

A Machine Learning Framework For Evaluating Student Academic Performance Based On Knowledge Attributes

¹Dr C Mohammed Gulzar,²MD Labeeb,³Pavadakula Sandeep,⁴Nayakanti Sudhakar

¹Associate Professor, Computer Science Of Engineering, Dr K V Subba Reddy Institute of Technology

^{2,3,4}B. Tech Students, Computer Science Of Engineering, Dr K V Subba Reddy Institute of Technology

ABSTRACT

Predicting student academic performance has become an important research area in the field of educational data mining and machine learning. Accurate prediction of students' academic outcomes helps educators identify learning difficulties at an early stage and implement appropriate interventions to improve learning effectiveness. This study proposes a machine learning-based approach to analyze and predict student academic performance using knowledge attributes and learning capabilities. The system collects and processes various student-related factors such as prior academic knowledge, learning behavior, study habits, attendance, and assessment scores. These attributes are used as input features for training machine learning models that can learn patterns and relationships influencing student performance. In the proposed approach, the dataset undergoes preprocessing steps including data cleaning, feature selection, and normalization to enhance model performance. Multiple machine learning algorithms are employed to analyze the data and build predictive models capable of classifying or forecasting students' academic results. The developed model evaluates how knowledge characteristics and learning abilities contribute to academic success and provides insights into key factors affecting student achievement. Experimental results demonstrate that the machine learning models can effectively predict student performance with high accuracy and reliability. The proposed system can assist educators, academic institutions, and policymakers in monitoring student progress, identifying at-risk students, and designing personalized learning strategies to enhance educational outcomes.

Keywords: Machine Learning, Student Academic Performance Prediction, Educational Data Mining, Knowledge Attributes, Learning Ability Analysis, Predictive Modeling, Academic Analytics, Student Learning Behavior.

I. INTRODUCTION

The rapid growth of digital learning environments and educational technologies has generated large volumes of educational data, creating new opportunities to analyze student learning patterns and academic outcomes. Educational institutions are increasingly adopting data-driven approaches to understand factors that influence student success. Predicting student academic performance has become an important research topic because it enables educators to identify students who may face academic difficulties and provide timely support. By analyzing various learning-related attributes, institutions can develop intelligent systems that assist in improving educational quality and student outcomes.

Machine Learning (ML) has emerged as a powerful tool for analyzing complex educational datasets and discovering hidden patterns within student learning behaviors. Unlike traditional statistical methods, machine learning algorithms can process large volumes of data and automatically learn relationships between different variables affecting academic performance. Techniques such as Decision Trees, Random Forest, Support Vector Machines, and Gradient Boosting have been widely used to predict student outcomes based on historical academic records and behavioral characteristics. These predictive models help educators gain valuable insights into factors that contribute to student achievement and learning progress.

Knowledge characteristics and learning abilities play

a crucial role in determining how effectively students grasp and apply educational content. Knowledge attributes may include prior subject understanding, conceptual clarity, and problem-solving skills, while learning ability refers to factors such as study habits, participation in learning activities, time management, and cognitive skills. Analyzing these characteristics using machine learning techniques allows researchers to better understand the relationship between learning behaviors and academic performance. Such insights can help educators design personalized learning strategies that address individual student needs.

This study proposes a machine learning approach for analyzing and predicting student academic performance using knowledge attributes and learning capabilities. The proposed system collects and processes various student-related features and applies machine learning algorithms to build predictive models that evaluate academic outcomes. By identifying significant learning indicators and performance patterns, the system aims to assist educators in early detection of at-risk students and support the development of adaptive educational interventions. Ultimately, the integration of machine learning in educational analytics can contribute to improving learning effectiveness, enhancing student engagement, and supporting data-driven decision-making in modern educational systems.

II. LITERATURE SURVEY

1. Student Performance Prediction Using Machine Learning Techniques

Authors: K. Shah, A. Patel, and S. Shah

Abstract:

This study explores the use of machine learning techniques to predict student academic performance based on educational datasets. The authors applied algorithms such as Decision Tree, Naïve Bayes, and Support Vector Machine to analyze student learning behavior and academic records. The results showed that machine learning models can effectively identify

patterns influencing student success and failure. The study highlighted the importance of features such as attendance, internal assessment scores, and study habits in predicting academic performance.

