

**A Study of Liver Disease Prediction Using Machine Learning Approaches**

**BY**

**MD. ABDUR RAHMAN BHUIYAN**

**ID: 181-25-645**

This Report Presented in Partial Fulfillment of the Requirements for the  
Degree of Master of Science in Computer Science and Engineering

Supervised By

**Dr. Md. Ismail Jabiullah**

Professor

Department of Computer Science & Engineering  
Daffodil International University



**DAFFODIL INTERNATIONAL UNIVERSITY**

**DHAKA, BANGLADESH**

**06 DECEMBER, 2019**

## DECLARATION

I hereby declare that, this project has been done by me under the supervision of, **Dr. Md. Ismail Jabiullah, Department of CSE, Daffodil International University**. I also declare that neither this project nor any part of this project has been submitted elsewhere for award of any degree or diploma.

**Supervised By:**



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**Dr. Md. Ismail Jabiullah**

Professor

Department of Computer Science & Engineering  
Daffodil International University

**Submitted By:**



---

**Md. Abdur Rahman Bhuiyan**

ID: 181-25-645

Department of CSE  
Daffodil International University

## **APPROVAL**

This Project/Thesis titled “A Study of Liver Disease Prediction Using Machine Learning Approaches”, submitted by Md. Abdur Rahman Bhuiyan, ID No: 181-25-645 to the Department of Computer Science and Engineering, Daffodil International University has been accepted as satisfactory for the partial fulfillment of the requirements for the degree of M.Sc. in Computer Science and Engineering and approved as to its style and contents. The presentation has been held on 06 December, 2019.

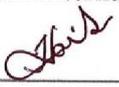
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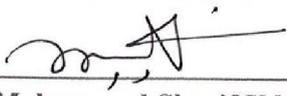
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## ABSTRACT

**Objective:** The key aspect of this study is to appraise the various Machine learning (ML) classifiers performance for forecast of liver disease.

**Methods:** In this work, I have used six supervised ML methods for the prediction of Liver Disease. For example: “SVM, NB, KNN, RF, DT and LR” were used for early prediction of Liver Disease. Therefore, I evaluated the Liver Disease dataset through sensitivity, specificity, f 1 measure and classifiers accuracy.

**Results:** The prediction performance of Liver Disease analysis shows that LR obtained the uppermost performance with utmost classification accuracy of 75%. Whereas, RF and DT has achieved the next highest accuracy by prediction.

**Conclusion:** My findings can be used to help reduce the occurrence of the Liver Disease through developing a machine learning based predictive system for early prediction.

## **CHAPTER ONE**

### **INTRODUCTION**

#### **1.1 Background**

For perdition of different disease to make the best feasible clinical care decisions it is very essential to take right decision form medical doctors. If the doctor takes their improper decisions, it could be likely to cause interruptions in medical action or drawn to loss of life. We know the medical services is a big commercial viewpoint in every time. The business stream always running in this fields. Patients are always searching there is a good platform for better services. But there is no 100% affordable platform for every patient. Therefore, in this domain need of one excessive platform for problem solves in healthcare and medical fields. Here is my main idea for improving healthcare services is to place more highlighting on early detection of chronic disease and less on treatment and live for better life.

#### **1.2 Motivation of the Research**

Liver Disease are the prominent causes of women death and disability in the global perspective. A report shows that the 508000 peoples died in 2011 by chronic disease, specially Liver Disease[1]. In 2015, around 17.7 million people were global death caused by CHD [2]. The World Health Organization (WHO) estimated that above 23.6 million peoples will be affected by 2030, because of such liver disease[3]. Very few peoples can get their treatment but most of scenario affected by chronic disease treatment is very expensive and complex [4]. Moreover, this reason to takes long time, mistaken or delayed decisions are possible to cause of death. However, the cost of Liver Disease diagnosis and replacement is very extreme and it can be calls as extreme level of financial expenses. A study reported that the Liver Disease cause the commercial benefits with cost over \$79 billion, and treating

people with end stage renal disease cost around \$35 billion [5]. Liver Disease is chronic in nature and take long time forcured. This causes most of the patients cannot afford the cost of the cure for Liver Disease. Furthermore, chronic disease prediction is most prominent matter for clinical practitioners and medical services center in order to take accurate decision of such disease. Therefore, machine learning based extensive platform can solve these Liver Disease problems through early detection and diagnosis. This works main aspect is to improve early treatment and diagnosis of Liver Disease for peoples of low-income and developing countries. Hence, our study can be a significant approach for the detecting Liver Disease outbreak with machine learning algorithms.

