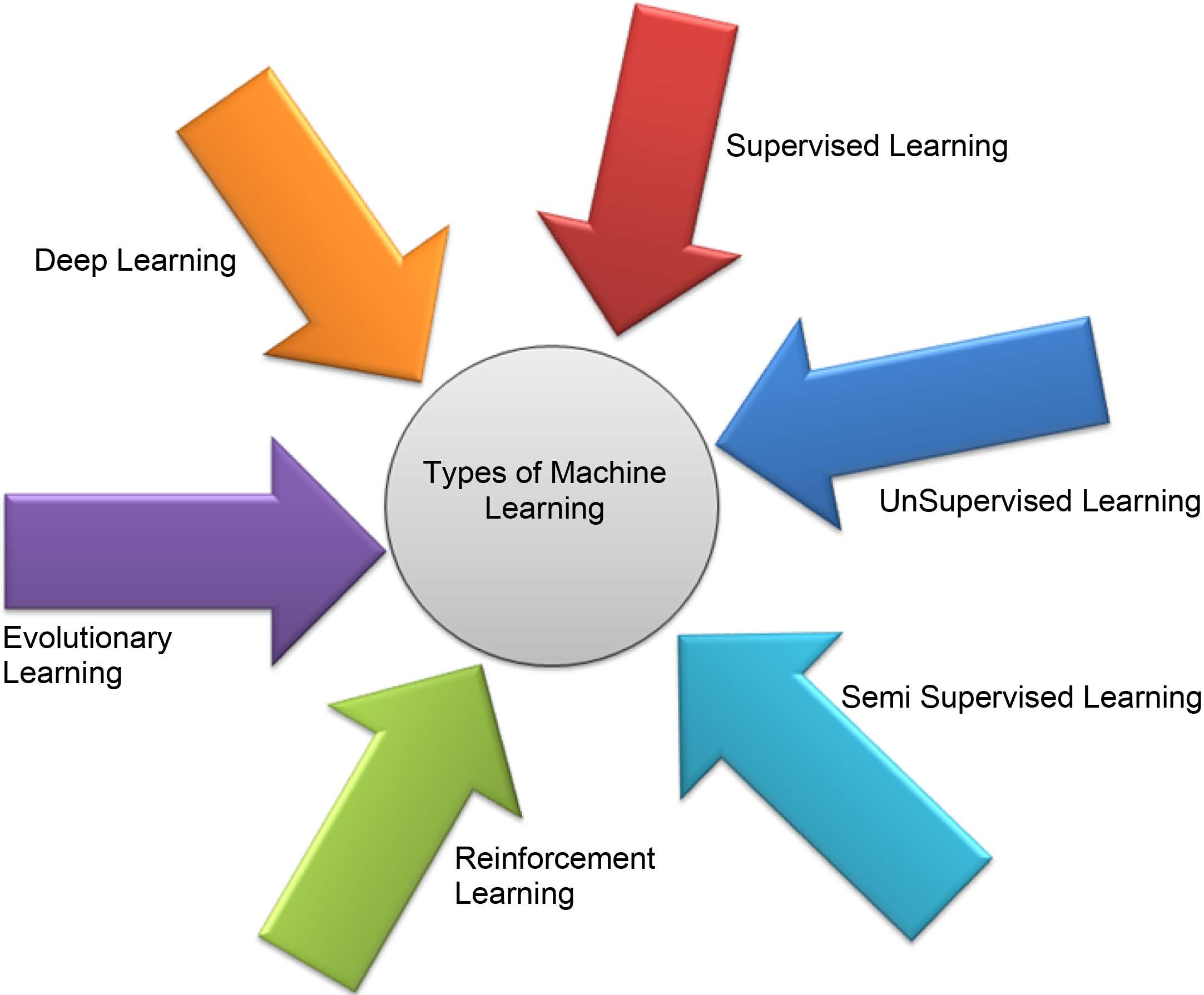
Artificial Intelligence can enable the computer to think. Computer is made much more intelligent by AI. Machine learning is the subfield of AI study. Various re- searchers think that without learning, intelligence cannot be developed. There are many types of Machine Learning Techniques that are shown in [**Figure 1**.](#_bookmark0) Su- pervised, Unsupervised, Semi Supervised, Reinforcement, Evolutionary Learning

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**Figure 1.** Types of machine learning techniques.

and Deep Learning are the types of machine learning techniques. These tech- niques are used to classify the data set.

* 1. Supervised learning: Offered a training set of examples with suitable targets and on the basis of this training set, algorithms respond correctly to all feasible inputs. Learning from exemplars is another name of Supervised Learning. Clas- sification and regression are the types of Supervised Learning.

Classification: It gives the prediction of Yes or No, for example, “Is this tumor cancerous?”, “Does this cookie meet our quality standards?”

Regression: It gives the answer of “How much” and “How many”.

* 1. Unsupervised learning: Correct responses or targets are not provided. Un- supervised learning technique tries to find out the similarities between the input data and based on these similarities, un-supervised learning technique classify the data. This is also known as density estimation. Unsupervised learning con- tains clustering [[1].](#_bookmark12)

Clustering: it makes clusters on the basis of similarity.

* 1. Semi supervised learning: Semi supervised learning technique is a class of supervised learning techniques. This learning also used unlabeled data for train- ing purpose (generally a minimum amount of labeled-data with a huge amount of unlabeled-data). Semi-supervised learning lies between unsupervised-learning (unlabeled-data) and supervised learning (labeled-data).
  2. Reinforcement learning: This learning is encouraged by behaviorist psy- chology. Algorithm is informed when the answer is wrong, but does not inform that how to correct it. It has to explore and test various possibilities until it finds the right answer. It is also known as learning with a critic. It does not recom- mend improvements. Reinforcement learning is different from supervised learn-

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ing in the sense that accurate input and output sets are not offered, nor sub- optimal actions clearly précised. Moreover, it focuses on on-line performance.

* 1. Evolutionary Learning: This biological evolution learning can be consi- dered as a learning process: biological organisms are adapted to make progress in their survival rates and chance of having off springs. By using the idea of fit- ness, to check how accurate the solution is, we can use this model in a computer [[1].](#_bookmark12)
  2. Deep learning: This branch of machine learning is based on set of algo- rithms. In data, these learning algorithms model high-level abstraction. It uses deep graph with various processing layer, made up of many linear and nonlinear transformation.

Pattern recognition process and data classification are valuable for a long time. Humans have very strong skill for sensing the environment. They take action against what they perceive from environment [[2].](#_bookmark13) Big data turns into Chunks due to multidisciplinary combined effort of machine learning, databases and statistics. Today, in medical sciences disease diagnostic test is a serious task. It is very important to understand the exact diagnosis of patients by clinical ex- amination and assessment. For effective diagnosis and cost effective manage- ment, decision support systems that are based upon computer may play a vital role. Health care field generates big data about clinical assessment, report re- garding patient, cure, follow-ups, medication etc. It is complex to arrange in a suitable way. Quality of the data organization has been affected due to inappro- priate management of the data. Enhancement in the amount of data needs some proper means to extract and process data effectively and efficiently [[3].](#_bookmark14) One of the many machine-learning applications is employed to build such classifier that can divide the data on the basis of their attributes. Data set is divided into two or more than two classes. Such classifiers are used for medical data analysis and disease detection.