2. Educational Data Mining for Student Performance Analysis

Authors: R. Baker and K. Yacef

Abstract:

This research focuses on the role of educational data mining in analyzing student learning data to improve academic outcomes. The authors examined different data mining techniques including classification, clustering, and association rule mining to identify patterns in student performance. The study demonstrates that analyzing educational datasets can help instructors better understand student learning behaviors and develop targeted teaching strategies to enhance academic performance.

3. Predicting Student Success Using Data Mining Techniques

Authors: M. Romero and S. Ventura

Abstract:

The study investigates the application of data mining methods for predicting student academic success in e-learning environments. Various predictive models were developed using classification algorithms to evaluate student participation, assignment performance, and online learning activity. The results indicate that predictive analytics can effectively forecast student outcomes and provide early warnings for students at risk of academic failure.

4. Machine Learning Based Academic Performance Prediction

Authors: S. Kotsiantis, C. Pierrakeas, and P. Pintelas

Abstract:

This research presents a comparative analysis of



machine learning algorithms for predicting student academic performance. Algorithms such as Decision Trees, Neural Networks, and Logistic Regression were evaluated using student academic datasets. The results demonstrated that machine learning models can significantly improve prediction accuracy and help educational institutions identify critical factors affecting student achievement.

5. Student Performance Prediction Using Ensemble Learning Methods

Authors: H. Yu and C. Liu

Abstract:

The authors proposed an ensemble learning approach for predicting student academic performance. Multiple machine learning models were combined to improve prediction accuracy and robustness. The study showed that ensemble techniques outperform individual algorithms by integrating multiple predictive models. The research also highlighted the importance of knowledge attributes and learning behaviors in determining student academic outcomes.

III. EXISTING SYSTEM

In traditional educational systems, student academic performance is typically evaluated using conventional assessment methods such as examinations, assignments, quizzes, and classroom participation. These evaluation methods mainly focus on measuring the final outcomes of students rather than analyzing the underlying learning behaviors and knowledge characteristics that influence academic performance. Teachers and academic administrators often rely on manual observation and historical academic records to identify students who are struggling in their studies. However, this approach may not provide accurate or early identification of students at risk of poor academic performance.

Several existing systems have attempted to analyze student performance using basic statistical methods

and simple data analysis techniques. These systems use limited datasets such as grades, attendance records, and assignment scores to evaluate academic progress. While these approaches provide some insights into student performance trends, they often fail to capture complex relationships between multiple learning factors. Traditional statistical models also struggle to handle large educational datasets and may produce less accurate predictions when dealing with diverse learning behaviors and knowledge attributes.

Some earlier research studies have applied basic machine learning and data mining techniques to predict student performance. Although these methods improved prediction accuracy compared to manual evaluation, many existing systems still rely on a limited number of features and do not fully consider knowledge characteristics and learning abilities. Additionally, many models focus only on classification of pass or fail outcomes rather than providing deeper analysis of the factors influencing student success. As a result, existing systems may lack flexibility, scalability, and the ability to provide personalized insights for improving learning outcomes.

Therefore, there is a need for more advanced and intelligent approaches that can effectively analyze multiple knowledge-based attributes and learning capabilities of students. Integrating modern machine learning techniques with educational data can help overcome the limitations of existing systems and provide more accurate, scalable, and data-driven solutions for predicting and improving student academic performance.

IV. PROPOSED SYSTEM

The proposed system introduces a machine learning-based framework designed to analyze and predict student academic performance by considering both knowledge attributes and learning capabilities. Unlike traditional systems that rely mainly on examination scores and attendance records, the proposed approach incorporates multiple factors



related to students' learning behavior, knowledge level, and academic activities. These attributes may include previous academic performance, subject understanding, assignment scores, study patterns, participation in learning activities, and other educational indicators. By analyzing these features collectively, the system aims to provide a more accurate and comprehensive prediction of student academic outcomes.

