to our dataset, we can apply Bayes' theorem in following way:

$$P(y|X) = \frac{P(X|y) P(y)}{P(X)}$$

where, y is class variable and X is a dependent feature vector (of size n) where:

$X = (x_1, x_2, x_3, \dots, x_n)$.

6.2.6 XGBoost

XGBoost is an implementation of Gradient Boosted decision trees. XGBoost models majorly dominate in many Kaggle Competitions.

In this algorithm, decision trees are created in sequential form. Weights play an important role in XGBoost. Weights are assigned to all the independent variables which are then fed into the decision tree which predicts results. The weight of variables predicted wrong by the tree is increased and these variables are then fed to the second decision tree. These individual classifiers/predictors then ensemble to give a strong and more precise model. It can work on regression, classification, ranking, and user-defined prediction problems.

Similarity Score = (Sum of residuals)² / Number of residuals + lambda

Gain = Left tree (similarity score) + Right (similarity score) - Root (similarity score)
Gain - gamma

Output value = Sum of residuals / Number of residuals + lambda

Best Algorithm is chosen based on the following

Precision: Tell us how many of the correctly predicted cases actually turned out to be positive.

$$\frac{TP}{TP+FP}$$

Recall: Tell us how many of the actual positive cases we were able to predict correctly with our model.

$$\frac{TP}{TP+FN}$$

F1-score: harmonic mean of precision & Recall, so it gives a combined idea about these 2 metrics. It is maximum when precision is equal to Recall.

Summary

The chapter describes about implementation process of proposed system. It describes about which platform, language, algorithms and processing techniques used for executing the project.

CHAPTER 7

TESTING

Testing is an important part of project where testing of each module is done. Testing guarantees that proposed system is well organized analyzed to meet require project goal. Testing is last stage of project which guarantees the system is error free and ready to give desired output. The goals of testing are given by:

- Give operational quality to system
- Search and remove errors.
- Best quality project is produced.
- To approve the product as a solution for the first issue.

The following are types of testing performed in proposed system.

7.1 Unit Testing

Each module of project is tested individually. Verification is done on each module. Module of project is tested individually. Testing is done in programming style. The unit testing for the proposed system is performed on initialization, Data collection, Genre classification, visualizing the output.

7.2 Integration

After unit testing is performed, integration testing takes place in each module of project. Integration is done on various classes of system. It is done on front and back end also.

- **Function into classes Integration**

Initially during code phase various functions is developed for development of system. Each function of system is tested and coded individually. As all the functions are verified, they are mixed into their particular classes.

- **Distinct classes Integration**

Based on functionality, testing of distinct classes is done independently. Verification of each class is done which gives good result and hence integration is performed again on different classes.

7.3 Integration Testing

Developing a programming framework is a sophisticated technique which is used by Integration testing. It solves various issues on dual verification problem and construct program which solves all related problems. Main objective of integration testing is to construct a program structure based on unit testing modules.

As modules of software are divided, testing is performed on each module. Later this separated module is tested as whole set. Here, to rectify errors is difficult as it has different isolated errors.

7.3.1 Top-down Integration

The up down integration deals with development of program framework in incremental way. Modules of each program are coordinated in descending order and it starts with primary module. This method is an incremental approach to the construction of program structure. Modules are integrated by moving downward, beginning with the main program module. Modules that subordinate to the main program module are incorporated into the structure in either a depth first or breadth first manner.

7.3.2. Bottom-up Integration

The below table integrated testing table is divided into integrated classes, functions of each classes, how test is performed and result generated. This method begins the construction and testing with the modules at the lowest level in the program structure. Since the modules are integrated from bottom to up, processing required for modules subordinate to a given level is always available. Therefore, in this case the need for stubs is eliminated. It is important to check whether the testing is error free or not for different classes.

7.4 Validation Testing

As completion of testing combination is done, writing computer program is put together in a package. Testing approval is described from various perspectives. Hence testing affirms items limits which are sensibly expected

7.5 User Acceptance Testing:

Key factor for success of any system is user acceptance. User acceptance testing is performed on users, show which will be successful based on user motivation and knowledge. At time of developing and making required changes, system under consideration along with prospective system, users undergo constant testing. The changes are made regarding to 3 points

- Input design
- Output design

7.6 Artificial Data

Artificial data is created for testing purpose. All combinations of formats and values are tested by generating the artificial test data. Data generating utility program in system information departments helps to prepare artificial data as quickly as possible. It is made possible to test all control paths, login through program.

