**A Study on Indians Diabetes Database with Analysis and Prediction Using Data Mining and Machine Learning Classification Approaches**

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**Abstract**

Machine learning classification constitutes a subset of both artificial intelligence and data science. It involves training models to categorize or classify data points into predefined classes based on their distinct attributes or features. The core objective of classification is to empower the model to discern patterns and correlations within the data, enabling it to correctly assign new and unseen data points to their appropriate classes. In this research, the datasets consist of several medical predictor (independent) variables and one target (dependent) variable, Outcome. Independent variables include Pregnancy, Glucose, blood pressure, Skin Thickness, Insulin, BMI, Diabetes pedigree function, Age, and Outcome in this research, analysis, and prediction using four different machine learning approaches, namely Linear Regression, SMOreg, Random Tree and REP Tree with accuracy parameters. Numerical illustrations are also provided to prove the results and discussion.

**Keywords:** Data Mining, Machine Learning, Decision Tree, Classifications and Diabetes.

**1.0 Introduction**

In classification scenarios, the input data encompasses a collection of features representing individual data points' quantifiable traits or characteristics. These features serve as the basis for the model to generate predictions about the category to which each data point belongs. The classes, in turn, signify the discrete categories or labels that the model strives to allocate to the data points. Constructing a classification model encompasses several pivotal stages: data collection and preparation, feature extraction and selection, model selection, model training, model evaluation, hyperparameter tuning, and model deployment.

Classification finds extensive utility in diverse domains, encompassing image recognition, natural language processing, fraud detection, medical diagnosis, sentiment analysis, and more. Its effectiveness hinges on factors such as the caliber and extent of the training data, algorithmic selection, and precise parameter tuning to attain accurate and dependable predictions. The term "Correctly Classified Instances" represents a concept used in evaluating machine learning models to assess their performance. Calculating Correctly Classified Instances is part of the overall model evaluation process. Incorrectly Classified Instances refer to the instances or data points in a machine learning model's evaluation or testing dataset that the model classifies incorrectly. In simpler terms, these are instances where the model's predictions do not align with the actual target or ground truth values.

**2.0 Literature Review**

The present study is to conduct a systematic review of the applications of machine learning, data mining techniques and tools in the field of diabetes research with respect to a) Prediction and Diagnosis, b) Diabetic Complications, c) Genetic Background and Environment, and e) Health Care and Management with the first category appearing to be the most popular. A wide range of machine learning algorithms were employed. In general, 85% of those used were characterized by supervised learning approaches and 15% by unsupervised ones, specifically association rules.  SVM arose as the most successful and widely used algorithm. Concerning the type of data, clinical datasets were mainly used. The title applications in the selected articles project the usefulness of extracting valuable knowledge leading to new hypotheses targeting deeper understanding and further investigation in DM [1].

There are many computerized methods to diagnose Diabetes Mellitus but the drawback of all these methods is that the patient still has to undergo various medical tests to provide the input values to the computerized diagnostic system and the overall cost for diagnosis will remain almost same for the patient but in our proposed model user itself sitting from home can diagnose whether he/she is suffering from Diabetes Mellitus or not. There is only need to provide some physical parameter values and on the basis of provided information's our model will detect whether that person is suffering from Diabetes Mellitus or not. In our work we have used Neural Network, for designing and testing Neural Network we used MATLAB software [2].

Past and recently recorded data are analyzed to extract patterns of behavior, discover new knowledge and provide explanations to the physician through the CP. Advanced tools in the CP allow the physician to prescribe personalized treatment plans and frequently quantify patient adherence [3].

The main goal of data mining is to discover new patterns for the users and to interpret the data patterns to provide meaningful and useful information for the users. Data mining is applied to find useful patterns to help in the important tasks of medical diagnosis and treatment. This project aims for mining the relationship in Diabetes data for efficient classification. The data mining methods and techniques will be explored to identify the suitable methods and techniques for efficient classification of Diabetes dataset and in mining useful patterns [4].