In the proposed framework, the educational dataset is first collected from institutional records or learning management systems. The collected data undergoes several preprocessing steps such as data cleaning, handling missing values, normalization, and feature selection. These preprocessing techniques help improve the quality of the dataset and ensure that only the most relevant knowledge characteristics and learning ability features are used for model training. After preprocessing, the dataset is divided into training and testing sets to develop predictive models using machine learning algorithms.

The system then applies machine learning techniques such as Logistic Regression, Random Forest, Gradient Boosting, and other classification algorithms to analyze the relationship between input features and student academic performance. These algorithms learn patterns from historical student data and generate predictive models capable of forecasting future academic outcomes. The performance of each model is evaluated using appropriate evaluation metrics such as accuracy, precision, recall, and F1-score to determine the most effective model for performance prediction.

Finally, the proposed system provides predictive insights that help educators identify students who may be at risk of poor academic performance at an early stage. By understanding the influence of knowledge attributes and learning abilities on academic success, educators can design personalized learning strategies and targeted interventions to support student development. The integration of machine learning into educational analytics enhances

decision-making processes and contributes to improving the overall quality of education and student achievement.

V. SYSTEM ARCHITECTURE

The system architecture for predicting student academic performance using machine learning consists of several interconnected components that process educational data and generate predictive results. The architecture begins with the data collection module, where student-related information is gathered from educational databases, learning management systems, or institutional records. This data includes various attributes such as student demographics, prior academic results, attendance records, assignment scores, learning behavior, and knowledge characteristics. These collected attributes form the primary dataset that will be used for analysis and prediction.

After data collection, the next stage is data preprocessing, which ensures that the dataset is clean and suitable for machine learning analysis. In this stage, data cleaning techniques are applied to remove duplicate records, handle missing values, and correct inconsistent data entries. Additionally, normalization and feature selection techniques are used to scale the data and select the most relevant attributes related to student knowledge and learning ability. This step improves the efficiency and accuracy of the machine learning models by reducing unnecessary or irrelevant data.

Following preprocessing, the processed dataset is passed to the machine learning model training module. In this stage, various machine learning algorithms such as Logistic Regression, Random Forest, Gradient Boosting, or other classification models are applied to learn patterns from historical student data. The dataset is typically divided into training and testing subsets to train the models and evaluate their performance. The algorithms analyze the relationship between knowledge attributes, learning capabilities, and academic performance to build predictive models.

The next component of the architecture is the

prediction and evaluation module. In this stage, the trained machine learning models are tested using unseen data to evaluate their predictive accuracy and reliability. Evaluation metrics such as accuracy, precision, recall, and F1-score are used to assess the effectiveness of the models. The system identifies the best-performing algorithm that can accurately predict student academic outcomes based on the given input features.

Finally, the result visualization and decision support module presents the predicted outcomes and analytical insights to educators or academic administrators. The results may include predicted student performance levels, identification of at-risk students, and analysis of key factors affecting academic success. This information enables educators to implement timely interventions, personalized learning strategies, and academic support programs to improve student performance and overall educational outcomes.

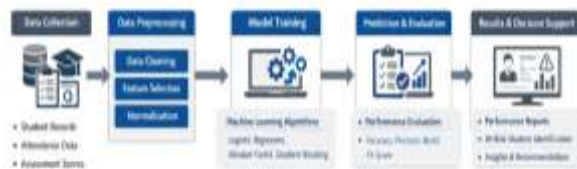


Fig 5.1: Structure of the Proposed System

VI. IMPLEMENTATION

```
df_num = df_port.select_dtypes(include = ['float64', 'int64']).drop(columns=['fname', 'lname', 'target'])
df_num.head()
```

