

Methods of Supervised Machine Learning for Forecasting Scholarly Performance of Scholars

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Abstract - Numerous supervised and unsupervised machine learning techniques are employed to uncover relationships and hidden information from data, which will ultimately assist leaders in the future in making the right decisions. a variety of robust algorithms that are employed in many spheres of daily life, including education. This work presents a model for predicting students' academic achievement that makes use of supervised machine learning techniques, such as logistic regression and support vector machines. When the outcomes of several tests conducted with varying technologies are examined, it becomes clear that the sequential minimum optimization method performs better than logistic regression by obtaining more accuracy. Additionally, by using the information gleaned from this study, educational institutions will be able to forecast students' conduct in the future and classify their performance as either excellent or poor.

The goal is not only to forecast students' future performance but also to offer the most effective method for identifying the most important aspects to focus on, such as student motivation and instructor effectiveness, which will ultimately lower the dropout rate.

Index Terms — Student performance prediction, multidimensional representations, relationship modeling, fine-grained aspects.

I. INTRODUCTION

The application of information analysis in the field of instruction and learning is crucial to improving the current educational system and enabling the identification of its fundamental issues from a variety of angles, including those of the administration, faculty, and students[1][2]. Data analytics discoveries aid in effective planning, comprehension of challenges in real time, and decision-making that contributes to system improvement [3]. Data analytics has made its journey into learning environments for tasks like information discovery [4] using existing databases, creating forecasts, text mining, machine learning, etc. because of the many benefits it provides. The main goal of learning analytics is to implement a technology paradigm shift in place of the current learning approaches..

Over the years, a large amount of information about learners has been collected and kept in databases; nevertheless, it is impossible to extract any meaningful information from this data and apply it for research purposes [3],[5]. In order to harness the power of insights in the discipline of education, the data found in several source systems must be merged [9]. Due to the abundance of study topics in the discipline of learning analytics, these factors have led to a surge in research activity [6].

Applying data analytics to educational data facilitates the understanding of each student's areas of weakness, the collection of faculty performance data, the efficient use of resources, and the formulation of appropriate plans to alleviate problems and implement the required changes to improve current circumstances [10]. It's critical to realize that while the analytics' findings aid in identifying problems from a variety of angles, improvements won't occur until all parties involved in the system—parents, teachers, students, and administrators—join forces to make the right decisions and carry them out[7]..

In order to comprehend the underlying circumstances and assist the instructors in taking the required steps to improve the academic achievement of the students, we have attempted to examine a variety of factors that are influencing the test scores of the learners in the instructional information set in this project [8]. Additionally, we have attempted to make this model available online so that instructors and students may use it to make predictions about the future and raise student performance [9]

II. SYSTEM ANALYSIS

Problem Statement:

The fast growth of technology has brought about a paradigm shift in the field of education. Most of the processes have been digitalized to increase the system's efficacy and efficiency. Though rarely used, plenty of information is being gathered these days for scholarly databases. Robust instruments are required to derive the intended benefits from such massive amounts of data. There is no way for traditional educational institutions to forecast student achievement.

No useful information can be taken from the data and utilized for analysis, even though a significant quantity of student data has been collected all through the years and stored in the database. In order for education to reap the benefits of analytics, it is necessary to aggregate data from many sources. These qualities have increased the quantity of research that has been conducted in the field of learning analytics since there are several avenues to pursue.

Aim of the Project:

The main reason machine learning has gained such popularity recently is that it can find hidden relationships between various parts of data, making choices more reliable and accurate. To achieve this purpose, supervised and unsupervised methodology techniques are used. As such, artificial intelligence may be used to almost any industry to support precise forecasting and decision-making. Our current research involves utilizing supervised machine learning methods such as the case of logistic regression, random forest modeling, support vector machines, and others to forecast student performance. In an attempt to make this model available to the academic community, we have also worked to provide it as a web application. Future student performance and grades may be predicted using the information that was extracted from the educational information collection and the insights gained from it. Predicting student achievement in advance and assisting with their subsequent academic achievement was the main objective of the study. We provided the suggested model, which is constructed using cutting-edge machine-learning techniques, the Kalboard data set.