Initially, algorithms of ML were designed and employed to observe medical data sets. Today, for efficient analysis of data, ML recommended various tools. Especially in the last few years, digital revolution has offered comparatively low- cost and obtainable means for collection and storage of data. Machines for data collection and examination are placed in new and modern hospitals to make them capable for collection and sharing data in big information systems. Technologies of ML are very effective for the analysis of medical data and great work is done regarding diagnostic problems. Correct diagnostic data are presented as a medi- cal record or reports in modern hospitals or their particular data section. To run an algorithm, correct diagnostic patient record is entered in a computer as an input. Results can be automatically obtained from the previous solved cases. Phy- sicians take assistance from this derived classifier while diagnosing novel patient at high speed and enhanced accuracy. These classifiers can be used to train non- specialists or students to diagnose the problem [[4].](#_bookmark15)

In past, ML has offered self-driving cars, speech detection, efficient web search, and improved perception of the human generation. Today machine learning is

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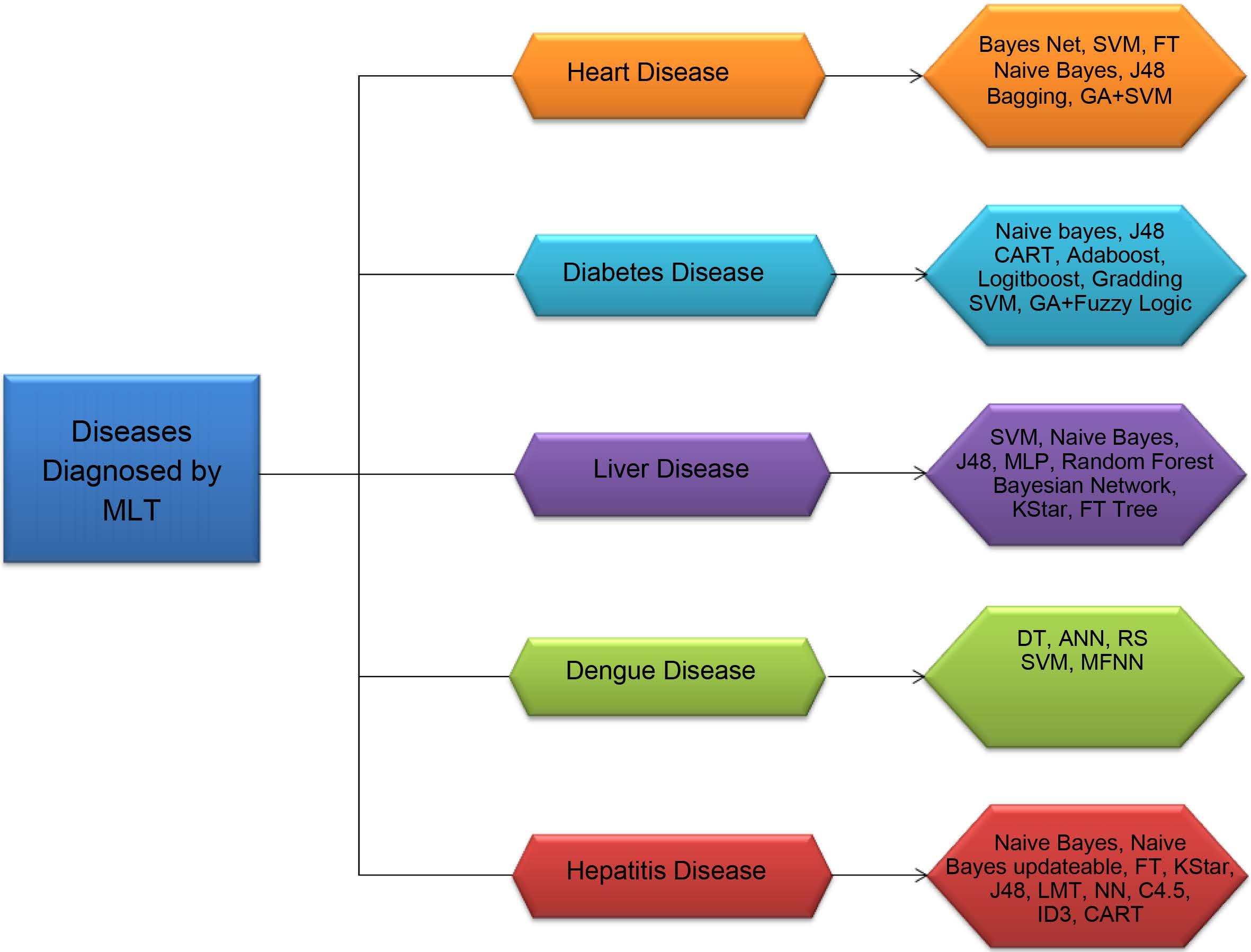
present everywhere so that without knowing it, one can possibly use it many times a day. A lot of researchers consider it as the excellent way in moving to- wards human level. The machine learning techniques discovers electronic health record that generally contains high dimensional patterns and multiple data sets. Pattern recognition is the theme of MLT that offers support to predict and make decisions for diagnosis and to plan treatment. Machine learning algorithms are capable to manage huge number of data, to combine data from dissimilar re- sources, and to integrate the background information in the study [[3].](#_bookmark14)

# Diagnosis of Diseases by Using Different Machine Learning Algorithms

Many researchers have worked on different machine learning algorithms for disease diagnosis. Researchers have been accepted that machine-learning algo- rithms work well in diagnosis of different diseases. Figurative approach of dis- eases diagnosed by Machine Learning Techniques is shown in [**Figure 2**.](#_bookmark1) In this survey paper diseases diagnosed by MLT are heart, diabetes, liver, dengue and hepatitis.

## Heart Disease

Otoom *et al*. [[5]](#_bookmark16) presented a system for the purpose of analysis and monitoring. Coronary artery disease is detected and monitored by this proposed system. Cleveland heart data set is taken from UCI. This data set consists of 303 cases and 76 attributes/features. 13 features are used out of 76 features. Two tests with three algorithms Bayes Net, Support vector machine, and Functional Trees FT are performed for detection purpose. WEKA tool is used for detection. After



**Figure 2.** Diseases diagnosed by MLT.

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experimenting Holdout test, 88.3% accuracy is attained by using SVM technique. In Cross Validation test, Both SVM and Bayes net provide the accuracy of 83.8%. 81.5% accuracy is attained after using FT. 7 best features are picked up by using Best First selection algorithm. For validation Cross Validation test are used. By applying the test on 7 best selected features, Bayes Net attained 84.5% of cor- rectness, SVM provides 85.1% accuracy and FT classify 84.5% correctly.

Vembandasamy *et al*. [[6]](#_bookmark17) performed a work, to diagnose heart disease by using Naive Bayes algorithm. Bayes theorem is used in Naive Bayes. Therefore, Naive Bayes have powerful independence assumption. The employed data-set are ob- tained from one of the leading diabetic research institute in Chennai. Data set consists of 500 patients. Weka is used as a tool and executes classification by us- ing 70% of Percentage Split. Naive Bayes offers 86.419% of accuracy.