### **1.3 Problem Statement**

In the last 10 years, the growth rate of medical data is going to large amount from enormousarenas [6]. From the art of Machine learning (ML) algorithms have portrayed that purpose to resolve various health and scientific problem [7]. An establishment of several studies show that MLmodels already have obtained dramatically excessive accuracies in disease based medical problems. However, supervised basedmodels are one of the utmost operative method for the academic and health products on clinical fields. [8]. This works main aspect is to improve early treatment and diagnosis of chronic disease for peoples of low-income and developing countries. Hence, our study can be a significant approach for the detecting chronic disease outbreak with machine learning algorithms.

### **1.4 Research Question**

In this work we have a specific research question for Liver Disease detection by prediction using computational techniques which is addressed below:

- (1) Does the different type of ML algorithms affect the performance between the various classifiers to prediction of Liver Disease?

## **1.5 Research Scope**

To date, machine learning classification techniques have created a significant impact and obligation in the chronic disease research society for primary discovery of chronic disease. Moreover, ML algorithms are given more accurate results in chronic disease prediction as compared to others data classification techniques [8][9]. Many of studies already shows that the supervised based classification techniques have obtained excellent accuracies in the field of disease prediction [10][4][11]Motivated by this, the authors have used six prominentML techniques for prediction and proper treatment of chronic patients. The main purposeof this work is to inspect the classifiers performance measurement of various prominent supervised methods and gained more efficient outcome by reducing extremely cost of diagnosis and dialysis of chronic diseases. For this study, six supervised learning techniques were used including “KNN, Support Vector Machine, Decision Tree, Random Forest, Naïve Bayes and Logistics Regression”. Moreover, the classifiers performance of selected learning techniques is evaluated using the confusion matrix and different statistical methods. Henceforth, the outperform classification technique will donated for the decision support system and diagnosis of chronic disease.

## **1.6 Thesis Organization**

The rest of the work is ordered as following, chapter one describes the objectives of this thesis, motivation behind this thesis, research scope and thesis organization. Chapter 2 depicted the literature review and related works in these clinical areas. And the materials and methodology are described with the evaluation benchmark of different classification techniques in Section 3. Therefore, the performance results and discussion are demonstrated in section four. Finally, the conclusions and future viewpointsof research and recommendations are deliberated in section five.

## CHAPTER 2

### BACKGROUND

Through related studies were done on applying and using several ML classifiers to determine early detection and prediction of Liver Disease using ML techniques. However, the outcomes of the previous work on machine learning used in Liver Disease prediction as follows: Jain et al. [12] presented a survey to attribute assortment and machine learning techniques for identification and forecast of chronic disease. This work focused on a comprehensive review of different feature selection methods and their advantage and limitation. The contribution of this study is to use adaptive with parallel classification techniques for chronic complaint prediction. Bartz-Kurycki et al. [12], introduced a new model to forecast “neonatal surgical site infections (SSI)” using diverse classification processes. The contribution of this study is to examine the ensemble model with more medication related attributes. Carvalho et al. [13] presented a new hybrid methods to sustainance the early verdict of Liver Disease. This study tries to find optimum accuracy to provide sustenance to verdict in circumstances whereas Bayesian Network does not provide a satisfactory outcome. The contribution of the study is to advance an automatic device to contribute a precise identification and prediction of Liver Disease. Kumari [14], presented a new prediction system that can predict the occurrence of Liver Disease at early stage by analyzing nominal set of attributes that has been selected from medical datasets. The KNN classifier obtain the best performance (99.28%) than others classifiers. The contribution of this study is to use the proposed system to predict the Liver Disease at early stage with greatly reduces the cost of treatment and improves the quality of life. Tapak et al. [15], introduced a comparative study between Naïve Bayes, Random Forest, AdaBoost, Support Vector Machine, LSSVM, Adabag, Logistics Regression and LDA to predict Liver Disease survival and metastasis. “LR and LDA” were achieved the maximum accuracy of (86%). The support vector machine and “LDA” have excellent sensitivity in assessment to other classifiers. The contribution of this study is to use SVM to predict existence of Liver Disease. Asri et al. [16] presented a comparative study between SVM, DT (C4.5), NB, K-NN to predict early stage of Liver Disease. The intelligent