Predictive model of short-term glucose homeostasis relying on machine learning is presented with the aim of preventing hypoglycemic events and prolonged hyperglycemia on a daily basis. Second, data mining approaches are proposed as a tool for explaining and predicting the long-term glucose control and the incidence of diabetic complications [5].

Various data mining techniques help diabetes research and ultimately improve the quality of health care for diabetes patients. This paper provides a survey of data mining methods that have been commonly applied to Diabetes data analysis and prediction of the disease [6].

The medical data is taken from UCI repository, consists of 9 input attributes related to clinical diagnosis of diabetes, and one output attribute which indicates whether the patient is diagnosed with the diabetes or not. The whole data set consists of 768 cases [7].

The purpose of this study is to develop a data mining model for which will predict a suitable dosage planning for diabetes patients. Medical records of 89 different patient records were used in this study. 318 diabetes assays were extracted using these patient records. ANFIS and Rough Set methods were used for dosage planning objective. According to the results of ANFIS and Rough Set methods, ANFIS is a more successful and reliable method for diabetes drug planning objective when compared to Rough Set method [8].

study aims to summarize and analyze knowledge on: (1) years and sources of DSM publications, (2) type of diabetes that took most attention, (3) DM tasks and techniques most frequently used, and (4) the considered functionalities. A total of 57 papers published between 2000 and April 2017 were selected and analyzed regarding four research questions. The study shows that prediction was largely the most used DM task and Neural Networks were the most frequently used technique. Moreover, T1DM is largely the type of diabetes that is most concerned by the studies so as the Prediction of blood glucose [9].

RF is a familiar machine learning decision tree algorithm that belongs to supervised learning methods. In these approaches working principles based on classification and regression. RF is generally called ensemble learning, which combines different classifiers to solve various problems with enhanced performance of the model. The Random Forests classifier compared to others is the best classifier for capable of precisely classifying the huge amount of data. RF decision tree approaches mainly focused learning procedure for classification and regression methods, it will be creating many decision trees and level of the tree at training time for outputs the class with classes output from single trees [10].

The destinations have been assented of solidness in paddy advancement and to expand the development of creation in a maintainable way to meet the nourishment prerequisite for the developing populace. In any farming fields, it for the most part, happens that at whatever point the choices in regards to different methodologies of arranging is viewed as, for example, season-wise rainfall, region, production and yield rate of principal crops, and so forth. In this paper, it is proposed to discover the forecast level of concentration in paddy improvement for different years of time series data utilizing stochastic model approach. Numerical examinations are outlined to help the proposed work [11]

In the recent times, there has been an increasing demand for efficient strategies in the field of data assimilation about groundwater. Data mining process is a discovery of hiding information that utilizes the prediction efficiently by stochastic sensing concept. This paper proposes an efficient assessment of groundwater level, rainfall, population, food grains and enterprises dataset by adopting stochastic modeling and data mining approaches. Firstly, the novel data assimilation analysis is proposed to predict the groundwater level effectively. Experimental results are done and the various expected ground water level estimations indicate the sternness of the approach [12].

**3.0 Methods and Background**

**3.1 Linear Regression**

Linear regression is a statistical technique employed to comprehend and forecast the connection between two variables by discovering the optimal straight line that most effectively aligns with the data points. It aids in ascertaining how alterations in one variable correspond to changes in another, proving valuable for predictions and trend recognition.

The core idea of linear regression is to find the best-fitting straight line (also called the "regression line") through a scatterplot of data points. This line represents a linear equation of the form:

**y = mx+b … (1)**

Where:

* y is the dependent variable (the one you want to predict or explain).
* x is the independent variable (the one you're using to make predictions or explanations).
* m is the slope of the line, representing how much
* y changes for a unit change in x.

b is the y-intercept, indicating the value of y when x is 0.

**3.2 SMO**

SMO stands for "Sequential Minimal Optimization," an algorithm used for training support vector machines (SVMs), machine learning models commonly used for classification and regression tasks. The SMO algorithm is particularly well-suited for solving the quadratic programming optimization problem that arises during the training of SVMs.