Use of data mining approaches has been suggested by Chaurasia and Pal [[7]](#_bookmark18) for heart disease detection. WEKA data mining tool is used that contains a set of machine learning algorithms for mining purpose. Naive Bayes, J48 and bagging are used for this perspective. UCI machine learning laboratory provide heart disease data set that consists of 76 attributes. Only 11 attributes are employed for prediction. Naive bayes provides 82.31% accuracy. J48 gives 84.35% of correct- ness. 85.03% of accuracy is achieved by Bagging. Bagging offers better classifica- tion rate on this data set.

Parthiban and Srivatsa [[8]](#_bookmark19) put their effort for diagnosis of heart disease in di- abetic patients by using the methods of machine learning. Algorithms of Naive Bayes and SVM are applied by using WEKA. Data set of 500 patients is used that are collected from Research Institute of Chennai. Patients that have the disease are 142 and disease is missing in 358 patients. By using Naive Bayes Algorithm 74% of accuracy is obtained. SVM provide the highest accuracy of 94.60.

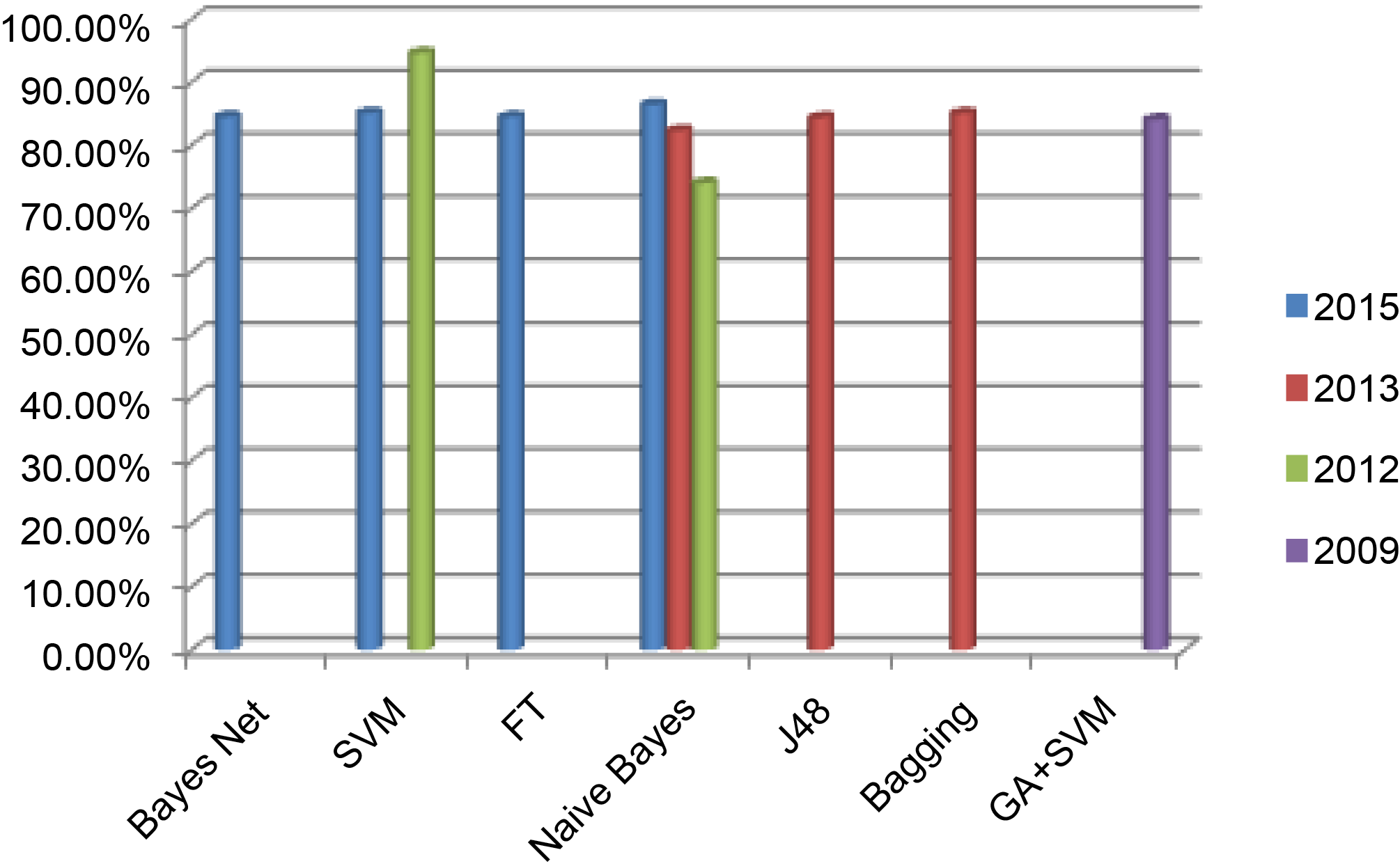
Tan *et al*. [[9]](#_bookmark20) proposed hybrid technique in which two machine-learning algo- rithms named Genetic Algorithm (G.A) and Support Vector Machine (SVM) are joined effectively by using wrapper approach. LIBSVM and WEKA data mining tool are used in this analysis. Five data sets (Iris, Diabetes disease, disease of breast Cancer, Heart and Hepatitis disease) are picked up from UC Irvine machine learning repository for this experiment. After applying GA and SVM hybrid ap- proach, 84.07% accuracy is attained for heart disease. For data set of diabetes 78.26% accuracy is achieved. Accuracy for Breast cancer is 76.20%. Correctness of 86.12% is resulting for hepatitis disease. Graphical representation of Accuracy according to time for detection of heart disease is shown in [**Figure 3**.](#_bookmark2)

### Analysis:

In existing literature, SVM offers highest accuracy of 94.60% in 2012 as in [**Ta-**](#_bookmark3)[**ble 1**.](#_bookmark3) In many application areas, SVM shows good performance result. Attribute or features used by Parthiban and Srivatsa in 2012 are correctly responded by SVM. In 2015, Otoom *et al*. used SVM variant called SMO. It also uses FS tech- nique to find best features. SVM responds to these features and offers the accu- racy of 85.1% but it is comparatively low as in 2012. Training and testing set of both data sets are different, as well as, data types are different.

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**Figure 3.** Machine learning algorithm’s accuracy to detect heart disease.

**Table 1.** Comprehensive view of machine learning techniques for heart disease diagnosis.

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Machine Learning Techniques** | **Author** | **Year** | **Disease** | **Resources of Data Set** | **Tool** | **Accuracy** |
| Bayes Net |  |  |  |  |  | 84.5% |

SVM

CAD (Coronary artery disease)

UCI WEKA

85.1%

FT 84.5%

Otoom *et al*. 2015

Naive Bayes Vembandasamy *et al*. 2015 Heart Disease Diabetic Research Institute Naive Bayes

in Chennai

WEKA 86.419%

82.31%

J48

Chaurasia and Pal 2013 Heart Disease UCI WEKA

84.35%

Bagging 85.03%

SVM Parthiban and Srivatsa 2012 Heart disease Research institute in

94.60%

WEKA

Naive Bayes

Chennai

74%

Hybrid Technique (GA + SVM) Tan *et al*. 2009 Heart disease UCI LIBSVM and WEKA 84.07%

### Advantages and Disadvantages of SVM:

Advantages: Construct correct classifiers and fewer over ﬁtting, robust to noise.