techniques are applied on WEKA data mining tool. Experiment results show that the SVM have the best performance accuracy, it is 97.13%. The contribution of this study to use SVM to predict the early stage of Liver Disease. Chougrad et al. [16] developed a deep convolutional neural networks based computer aided treatment system. The CNN model achieved the best performance, it is 98.94%. And they tested the CNN model on independent database and they've got the accuracy 98.23% and 0.99 AUC. The contribution of this study to use the high performer classifiers within the proposed structure and that can be used to forecast the patients are "benign or malignant". Wang et al. [17], presented a new model to use Liver Disease diagnosis based on patient's historical data from clinical data. The proposed WAUCE model reduces the variance by around 97.98% and increase accuracy by 33.34%. The contribution of this study can be further applied to safer, more reliable illness diagnosis process. Madhuri & Bharat et. Al [18], present a comparative study to diagnosis the Liver Disease patients through supervise machine learning techniques. They applied multiple machine learning algorithms including LR, RF, DT and Multi-layer perception. Multi-layer perception gives high performance compare to others algorithms. Layla & Hana [19], implement a feed forward back propagation network (FFBPN) to classify the "benign Liver Disease or malign Liver Disease". They showed that for an Artificial Neural network best design for classification is that three hidden layer and twenty-one neurons in every hidden level. Propose design gives highest accuracy 98%. Amrane et al. [20] presented a comparison between two machine learning classifier NB and KNN to provide an effectively diagnosis Liver Disease patients. The comparison result was KNN gives high accuracy at 97.51% and NB has 96.19% accuracy. Al-hadidi et al. [21], presented a model to detect the liver disease with a higher accurateness. Their model was divided into two part, first one was processing the images for extract the features and second one was used two supervise machine learning techniques to get the accuracy. Xiao et al. [22], introduce a new strategy for gene expression analysis to five different classification algorithms with deep learning methods. The contribution of this study is shown to be accurate and effective prediction results for Liver Disease prediction.

## CHAPTER THREE

### MATERIALS AND METHODS

#### 3.1 Experimental Setup

In this work, this section represents the working process (figure 3.1.1) of the experiment including machine learning techniques. Liver Disease data sets have been considered in this work. Firstly, we focused on preparing and combined data from the main datasets. Moreover, I extracted features from the liver datasets. Then, we have checked the missing values and correlated values. Secondly, Data set splitting is an important task of this machine learning based fields. In this dataset, I have not found split and test datasets. Figure 3.1.1 shows the Parkinson data set has split into trainset and test sets. After that, 6 supervised based classifiers performed the operation. After successfully executed these algorithms LR obtained the highest performance.