7.7 Assurance of Quality

Quality assurance is testing and analysis of administration element. The main objective is to give knowledge about item quality to administration. Assurance of quality involves:

7.7.1 Quality Factors

The main goal of confirmation value is track product quality and observes procedures to enhance programming.

Quality factors are described in following two categories:

- Directly measured factors.
- Indirectly measured factors.

Quality factors mainly focus on following three things:

- Operational attributes
- Experiences capacity changes
- Versatility.
- Effectiveness
- Time duration.

7.7.2 General Risks

Risk is nothing but which gives negative results at undesirable incident. The following three things are considered to recognize risk in other projects:

- Damage occurs during an occasion.
- Probability of occurring an occasion.
- The result that modified at certain level.

7.7.3 Security Technologies & Policies

The seven major activities the software quality is comprised of which are follows:

- Conduct of formal specialized audits
- Measurement
- Testing Software
- Application for specialized techniques.
- Control of progress
- Record keeping and announcing
- Enforcement of measures

Summary

The chapter describes about various testing techniques such as unit testing, integration testing, validation testing, output testing and etc.

CHAPTER 8

RESULTS

8.1 Dataset

The Dataset which has been use in this project is Chronic Kidney Disease Dataset. ThisDataset contains 400 records of different parameter from patients and non-patients.

8.2 Results

8.2.1 CONFUSION MATRIX OF LINEAR CLASSIFIER

True Negative	TN	00	69
True Positive	TP	11	46
False Negative	FN	10	04
False Positive	FP	01	01

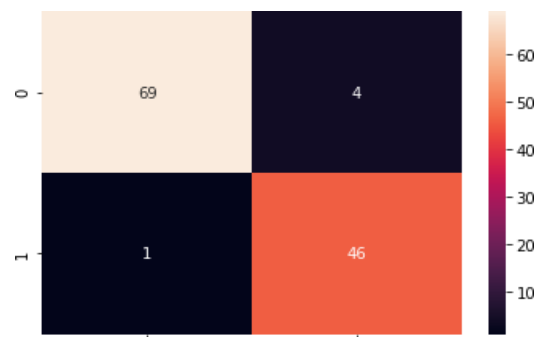


Fig 8.2.1: Confusion matrix of Linear Classifier

Calculation:

$$\frac{TP+TN}{TP+TN+FN+FP} = \frac{46+69}{46+69+04+01} = \frac{115}{120} = 0.9583$$

CLASSIFICATION REPORT OF LINEAR CLASSIFIER

	Precision	Recall	F1-score	Support
0	0.99	0.95	0.97	73
1	0.92	0.98	0.95	47
Accuracy			0.96	120
Macro average	0.95	0.96	0.96	120
Weighted average	0.96	0.96	0.96	120

Table 8.2.1 Classification report of Linear classifier

8.2.2 CONFUSION MATRIX OF KNN

True Negative	TN	00	58
True Positive	TP	11	53
False Negative	FN	10	05
False Positive	FP	01	04

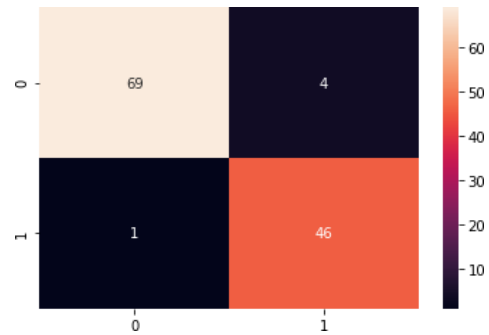


Fig 8.2.2: Confusion matrix of KNN

Calculation:

$$\frac{TP+TN}{TP+TN+FN+FP} = \frac{53+58}{53+58+05+04} = \frac{111}{120} = 0.925$$

CLASSIFICATION REPORT OF KNN

	Precision	Recall	F1-score	Support
0	0.94	0.92	0.93	63
1	0.91	0.93	0.92	57
Accuracy			0.93	120
Macro average	0.92	0.93	0.92	120
Weighted average	0.93	0.93	0.93	120

Table 8.2.2 Classification report of KNN

8.2.3 CONFUSION MATRIX OF SVM

True Negative	TN	00	65
True Positive	TP	11	51
False Negative	FN	10	03
False Positive	FP	01	01