Disadvantages: It is a binary classiﬁer. For the classification of multi-class, it can use pair wise classiﬁcation. Its Computational cost is high, so it runs slow [[10].](#_bookmark21)

## Diabetes Disease

Iyer *et al*. [[11]](#_bookmark22) has performed a work to predict diabetes disease by using deci- sion tree and Naive Bayes. Diseases occur when production of insulin is insuffi- cient or there is improper use of insulin. Data set used in this work is Pima In- dian diabetes data set. Various tests were performed using WEKA data mining tool. In this data-set percentage split (70:30) predict better than cross validation. J48 shows 74.8698% and 76.9565% accuracy by using Cross Validation and Per- centage Split Respectively. Naive Bayes presents 79.5652% correctness by using PS. Algorithms shows highest accuracy by utilizing percentage split test.

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Meta learning algorithms for diabetes disease diagnosis has been discussed by Sen and Dash [[12].](#_bookmark23) The employed data set is Pima Indians diabetes that is re- ceived from UCI Machine Learning laboratory. WEKA is used for analysis. CART, Adaboost, Logiboost and grading learning algorithms are used to predict that patient has diabetes or not. Experimental results are compared on the behalf of correct or incorrect classification. CART offers 78.646% accuracy. The Adaboost obtains 77.864% exactness. Logiboost offers the correctness of 77.479%. Grading has correct classification rate of 66.406%. CART offers highest accuracy of 78.646% and misclassification Rate of 21.354%, which is smaller as compared to other techniques.

An experimental work to predict diabetes disease is done by the Kumari and Chitra [[13].](#_bookmark24) Machine learning technique that is used by the scientist in this ex- periment is SVM. RBF kernel is used in SVM for the purpose of classification. Pima Indian diabetes data set is provided by machine learning laboratory at University of California, Irvine. MATLAB 2010a are used to conduct experiment. SVM offers 78% accuracy.

Sarwar and Sharma [[14]](#_bookmark25) have suggested the work on Naive Bayes to predict diabetes Type-2. Diabetes disease has 3 types. First type is Type-1 diabetes, Type-2 diabetes is the second type and third type is gestational diabetes. Type-2 diabetes comes from the growth of Insulin resistance. Data set consists of 415 cases and for purpose of variety; data are gathered from dissimilar sectors of society in In- dia. MATLAB with SQL server is used for development of model. 95% correct prediction is achieved by Naive Bayes.

Ephzibah [[15]](#_bookmark26) has constructed a model for diabetes diagnosis. Proposed mod- el joins the GA and fuzzy logic. It is used for the selection of best subset of fea- tures and also for the enhancement of classification accuracy. For experiment, dataset is picked up from UCI Machine learning laboratory that has 8 attributes and 769 cases. MATLAB is used for implementation. By using genetic algorithm only three best features/attributes are selected. These three attributes are used by fuzzy logic classifier and provide 87% accuracy. Around 50% cost is less than the original cost. [**Table 2**](#_bookmark4) provides the Comprehensive view of Machine learning Tech- niques for diabetes disease diagnosis.

### Analysis:

Naive Bayes based system is helpful for diagnosis of Diabetes disease. Naive Bayes offers highest accuracy of 95% in 2012. The results show that this system can do good prediction with minimum error and also this technique is impor- tant to diagnose diabetes disease. But in 2015, accuracy offered by Naive Bayes is low. It presents 79.5652% or 79.57% accuracy. This proposed model for detec- tion of Diabetes disease would require more training data for creation and test- ing. [**Figure 4**](#_bookmark5) shows the Accuracy graph of Algorithms for the diagnosis of Di- abetes disease according to time.

### Advantages and Disadvantages of Naive Bayes:

Advantages: It enhances the classification performance by eliminating the un- related features. Its performance is good. It takes less computational time.

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**Table 2.** Comprehensive view of machine learning techniques for diabetes disease diagnosis.