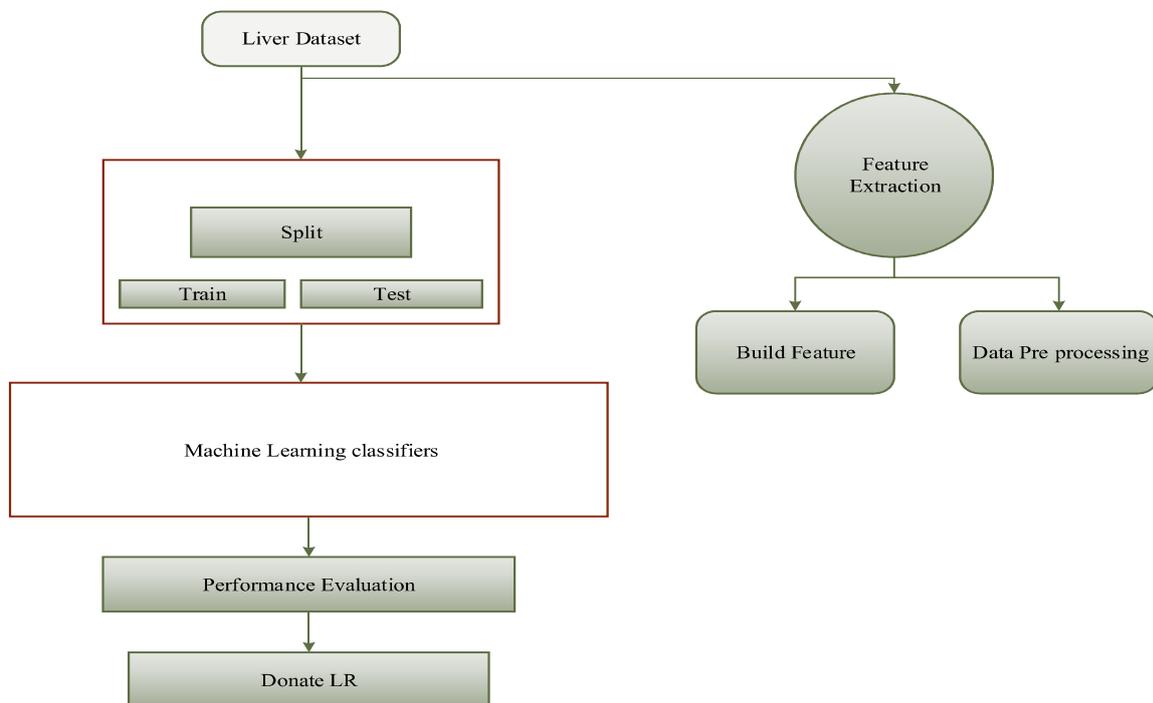


Figure 3.1.1. The experimental setup

### 3.2 Data Collection

In this study, we use the liver patient data from provided by the “University of California, Irvine”[23]. The objective of the dataset is to diagnostically predict whether or not a patient has liver disease based on certain diagnostic parameters. Moreover, in the dataset contains 583 liver patient’s data whereas 416 samples are liver patients and 167 samples are non-liver patient. This data set contains 75.64% samples are male patients and 24.36% samples are female patient. However, I chose the particular parameters for data analysis which are depicted in table 3.2. 1

Table 3.2.1 Parameters for data analysis

No	Attributes	Indication	Description
1	Age	Numerical	Years
2	Gender	Male or Female	Sex of the patients
3	Total_Bilirubin	Numerical	mg/dL
4	Direct_Bilirubin	Numerical	mg/dL
5	Alkaline_Phosphotase	Numerical	ALP in IU/L
6	Alamine_Aminotransferase	Numerical	ALT in IU/L
7	Aspartate_Aminotransferase	Numerical	AST in IU/L
8	Total_Protiens	Numerical	g/dL
9	Albumin	Numerical	g/dL
10	Albumin_and_Globulin_Ratio	Numerical	A/G ratio
11	Dataset	Yes or No	patient has liver disease or not

### 3.3. Data Pre-Processing

Our dataset contains 416 samples diagnosed with liver disease and 167 samples not diagnosed with liver disease. Figure 3.3.1 shown that the ratio of the liver patients. In this dataset, there is most of the patient are male, Fig 3.3.2 shown that the ratio of gander the liver patients (the number of male patients is 441 and 142 are female).