1. **Initialization:** Start with all the data points as potential support vectors and initialize the weights and bias of the SVM.
2. **Selection of Two Lagrange Multipliers:** In each iteration, the SMO algorithm selects two Lagrange multipliers (associated with the support vectors) to optimize.
3. **Optimize the Pair of Lagrange Multipliers:** Fix all the Lagrange multipliers except the selected two, and then optimize the pair chosen to satisfy certain constraints while maximizing a specific objective function.
4. **Update the Model:** After optimizing the selected pair of Lagrange multipliers, update the SVM model's weights and bias based on the new values of the Lagrange multipliers.
5. **Convergence Checking:** Check for convergence criteria to determine whether the algorithm should terminate.
6. **Repeat:** If convergence hasn't been reached, repeat steps 2 to 5 until it is.

**3.3 Random Tree**

A "Random Tree" could refer to different things depending on the context. It might refer to a decision tree that has been built using some form of randomness, or it could be a term used in a specific domain or framework. Without more context, it's challenging to provide a precise answer. However, I can offer a couple of interpretations that might be relevant:

1. **Randomized Decision Tree:** A Random Tree might be referring to a decision tree constructed using randomness, like how Random Forest uses random sampling of data and features.
2. **Specific Framework:** Depending on your machine learning or data analysis framework, "Random Tree" could be a specific term or concept introduced within that framework.

**3.4 REPTree**

REPTree, short for "Reduced Error Pruning Tree," is a decision tree algorithm primarily used for classification tasks in machine learning. It is designed to create decision trees while incorporating a reduced-error pruning technique to avoid overfitting. The algorithm was introduced as a part of the WEKA machine learning software. Here's how the REPTree algorithm works:

1. **Tree Construction:** REPTree follows a recursive approach to build a decision tree. It starts by selecting the best attribute to split the data based on metrics like information gain or gain ratio.
2. **Recursive Splitting:** The algorithm examines potential attribute splits at each node and chooses the one that maximizes the selected splitting criterion.
3. **Reduced Error Pruning:** After the tree is fully grown, REPTree performs reduced-error pruning to eliminate branches that do not contribute significantly to the tree's accuracy.
4. **Prediction:** Once the tree is pruned, it can be used for making predictions.

**4.0 Numerical Illustrations**

**4.1 Dataset**

dataset is originally from the National Institute of Diabetes and Digestive and Kidney Diseases. The objective of the dataset is to diagnostically predict whether a patient has diabetes, based on certain diagnostic measurements included in the dataset. Several constraints were placed on the selection of these instances from a larger database. All patients here are females at least 21 years old of Pima Indian heritage. The datasets consist of several medical predictor variables and one target variable, Outcome [13]. Predictor variables includes the number of pregnancies the patient has had, their BMI, insulin level, age, and so on.

**Table 1: Sample Data**

|  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- |
| **Pregnancies** | **Glucose** | **Blood Pressure** | **Skin Thickness** | **Insulin** | **BMI** | **Diabetes****Pedigree****Function** | **Age** | **Outcome** |
| 6 | 148 | 72 | 35 | 0 | 33.6 | 0.627 | 50 | 1 |
| 1 | 85 | 66 | 29 | 0 | 26.6 | 0.351 | 31 | 0 |
| 8 | 183 | 64 | 0 | 0 | 23.3 | 0.672 | 32 | 1 |
| 1 | 89 | 66 | 23 | 94 | 28.1 | 0.167 | 21 | 0 |
| 0 | 137 | 40 | 35 | 168 | 43.1 | 2.288 | 33 | 1 |
| 5 | 116 | 74 | 0 | 0 | 25.6 | 0.201 | 30 | 0 |
| 3 | 78 | 50 | 32 | 88 | 31 | 0.248 | 26 | 1 |
| 10 | 115 | 0 | 0 | 0 | 35.3 | 0.134 | 29 | 0 |
| 2 | 197 | 70 | 45 | 543 | 30.5 | 0.158 | 53 | 1 |
| 8 | 125 | 96 | 0 | 0 | 0 | 0.232 | 54 | 1 |
| 4 | 110 | 92 | 0 | 0 | 37.6 | 0.191 | 30 | 0 |
| 10 | 168 | 74 | 0 | 0 | 38 | 0.537 | 34 | 1 |
| 10 | 139 | 80 | 0 | 0 | 27.1 | 1.441 | 57 | 0 |
| 1 | 189 | 60 | 23 | 846 | 30.1 | 0.398 | 59 | 1 |
| 5 | 166 | 72 | 19 | 175 | 25.8 | 0.587 | 51 | 1 |
| 7 | 100 | 0 | 0 | 0 | 30 | 0.484 | 32 | 1 |
| 0 | 118 | 84 | 47 | 230 | 45.8 | 0.551 | 31 | 1 |
| 7 | 107 | 74 | 0 | 0 | 29.6 | 0.254 | 31 | 1 |
| 1 | 103 | 30 | 38 | 83 | 43.3 | 0.183 | 33 | 0 |
| 1 | 115 | 70 | 30 | 96 | 34.6 | 0.529 | 32 | 1 |