**Machine Learning Techniques Author Year Disease Resource of Data Set Tool Accuracy**

Naive Bayes

Iyer *et al*. 2015 Diabetes Disease Pima Indian Diabetes dataset WEKA

79.5652%

J48 76.9565%

CART 78.646%

Adaboost 77.864%

Sen and Dash 2014 Diabetes Disease Pima Indian Diabetes dataset from UCI WEKA

Logiboost 77.479%

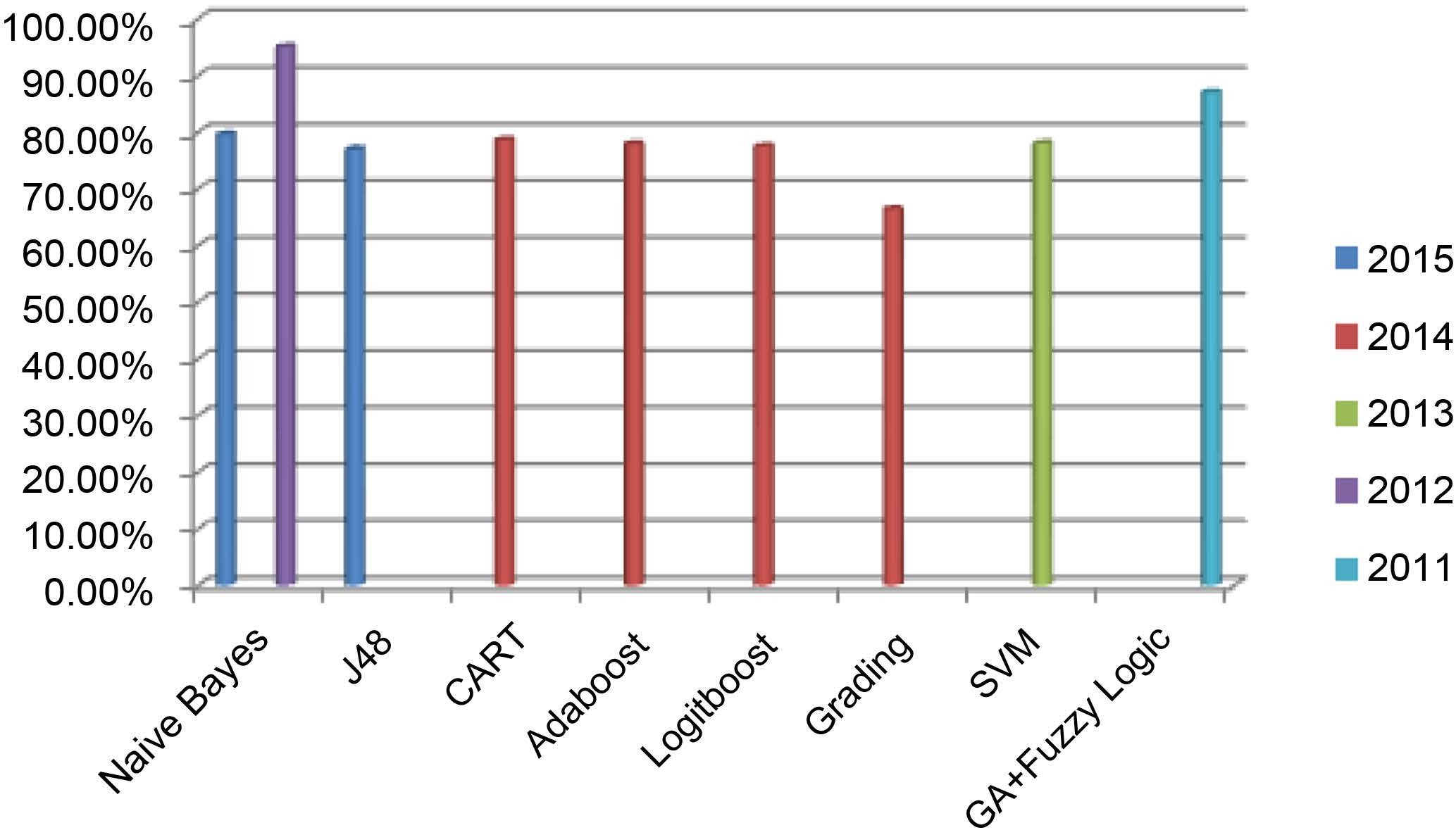
Grading 66.406%

SVM Kumari and Chitra 2013 Diabetes Disease UCI MATLAB 2010a 78%

Naive Bayes Sarwar and Sharma 2012 Diabetes type-2 Different Sectors of Society in India MATLAB with SQL

Server

95%

GA + Fuzzy Logic Ephzibah 2011 Diabetes disease UCI MATLAB 87%

**Figure 4.** Accuracy of machine learning algorithms to detect diabetes disease.

Disadvantages: This algorithm needs large amount of data to attain good out- comes. It is lazy as they store entire the training examples [[16].](#_bookmark27)

## Liver Disease

Vijayarani and Dhayanand [[17]](#_bookmark28) predict the liver disease by using Support vector machine and Naive bayes Classification algorithms. ILPD data set is obtained from UCI. Data set comprises of 560 instances and 10 attributes. Comparison is made on the basis of accuracy and time execution. Naive bayes shows 61.28% correctness in 1670.00 ms. 79.66% accuracy is attained in 3210.00 ms by SVM. For implementation, MATLAB is used. SVM shows highest accuracy as com- pared to the Naive bayes for liver disease prediction. In terms of time execution, Naives bayes takes less time as compared to the SVM.

A study on intelligent techniques to classify the liver patients is performed by the Gulia *et al*. [[18].](#_bookmark29) Used data set is picked up from UCI. WEKA data mining tool and five intelligent techniques J48, MLP, Random Forest, SVM and Baye- sian Network classifiers are used in this experiment. In first step, all algorithms are applied on the original data set and get the percentage of correctness. In

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second step, feature selection method is applied on whole data-set to get the sig- nificant subset of liver patients and all these algorithms are used to test the sub- set of whole data-set. In third step they take comparison of outcomes before and after feature selection. After FS, algorithms provide highest accuracy as J48 presents 70.669% accuracy, 70.8405% exactness is achieved by the MLP algorithm, SVM provides 71.3551% accuracy, 71.8696% accuracy is offered by Random forest and Bayes Net shows 69.1252% accuracy.

Rajeswari and Reena [[19]](#_bookmark30) used the data mining algorithms of Naive Bayes, K star and FT tree to analyze the liver disease. Data set is taken from UCI that comprises of 345 instances and 7 attributes. 10 cross validation test are applied by using WEKA tool. Naive Bayes provide 96.52% Correctness in 0 sec. 97.10% accuracy is achieved by using FT tree in 0.2 sec. K star algorithm classify the in- stances about 83.47% accurately in 0 sec. On the basis of outcomes, highest clas- sification accuracy is offered by FT tree on liver disease dataset as compared to other data mining algorithms. [**Table 3**](#_bookmark6) presents the comprehensive view of algo- rithms for the detection of liver disease.

### Analysis:

To diagnose liver disease, FT Tree Algorithm provides the highest result as compare to the other algorithms. When FT tree algorithm is applied on the dataset of liver disease, time taken for result or building the model is fast as compared to other algorithms. According to its attribute, it shows the improved performance. This algorithm fully classified the attributes and offers 97.10% correctness. From the results, this Algorithm plays an important role in determining enhanced classi- fication accuracy of data set. Accuracy graph of algorithms are shown in [**Figure 5**.](#_bookmark7)

### Advantages and Disadvantages of FT:

Advantage: Easy to interpret and understand; Fast prediction.

Disadvantage: Calculations are complex mainly if values are uncertain or if several outcomes are linked.

## Dengue Disease

Tarmizi *et al*. [[20]](#_bookmark31) performed a work for Malaysia Dengue Outbreak Detection