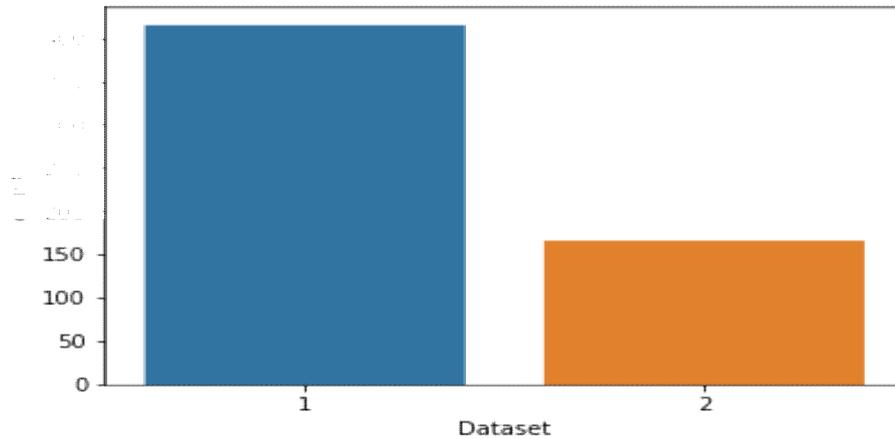


Figure 3.3.1 Count plot shows the ratio of the liver patients

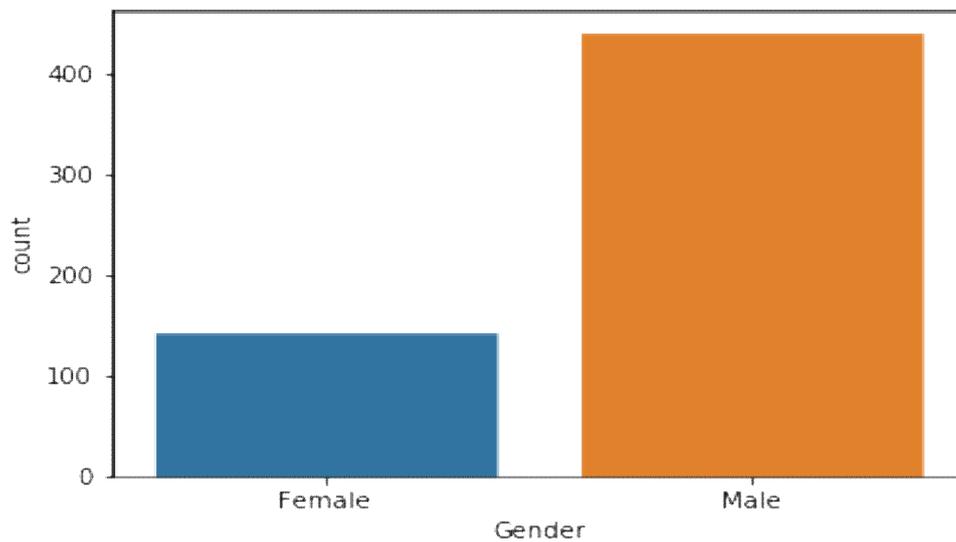


Figure 3.3.2 Count plot shows the ratio of gender the liver patients

This dataset has ten features column and output column. In the output column, value (1) specifies that the sample has a negative condition and '0' indicates the sample has a positive condition. Therefore, we examine between columns to find the correlated column (figure 3.3.3), then we found some correlated features from a heat map.

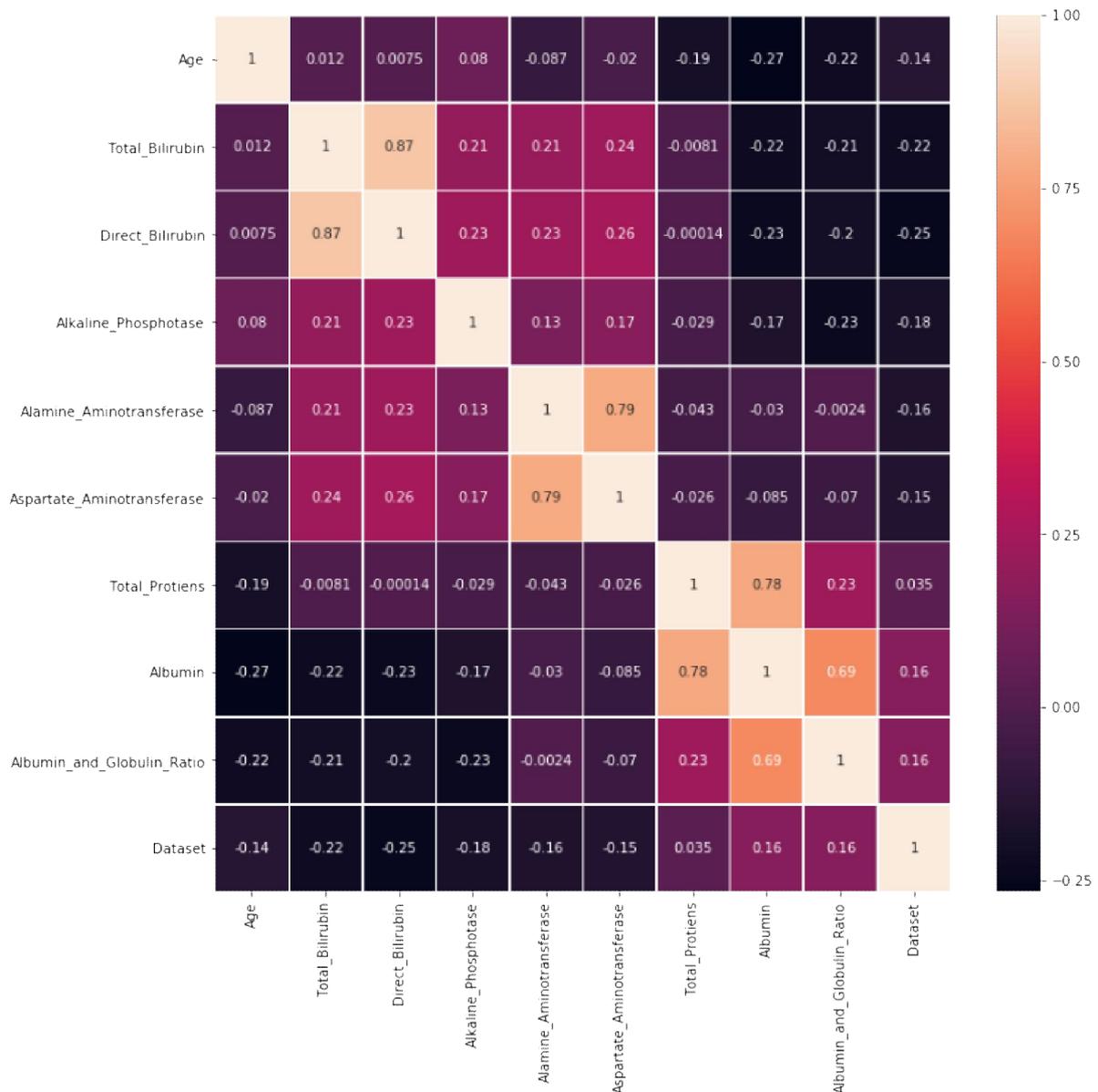


Figure 3.3.3. Heat map for checking correlated columns for Liver Diseases

### 3.4 Evaluation Criteria

In this thesis, we have depleted 6 ML methods for the forecast of Liver Disease. Therefore, the performance measurements of the ML methods are appraised by diverse statistical measurement techniques. For example, we have chosen the confusion\_matrix (True\_Positive, False\_Positive, True\_Negative, False\_Negative), Recall, Precision, f1-measure etc. [24].