Scope of the Project:

The project's only objectives are to forecast student performance and determine how accurate the suggested model is. The suggested model is trained using training data and tested using test data by the system administrator. Both instructors and students may register on the site and use the prediction model as users.

Proposed System:

Our proposal involves creating and implementing an online application that can forecast a student's academic achievement by taking into account many factors. The web version of the model created using machine learning would be made available to students so they could upload, view, and get forecasts regarding their academic parameters. Our goal is to create an automated prediction system that can

correctly categorize a student's academic grade and provide teachers, parents, and students with access to this information.

Advantages:

This automated method makes use of the vast quantity of data produced by educational systems and can forecast student grades before the start of the course.

III. PROPOSED MODULAR IMPLEMENTATION

The technological solution to the issue is given below:

1. Identification of Dataset
2. Data preprocessing
3. N-gram analysis of reviews
4. Fine-grained aspect extraction
5. Creation of Model
6. Testing the model.

The project's suggested modular implementation is shown below. There are two modules in it:

1. Administrator
2. Learner

Admin Module:

The system administrator is in charge of the following tasks:

1. Putting the dataset online
2. A study of the dataset's data
3. Dividing the training and testing datasets
4. Logistic regression and SVM model training
5. Examine how well the algorithms performed using the provided dataset.
6. Examine student information and forecasts.

Student Module:

The following machine learning services are accessible for the system's user to utilize.

7. Logging into the system.
8. Receiving projections for academic performance
9. Saving educational parameters to the database

IV. PROJECT EXECUTION

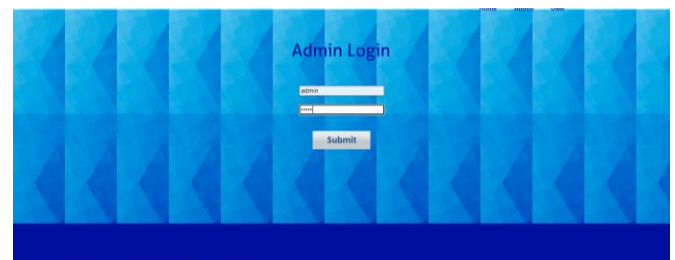
Home page:

This is the starting page of the application when the application is executed on Pycharm, the application is hosted on a web server and URL is generated to access the application once the user clicks on the URL the below page is opened on the browser.



Admin Login:

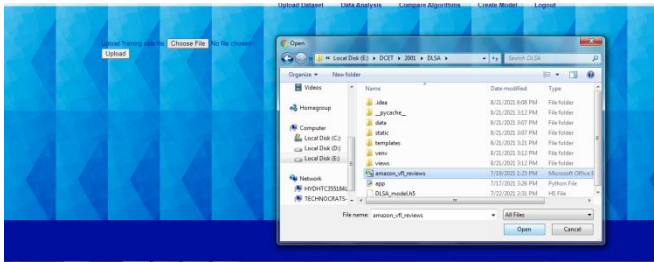
This is the admin module login page. To perform tasks such as uploading, training, exploratory data analysis, feeding the dataset to various algorithms for machine learning to determine which algorithm can meet the best accuracy, and so on, the administrator must log into the system using his credentials. Make a model that users may utilize by hosting it on the Flask application.



Upload Dataset:

The system administrator can upload the datasets needed to train machine learning algorithms on this page. When transferring the document to the server, the administrator must first choose the document by clicking the Select file button and then click the Upload button. The file has been successfully uploaded, according to the success message that appears once the upload is finished. We are utilizing the performance of

students' reviews as the dataset for this project.



Data Analysis:

Using graphs, statistics, and other visual aids, data exploration is used to the dataset to remove any missing data, spot trends, and determine the connections between different output characteristics.

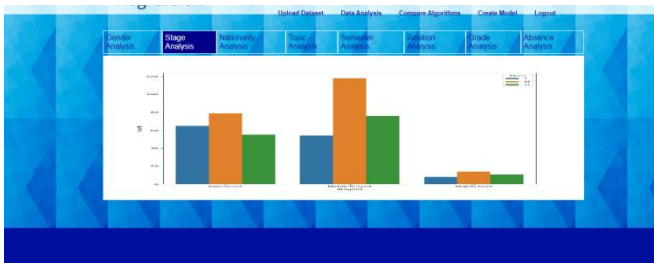
Gender Analysis:

The gender analysis is displayed in the graph below.