**Table 3.** Comprehensive view of machine learning techniques for liver disease diagnosis.

**Machine Learning Techniques Author Year Disease Resource of Data Set Tool Accuracy**

SVM

Vijayarani and Dhayanand 2015 Liver Disease ILPD from UCI MATLAB

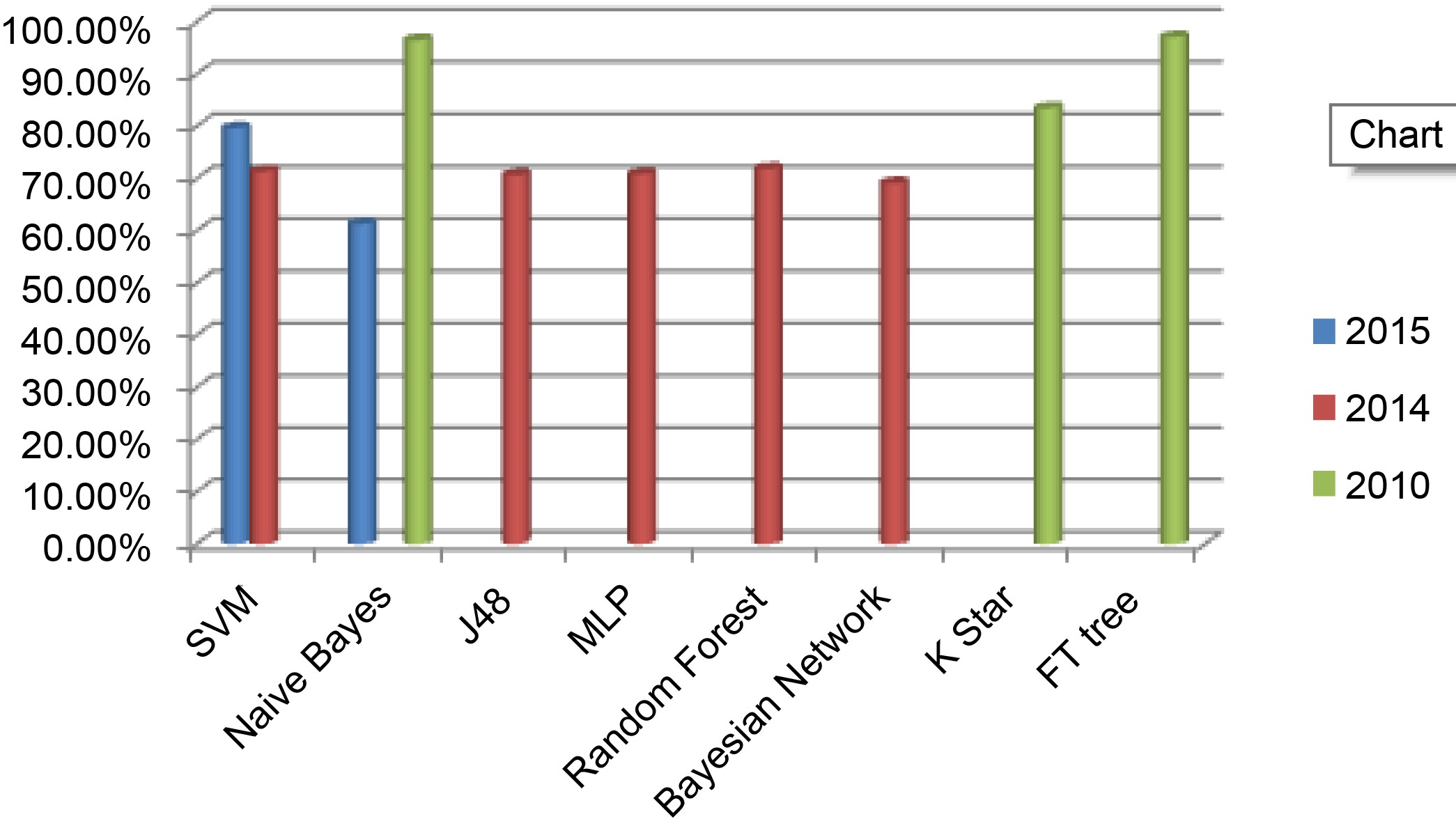
|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| Naive Bayes |  |  |  |  |  | 61.28% |
| J48 |  |  |  |  |  | 70.669% |
| MLP |  |  |  |  |  | 70.8405% |
| Random Forest | Gulia *et al*. | 2014 | Liver Disease | UCI | WEKA | 71.8696% |
| SVM |  |  |  |  |  | 71.3551% |
| Bayesian Network |  |  |  |  |  | 69.1252% |
| Naive Bayes |  |  |  |  |  | 96.52% |
| K Star | Rajeswari and Reena | 2010 | Liver Disease | UCI | WEKA | 83.47% |
| FT tree |  |  |  |  |  | 97.10% |

79.66%

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**Figure 5.** Accuracy of machine learning algorithms to detect liver disease.

by using the Models of Data Mining. Dengue is becoming a severe contagious disease. It creates trouble in those countries where weather is humid for example Thailand, Indonesia and Malaysia. Decision Tree (DT), Artificial Neural Net- work (ANN), and Rough Set Theory (RS) are the classification algorithms that are used in this study to predict dengue disease. Data set are taken from Public Health Department of Selangor State. WEKA data mining tool with two tests (10 Cross-fold Validation and Percentage split) is used. By using 10-Cross fold vali- dation DT offers 99.95% accuracy, ANN presents 99.98% of Correctness and RS shows 100% accuracy. After using PS, Both Decision tree and Artificial Neural Network gives 99.92% of correctness. RS achieves 99.72% accuracy.

Fathima and Manimeglai [[21]](#_bookmark32) performed a work to predict Arbovirus-Dengue disease. Data mining algorithm that are used by these researchers are Support Vector Machine. Data set for analysis is obtained from King Institute of Preven- tive Medicine and surveys of many hospitals and laboratories of Chennai and Tirunelveli from India. It contains 29 attributes and 5000 samples. Data is ex- amined by R project version 2.12.2. Accuracy that is achieved by SVM is 0.9042.

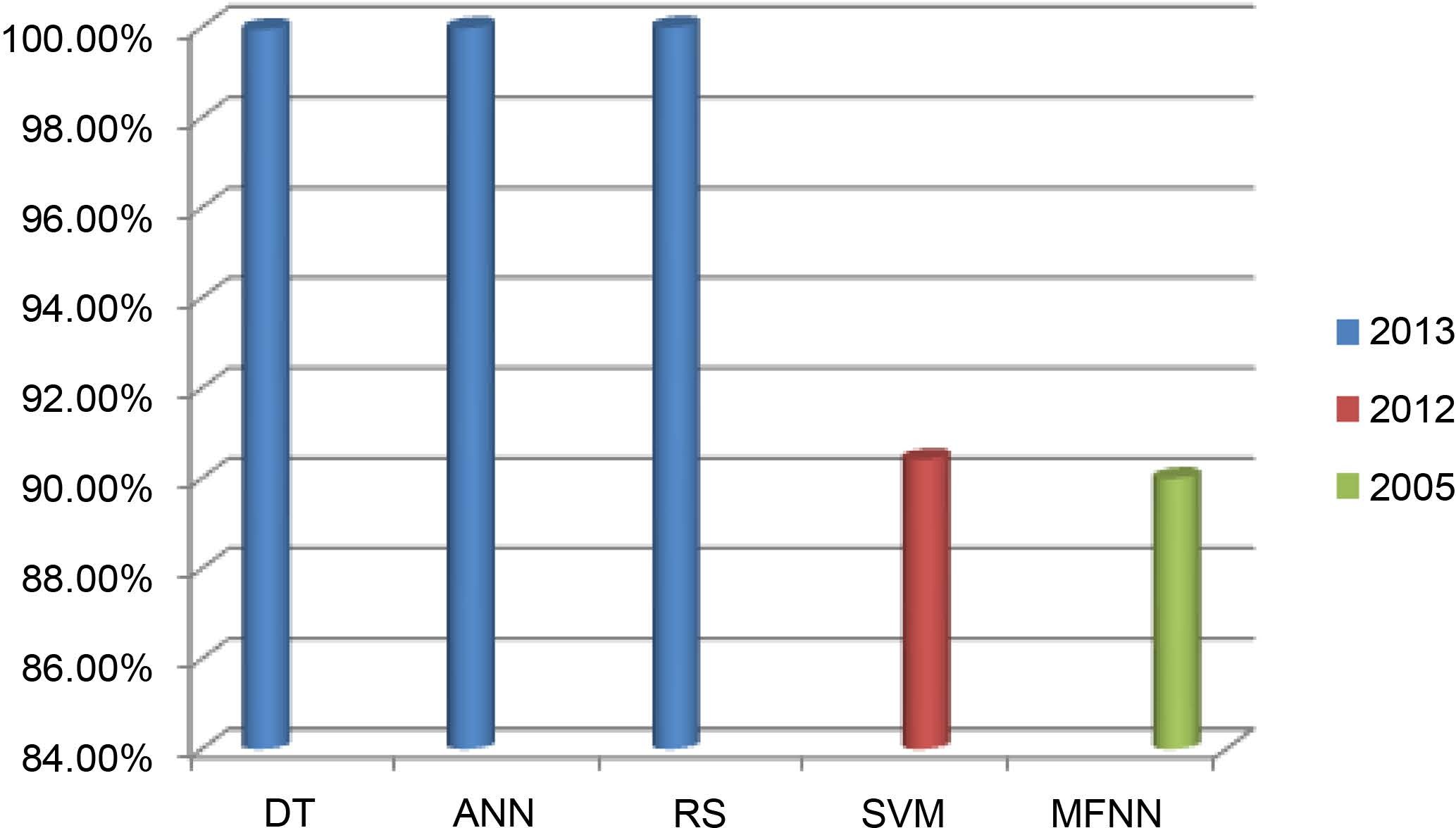
Ibrahim *et al*. [[22]](#_bookmark33) suggested a system in which Artificial neural network is