**Table 2:** **R2 Score or Correlation coefficient**

|  |  |
| --- | --- |
| **Function and trees** | **Correlation Coefficient** |
| **Linear Regression** | 0.7322 |
| **SMOreg** | 0.6266 |
| **Random Tree** | 0.4552 |
| **REP Tree** | 0.7018 |

**Table 3: Mean Absolute Error (MAE) and Root Mean Squared Error (RMSE)**

|  |  |  |
| --- | --- | --- |
| **Function and trees** | **MAE** | **RMSE** |
| **Linear Regression****SMOreg****Random Tree****REP Tree** | 0.3366 | 0.4036 |
| 0.3248 | 0.4144 |
| 0.3268 | 0.5717 |
| 0.3182 | 0.4187 |

**Table 4: Time taken to build the ML modeling (seconds)**

|  |  |
| --- | --- |
| **Function and trees** | **Time Taken (seconds)** |
| **Linear Regression** | 0.2200 |
| **SMOreg** | 0.4000 |
| **Random Tree** | 0.0400 |
| **REP Tree** | 0.0900 |

**Fig 1: R2 Score or Correlation coefficient**

**Fig 2: Mean Absolute Error and Root Mean Squared Error**

**Fig 3: Time taken to build the ML model (seconds)**

**5.0 Result and Discussion**

This research focused on Indians Diabetes recommendation systems, including Pregnancy, Glucose, blood pressure, Skin Thickness, Insulin, BMI, Diabetes pedigree function, Age, and Outcome. The related sample dataset is indicated in Table 1.

 Based on Table 2 and Fig. 1, the R2 score is the most essential technique in machine learning, which is used to find the relationship between independent and dependent variables. In this case study, linear regression return a strong positive correlation base on different parameters. REP Tree return nearly 70% which means linear regression return a strong positive correlations. The related results and discussions are shown in Table 2 and Fig. 2.

Mean Absolute Error (MAE) is a metric commonly used to measure the accuracy of a predictive model, particularly in the context of regression tasks. In this case, all the weather and nutrient parameters have nearly 0 error rates for using MAE test statistics. Similarly, the Root Mean Squared Error (RMSE) is another standard metric used to evaluate the performance of predictive models, particularly in regression tasks. Like Mean Absolute Error (MAE), RMSE measures the accuracy of predictions. In these cases, the error rate is also nearly 0. Both the MAE and RMSE also returned almost 0. Table 3 and Fig. 2 show the related results and discussions.

Time complexity one of the important parameters for analysis and prediction using machine learning approaches. In this case, for analysis of Diabetes recommendation systems Random tree taking very less time for analysis and prediction and next position REP Tree. The similar results and discussion shown in Table 4 and Fig. 3.

**6.0 Conclusion and Further Studies**

Based on results and discussion, most ML approaches return better results with test statistics. However, it's essential to acknowledge the limitations of our study. Our analysis was constrained by the available dataset's, which occasionally hindered a more nuanced exploration of certain factors. Furthermore, the study primarily focused on specific predictions for all the parameters. urban context may not be fully generalizable to diverse geographical and cultural settings. In the future, consider other machine learning approaches with test statistics to improve the accuracy and reduce the time complexity.

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