	age	Medu	Fedu	traveltime	studytime	failures	lamedu1	lamedu2	lamedu3	lamedu4	lamedu5	lamedu6	lamedu7	lamedu8	lamedu9	lamedu10	lamedu11	lamedu12	lamedu13	lamedu14	lamedu15	lamedu16	lamedu17	lamedu18	lamedu19	lamedu20	lamedu21	lamedu22	lamedu23	lamedu24	lamedu25	lamedu26	lamedu27	lamedu28	lamedu29	lamedu30	lamedu31	lamedu32	lamedu33	lamedu34	lamedu35	lamedu36	lamedu37	lamedu38	lamedu39	lamedu40	lamedu41	lamedu42	lamedu43	lamedu44	lamedu45	lamedu46	lamedu47	lamedu48	lamedu49	lamedu50	lamedu51	lamedu52	lamedu53	lamedu54	lamedu55	lamedu56	lamedu57	lamedu58	lamedu59	lamedu60	lamedu61	lamedu62	lamedu63	lamedu64	lamedu65	lamedu66	lamedu67	lamedu68	lamedu69	lamedu70	lamedu71	lamedu72	lamedu73	lamedu74	lamedu75	lamedu76	lamedu77	lamedu78	lamedu79	lamedu80	lamedu81	lamedu82	lamedu83	lamedu84	lamedu85	lamedu86	lamedu87	lamedu88	lamedu89	lamedu90	lamedu91	lamedu92	lamedu93	lamedu94	lamedu95	lamedu96	lamedu97	lamedu98	lamedu99	lamedu100
0	18	3	2	1	1	1	2	1	1	2	2	5	0	0																																																																																												
1	19	1	1	2	1	1	4	1	2	1	3	5	0	0																																																																																												
2	18	1	1	1	1	2	2	1	5	2	5	4	0	0																																																																																												
3	18	2	2	2	1	1	5	5	5	1	1	3	0	0																																																																																												
4	17	4	2	1	2	0	5	5	5	1	3	5	0	0																																																																																												

Fig 6.1: Dataset Loading

```
In [28]: from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, random_state=42)

# Show the results of the split
print ("Training set has {} samples.".format(X_train.shape[0]))
print ("Testing set has {} samples.".format(X_test.shape[0]))

print ("Accuracy rate of the training set: {:.2f}%".format(100 * (y_train == y_train).mean()))
print ("Accuracy rate of the testing set: {:.2f}%".format(100 * (y_test == y_test).mean()))

Training set has 3039 samples.
Testing set has 995 samples.
Accuracy rate of the training set: 2.72%
Accuracy rate of the testing set: 2.82%
```