used for forecasting the defervescence day of fever in patients of dengue disease. Only clinical signs and symptoms are used by the proposed system for detection. The data are gathered from 252 hospitalized patients, in which 4 patients are having DF (Dengue fever) and 248 patients are having DHF (dengue hemorr- hagic fever). MATLAB’s neural network toolbox is used. Algorithm of Multilayer feed-forward neural network (MFNN) is used in this experiment. Day of defer- vescence of fever is accurately predicted by MFNN in DF and DHF with 90% correctness.

[**Figure 6**](#_bookmark8) shows the accuracy graph of all algorithms for the diagnosis of Den- gue disease.

### Analysis:

Different Machine learning techniques are used to diagnose dengue disease. Dengue disease is one of the serious contagious diseases. As in [**Table 4**,](#_bookmark9) for de- tection of dengue disease, RS theory shows the highest result as compared to the

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**Figure 6.** Accuracy of machine learning algorithms for dengue disease.

**Table 4.** Analysis of machine learning techniques for dengue disease detection.

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Machine Learning Techniques** | **Author** | **Year** | **Disease** | **Resource of Data Set** | **Tool** | **Accuracy** |
| DT |  |  |  |  |  | 99.95% |
| ANN | Tarmizi *et al*. | 2013 | Dengue Disease | Public Health Department of Selangor State | WEKA | 99.98% |
| RS |  |  |  |  |  | 100% |
| SVM Fathima and | | 2012 Arbovirus-Dengue King Institute of Preventive Medicine and R project Version 90.42% | | | | |

Manimeglai

disease

surveys of many hospitals and laboratories of Chennai and Tirunelveli from India

2.12.2

MFNN Ibrahim *et al*. 2005 Dengue disease From 252 hospitalized patients MATLAB neural

network Tool box

90%

other algorithms. In 2005 and 2012, researchers used different algorithms but did not attain highest result and improvements. In 2013, accuracy is improved by using RS. It is capable to manage uncertainty, noise and missing data. For the purpose of classification, Developed RS classifier is based on the Rough set theory. Selection of attribute empowers the classifier to surpass the other models. RS is a promising rule based method that offers meaningful information. RS is also best from neural network in term of time. NN takes much time to build model. DT is complex as well as costly algorithm. RS does not need any initial and additional information about data but Decision tree needs information.

### Advantages and Disadvantages of RS:

Advantages: It is very easy to understand and provides direct understanding of attained result. It evaluates data significance. It is appropriate for both qualit- ative and quantitative data. It discovers the hidden patterns. It also finds minim- al set of data. It can find relationship that cannot be identified by statistical me- thods.

Disadvantages: It has not so many limitations still it is not widely used.

## Hepatitis Disease

Ba-Alwi and Hintaya [[23]](#_bookmark34) suggested a comparative analysis. Data mining algo- rithms that are used for hepatitis disease diagnosis are Naive Bayes, Naive Bayes

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updatable, FT Tree, K Star, J48, LMT, and NN. Hepatitis disease data set was taken from UCI Machine Learning repository. Classification results are meas- ured in terms of accuracy and time. Comparative Analysis is taken by using neural connections and WEKA: data mining tool. Results that are taken by using neural connection are low than the algorithms used in WEKA. In this Analysis of Hepatitis disease diagnosis, second technique that is used is rough set theory, by using WEKA. Performance of Rough set procedure is better than NN specially in case of medical data analysis. Naive Bayes gives the accuracy of 96.52% in 0 sec. 84% Accuracy is attained by the Naive Bayes Updateable algorithm in 0 sec. In 0.2 sec FT Tree presents the accuracy of 87.10%. K star offers 83.47% Cor- rectness. Time taken for K star algorithm is 0 sec. Correctness of 83% is achieved by J48 and time that J48 takes to classify is 0.03 sec. LMT provides 83.6% accu- racy 0.6 sec. Neural network shows 70.41% of correctness. Naive Bayes is best classification algorithm used in the rough set technique. It offers high accuracy in minimum time.

Karlik [[24]](#_bookmark35) shows a comparative analysis of Naive Bayes and back propagation classifiers to diagnose hepatitis disease. Key advantage of using these classifiers is that they require small amount of data for categorization. Types of hepatitis are “A, B, C, D and E”. These are generated by different viruses of hepatitis. Rapid Miner open source software is used in this analysis. Hepatitis data set is taken from UCI. Data set include 20 features and 155 instances. 15 attributes are used in this experiment. Naive Bayes classifier gives 97% accuracy. Three-layered feed- forward NN are used and trained with Back propagation algorithm 155 instances are used for training. Correctness of 98% is attained.

Sathyadevi [[25]](#_bookmark36) employed C4.5, ID3 and CART algorithms for diagnosing the disease of hepatitis. This study uses the UCI hepatitis patient data set. WEKA, tool is used in this analysis. CART has offered great performance handling of missing values. So, CART algorithm shows a highest classification accuracy of 83.2%. ID3 Algorithm offers 64.8% of accuracy. 71.4% is attained by C4.5 algo- rithm. Binary decision tree (DT) that is generated by CART algorithm has only two or no child. DT that is formed by the C4.5 and ID3 can have two or more children. CART algorithm performs well in terms of Accuracy and time com- plexity.

### Analysis:

Many algorithms have been used for diagnosis of different diseases. [**Table 5**](#_bookmark10)gives the comprehensive view. For the detection of Hepatitis disease, Feed for- ward neural network with back propagation shows highest accuracy of 98%. Be- cause in this model, three layered feed forward neural network is trained with error back propagation algorithm. Back propagation training with the rule of delta learning is an iterative gradient algorithm planned to lessen the RMSE “root mean square error” between the real output of a multilayered feed-forward neural net- works and a desired output. Every layer is connected to preceding layer and having no other connection. Second best result is offered by Naive Bayes. But in terms of time to build model, Naive Bayes runs fast as compare to neural network. Fi-

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gurative approach for the detection of hepatitis is shown in [**Figure 7**.](#_bookmark11)

### Advantages and Disadvantages of NN:

Advantages: Adaptive Learning, Self-Organization, Real Time Operation Fault Tolerance via Redundant Information Coding.