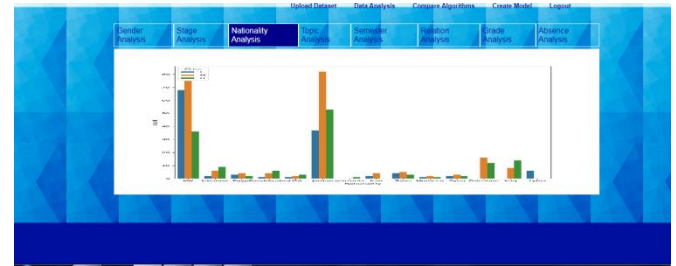
Stage Analysis:

The Stage Analysis is displayed in the graph below.



Nationality Analysis:

This graph displays the analysis of nationality.



Topic Analysis:

The Topic Analysis is displayed in the graph below.



Compare Algorithms:

The administrator can train different algorithms and determine the test's accuracy for each method by feeding the dataset to them on this page.

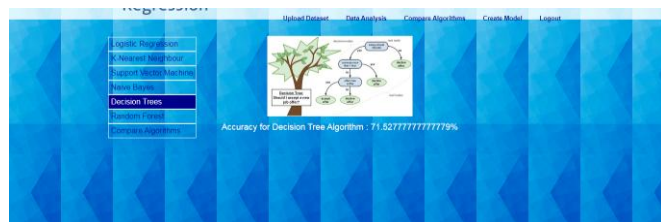
Logistic Regression:

We find that the test's accuracy is 75.0% if the dataset is fed into the logistic regression technique.



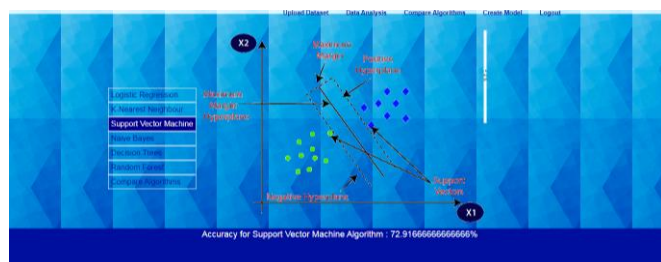
Decision Trees:

The Decision Trees method yields a test accuracy of 72.16666666666666% whenever the dataset is fed into it.



Support Vector Machine:

The test accuracy, as observed when the dataset is fed into the Support Vector Machine method, is 72.1666666666666%.



Naive Bayes Algorithm:

The test accuracy, as determined by feeding the dataset into the Naive Bayes method, is 35.4166666666667%.



Create Model:

The generated model's accuracy is displayed on this screen as 85.4166666666666%.



V.CONCLUSION

A student's ability to succeed in their job is mostly determined by their academic performance. The majority of nations switched from traditional classroom settings to online learning environments. As a result, it can be challenging to evaluate how students behave while they are studying remotely, including when they take part in class activities, participate in discussion groups or exams, and interact with teachers. Academic institutions may profit greatly from the analysis and extraction of the hidden knowledge gained through these systems of learning management. Also, by projecting future performance and making the appropriate adjustments, this kind of knowledge may be used to increase the academic brilliance of the student. In the study that follows, we looked at a student model. Support vector machines and logistic regression are both classifiers that we used to assess the model's effectiveness. Based on the characteristics chosen via the feature selection technique, these classifiers are chosen using two distinct technologies. The three main categories—the degree of student interaction satisfaction with the system, classroom punctuality, and system satisfaction—are what determine a student's academic grade. Support vector machines and gain ratio selection of features are used to get the results for about three factors. If we demonstrate the help vector machine technique utilizing sequential minimum optimization, the results for forecasting of students' future performance will be more accurate. In an effort to reduce dropout rates and assist students in their areas of weakness, we also suggest implementing the model into an online application and making it accessible to instructors, parents, and students.

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