Fig 6.2: Dataset Preprocessing

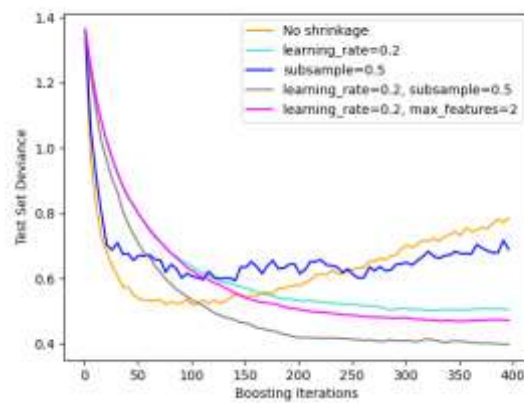


Fig 6.3: Model Training

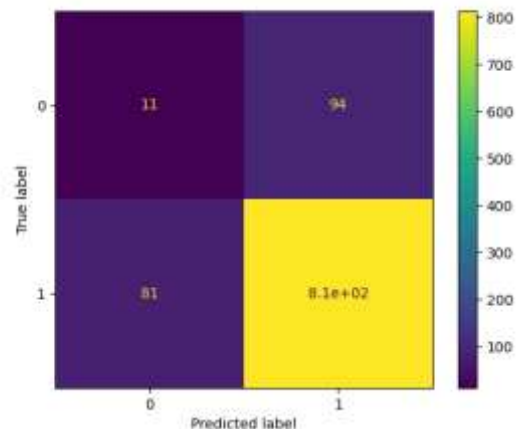


Fig 6.4: Model Evaluation and Accuracy Results

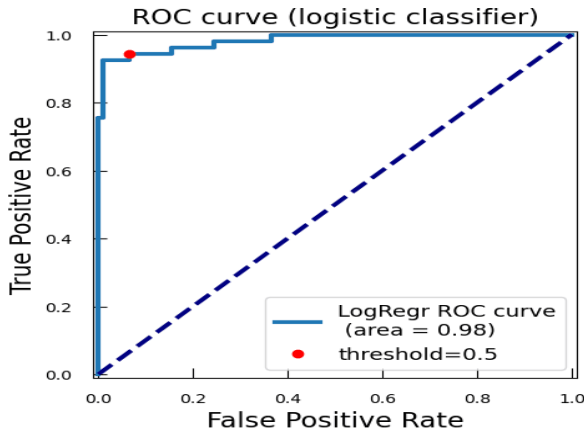


Fig 6.5: Prediction Output For Student Performance

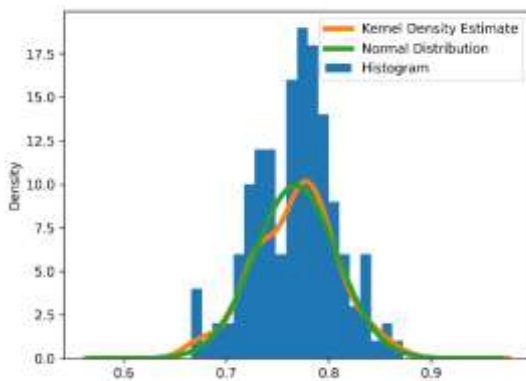


Fig 6.6: Visualization of Student Performance
Analysis

VII. CONCLUSION

This study presented a machine learning-based approach for analyzing and predicting student academic performance using knowledge attributes and learning capabilities. The proposed system focuses on identifying important factors that influence student learning outcomes by analyzing various academic and behavioral attributes. By integrating educational data with machine learning techniques, the system is able to identify patterns and relationships between student knowledge characteristics, learning behaviors, and academic performance.

The implementation of machine learning algorithms enables the system to process large educational datasets efficiently and generate accurate predictions of student academic outcomes. Through data preprocessing, feature selection, and model training, the system can effectively evaluate different factors that contribute to student success. The experimental results demonstrate that machine learning models can significantly improve the accuracy of predicting student performance compared to traditional evaluation methods.

Furthermore, the proposed system can support educators and academic institutions in identifying students who may be at risk of poor academic performance at an early stage. By providing predictive insights and analytical reports, the system helps instructors design personalized learning strategies and targeted interventions to improve student learning outcomes. Overall, the integration of machine learning in educational analytics provides a powerful tool for enhancing teaching effectiveness, supporting data-driven decision-making, and improving the overall quality of education.

VIII. FUTURE SCOPE

The proposed machine learning approach for predicting student academic performance can be further enhanced by integrating more advanced techniques and larger educational datasets. In future work, deep learning models such as Artificial Neural Networks (ANN), Convolutional Neural Networks (CNN), or Long Short-Term Memory (LSTM) networks can be applied to capture more complex learning patterns and improve prediction accuracy. These advanced models can analyze large-scale educational data more effectively and provide deeper insights into student learning behaviors and academic performance.

Another possible improvement is the integration of real-time data from learning management systems, online educational platforms, and student activity tracking tools. By collecting real-time information such as online participation, assignment submission patterns, and interaction with digital learning



resources, the system can provide continuous monitoring of student performance. This would allow educators to detect learning difficulties earlier and provide timely support to students who require additional academic assistance.

In addition, the system can be extended to develop personalized learning recommendation systems that guide students toward suitable study materials, learning strategies, and academic resources based on their individual performance and learning abilities. Future research can also focus on incorporating psychological and behavioral factors such as motivation, engagement, and learning styles to create a more comprehensive predictive model. By integrating these enhancements, the system can evolve into a more intelligent educational analytics platform that supports adaptive learning environments and improves overall educational outcomes.

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