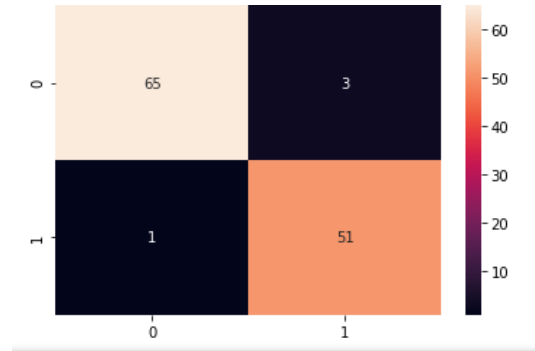


Fig 8.2.3: Confusion matrix of SVM

Calculation:

$$\frac{TP + TN}{TP + TN + FN + FP} = \frac{51 + 65}{51 + 65 + 03 + 01} = \frac{116}{120} = 0.966$$

CLASSIFICATION REPORT OF SVM

	Precision	Recall	F1-score	Support
0	0.98	0.96	0.97	68
1	0.94	0.98	0.96	52
Accuracy			0.97	120
Macro average	0.96	0.97	0.97	120
Weighted average	0.97	0.97	0.97	20

Table 8.2.3 Classification report of SVM

8.2.4 CONFUSION MATRIX OF Random Forest

True Negative	TN	00	66
True Positive	TP	11	32
False Negative	FN	10	00
False Positive	FP	01	02

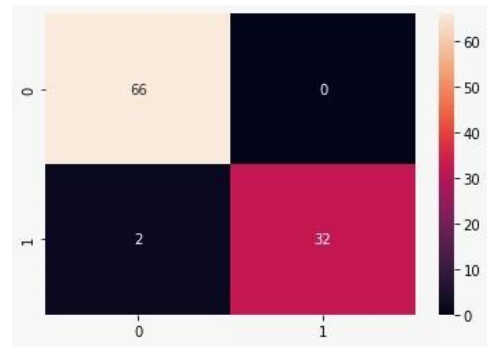


Fig 8.2.4: Confusion matrix of Random Forest

Calculation:

$$\frac{TP + TN}{TP + TN + FN + FP} = \frac{32 + 66}{32 + 66 + 00 + 02} = \frac{98}{100} = 0.98$$

CLASSIFICATION REPORT OF RANDOM FOREST

	Precision	Recall	F1-score	Support
0	0.94	1.00	0.97	66
1	1.00	0.88	0.94	34
Accuracy			0.96	100
Macro average	0.97	0.94	0.95	100
Weighted average	0.96	0.96	0.96	100

Table 8.2.4 Classification report of Random Forest

8.2.5 CONFUSION MATRIX OF XGBOOST

True Negative	TN	00	66
True Positive	TP	11	32
False Negative	FN	10	00
False Positive	FP	01	02

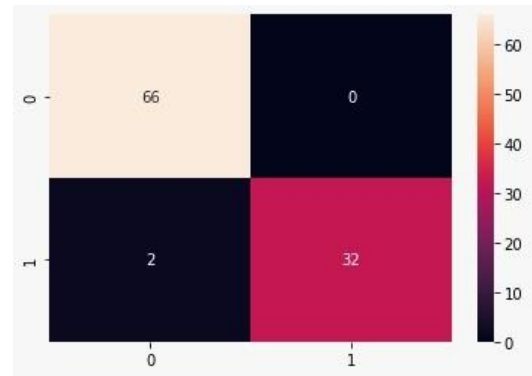


Fig 8.2.5: confusion matrix of XGBoost

Calculation:

$$\frac{TP + TN}{TP + TN + FN + FP} = \frac{32 + 66}{32 + 66 + 00 + 02} = \frac{98}{100} = 0.98$$

CLASSIFICATION REPORT OF XGBoost

	Precision	Recall	F1-score	Support
0	0.97	1.00	0.99	66
1	1.00	0.94	0.97	34
Accuracy			0.98	100
Macro average	0.99	0.97	0.98	100
Weighted average	0.98	0.98	0.98	100

Table 8.2.5 Classification report of XGBoost

8.2.6 CONFUSION MATRIX OF NAÏVE BAYES

True Negative	TN	00	62
True Positive	TP	11	34
False Negative	FN	10	4
False Positive	FP	01	0