Disadvantages: Less over fitting needs great computational effort. Sample Size must be large. It’s time consuming. Engineering Judgment does not develop the relations between input and output variables so that the model behaves like a black box [[26].](#_bookmark37)

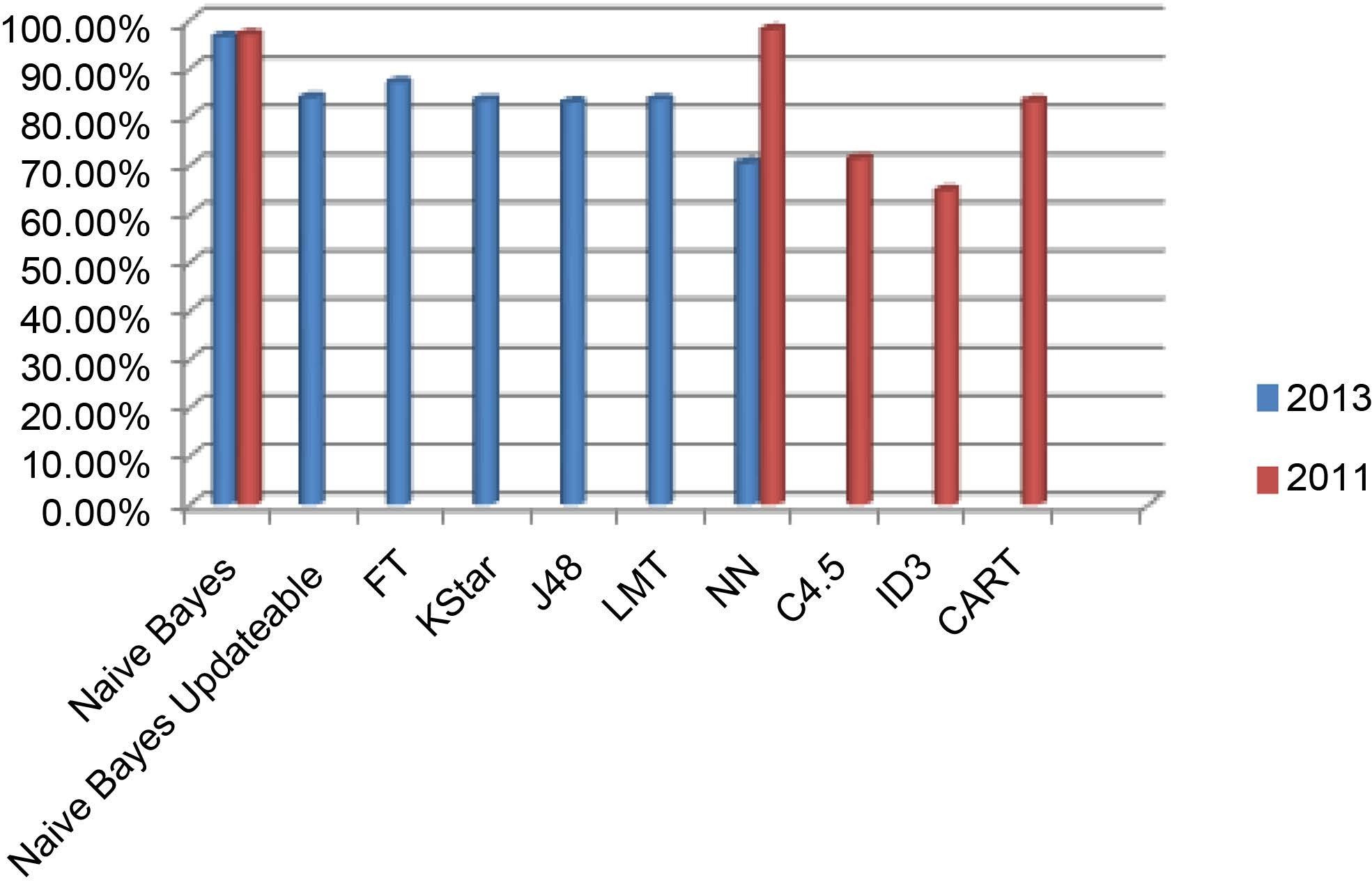
# Discussions and Analysis of Machine Learning Techniques

For diagnosis of Heart, Diabetes, Liver, Dengue and Hepatitis diseases, several machine-learning algorithms perform very well. From existing literature, it is observed that Naive Bayes Algorithm and SVM are widely used algorithms for

**Table 5.** Comprehensive view of machine learning techniques for hepatitis disease.

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| **Machine Learning Techniques Author** | **Year** | **Disease** | **Resource of Data Set** | **Tool Accuracy** |
| Naive Bayes Naive Bayes updateable  FT  K Star Ba-Alwi and Hintaya, | 2013 | Hepatitis Disease | UCI | 96.52%  84%  87.10%  WEKA 83.47% |
| J48 |  |  |  | 83% |
| LMT NN |  |  |  | 83.6%  70.41% |
| Naive Bayes  Karlik  Feed forward NN with Back propagation | 2011 | Hepatitis Disease | UCI | 97%  98% |
| C4.5 |  |  |  | 71.4% |
| ID3 Sathyadevi | 2011 | Hepatitis Disease | UCI | WEKA 64.8% |

Rapid Miner

CART 83.2%

**Figure 7.** Machine learning algorithm’s accuracy to detect hepatitis disease.

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detection of diseases. Both algorithms offer the better accuracy as compare to other algorithms. Artificial Neural network is also very useful for prediction. It also shows the maximum output but it takes more time as compared to other algorithms. Trees algorithm are also used but they did not attain wide accep- tance due to its complexity. They also shows enhanced accuracy when it re- sponded correctly to the attributes of data set. RS theory is not widely used but it presents maximum output.

# Conclusion

Statistical models for estimation that are not capable to produce good perfor- mance results have flooded the assessment area. Statistical models are unsuc- cessful to hold categorical data, deal with missing values and large data points. All these reasons arise the importance of MLT. ML plays a vital role in many ap- plications, e.g. image detection, data mining, natural language processing, and disease diagnostics. In all these domains, ML offers possible solutions. This pa- per provides the survey of different machine learning techniques for diagnosis of different diseases such as heart disease, diabetes disease, liver disease, dengue and hepatitis disease. Many algorithms have shown good results because they identi- fy the attribute accurately. From previous study, it is observed that for the detec- tion of heart disease, SVM provides improved accuracy of 94.60%. Diabetes dis- ease is accurately diagnosed by Naive Bayes. It offers the highest classification accuracy of 95%. FT provides 97.10% of correctness for the liver disease diagno- sis. For dengue disease detection, 100% accuracy is achieved by RS theory. The feed forward neural network correctly classifies hepatitis disease as it provides 98% accuracy. Survey highlights the advantages and disadvantages of these algo- rithms. Improvement graphs of machine learning algorithms for prediction of diseases are presented in detail. From analysis, it can be clearly observed that these algorithms provide enhanced accuracy on different diseases. This survey paper also provides a suite of tools that are developed in community of AI. These tools are very useful for the analysis of such problems and also provide opportunity for the improved decision making process.

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