The calculation approach of the measurement techniques are following,

$$\text{Accuracy}_i = (TP_i + TN_i) / (TP_i + FP_i + TN_i + FN_i) \quad (1)$$

$$\text{TPR}_i \text{ or Sensitivity}_i \text{ or Recall}_i = TP_i / (TP_i + FN_i) \quad (2)$$

$$\text{Specificity}_i = TN_i / (TN_i + FP_i) \quad (3)$$

$$\text{Precision}_i = TP_i / (TP_i + FP_i) \quad (4)$$

$$f1_i = 2 * (\text{Recall}_i * \text{Precision}_i) / (\text{Recall}_i + \text{Precision}_i) \quad (5)$$

$$\text{False Positive Rate} = 1 - \text{Specificity}_i \quad (6)$$

The f1\_measure is denoted by the weighted norm of the recall<sub>i</sub> and precision<sub>i</sub>. To classify as a better classifier this the value will be 1 and for the lowest performance, it will be 0.

### 3.5 Software and Tools

This work was employed in prominent programming language, i.e. python version 3.7.1 and using Anaconda Distribution including Jupyter Notebook. The version of the notebook server is: 5.6.0-3badce9.

## CHAPTER 4

### RESULT AND DISCUSSION

#### 4.1 Analysis of the Results

In this thesis, I have accompanied several analyses to examine the six ML based supervised techniques for diagnosis and forecast of Liver Disease. The performance comparison and performance measure of six machine learning classifiers for chronic disease prediction as detailed in the following.

In this research, we have taken some investigation to inspect the performance of six ML methods for the detection of liver disease. Figure 4.1.1 indicates that the performance analysis of six supervised based approaches for liver disease data. With respect to precision LR achieved the highest score, 91% and NB perform worst (i.e. 0.36%). However, when considering the sensitivity, SVM attained the uppermost value (i.e. 0.88%) and KNN obtained the poorest value (i.e. 0.76%). Logistics Regression was also the best performer in terms of f1 measure (“i.e. 0.83%”), and NB gained the poor performance (i.e. 0.53%). In the terms of accuracy, LR attained the maximum accuracy (i.e. 0.75%), and NB accomplished the bad performance (i.e. 0.53%). The LR suggests that this ML method is more operational than the other algorithms for predicting Liver Disease. Moreover, the confusion matrix of this analysis is presented in figure 4.1.2.