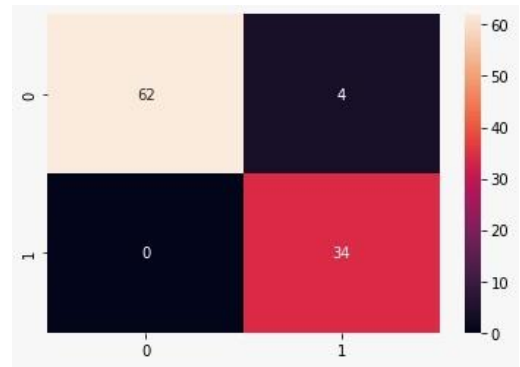


Fig 8.2.6: Confusion matrix of Naïve Bayes

Calculation:

$$\frac{TP + TN}{TP + TN + FN + FP} = \frac{34 + 62}{34 + 62 + 04 + 00} = \frac{96}{100} = 0.96$$

CLASSIFICATION REPORT OF NAÏVE BAYES

	Precision	Recall	F1-score	Support
0	1.00	0.94	0.97	66
1	0.89	1.00	0.94	34
Accuracy			0.96	100
Macro average	0.95	0.97	0.96	100
Weighted average	0.96	0.96	0.96	100

Table 8.2.6 Classification report of Naïve Bayes

8.3 COMPARISION TABLE

	Accuracy	precision	recall	F1-score
	(%)	(%)	(%)	(%)
Linear Classifier	96.0	100	88.2	93.7
KNN	96.0	96.8	91.1	93.9
SVM	95.0	93.9	91.1	92.5
XGBoost	98.0	100	94.11	92.5
Random Forest	98.0	100	94.11	96.9
Naïve Bayes	96.0	89.4	100	94.4