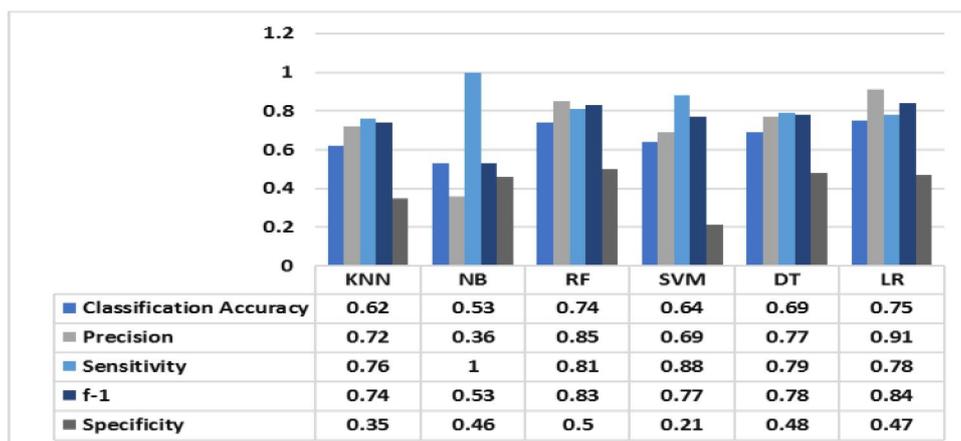


Figure 4.1.1 The performance comparison of six supervised learning techniques

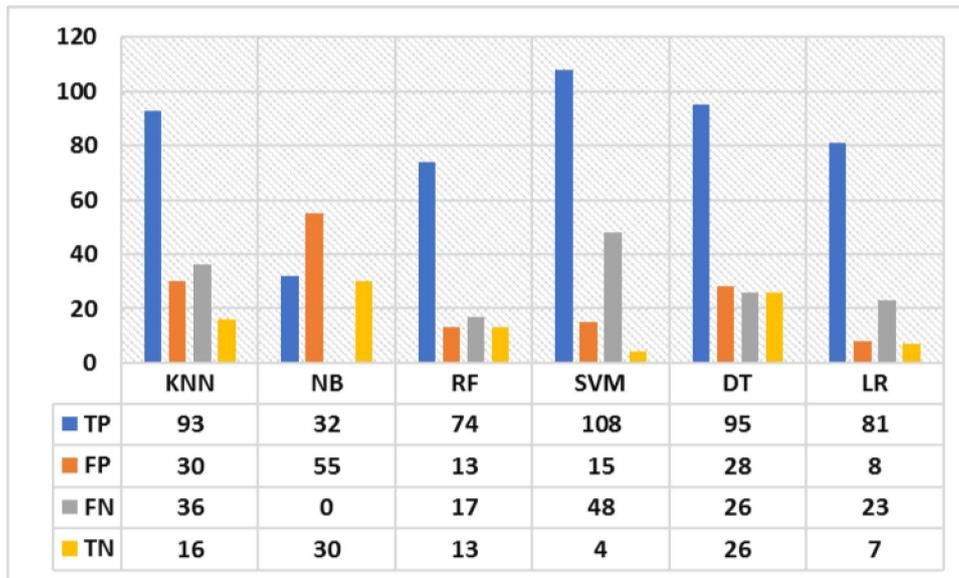


Figure 4.1.2 The confusion matrix of prediction results

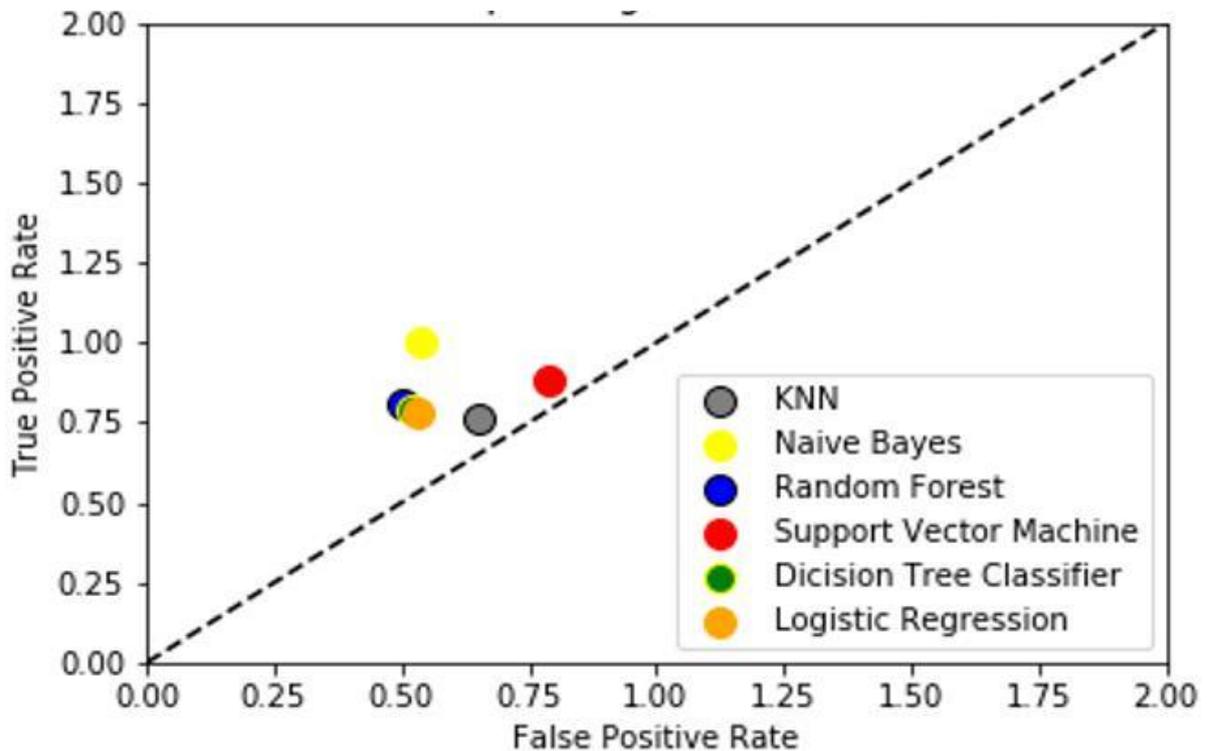


Figure 4.1.3 Receiver Operating Characteristics curve for Liver datasets

The prediction result shows the classifiers outcome above 70% for Liver Disease detection. Moreover, the six machine learning algorithms are doing very good for Liver Disease prediction. In our experiment, it is very essential to recognize the “receiver operating

characteristics (ROC) curve”, which is justified on “true positive rate (TPR) and false positive rate (FPR)” of these prediction outcomes. According to ‘ROC curve’ (figure 4.1.3.) achieved highest AUC (area under curve) for ROC.

## **CHAPTER 5**

### **CONCLUSION AND RECOMMENDATION**

#### **5.1 Findings and Contribution**

In this study, I have depicted several ML based classification techniques. Therefore, I deliver an experimental process on ML based system for the early prediction of Liver Disease. Therefore, I completed the presentation of the six algorithms which are depleted in the forecast of Liver Disease diseases and assessed by their results using a statistical technique namely is confusion matrix. The experimental performance shows that the NB has achieved the outperform than the other classifiers within Liver Disease datasets. This inspection has usage six ML methods for the prediction of liver disease with some clinical attributes. In addition, this study is part of a project that has the purpose to create an advanced tool to give more precise treatment to normal events and make a superior decision to complex situations. The application will be able to early detect in Liver Disease in a few minutes and notify the real

condition with extreme likelihood of having disease. This application can be remarkably beneficial in low-income countries where is lack of medical institutions and as well as specialized practitioners.

## **5.2 Recommendation for Future Work**

In my experiments, related to most work in the study, each classification algorithms were trained and evaluated on a training dataset that comprises both of positive values and negative values. Moreover, the work can be helpful for chronic disease diagnosis and detection by collecting data from different devices and health related sensors, clinical and medical center and can deliver more accurate results for disease prediction and diagnosis. In my research perspective, there are several directions for the future work in this area of research. I only investigated to some popular supervised machine learning algorithms, it can be choosing more algorithm for build the accurate model of these chronic disease prediction and performance can be more improved. Additionally, I have underlined the study trend and possibility in comparative to liver disease research and medical data analysis through ML based techniques.

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## **Appendix**

### **Appendix A: Liver Disease Dataset**

#### **Sample of Indian Liver Patient Data**

Age	Gender	Total_Bilir	Direct_Bil	Alkaline_P	Alamine_P	Aspartate	Total_Prof	Albumin	Albumin_	Dataset
65	Female	0.7	0.1	187	16	18	6.8	3.3	0.9	1
62	Male	10.9	5.5	699	64	100	7.5	3.2	0.74	1
62	Male	7.3	4.1	490	60	68	7	3.3	0.89	1
58	Male	1	0.4	182	14	20	6.8	3.4	1	1
72	Male	3.9	2	195	27	59	7.3	2.4	0.4	1
46	Male	1.8	0.7	208	19	14	7.6	4.4	1.3	1
26	Female	0.9	0.2	154	16	12	7	3.5	1	1
29	Female	0.9	0.3	202	14	11	6.7	3.6	1.1	1
17	Male	0.9	0.3	202	22	19	7.4	4.1	1.2	2
55	Male	0.7	0.2	290	53	58	6.8	3.4	1	1
57	Male	0.6	0.1	210	51	59	5.9	2.7	0.8	1
72	Male	2.7	1.3	260	31	56	7.4	3	0.6	1
64	Male	0.9	0.3	310	61	58	7	3.4	0.9	2
74	Female	1.1	0.4	214	22	30	8.1	4.1	1	1
61	Male	0.7	0.2	145	53	41	5.8	2.7	0.87	1
25	Male	0.6	0.1	183	91	53	5.5	2.3	0.7	2
38	Male	1.8	0.8	342	168	441	7.6	4.4	1.3	1
33	Male	1.6	0.5	165	15	23	7.3	3.5	0.92	2
40	Female	0.9	0.3	293	232	245	6.8	3.1	0.8	1
40	Female	0.9	0.3	293	232	245	6.8	3.1	0.8	1
51	Male	2.2	1	610	17	28	7.3	2.6	0.55	1
51	Male	2.9	1.3	482	22	34	7	2.4	0.5	1
62	Male	6.8	3	542	116	66	6.4	3.1	0.9	1
40	Male	1.9	1	231	16	55	4.3	1.6	0.6	1

# A Study of Liver Disease Prediction Using Machine Learning Approaches

*by* Abdur Rahman Bhuiyan

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