Table 8.3.1: Comparison of Algorithms

8.4 COMPARISION GRAPHS

Accuracy

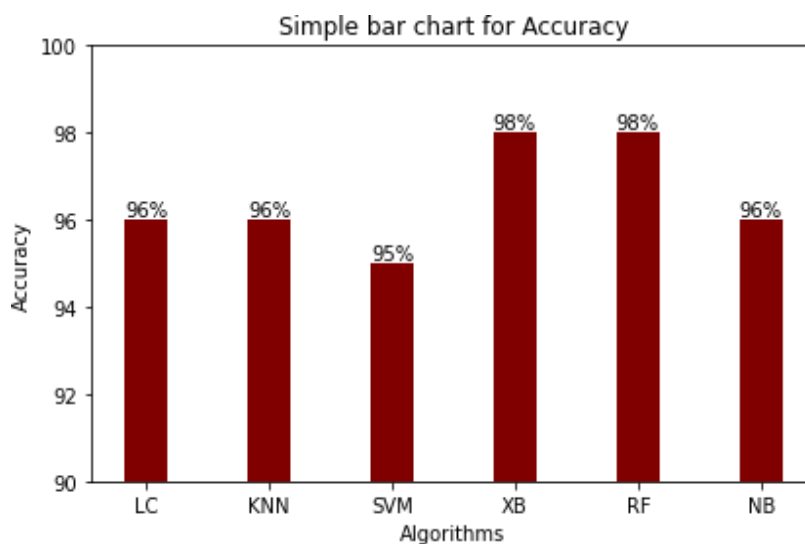


Fig 8.4.1: Accuracy Bar Chart

Precision

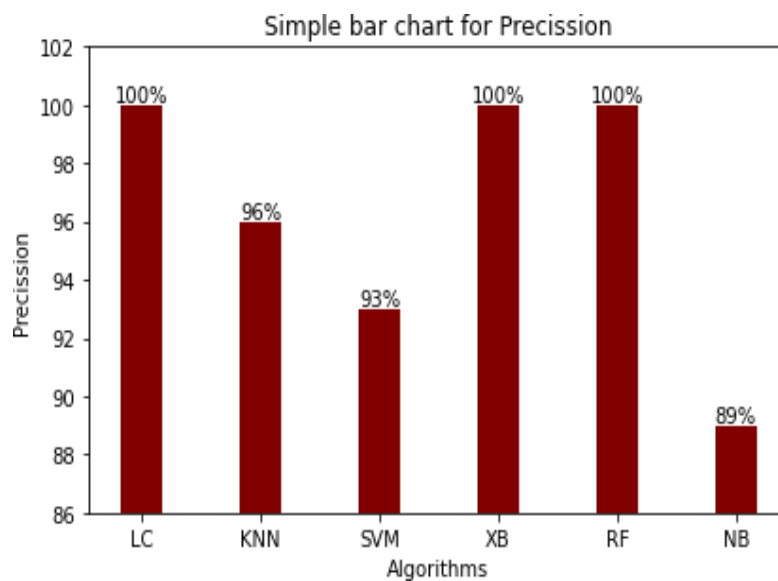


Fig 8.4.2: Precision Bar chart

Recall

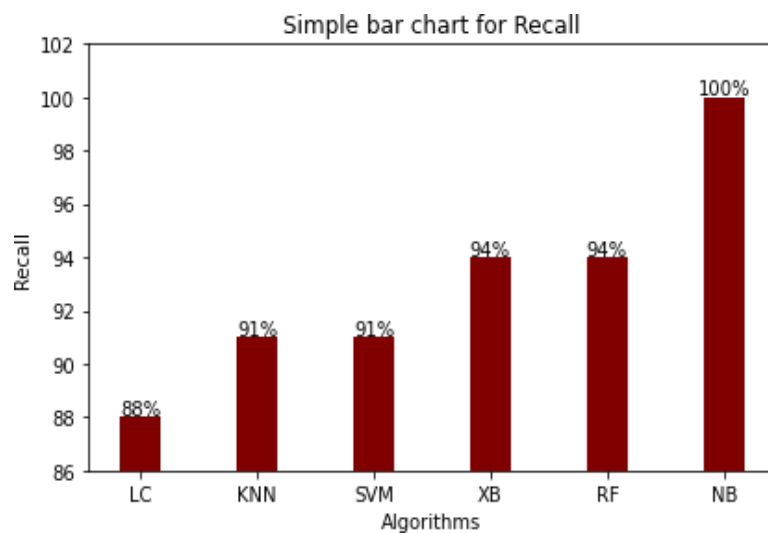


Fig 8.4.3: Recall Bar chart

F1-Score

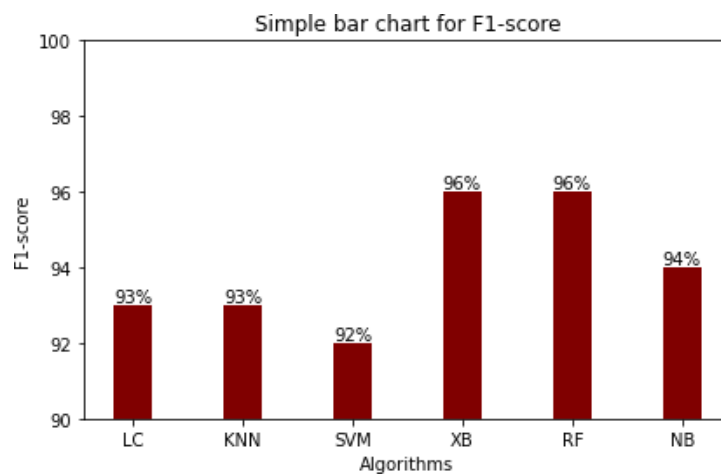


Fig 8.4.4: F1-score Bar chart

Summary

The chapter describes about various screenshot of results, step by step process to execute proposed project and advantages of proposed system

CONCLUSION

One of the disorders that affects a lot of individuals is chronic kidney disease (CKD). The disease is frequently only discovered once it has progressed to an advanced state because the symptoms of CKD are not noticeable in the early stages. This could result in kidney failure and death. Machine learning classifiers offer an effective means of early disease prediction. The Linear Classifier, Support Vector Machine (SVM), K-Nearest Neighbors (KNN), XGBoost, Naive Bayes, and Random Forest machine learning techniques were utilised. Several measures were used to assess how well these classifiers performed. The XGBoost fared better with 98 percent accuracy. However, as the dataset was a little unbalanced, accuracy cannot be the only factor taken into account when evaluating the parameter. Based on Precision, XGBoost and Random Forest both displayed 100%, whereas Linear Classifier also displayed 100%. Based on recall, XGBoost demonstrated 94%, Random Forest demonstrated 94%, and Naive Bayes demonstrated 100%. Based on the F1 score, XGBoost and Random Forest both displayed 96 percent. After examining every metric, we can say that accuracy is a good indicator if the datasets are fairly balanced and all types of outputs are valued equally. However, F1-score combines precision and recall and can be used in cases where the datasets are unbalanced because it requires both precision and recall to have a reasonable value. The formula will weigh the metric value down if the precision or recall of the positive class is low, even if there are few positive cases compared to negative cases. So, we are concluding our accurate algorithm by comparing with the metric F1-score. Based on F1-score XGBoost and Random Forest showed 96%. By carefully analyzing about the bagging and boosting algorithms, boosting algorithms can be more precise than the bagging algorithms. Based on the evaluation XG Boost was the best classifier when compared with all the other algorithms.

Therefore, it is important to investigate how the KNN imputer-based method may be used in the future to handle missing values in data sets relating to various diseases.

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